.of: 166.4 One impt idea is that I have (via MCT) many modes of lrng possl.,
So I can take an advanced part of a TSQ, — decide what info it needs to precede it, then get that info in to TM — in any of a variety of ways.

RE: "T. Discovery of t. laws of Algebra" This turned out to be diffl.! Perhaps it is because t. laws of Alg. were invented rather than "Discovered"? No! T. operations of Alg. were defined before t. Laws were descd. I think t. discovery was then (perhaps) that t laws could be extrapolated to Literals.

A nice way to do a TSQ: First write an ordered list of problems & concs I want TM to know. Then devise a machine that can "do" that list. Each part of t. list can be done in t. most convenient way" — w. attempts to do it via lrng. Then I begin to replace "expert-system-like" parts w. true lrng. As part of t. list", I do include impt. concs. So even after TM has acquired t. info in a rather AH. way, it can learn rather well!

One TSQ direction (ANL) Conc. of "Quantity" & simple linear eq. soln. TM owes a more deftly in untangling eqns in 1 unk.
2 linear eqns in 2 unks; then → M eqns in n unks,
also any quadratic, cubic, quartic sequences. Then simple heuristic.
Also, very complicated eqns in 1 or 2 unks. Must be solvable as linear or quadratics or . . . .

.24 So: Get QA (Quantity?) meaning clear: lots of disparate examples: → Also Note: 173.10–.16
QA(3) = 3? answer 3. : QA(7+5) = 12

.26 If X=3 Then QA(X)=3. Note 2 meanings of ff, Then:
1) .26 c 2) If X=3 Then Y←Y+1 : Give them distinct notations
QA(3×(7+3)) = 33 : Learn about parens; Also how related to stack.

Then: If X+3=5 Then X = 5−3 = 8. (Soln of simple eqns).
If (X+3)×5 = 50 Then X = 50/5 − 3 = 7 (more complex eqns)
Laws of Alg? If X=Y AND Y=Z Then X=Z
How to teach this. Meanings of AND, ON, NOT.

~~.34~~ If X=3 Then Not X=6 ; If X=3 AND Y=7 Then X+Y = 3+7 = 10 →168.21
→ .24 — ~~.34~~ should keep me busy for quite a while!
[common notation = is ← assignment and == is equality]

So first limit to t. TSQ, (perhaps telling how many examples of each kind of eqn.) What I want here is an intuitive idea of what t. TSQ is supposed to convey.

Next, do a detailed defn. of what each concept "means". Its properties, — The meaning of any expressn. containing it.

.02 To deal w. roundoff error: for linear equs only, use for rational fractions instead of floating pt. (ratios of integers). TM has builtin routine for removing common factors of Numerator or denominator.

.05 — But this will not work for irrationals. — So maybe use idea of approximate equality. → (.18)

.07 "Discovery of Laws of Alg." was dif'lt to formulate for TM because arith w. large nos. had so little cc for it. We can, hvr, reproduce condition of Human discovering the laws, by ad-hoc making cc of arith >> cc of simple logic of the "laws of Alg." [see 169.37 for remark on this]

.12 For Humans, large cc of arith of large nos., was a [Heuristic]! — we have

.13 to simulate this heuristic for TM! But may be no need! → (.23)

An analogous situation: In Polya an impt heuristic "state the problem exactly": For TM, this can't be done. The problem is stated exactly already! For TM the heur is to "state the problem in its vague, original form, in which associations w. other ideas are expressed in the form of the statement.

.18 (.05) Re: .02–.05: At advanced state, we can let TM use floating pt. & "discover"

.19 that if 2 nos are very close they are "probably equal". → (.33)

.21: 167.34: An additional concept: "Truth": T(3=33/11) = 1 (true)

.22 T(3=5) = 0 (false). → (.36)

.23 (.13) If we use floating pt. in 486 or Pentium, arith takes ~100 times as long as simple logical reasoning! Hvr, it might be well to simulate Mult & Div taking much longer than Add, subtract: In floating pt, they take about same time: If we don't do floating pt, Add, subtract are as fast as Logical statements, but Mult/div takes longer.

If +, − takes as long as logic, then $x+y=y+x$ would not be discovered.

Hvr, in SAARB, I found a way for TM to Discover laws of Alg, because

.31 they were useful in solving Equations. This is a somewhat difrnt

.32 path from Human Discovery of the laws.

.33: (.19): Well, in most cases if we use floating pt and $|A-B| < \frac{10^{-5}}{A}$, then A=B.

. "will work "almost always": for even fewer errors use double precision & smaller threshold.

.35 (.33R) is probably good enuf for most early work.

.36 (.22) The 167.24–.34 would be an ext Alg. (perhaps w. True( ) added) ← 168.21
for me to begin to teach English, it would be easy to extend that algebra

.38 in many ways: quadratic, cubic, quartics; e^x, ln x, complex nos, trig. and h trig., & trig identity proving. Also several equs in several unks linear/non-linear; Matrices, Vectors, etc.

Anyway: for time being, assume that I can deal w. equality (168.02–.06, .18, .19, .33–.35) and assignments of arby cc. to various Arith. Operations (168.07–.13; .23–.32)

.03 [I want to avoid that much detail, at this pt in my Analysis] → 180.24

Ok. so start TSQ!

Def: "Q" mean "Q." mean "Quantity of" = "evaluation of". Normally it is a number: (real or complex): whether I want it to be able to be true, false or a string (that mite be t. output of an algm) is not at present, clear.

So Q(4) = 4 ∴ 4 is correct reply to "Q(4) = ?"

Initialy TM knows numbers v.s. non-numbers (= strings).

Defs { Q(4) = 4 has 6 symbols; 2 are nos, 4 are non-nos. Hence forth "=" means logical equality "←" is an assignment statement }

Q(4+5) = 20, Q(5*8) = 40, Q(9−3) = 6, Q(7÷3) = 2.$\dot{3}$, etc.

We may want to teach one operation at a time or a mix.

With Q(4+5) = 20 TM learns that Q(4+5) = ? has answer 4, 5, add. (op code) This is obtained by taking t. 2 numbers in "Q(4+5) = ■" and adding them. T. rest of t. symbols are irrelevant.

Then Q(8−2) = 6 .... Q(4−7) = ? TM says "11" since 4+7 = 11,

This is, hvr, wrong. After Q(8−2)=6, hvr, it is clear that simply adding t. nos. in "Q(8−2) =" will not always work. After a few examples of "−", or one example of hy precision, TM finds that Q(8−2) = gives 10 a fraction of time and 6 another fraction. This is t. best model, until we do more search and find whether its "8, 2 sum" or "8, 2 sub" depends on whether a "+" or "−" occurred in t. data of t. corpus.

(Th. detaild mechanics of this decision is as yet, undecided (unclear)).

After Q(8±2) is learned w. 100% accuracy (: very short code)

We try Q(8*2) = 16 & ~ data: at first TM looks at the codes representing "−", "+" "*": If "*" is most similar to "−" code, it will give hy pc to 8*2, sub than to 8, 2, add. Hvr, as we give more examples of "*" (a/o more precision), TM looks for a better code and finds that t. symbol * makes 8, 2 mult t. correct response.

Similarly, it finally lrns Q(8/2) = 4.

.32 One thing is unclear: Just how much cc TM spends on these srches. After being give Q(8+2) etc, Q(7−3) etc., it could stay with 50% accuracy. Perhaps I can arby give it a Threshold of p = .9 or .1; or p = .99 or .01 for its predictions.

.35 T. .99, .01 can be obtained w. (a 5+2 or more precision (no. of bits) → 171.31

.36 in examples of corpus. This seems to solve it for Algebra & much of Math! → 170.01

.37:168.12 [SN] Cheapness of Ram & disc storage make for a negative heuristic for psrchs w.r.t. compression techniques. Cheapness of hy speed computers is neg. heuristic for learning efficient use of machines.

8/3/00 Bulg

(.37) Negative Case Counts

(.06) TSQs & possl. Extensions of them

.01: 172.37 "oneway" connection form A to B could be a ckt. w. A as input, that "drives" B in a way that a cap of size C would drive B from A. To simulate t. cap acting from B to A, we would need another ckt of this kind. The value "C" of t. cap. is a voltage input of.

.05 simulating ckt.

.06: 172.17 : 167.24 — .34 (173.10) is initial TSQ [redacted] Some other relevant notes, extensions, additions:

167.34 — 168.05 ; .18 — .19, .33 — .34 || 168.07 || 168.21 — 22

.08 Extensions: 168.36 — 169.03, 180.24, 18?.5

.10 : 167.24 Another possl part of early TSQ:

3 = 3 ; 3 = ? x = 3 ; x = ? 3 = ? [y = x, x = 3, y = 3 { y = x, x = 7, y = ?

.12 Would it quickly learn x = y, z = y, w = z, x = ? ; w = ? i.e. a long string of equalities. [Advanced Algebra (2, 3, n unknowns linear; Matrix algebra) quad ... / ex dnx, complex nos, trig identity proving, → 180.34

Also I may want to introduce t. "If, Then" logical notation. (not control If, Then)

What is another good name for it.? Cond imp.

.16 Condition, implies. Cond, Imp. / Cond means "for this problem only".

Hvr, Just what is "for this problem only"? How much can be transferred (learned) to other situations?

A real possy, that Ordinary Human induction (even in a pure math setting) involves very many assumptions (≈ "Common Sense"). While it may be quite important & very pervasive, it may be possl to List the assumptions made.

e.g. in Cond x = 3 [redacted], y = x imp y = 3

we carry x = 3, y = x into t. imp side of t. cond, imp. —

[~~Future do not carry~~ We then want TM to carry the transitivity of =, that was lrnd here, into future problems, but not carry x = 3, y = x etc into future probs.

How serious a problem is this? It must be "clear" (or inducable) "By induction" i.e.

.28 ~~from~~ from t (corpus) (examples) alone. → 174.04

Each induction problem has to be defined either by CB or by compresn Expected. — Otherwise TM doesn't know when to stop searching.

On OSL: A very Common use of OSL is in Heuristics — say one successfully worked a problem a certain way in t. past. This is about t. same as "Case Based (Learning) Reasoning". The practical aspect is indexing events / processes, so they can be retrieved for (CBR / OSL)

.37 SN On Negative ssz: using t. D.F. $p^A(1-p)^B$ as a prior. is equivt to ssz of A + B: This D.F. w. A &/o B < φ could be meaningful. It would mean concentration of pc at φ &/o at 1 (Not unusual: Particularly in MoR). → 174.01

8/9/00 Bulg

.01: 178.40 Another kind of "Hint": Hard or soft (= probabilistic) restrictions on set of concs a/o methods of combination of concs used to solve a problem. An extreme case would restrict TM to search for a known (to teacher) soln.

In general the value of TM's srch for a soln. ~~to TM~~, is reduced (sometimes considerably) by the giving of "Hints" — so when we give hints, we should try to minimize this Bad aspect of ~~[illegible]~~ the hints.

Also possl. Soln. is MAD (Mutually Assured Destruction). Each opponent is unsure that he can completely destroy all all adversarys, so it he tries & fails, the remains of the Adversary(s) will try to destroy him. — In this case, best concentrate on defense & development of more powerful ~~[illegible]~~ weapons; (~~= Arms/race~~) = (offense/defense) race (Arms offense/defense) race.

[margin: Mrs. written Remarks — "Crap" — copy to "Crap".]

A search is "Of Value" if (1) It discovers/new useful heurs/concs. (2) or it changes params of old concs. (3) Discovery of solns to probs that are different from those obtaind when Hint(s) were usd. (This is maybe a subclass of (1))

.17 "Am. Anthony" I don't expect ~~[illegible]~~ searches will be at all "interesting":
.18 TM will simply find concs that are compositions of other concs. Later, I expect that TM will watch (another person or itself) solve a problem ~~[illegible]~~ & induce (by OSL or ~~[illegible]~~ Multishot lrng) a way to do a search or a
.20 set of rules or heurs for ~~[illegible]~~ solving probs of that type.

3 Big Complaints I had about the TSQ's designed at SAARB:
.22 1) Hard to write (MCT is supposd to fix that!) (3) It was beginning to "scale" badly; pc's of concs were getting too small as TM had more of them; Poor organization on TM's part!
2) It only found concs that were solns to problems in the TSQ.

This second complaint is (partly at least) dealt w. in (.17). I need more complex probs before I begin to notice useful sub-trees in solns. — also I need more complex probs before TM begins to discover stuff like .18 - .20.

A big thing would be for TM to discover what a "Proof" is! Ramanujan's "education" George Shoobridge Carr top of 176
And, of course, if TM could usefully go thru Carr's book(s) (or just part of one!) it would be a Major Break thru.

.37 List of Recent elementary TSQ refs: 178.09 - .27 (176.21 ff. is "how TM works"); 175.28 - .40
174.30 - .40; 173.10 - .16 (includes advanced stuff); 173.06 - .08 (gives refs from 167.24 & 169.03)

[margin: characteristic operator v.s. unordered set of finite objects]

From .37 ff I should make a list of (conc, example sets) and ~~[illegible]~~ Try to order them in a reasonable way. I have been concerned w. the exact details — then DON'T BE. Just write the TSQ & put TM to work on it. TM should be able to work it even if it's not exactly rite! I may discover some very serious deficiency (180.01)

8/10/00 Bolg

T! ~~[illegible]~~

.01: 179.40 in t. TSQ's! But O.K.: Then maybe fix it. ← (But Perhaps Generalize it first — like .11 ff)

.02 → Also, I want to Address t. "scaling" problem (#3 of 179.22R) | ← (A very Serious Problem)

(.02) involves finding ways to categorize conc., ways to know when they are likely to be used. I should be able to get ideas about This by examining my own workings of t. probs. in t. TSQ.

.06 So make a list of actual problems in 179.37: Hvr, do it on a text editor, so I can insert lines. In Reso notes, write

.08 comments on t. Set of Trg. Examples.

T. Q of Operator v.s. Unordered set of sets seems impt: Think about it. ? [ABcde]

.11 Otherwise, Just write t. TSQ and worry about .02, when it occures.

But when it occures: Try to get a good perspective of t. Problem — "IN ENGLISH" (ab/root) — But look at it in t. most GENERAL Possl. way.

T. way it looks Now: That t. details of t. TSQ are not very

.16 impt. — That it does, hvr, have to have t. needed info in it. If it

.17 does not, this will become apparent in t. mode of failure of TM's attempts to predict/"teach" it.

Hvr, the "scaling" problem of .02 can be serious.

Also note that ~~[illegible]~~ if we run into .16 – .17 (could not conf info in corpus), t. MCT should give us/ several ways to put ~~[illegible]~~ t. needed info into t. corpus!

.24: 169.02 (SN) on Extending t. TSQ! ~~x~~ x^2, $x^{\frac{1}{2}}$, $x^{\frac{3}{10}}$, x^{real}

exp(x); ln x; hyperbolic trig functs., (Laplace xfm?)

complex nos. " " ~~t.~~ (" ") (F xfm)

So before I introduce complex nos, hyperbolic trig & Laplace xfm are definable.

(Also for Laplace, I need to define Integrals ←? — maybe not — use real analog of Fourier Series.) ↳ The for both Laplace & F xfms There are "Σ" rather than "∫" versions

⟶ Also note: Solns of polynomials can be extended via complex nos, This does not need e^x or ln x. So maybe introduce complex nos after "polynom. soln". Then e^x, ln x etc.

.34; 173.12 ⟶ Q: Could I Get TM to invent Complex nos, etc. ? This is in T. direction of t. "creative" Mi. Research (Conat's "AM") → (176.01 –.13) (I want Ramanujan)

Perhaps Think about How TM could invent Complex nos: What would be Motivation & Heuristics to direct it to To best known way to do Complex nos.? Any other way(s) t. idea of a "soln" of a poly, can be extended?

8/10/00 Bulg.

→ one way I dealt w. my personal "scaling" problem: when I ran a new, useful conc. (= idea/o category), I try to think of all the areas of inquiry in which it could be useful. This could be very narrow area or very broad problem area. This process of suggesting applic areas [redacted] is solvable by Inductive Inference. → [redacted] Infant TM has too small SSZ for induction in this area — But mature TM does it w/o the "scaling" problem ↳ 183.07

.01 (80.90): Re: transcendental funcs: There is a nice, fast computer way of calculating them all (& ≈ the Cordic Method). Another (or perhaps the same) way, generates e^x by $e^{x+a} \approx e^x(1+a)$ loops are replaced by expansion of binary nos. Multiplication & division are done by a kind of successive approx used in A to D converters.

.06 More on "Scaling": The scaling problem is one aspect of "Too large CJS": In the past, I'd dealt w/ this by PC of Concs by factoring them, then using TSQ to discover the factors & make them of by PC. However, even this technique will fail for "Scaling" reasons — i.e. if we don't find ways to categorize, to narrow down the search for a conc (the "narrowing down" should be (stochastic) ~~Stat~~ Soft rather than Hard. [redacted] = deterministic)

.12 However, it may be OK for a young TM, there is not enough SSZ for useful

.13 categories, to do this.

.14 Note that this "Narrowing Down the Search Space" (cf. .10–.12) is one of the commonest types of Heurs. — But for a young TM, we will probably have to insert the needed heurs. — because (.13–.14) TM will not have enough experience to discern the relevant heurs. (Circ.)

.19 — So the "Scaling Problem" may be equivt. to "Having Adequate Heurs" → Adequate soln. to scaling problem implies that the CJS at most problems stays about the same size as the corpus grows

So start of TSQ: file c:\[redacted] (on Zenith laptop) see 181.5 TSQ1

167.24–.34 I had idea that the Conc. of "Quantity" would be useful. but I'm not so sure of this now! — It may be that TM could run these TSQ's "about as well" w.o. that "conc." Part of my objection to it was that I don't know exactly how I want to define it.

Qa(3)=3? : Cond X=3 imp Q2(X)=3 But also, "X=3" So I'm not sure what Q2(.) is about! Usually, if not always, if "Qa( )" are removed, the expression is still true (a wff? worse): so it looks like it's redundant. — The redundancy can have heur values. This/one seems vacuous! (tho I'm not really sure).

Anyway: [redacted] l, m, n are say 32 random bit nos.

HVR: Many of the properties of =, +, − are discoverable by TM, empirically.

.33 The way I'd considered doing this (most recently) was to have +, −, ×, ÷ be very expensive variables to logical reasoning.

In the present type pgm "TSQ1." TM "learns" "laws of Alg" but the learning is of a "Shinnarian" kind (small CJs, very little cleverness)

.37 In .33 one way to do this is to have TM try to find ways to do arithmetic "fast". Probably, it would be a good idea to try Both (.33, .37) and the method of "TSQ1" (191.5)

182.01

8.10.00 Bolg.

TSQ1

181.5

This will be an ordered list of problems, tasks, definitions, ...
--- toward the construction of an initial TSQ for TM.
167.21-.34
l, m, n are 32 bit random numbers (or IEEE 80bit reals ?) OBASIC 32 Bits per single prec. 64 " " double precision.
u, v, x, y, z are variables
":" separates examples. "," separates data within an example
cond means "conditions for this problem"
imp means "what is implied by these conditions"
n=n e.g. 3=3, 7=7
cond x=n imp n=x
"[" and "]" are "metasymbols" [3+5] means "8"
l+m=[l+m] e.g. 4+5=9 --- learning Addition 4+5 = ?
[l+m]=l+m e.g. 9=4+5 --- Equality Commutes ? = 4+5
cond x=y imp y=x Equality Commutes
l+m=[m+l] Addition Commutes
x+y=y+x Addition Commutes
x+m=m+x Addition Commutes
l+(m+n)=(l+m)+n Addition is Associative
cond x=m imp x+n=m+n --- if equals are added to equals the sums are equal
cond x=y, u=v imp x+u=y+v --- as in previous example
l-m=[l-m] learning Subtraction
m-m=0 meaning of Zero
x-x=0
cond x=m imp x-n=m-n -- if equals subtracted from equals, remainders are equal

7/26/00 Bulg [redacted] ALP (.01) Convergence of $\ln p'$ to $\ln p$

Optim Techniques: (08) 1205 : Jack

.01 In 578 I should ~ $E \sum (p_T - p)^2$ was < minimum $\ln \frac{p}{p_T}$

r is max value of ratio of

.03 A Q was what about $E \cong \left(\ln \frac{p}{p_{true}}\right)^2$?

.04 Actually, [redacted] 578T3 proof showed $E \cong \ln \frac{p}{p_{true}} < \ln \frac{p}{p_T}$

which was stronger than .01. The .04 is not interesting (apparently)

.06 comparable to .03, if does suggest that $\ln \frac{p}{p_T} \to 0$ rapidly.

ff 155.30 ff on ordering of trials for (OT (02)) probs.

.08 152.24 ff On finding peaks (Max) of a function. Assume a certain value

.09 for ϵ: $(f(x+\Delta) - f(x))/\Delta < \epsilon$. (or by Δ)

Using this ϵ value, do a search w. mesh edge δ. Find f_{max} for that Mesh. Because of .09, certain regions of the mesh, we know can't have pts w. $f > f_{max}$

.12 Do search w. mesh edge $\frac{\delta}{2}$. Over regions of not for which f could be $\geq f_{max}$. This $\frac{\delta}{2}$ search yields new f_{max} (maybe = old f_{max}) and gives (perhaps) new regions where f can be $\geq f_{max}$.

To do the search: Make a "linked list" of regions in which $f \geq f_{max}$ can occur: Each such region has a Δ value and 2 end values f_1 & f_2;

Also, peak poss'l = $\max(f_1, f_2) + \frac{1}{2}\Delta\epsilon$: We elim regions from our linked list when that peak poss'l becomes < latest f_{max} found.

So, we do the whole space until all pts are covered that could have $f \geq f_{max}$.

Then we do $\epsilon \to \frac{1}{2}\epsilon$, & we redo the whole search.

We re search for new ϵ then $\epsilon \to 2\epsilon$ & redo the system.

.23 This continues until we find no new peaks w. $\epsilon \to 2\epsilon$. (.32)

.24 MORE GENERALLY: R.E. OT's: An OT is impossible over a continuous space unless one makes some constraints on $G(x)$. Otherwise $G(x)$ can be anything at any point and info about some pts. doesn't [redacted] constrain the value at other pts., so no search method is poss'l. (other than random or exhaustive search (at hyperlinear rate)).

In Approximating Functions $f(x)$ using linear sums of other functions: It is well to know the constraints on $f(x)$ & design the "Basis" functions. This is true for Basis for those constraints

.31 Neural Nets approximate as well.

.32 : (.23) A weaker assumption than $f(x+\Delta) - f(x) < \Delta\epsilon$: Constraint on ≈ "second drive" of f. So if ΔF has been large in an adjacent mesh region, we allow ΔF to be $\epsilon\Delta$ larger than that in the present Mesh.

"Second Drive" may have to be "Second Difference": The details of how to use a "Linked list" (or some similar data str.) would have to be worked out

PSG dizzy

.01: 149.40 This defines R_s ≡ R. Symbol set that precedes S.

Next: Define the Symbol set that follows R_s (R will be a nonterminal).

Call it T_F : So we can make the concat product set $R_s \frown T_F$.

So, we have fully ways to create (recursively) nonsets:

1) All prim symbols are nonsets w. 1 member.
2) T. set of symbols (or nonsets) that precede or follow a gen. nonset.
3) The "OR" of 2 nonsets.

Note! These "nonsets" are really "BAGS". So it's easy to "OR" them but "AND" has no obvious def'n. (I wrote about this in a past)

Maybe see notes on G. Wolff.

To define α_F (fr. set of symbols that follow symbol α):
we write αF (F a special symbol) before coding the corpus: every time α occurs, we add the following symbol as a separate Barrouilli seq. If any method of coding the corpus, then α_F is a legit nonset.

How we define & evaluate R_s (fr. symbol set that precedes precedes R_s is unclear

---

SN I do want to find ways to effectively "undo" part of the Grad. Grammar. Otherwise, I will simply accumulate errors. Could "Reparsing" be of help?

PSG rules are of 2 types only:

A = C ⁀ D	concatenation of 2 T = concat of 2 NT's.
A ⇒ C D E → FG	NT = Boolean AND of several prior NTS or OR of several concats.

So: A Grad could be a small (recursive) set of Grammar rules.

A mut/recursive (ici crossover) could remove Grammar rules as well as add or modify them.

A loop A → C D A → D · A · E is same as any other Gramm rule.

Hvr. Every loop/NT should have at least one production that is not in that loop!

eg. A ⇒ = A · D
E

But in general when there are many loops, fr. nasty constraints (to assure meaningfulness), are complex.

What I want to know rules so I can avoid making trial modifns (Grads) that are meaningless.

IN GENERAL! Say we have obtained a target

LATER! Read 149.15 - .40, 161.01 ff. When Starthagen PSG dizzy

I think there are some Good ideas here: Some weird developments here:

I vaguely want fr. idea, hvr, that PSG dizzy may not be critical to TM or TM's (regs. of lang users) (since it will not (run comp. by looking for regs in a large corpus: It will learn by associating English Qs, A's w. things in a Domain it knows About.

.01 : 159.40 : 153.01 ..... 157.19 Discusses ways to get Rosse 2 kinds of approxns of G(x) &/or P(G,x).

154.01 : On relation of Mutations/Crossover to t P(G,x) & G(x) approxns.

157.24 – 157.28 : That t. Treatment of "Peak Finding" of 152.13 – 156.05 Doesn't do "Experiments" : How to deal w. This (a bit).

157.28 – 158.06 Suggests possl. Good approach to "EXPERIMENTS" in t. treatment of 152.13–156.05

160.01–.06 ON Sol 78 TZ : How it can deal w. χ $(\ln \frac{p'}{p})^2$ error criterion !

161.01–.40 PSG discovery : (Continuation 149.15–.40)

Review of Reviewers. T. pp of reviewers: 140, 141, 148, 158, 159, 162.

140 details from 117.21 to ~139.40.

159, 162 " " 142 to 161

141 Discusses several Models for G(X). (easy to find K ⇒ G(K) is χ Max.)

148 Gives an allover view of Motivation : Why this stuff is impt.

158.13 : A repeat of χ 148 : Perhaps slightly different.

.01 Now, I want to outline a ≈ final TM., based on MCT. is (probly perhaps) Lsrch.
First a rough outline, then progressively more detail.

T. main idea is to do a T.S. of human starting w. Algebra. Using various techniques to reduce CJS. Mainly [illegible] definitions by induction: then giving problems in which those defns. are useful.
[Also, we want TM to be able to devise complex (perhaps recursive) concepts to solve problems. [look at Koza's methods for some ideas on how to do this - (it is not obvious!)]

After TM. understands some Algebra, begin teaching English, by asking Q's in English & expecting Answers (in TM's normal mode of thought, — but later, in English). See how far we can go w. this — & see if TM can work hard probs (eventually). Since Koza seems to be able to work hard probs, using a rather unsophisticated search system (i.e. it doesn't learn heuristics to guide search), I think that TM should be able to do far better.

.16 A VERY IMPT. fundamental idea: That I want TM to learn things the way I think humans do. — So if there is some reasonable, good, way to solve a problem, perhaps using heuristics based on experience of TM &/or the Scientific Community — I want TM to learn to use those heuristics. — ideally, it should learn them from its own experience — but if the CJS is to happen, I will give it "hints" to the CJS.
— The default "hint" will be "opening" TM is arranging for it to "discover" the desired conc. I will probably be giving lots of "hints", but I expect that eventually TM will be able to discover some serious, large CJS concs, "on its own".

At first, TM will solve all induction problems by simple Lsrch. However, as TM matures, it should be able to do "proofs" & "finding Solns" that it does by inducing from examples. This induction is done by Lsrch, then the Algms (found by Lsrch) are used to solve the problems.
— A simple case would be Soln. of linear eqns. — or perhaps numerical &/or literal evaln. of an alg. expression.
→ Perhaps Look at latest version of the "Paul-Solomonoff" paper for some useful ideas.

Re: Alg Notation Lang: A new idea was to teach the idea of "Quantity" — which makes ALN a lot easier, & it's a conc. that is very useful later.

Perhaps Review Search Notes.

ANOTHER impt idea: Be sure to RETAIN PERSPECTIVE. Don't get bogged down in picky details. Keep discussn. in BROAD, "English" terms as long as possl.

GA

.01: 162.40 On speeding up GA: Paper in Recent IEST on Evolutionary Computing / E.C. July 2000 p. 188
Uses "Momentum" concept from Back Prop in ANN. to speed up GA srch: In the cases they report "The speedups enormous, — but of course they present only the cases
.04 where it worked. Still it would seem like a v. good idea. → .09
They suggest that other heuristics from Back prop. be used for GA. Perhaps Partial derivative idea?

Look at Thesis

Pogio wrote a paper showing how ANN, radial Basis Functions, Fuzzy (logic controllers) were all very similar. Also paper by Farmer (?) on similar theme.

.09: .04 Re: "Momentum" This is essentially a derivative of G. They multiply it by a constant, μ, to obtain next jump size. The constant μ is chosen by user. If they had second derivatives they could do optimum jump. However second derivs are very noisy. One has to pool data to get more: 3 kinds of "pooling" ① Over time of same vector component of $\vec{x}$ ($\vec{x}$ is the description of the candidate) ② over different components of $\vec{x}$. ③ Between $\vec{x}$ components in near by or more distant regions of $\vec{x}$ space. (We are doing a STEINish analysis).

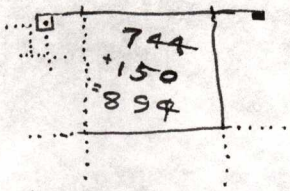
Also Two books on GA Theory: ① Michael D. VOSE M.VOSE@M.VOSE One reviewed in recent July 2000 p 191 IEEE Trans on Evol. Comp. (by reviewer) was referred to. ② Hans-Georg Beyer Springer V. 2000

Also look at a book I have called "Searching w. Probabilities" Andrew J. PALAY 1985
↳ This uses Chess as Example Source of Problems.

172.19-173.05

185.04

7/30/00 Bulg.



#64 is a "Speeding up GA" See coarser Bulg files

.01 : 163.40 E.G. : I want TM to learn meanings of various (terms definitions) in Algebra. In many cases, exact definitions of what I want, are quite diff't. ("Set" is a good example : even "set of numbers")
In these cases, I could just use ruff definitions: give some reasonable examples, w.o. my really understanding how to complete defn. could be understood by T.M.

So: TM would at first have very narrow ideas of what a "set" was, & would later have to "Generalize" this concept. The simplest way to Gen. is to remove constraints; But this is certainly not the only way ◎.

I might just give a few examples of a concept, then see how
.11 T.M. has done on it by asking about new ~~cases~~ cases. — Then I could see if it (T.M.) made any errors in the concept. This could go back and forth w.th errors by TM. Hrr., if it does go "back & forth" many times, I'm afraid of Getting close to "Skinnerian Irng" — w. T.M.
.15 ending up w. a long defn. w. many "special cases".

What To do: I do the "Bad" Irng of .11 - .15 : This teaches me how my examples are inadequate. I then mind-wipe TM of that part of the TSQ (if this is possible!) → Then I could stage TM's mind before I do this section of the TSQ. — So easy to erase subseq. [redacted] "malthing". ← ("mal-training"?)

I want to keep my defns "in English" so they are as "vague" (= general) as poss'l. In the case of "set", it may be hard to define it w.o. using the concept itself! — So maybe it is a "Primative Concept".
So — if I run into trouble devising an "ENGLISH" definition, Consider the possibility that it may be a Primative concept!

Another Impt. idea is that I needn't be very careful about introducing concepts that may have to be discarded or grossly modified, later. TM should be able to deal w. this, & not "get stuck" / somewhere by "Painting itself into a Corner" ◎.

.32 So what are some prelimy concepts / defns? — What are some prelimy probs?
First: Quantity: ≡ [redacted] Q_3 [redacted] 3. $Q_3(3)$ is 3
[redacted] $Q_3(3 \times 7)$ is 21.

If X = 3 Then $Q_3(X)$ is 3. The If, Then idea

The If gives a scrnd universe. "Then" gives some Conclusions about it.
The idea of Parentheses. Also idea of a "Stack" T.M. should Im to (166.01

~~Stack~~ stack **STACK** THE PRIME HEURISTIC → .21

.01:165.40 relate a stack to parenths.

If x = 3 Then 4 × x = 12

If 5 × x = 12 then x = 12 ÷ 5.

Perhaps just try teaching numbers, + but perhaps − .) Then,

~~[illegible]~~ Q2(3) is 3. If x = 3 then Q2(x) = 3

~~[illegible]~~ Then eventually: If x + 1 = 5 then x = 5 − 1 = 4.

.07 Maybe use "=?" as way to ask question. This is a reasonable induction question form.

So list some ~~[illegible]~~ Concs I'd want TM to recognize;

Some kinds of Behavior & kinds of lrngs.

Q2: The idea of "Quantity" or "evaluation" I'm not entirely sure what I mean!

The idea of Parenthesis & perhaps related [stack] ← This could be very impt! very useful in ordering tasks.

165.32 — 166.07.

[redacted] Ability to solve simple eqns, then more complex eqns. After eqns w. +, − are solved, ×, ÷ are introduced & more complex eqns are solvd.

Another trick: TM learns to solve x + 3 = 7;

then we teach it mult, div. & see if it learns to solve ~~[illegible]~~ x · 18 = 13 faster. Then it lrns to solve x + 3 = 7. The Q is: does it recognize t. similarity? (A variety of "(short lrng")

.21 The **PRIME HEURISTIC**: "All t. info is in t. p.d. ⟷ ~~L~~ srch is t. best way to solve t. problem" (quick sort)

■ Apparent Exceptions (like Quick sort) can be dealt w. by properly defining t. problem. In this ↑ case, one redefines t. problem to be ~~[illegible]~~ lo cc solns of problems (not necly max pc/cc). Hmm, Note that Most INV probs. are solved as OZ probs.! ~~a mount~~.

MCT makes it relatively easy to put any kind of info into t. P.D.

.27 Perhaps: First write fair ammount of ruff t. S.Q. Do this before working out any "details". Stuff like 165.32 ff, but more "complete" Something close to what I expect to feed to TM.

For .27 ff: In later (or even perhaps earlier) sections of t. TSQ, Tell just what Heurs are expected to be used (if any!).

Look at those 3 or 4 TSQ's I worked on in SAARB! To what extent Can I use them or (parts of / extrapolns of) them?

Promising Approach! Write TSQ's for an Advanced TM, first. — Then work backward to get to Primitive Concs, a/o use many "Hints" to ↓ CJs of ~~Advanced~~ solns. to "Advanced" problems.

Sol 89 had some reasonable ideas about TSQ's: perhaps reread it. I had diffnt "Phases" of t. expected TSQ — culminating in Ability to read/understand English text books.

167.01

.01: 169,36 ! So, next, Q(3*(4+2)) = 18. "feel"

The idea of introducing Q( ) notation was to give TM a "feel" for "Quantity" — but my success, in that ex. examples up to this point, would not do that, so ex. concept of "Quantity" would not yet be used as a theorem in (.01)

So try this: "If X=3 Then Q(X)=3."

If X=Q(5) Then X=5

Q: Do I want to give occasional negative cases? A neg case means that ex. code should not give my pc for those examples. — T. idea of neg cases may have attrition wherever we are fitting "concepts" (in Valiant's sense) to data. Winston used it in his "Arch" v.s. "not an Arch" paper. Also Grammars to [redacted] compl.

A neg ex example is necessary for [redacted] pure Math. A neg. example is a

.12 pos. ex case of "Not (ex. concept)". Pos & Neg cases have interesting property with → (172,03) Maybe getting too complicated!

Maybe getting too complicated → Well, T(Q(7,7)=1) = [redacted] (Truth) of (Q(7,7)=1) is "false".) ← Value

.15 4*Q(4+5) = 36 : So Actually "Q" acts like parenthesis.

in (.15 ↙) → omitting ex. ommitting ex. Q would give correct expression.

What I've been trying to do is get TM to understand what a "Quantity" is.

We say X=3+5 what this "really" means is that Q(X) = Q(3+5)

So: X is a symbol: Q(X) is its "value" 3+5 is a symbol. Q(3+5), its value.

Would things be easier in RPN? T. system is certainly less ad-H. than normal Alg. notation.

∑ An advanced problem would be for TM to learn to X/X from RPN to normal Alg to Polish notations.

Remember: T. pt. of all of this was to reduce ex CJS -(≡ cc/pc = L_cost^-1) of learning Alg-notation. I feel that Quantity was an impt conc. that was more generally useful: — say in explaining to TM what "solving an eq." means. — And this is why I want to "go over" a large part of ex. TSQ, to see when certain cases are important in many places of ex TSQ

.32 Perhaps "Built-in" concepts: T. idea of "localization" ≡ [Subnet] subnet

Another possy is that "idea of Quantity" is inherently difficult to teach.

— ~~Perhaps~~ So — (.32) — need to "build it in" somehow.

Or, I could just write out ex. soln. that I want TM to acquire & go on from there. — In fact, it would be well for me to write a T.S. Q., Write ex. series of like TM to acquire, Then look at ex. CJS's involved, Then see what I can do to reduce them.

What I much want is the CJS that seems Small to a Human,

is also Smallite TM.

{{ Thus it could be that the reason certain CJS's are small for a Human is that he has other Ing instances that are not immediately apparent.

Re: ANL: Say I used RPN or / consistently w. parens. (like LISP): If TM learns #(3,4), *(7,5) etc, then the idea that sets is parts of strings (subtrees) could have been much used in the past, have hype for future use! This might be an adequate lexor for ANL. That the idea of "Quantity" might be as basic as the "idea of" sub-tree. — which will (probably) be a built-in heuristic.

SN If I use 1 example w. hy res. v.s. many examples of low res, the pc of discovery of the concept will end up the same: But The pc of discovery of the conc will be the same .... but for combining w. other concs! — The fact that a given conc. was used many times in the past (even w. low res on examples) is important — it would seem to give it higher pc v.s. use w. 1 hy res Esp. examples in the past.

Actually ANL can be expressed as a stack operator.
f '4,3 → add 4,3. ; (the mechanics of this is unclear!)
Anyway, after it acquires the operators for +(4,3), it will be able to acquire the op for combinations by combining sub-operators.

It may be that in my early approach, I used a common "Memory" for concs, & so I did not really get from proper pc's —
— In particular I did not have Recon conditional pcs (which functioned so as to narrow down Recon expected applicas & ↓ Recon pc's in a ga. applicn. (re idea of "directories")

Hwr., I'd like to avoid that much detail — but I do want to be sure I get pc's of conc's right, so they are indeed much more (theory to be used as parms for candidate Concn

.31; 169.35 1) On Re theorem 10% or 1% pc of error of 169.32–.36 : In general, our ability to get 1% certainty in predn. will depend on 552 of examples used for new Conc.

.34 2) In discovering new concs.: A (Human-type level) "soln" is not complete until I have some way to narrow down the conc, that were used in the new cand conc. This can (perhaps usually will) involve categorizations of concs, and for each category, conditions in which it is likely to be relevant. This (category/condition) set will be a heuristic & it will operate in a Bernoulli seq — (like manner.) → The utility of a (i.e. the way I coded a Bern seq.)

8/3/00 Bvlg

MAIN METHODOLOGY for TSQ WRITING .11 (also see 171.34)

.01: 171.40 Theorem of this kind will depend on it's efficacy in pct of past
.02 solns. of induction problems. → .11

.03: 170.12 : Each ~~any~~ thing must be one or the other but not both. These are
"category" concs — The kind defined by Computational Complexity Theory ( ).
Now These concs are e.g. Number, v.s. not a number, even v.s. not even, odd, etc.

[I use, here, normally, a more General def. of a conc. It is a tree or subtree. It is a part or whole of a computer pgm. [It is a string of op-codes.

Give special names to concs. That partition & increase entropy parts
Def like this: Call them P concs. (PARTITIONING concs) — They must
do > 2 partitions. [A "P conc". may correspond to an "OB"] eg. OB, OP Algebra.

.10: .02 — So: Re hours! I want to first design a TSQ, & I have for
each problem set a Human heuristic soln. — So it seems reasonable to
that a Human could make the necessary epistemological jump — so it seems
like the CTS for Humans is acceptable/reasonable. Then I have
the problem of converting (apparent) human hours into TM-type hours. If
I have trouble w. this, it is clear that I don't completely understand
.17 6. Human-hour! → 173.06

Put work to this above end of work on GA

.19: 164.40 SN On GA: Design of filters/electronic chts could be much speeded
up by using RW. components (RW chts) for Caudy.
This is because (A) Spice is very slow (B) RW can implement things Spice wouldn't
be able to think of. Putting components together could be done by recheys
at scaled-down speeds. For faster trials, we would have to find
electronic ways to simulate hoza's trials.

NB At present, there is research afoot to design devices components
that can easily be switched from being a Computer say to a
fm radio, to a GPS system, etc. — find out how they works — it would
be great for GA work!

Another idea in this from is use of Analog signals
on digital components ∴ GA's w. analog output but digital
switching.

Another possy is to use GA for Analog inputs as well as
digital inputs to circuits.

It might be expensive to have Variable L C R, in chts,
but we can Make analog chts that convert from voltage to
L, C, R. connecting a C, say, of such a type
betw. 2 pts, can be not so easy!

.37 — Maybe not so difficult! Say we want a cap betw. pts A & B. → 173.01

.01: 173.40: Hvr, if $A < \phi$ then $A \to A+1$ will $\downarrow$ pc of $p^A(1-p)^B$ · which is wizard [$\frac{p^{A+1}}{p^A} = p$]

Hvr, it's true even if $A > \phi$! Perhaps it's wiggy to Norma.

.04: 173.28: 173.10–.28 seems to be a real dirty — in a sense that I will often have an induction problem in Mind, so I think I know what it is, & I know what answer I want — But in fact, I really haven't thot out just what the constraints on the problem are.

This is Like the "Robot's Dilemma" : the "FRAME Problem" in A.I.

When Given a new problem in a sequence, the Robot doesn't know what features have Changed & what ones have remained the same.

In 173.10–.28 the Q is: "Just what are the constraints in each new problem?" If may be that this can be answered by a more GLOBAL Induction. — But "am sanguine" There isn't much corpus to make "Global Induction" on. — So I have to be careful starting out:

Possl. troubles: ① I don't put in vital constraints so it fails: ② I inadvertantly put in extra info, so it works, but it contains seeds of future failure.

.20 ANother Way to write (Neuman-ish) TSQ's: Make up a set of Problems: For each Problem & separate

.21 write a fancy apparatus needed as well as the search procedure.

If prime diddy, (W.20) it that we will not have any idea of how unsure & what assoc. pc's, Bayes are for the other (not "correct") concs., so we can't have estimates for Cost of the expected "soln." ⇒ On the Other hand, we could make as part of the "Problem soln" various heads to the pc's (for that particular problem) of the concs used in the soln. — This base is certainly an important part of the soln. of any (largish) Problem.

→ Hvr, even w.r.t. (.245) .20 ≈ .21 would be a useful way to organize a Tsq. 10:16 17 18

.31 Some small TNG Tasks:

1) Equality: $3=3$; ~~cond $3=3$ imp $3=3$. cond $3=3$ imp $3=4$ low pc.~~

$3=3$ hype $3=4$ low pc

2) Introduce of cond, imp formalism: See 178.09 for discn. cond($3=x$ / $x=3$) imp($3=x$ / $x=3$) "cond, imp" diagram

cond $x=3$, $y=x$ imp $y=3$ { other problems involving transitivity of "=" }

3) $3+4=7$ // cond $x=3$ imp $x+4=7$

4) $4-1=3$ // (cond $x+3=8$ imp $x=8-3$) ← Down-ward move TNG better this?

~~also cond $x=3+4$ imp~~ $3+4=\phi$ $3-3=\phi$ $3+\phi=3$ ∴ $\phi+3=3$ ⇒⇒⇒

Local v.s. Global ideas

175:01

.01 : 174.40 : Possl. Notation note: L, M, N are [redacted] hyperprecision (32 bits say) random nos.
— So t. single exemplar L=L is equivt to t. exemplar set $[L_i = L_i]$ i=1 (2^{16}) (for L_i's 32 bit nos)

.02 For realization of a Lrng Algm; TM can take 2 general forms: ① In one it hunts for (low cost) operators → w. input string "N = ?", t. output will be "N."
② In t. second, It tries to find various Nos for "?" that will give entire corpus hy pc.
— #2 is, perhaps, t. more general soln. to t. problem. Usually, #1, t. operator

.06 form, is insensitive to regularitys in t. "input" [redacted] part of t. corpus.

For t. N=N corpus, t. N=? problem has one simple soln: Look at input string: find first number: this will be output.

For #2, we find regularitys in input: This could be useful for predicting [redacted] long term trends in t. types of problems being Given. — it could be used to guess at types of future problems that will be given — up to the future Limit of the "Horizon". This could help in t. long-term optimization of TM's Behavior.

What are t. MAIN IMMEDIATE PROBLEMS?

1) Choice betw. ① & ② Model types in .02 – .06
2) Choice of what parts of 174.31 – .40 to use, to start on.
3) Fear that Choice of ② would get me "stuck" in an "Non-English Level" Aspect of TM. — That I would be getting "Too Specific" for this phase of TSQ writing.

Perhaps for t. Probs of 174.31 – .40: For each problem, write out t. kind of behavior I'd like to be elicited from each ~~Arithmetic~~ Material Concept introduced.

In Some Cases, I don't know just which properties I want to introduce "at t. beginning": e.g. Commutativity of Addition? (probably)

.28 $x+x=2x$ (probably not: This involves [$x+x = 1\cdot x+1\cdot x = (1+1)x = 2x$.] ← .28R
If TM learns x+x this way, will it really be able to "count" objects? Perhaps I could introduce t. counting process (later or earlier or in a "separate stream") & then have it somehow integrated w. t. idea of (.28R).

For "counting" TM could do "j←j+1" every time a new "x object" was found.

Also, I could introduce Multiplication via $3+3 = 2\times 3$, $2\times 3+3 = 4\times 3$, no counting

I want T.M. to know all of these aspects of arithmetic operators.

i.e. to get sum of a bunch of 3's: count the no. of 3's, then 3× t. counter no.
For this trick to save any CC, "adding 1" must have << cc than "adding 3". We can use "adding N" where N is large binary no.

8/6/00 BUG

.21 HOW TM CODES NEW DATA
(steady state learning)

Ramanujan: Synopsis of Elementary Results in Pure and Applied Mathematics. 2 vols (1880-1886) 6,000 Forms By George Shoobridge Carr (Maybe Camb. Univ. Press) up to 1860

.01 On the other hand perhaps TM (as a "curious" scientist) should get a lot of emotional excitement from discovering these facts, rules, laws about Arithmetic. — A Ramanujan-like Goal! Srinivasa Ramanujan 1887-1920

.04 Evaluating the values "of these discoveries" is diff't! Lenat's "AM" did (hvr, in a not unreasonable way). Its evaln. rules [for "interestingness"] would develop as it had hours for developing them.

A TM w. many (eg. TSQ would itself discover heuristic Rules for telling whether a conc. was "interesting" or not. For an infant TM, hvr, we will have to insert some of these heuristic Rules

.10 "am anfang."

Q: How far can TM go w.o. hearing Lenat's hours of .04 ÷ 10? I'd sort of like to make a "Minimal TM" to

.13 start off.

I could just pick a reasonable no. of concs int. corpus & see

with a suitable TSQ, I could use GA to find a population of low K cost. As new examples appeared, the population would evolve to track the new data (as well as better fit the old data). This might be a good way to start TM.

.21 → Re: ↑ The way I envision TM normally working: Say it's looking for Operators (I think it's about the same for Nous' "operator" inductions) that map input problem into output soln. We start, doing an L-srch for an operator that solves past problems well. When we augment the corpus, I expect that small/trivial changes will be made in the defn of the operator (≡ "soln."). This is essentially pure mutation.

.26 [Re: .21 In coding Sequential Corpi, when the corpus is Augmented,
.27 normally, the code is simply augmented w. minimal search needed.
.28 (Hvr, occasionally, when a new sacrifice of the corpus comes in, a large portion pairs (sometimes large) of the code must be rewritten. BACKTRACKING This can be done by using parallel codes that were not the shortest for the "past" but were recent corpus augmentations. (This is an aspect of "Theory Revision").
.31 A real technique is OSC. OSC is usually simpler than other kinds of "Theory revision"
.32 In .26-.27 we add New Definitions, but in .28-.32 (including OSC) we add new defns a/o delete/modify old ones.

deletion, revision, addition of "Definitions"

A common type of Theory Revision, (of which OSC is an Extreme Case) occurs when the SSZ for a given definition in the existing corpus is too small to warrant its definition, but is increased by recent augmentation of the corpus ↑ the SSZ for new data, so it becomes Legit.

.39 Another kind of simple "Theory Revision" occurs when we notice

177.01

.01 176.40 regularities in t. Code of t. corpus (so we write a shorter code for t. code). ← A/O we may break code down (variable) into smaller pieces & try to represent it in a better (hyper PC) way. previous.

.02 An indication that we need "Serious Theory Revision" — That t. new corpus has many parts that cannot be predicted w. t. accuracy that we had obtaining previous corpus. (— i.e. Our old theory doesn't seem to be working w. t. new data). ← This problem can occur in t. very beginning of t. TSR! see (205.08) very vague criterion!

.07 In such a case, (if it is possl.) we should (try to) arrange to get more of t. aberrant-type data (= "experiment"), to ↑ its ssz. — also try to get more diversity in that data. → (.207)

In 176.28 ff I talk about (re) coding a "new section" (or new augmentation) of t. corpus. The corpus hyr, may not be so "sequencially" organized, so a "t. new section of t. corpus" could include "all problems involving triangles" rather than just "t. last 1000 bytes of t. corpus". Slightly more generally "All parts of t. corpus that used a particular defn. or set of defns". T. code for those parts

.15 has to be revised.

176.21 ff (& .26 ff in particular) Is a VERY HIGH LEVEL "ENGLISH" defn. of TM's operation! While it's fairly good, it probly doesn't extend very far into t. Varieties of TM behavior. Hyr 176.21 ff is very IMPT! — I do want to Continue it.

Varieties of Coding techniques:

.20 (.07) Also note .02 — .07: Before trying to get new data relevant to t. new phenomena: — We look at t. old corpus for "sections"/"parts" that are usefully similar to t. new intractable subcorpus. — This is usually not an easy thing to do.

Note: 176.21 ff is on Coding a Corpus (= [Induction; Prediction] (Presumably Sequences) but I should think a corpus of unordered finite objects would not change t. discussn. much (if any!).

.26 Hyr, TM's other usual probs are OZ probs & INV probs. INV problems can be solved via Lsrch (see Principle 166.21). Hyr, note that most INV probs are solved as OZ probs.

T. MCT does link Lsrch (probabilistic) to Optimization (≈ OZ probs): Enables one (in Theory) to give probly that a gn. O.T. will do best, w. a gn. OZ problem. — So Lsrch can be done on OZ probs in a "practical" way.

Hyr., Note that t. MCT treatment of OZ is entirely Empirical: It looks at the G results of each OT on whatever corpus was used & from this it predicts t. likely goodness of any one of those OT's giving a max G of t. given OZ problem w. t. stated CC. limit. MCT does not look at the structure of each OT to determine t. feasibility of each OT for t. given OZ problem. So MCT's analysis would not suggest ways to design new, good O.T.'s Perhaps "TM$_2$" would do it?

178.01 to some extent But not much!

187.25 space

8/7/00 Bugs:

$(1+i)^2 = 2i$ co ; Imp

ucsos.javanet.com/~Bob mere co : imp

Logical Negativism

Illogical Negativism

ElmoreElevators

So, ritanow, Major Problem areas:

1) Worry about the Learning Algm needed for General INV, 02 probs

177.26 -.40 discusses this abt. T. General Process of 176.21 seems Reasonable.

2) Concern about the epist. details of the TSQ: What do I expect F.M. to assume about the examples in the corpus. The "global" v.s. "local" variables, concs, defined by "cond, imp" for each problem: but there may be ambiguity. .09

.06 operators broadcard set of finite strings.

3) Which models of induction: {175.02 or .06} to use.

.09 .06 In the cond probs w. {cond, Imp}, we want certain parts of the Irng. to be global, yet certain info applies to one problem only. (= "local")

In mathical langs like Maple, Mathematica, MacCode — there are tricks they use to make things unambiguous.

Equality: $3=3$ $4=?$ | cond $X=3$ imp $3=X$ (Symm) | cond $X=Y$, $Y=Z$ imp $X=Z$ (trans)

Addition: $3+4=4+3$; (Assoc? $(3+4)+5 = 3+(4+5)$) (Absorbed?)

Subtraction: $3+\phi=3$; $X+\phi=\phi+X=X$; $3-3=\phi$; $X-X=0$ cond $X=3$ imp $X-3=\phi$ $3-X=\phi$

$3-5 \neq 5-3$ $3-5=-2$; $5-3=2$; $2 \neq -2$. This idea of $2, -2$ is unclear. $N \neq -N$ unless $N=\phi$.

Well: I can teach these relatively A.M. things about the fund. concs of Algebra! Will TM be able to induce/deduce certain of these interesting things)

Perhaps the main goal of the Learning of Algebra is to enable co. for TM to acquire a complex enuf domain of knowledge, so that I can get it to learn usefully Irn to understand English S's about that domain, & get it to generalize. English to other domains.

.27 Also, to bother to deal w. "English" that has errors in it.

.28 Perhaps write down a large bunch of things about =, +, ¬(=), imp), ×, that I want TM to Learn; see if I can order the examples so it can learn in a useful way.

Then, try to see if TM can Irn. it. If there is trouble, return to .28

Don't get too deeply involved. At present, the problem is mainly one of adequacy of the info in the TSQ, and not on just how TM "Irns".

So try juggling around/backtracking Irng Sets & modifns of the Irng Algm, & other ideas for Irng.

Re: T. "Irng Algm": One imp. part of it tells what parts of the corpus to test when a Modification of the Coding algm. is made.

One kind of "Hint" could help TM with this problem.

(A "Hint" is any kind of info that U CDs for the "Student".)

.01: 181.40 : I should outline a bit more on 181.33, .37! Some sample problems.

Cond $m \overset{*}{+} n = [n \overset{*}{+} m]$ imp $n \overset{*}{+} m = ?$

$m - m = \phi$; $m \div m = 1$!

$m + \phi = m$: $m - \phi = m$; $m * 1 = m$: $m \div 1 = m$

After these have been learned; TM should be able to quickly "learn" to work

$m + n - n = m$; $m * n / m = m$ (in last, there is trouble if $m = 0$!)!

Associative law could save lots of time (w. commutative law)!

.08 $(m+n) - m = (n+m) - m = n + (m-m) = n$.

.09 Hvr, TM has to find ways to do this reasoning fast otherwise it doesn't "pay" —

.10 — (Say TM had to do a large srch to find ribo xforms in .08.) → .20

One hour trick: If one has a seq. of "+" or "*" operations, ommit t. parens, because one can put them back in an arb. way.

Also: a string of $(\div a)$; $(*b)$ type operations can be rearranged in arb. order because of commutation & assoc.

Similarly w. a string of $(+a)$, $(-b)$ type operations.

Some rules, hvr: [It may be necy to insert a leading ϕ in an $+, -$ string
" " " " " " " " " 1 " " $*, \div$ string

$3 + 4 - 7 - 3 = \phi - 7 + 4 - 3 + 3 = 0 - 7 + 4 = 4 - 7$ } "calculator
$3 * 4 / 7 / 3 = 1/7 * 4/3 * 3 = 1/7 * 4 = 4/7$. } Arithmetic".

.20: .10 so this sort of thing motivates TM to find fast ways to do searches

.21 & any other logical processes.

.23 A Third way to get TM to discover "Laws of Alg" would be in t. direction of what I did in SAARB on this problem: That T. laws are useful in solving (linear) equs. I may be able to do this better, now, using MCT!

.27 So 3 Ways (1) T. method of c:\TSQ1 (2) 182.01 - .21 (speed in calcn) (3) 182.23 (use in solving linear equs.)

(N.B.) In solving equs, like $x + 51 = 43$, TM could simply try integers to find x; But if we make addition expensive, it will be motivated to find faster ways to solve linear Eqns

.32 **HEURISTICS** can be a fact (usually probilistic), technique devn, ... it can be computed locally (devized for a particular problem): it may involve much "logical" reasoning to derive a/o apply it in a particular situation. OSL could be important: Since t. solution in t. past that is relevant may have only occurd Once!

For OSL (& other applicas) (indexing / Categorizn) is very impt. When an event occures, we want to be able to recall it as soon as possl. when an associated (perhaps relevant) event occures.

.01 This idea of "association indexing" is very simple, but there are probly very many distant methods used in Human "indexing" for (relevant) recall. One very impl. way that TM "Matures" is in its learning of many new important, original, idiosyncratic methods to categorize situations / events / problem-solving methods for retrieval.

In General, everytime one solves a problem, or gets some impl. info from an experience (this can be during the soln. of a problem), one indexes that event in many ways. "Many" is impt. because (a) I want it to be accessed in many distant "similarity dimensions" (b) If the present / new event is similar in exact "dimensions" to an event in the past, it will have exact "code bits", "describing bits" in common w. that event to "qualify" for OSL. (Also see top of P181)

How to discover these "categories": When I am working on problems that TM is working on, observe the associations that I make & try to discover categories that would help in these (as well as other) useful associations.

Its not clear that the forgg. stuff will be very useful for current TM — & early TS Q's. I could just assume larger IPC available: Say 10^{12} ops/sec. Later, when the "scaling problem" becomes serious, I will certainly need "categories" to narrow down search.

.18 So: This is an impt. idea: Categories are hard to devise for early TM work, but will not be very fancy. Later, tho, there will be very many choices of concs to combine & categories (& other heurs) will be needed.... but they should be available (from data a/o introspection) at that time.

Another Aspect of Categories & Heurs: Whenever they are used, they not only block such trains, but they narrow the Prob. distribn. for the Answer. For MTM this may not be critical — but in NMTM it can be very useful!

The Category/heur idea is Very Impt. in CREATIVE solns to Probs. A good set of categories/heurs will enable rapid solns. of otherwise unsolvable probs. Hvr, it will ↓ diversity (unless it is > e.g. set of heurs). A novice researcher will have fewer heurs & could have more diversity in search. Search will take much longer (than w "Good" heurs), but could yield a very "Creative", "Unexpected" soln. — could be much better than the more rapid heur. search.

The Methods of Conventional (not Super GA) GAs may be an example of a slow, by diversity srch, that can find v.G. solns.

[7.6.2] ff is a reasonable way for TM to Operate for Induction problems. For a seq. of INV. problems, we solve early probs. w. simple srch, using our Approach that we insert into our Universal Reference Machine. As we solve more problems, we add constructive concs, to the (UMC/pd) & we change pc. of older concs. Occasionally we do. New problems are usually involve relatively small incremental Modifns. of the pd. Occasionally, like (84.01

8-12-00 Bulg.

1867
(Stuff on GA:
(Mutation & Recombination)
A "Creativity"

.01: 183.40 on 176.28, we will have to backtrack", and redo the recent string of concs. in a better way. Task: what Could, we decide to do this, & just how we do it, is unclear. 176.39 ff has some ideas on "Theory Revision". 177.02 - .15 is of particular interest — If can deal w. "Non-Sequenced" data

.06 On 176.21 ff I have this sort of "System" by which I am working:

.07 Contrast this w. MCT analyses, in which I use at all times a "Conditional" Pd — with each new problem/item being the "condition" on the pd to be used in Looking Next problem.

Well, MCT ~~is~~ isn't really disturbed by .07: MCT simply Gives a way to ~~do~~ take a corpus that consists of several kinds of probs. (Induction, Inv, Oz) & obtain a pd from that corpus. That can be used to get probs/pd's for any aspect of that corpus. .... I need to get the Q of .06 resolved!

I think MCT & .06 are really "orthogonal" They are on different parts of the TM problem, & they don't really interact. MCT ~~[illegible]~~ tells what needs to be coded — what Code need to be mixed. .06 tells a heuristic way for minimizing that Code. /// MCT defines the problem; .06 tells an approach.

.15 Way to solve the problem..

.21 On the Q of whether to Model using ① a "Stock operator" or use the ② "Finite unordered set" model. Actually, the 2nd model has the same final form as the first, but uses more info — it uses info about regs & in & out put set to the stock operators. It may be that in both cases we end up w. a stock operator, but that the Gen's ~~[illegible]~~ over the universe of

.26 Such operators are different in the 2 cases.

.01: 184.40 On t. "Learning" part of TM!

Look at TSP(1): (18.5)!

NEW (a set of numbers w "=" connecting them)

The problem is to make a "program" of the primitive concs, so that

.05 input n_1 = Give n as output.

[I will describ. what TM does in English; Then later isolate out the primitives needed.]

.08 (1) In t. case of .05 a soln. w. to be: — Since output is a number, t. final operator must have a no. as output. (useful in "backward chaining")

The only way to get n (output) is from n (input), so t. soln is to take t. number in t. input & present it as output. So t. pgm picks t. no. on t. input & transfers it to output. (or transmits it as output)

.13 (2) try $3.7 + 4.1 = 7.8 \cdots$ ect. a long seq. of examples (like this), or a few by precision crossovers.

Again output is no. so final operator must map to nos, +, −, ×, ÷ or transfers.

Transfers don't work: + − × ÷ all have 2 inputs — num. inputs, 1 num output.

try all 4 possys: $m+n = [m+n]$; $m+n = [n+m]$ both work o.k.

.17 ? ⇒ Hvr. .08 no longer works on t. new examples. (what is meaning of this remark?) → 204.24

(SN) QS: ① How long would it take to Irn these Rules w.o. t. Hours?

② Under what circumstances would it Irn these Hours (itself)?

When I put a Hour into TM: Imagine a TSP that would result in TM (learning / learning) that Hour, & put it into TM in t. form that it would have Irn'd it.

I think this will keep TM's structure very simple! Easy to pgm, Easy to Understand (Easy to Debug. ?)

.25 : (176.21 / 177.40) A kind of Irng Model I've had in mind (in linacw. 176.21 – 177.40)

This "concept net": No feed back in net; No loops in net, but "loops" are a type of Functional that is used to combine functions to obtain a new function.

A Net of this kind is probly limited to Primitive Rec. funcs. — This is not ordinarily a serious limitation (I Think!)

[176.05 — Probly Total Rec. Funcs w.o. diffty]

The language assigns pc's to subnets of functions. At any time, the system is a Bernoulli seq. of t. functions (≡ primitives + subnets that have been learned)

.32 a Composition of these functions.

.33 We also have a Class of this sort for not "operators" but "observers"

The combin of t. 2 is t. ob-op algebra. An ob is a function w. output of T/F or a Pd on T/F (= Boolian).

ob: Boolian, Numeric or String input; Boolian output.

op: " " " " : Boolian or numeric output (no Boolian output).

ob, ob → ob; op, op → op; ob, op → op

The Boolian output controls which of several funcs is to be used.

I think one of the main functions of the ob-op Algebra was to get Comb Procs — to deal w. t. "Scaling" problem! 228.36

.11 187.40: This "Selection" function of t. obs. is mindful of certain primitive functions used in Rec. func. Theory. — So it may be possl. to combine t. ob. ops into something near "Universal", — But more important — t. "Algebra" may be adequate to express all of t. kinds of "regularities" in a corpus (including Hours) that I want to express.

T. way induction is done using t. Rec. def: We normally try operators in pc order (w. of. the "loop" functional). Occasionally we introduce function that involves several repeatings of a function: We then try to find a lower pc version of it using t. "Loop functional"

.10 The loop functional is a combination of a ob and an op. The op is used repeatedly; the ob looks at some result of t. op & tells it. repeat or when to stop. The op is usually a vector function: w. 2 components.

.13 One component is used for stop criterion. This is an "Until" loop. [Also, one need initialization of both vector components initialized

One very common component of the vector argument is initialized with x=0; and updated with x ← x+1. The threshold for termination of t. loop is n & t. loop terminates when x = n (an "ob" function).

.17 It is probly possl. to devise functionals that could define functions beyond Prim Rec. Functs, using a formalism ~ to .10→13, but I will not Go into this just now.

So, using t. 2-141 model for assigning pc's to new ones; (the details of coding the subtrees will have to be worked out) and t. ideas of 187.25 – 188.17: Is this enuf for a more or less complete model of "practical learning"? [It solves (or is supposed to solve t. "Symbolic Regression" problem

What it does seem to do is give a model for learning (functions/operators /stochastic operators). This is a Very Large area of induction: it means that we show TM examples of desired I/O behavior & it extrapolates,

.27 probabilistically,

.28 → 189.01 space

its use in solving CNU or OZ problems. It could "Learn" to do so. Say we (or bet all Env. problems; then gave "solns" (e.g. traces) for solving them. From many examples of this kind & "building up": (starting w. simple env. problems then building up toward more difficult Env probs.)], TM builds up a good Pd to use for t. standard (Strict Soln of Env problems.

Mixed Complexity Thm.

Similarly OZ prob. soln. could be taught.

On t. other hand, once TM can do "induction" (i.e. 187.25 – 188.17) TM can use MCT to compute updates to cpd's needed to for such as Env. & OZ probs.

(SN): Perhaps I/NOW have a "Critical Mass" of ideas for TM: That for any TM sub-problem that arrises, I have some ideas (that I've developed in some detail) that we could try out to solve that sub-problem.

189.01

4:32 pm optimum start.

188.40

.01: 188.27 So; ok: Then — so does Z141 + 187.25–188.17 give adequate model for learning stochastic Operators? (≡ functions). ∵ Consider also, lrng. of Neurs.

[Dropping, for t. Moment, t. Q of (or t. details of!) 184.21–.26]

One problem is t. Ordering of Function Trials for Lsrch : See 130.19 ff on how to count t. no of operators using n functions of a difnt type, w. the i^{th} function type having $c(i)$ inputs (& 1 output) | 136.21 tells how to construct them in BASIC

One diffy is that there will be many duplications of functions in the listing of 130.25ff! (see 131.11 ff for eqns). This is because the list of functions to be used, will contain compositions of some of the primitive functs. If we list them in pc order, the hyer pc versions (that come first) will contain t. composed form of t. primitives. — So we will get dups, but later in t. listing. — This is t. usual diffy of Lsrch being controlled by (≈ K cost) rather than p cost. My impression is that usually this doesn't ↑ cost of Lsrch by very much. (perhaps a tollerable factor of debility!)

"Symbolic Regression".

In general, the trial operator will not contain more than 3 or 4 functions, because t. pc ↓ so fast & as t. no. of functs considered ↑ !

To order t. Macro functions! Assign to each funct, a pc =

.17 $\frac{pc_i}{d^{c(i)}}$! pc_i is t pc of drawing t i^{th} function. d is t no. of inputs to t. Macro function

$c(i)$ is t. no of inputs of t i^{th} (micro) function.

Actually, t. factor $(d+n)^{c(i)}$ should be used, n being the order in which this micro function is chosen — but assume that $d \gg n$ (n will be 1 or 2 or 3) and usually d will be > that. (I Guess!)

Anyway, use .17 for the effective pc of t. i^{th} function; Then order the listings of t functions via Huffman (142.01ff)

Normally, there will be a long list of micro functions to be combined. Also "Normally" there will be heuristics that Modify t. pc's of t. μfunctions, in line w. t. "Nature of t. present problem".

So! T. induction process for "Symbolic Regression" consists of; trying to find the "Best" Macrofunction to fit t. data! We do this by first assigning pc's to t. component functs to be combined; we then order strings of them wrt. $\frac{pc_i}{d^{c(i)}}(.17)$ using t. Huffman Code technique. [The pc's of t. trial strings are first Modified by any available heuristics.]

.30

We test t. ordered strings of .30 using Lsrch.

We finally find a good fit:

The Function grows along w. t. corpus — just as t. data in t. "Procorpus" grow in Z141. Hvr, t. difference is, that in t case of "Symbolic Regression" t. only Functions that ever get defined are ones that are, for a certain length of t. corpus, complete Macro functions for that chunk of corpus.

I ran into this trouble w. t. TSQ's I wrote in SAARB, but 190.01

8.18.00 Bldg

ABCDE

.01: 189.40 I thot that this was a fault of the TSQ's! — But it was not (at least not entirely) because the Induction model used Implied this "Fault"!

One diffy seems to be, that there is only one Macro function: So only 1 "conc net." If a particular ~~[illegible]~~ primative or defined funct is used several times in the net, I think it automatically get higher pc. Hvr, it may be possl. to examine the Macro funct & find subtrees that are used ≥ 1 time, so it pays to define them. This amounts to a Recoding of the Macro funct (≡ conc. net)

I'm not sure the conc net & the Macro funct are really the same. They are not identical, but the ordering of concs & of function definitions have (1 to 1) ↔ correspondence.

.14 A vague idea: that somehow the pc of a micro funct should be related to how often it is used in coding the corpus.

Another idea: That the pc's of micro functs should be related to the frequency with which they were successful ~~in deriv~~ in helping to devise (temporarily) successful Macro Functs.

Another "idea": we have obs that switch ops on & off. Perhaps the fraction of the time that a funct is switched on is csmpt.

I feel uneasy about the "statistics" of the components of a small Macro funct. In a stoch PSG, the frequency with which various paths are used, determines their pc's. Hvr, in MTM, this may present a problem. None-the-less, even in MTM, I will have a SSZ for each "decision" — even if it's always "like" (13/0) (0/10).

Also, in linear regressn, the values & precision of the coefs are critically dependent on SSZ.

Hvr, the functional forms for linear (or nonlinear) regressn: what is the SSZ for these? Well one chooses a functional form for each time series predicted, so SSZ = no. of Time Series predicted

What is SSZ of "components" that help derive functional forms for regression? Say we used many difnt functional forms for regressn. Then the SSZ of the components of these functional forms, would be clearly defined. Hvr, suppose we ended up with 5 difnt forms for regression & we used them all the time hereafter. Then the max SSZ ~~of~~ for any component function would be ~5 only!

So: Consider LARGE, MCT-type, system, w. many difnt kinds of problems / modes of behavior {Inv, OZ, induction of time series / unordered objects, Heuristics} → 191.17 spec

abcdefg
ABCDEFG abcdefg.

abcd. (size!) kind

It may be that my "intuition" is more oriented toward a system of that kind. This is an impt problem: In particular, the Q of SSZ for Hawaiian is impt

In Z141, [redacted] (w. ntopdeth for n≥2), an [redacted] ngm. can be used either for prodn or for defn. of another ngm. (In general, the posets of the 2 applicns can be distinct, but they have been the same in the 4 systems I've analysed.)

In the SAARB TSQ work, each function defined was defined once, used in prodn once & could be used as a part to help define another function, several times

[redacted] In Z141; each ngm is defined once, can be used several times to define other ngms; can be used many times as a predicted object.

SAARB & Z141 differ in that in Z141 the items were both (a) part of the corpus (b) objects that could be used to define more ngms. (69)

In SAARB: the functions were used to define other functions; which could be part of the definition. One Macrofunction that did the prediction.

.17: (spccc 190.40) In MCT prediction, each subcorpus usually has its own prediction function.

Sub corpi can be: defined time scores, different in the case of unordered objects perhaps, & each set will have a distinct stochastic. Each OZ problem will have its own weighted set of OT's. However, the total of all predictors are able to share subconcepts concepts. We want to minimize the total bcast of the

.23 Entire prediction "object". A later development will be the integration of the various sub-systems, by system - at first by sharing functions - later by a unified system on that modified itself in response to the nature of the problem [redacted] given to it. [redacted]

Say we had a "Unified System" like (.23). How would we (modify/improve) it? Well, suppose the system was a person, & was used to run his life. Part of the System will be dealing w. physics. So we have these "laws of physics": we observe some new phenomena: "the error of prediction is much larger than was expected" (I don't exactly know what that means!) But anyway, we decide to try to revise "the laws of physics" We do this in the "standard ways", by backtracking as little as possible & trying to modify the part of "the Laws of Physics" that seems most relevant. If we are successful, almost all of the laws will be about the same & the rest of the "human response operator" will not change. (Later we may try to integrate the "modified laws of physics" into the rest of the "human response function")

Perhaps Go into more detail on "Theory Revision". How it is done; How PC's are assigned to concs.

A common way to learn: Each sub corpus has a particular

8-19-00 Bulg.

80x64 = 5120 = 5M

90M / 40M " / 64 / 16 = 600k " = 70days,

Method of solving its problems: Say a function (≡ "op"). Also we have a way (OB) to recognize that sub corpus, so we can apply this Op to it.

The main process, then, is developing obs & ops for various sub corpi.

After we have lots of sub-corpi in assoc obs, obs; we try to integrate all of this, by finding similarities in obs, ops, for distinct sub corpi.

These similarities can be due to "sameness" in much of the functions involved, or it could be due to "Analogy" x-form mapping from one to the other.

Think of the development of Sci, & Theory revision: Do this in English. Get forms of heuristics. Get ideas for Design work of PC.

Consider the GRUE problem-types

[Now, I should be able to do some Analysis on the Elementary Algebra I >< directly! — Perhaps do both in II.]

.14 One kind of activity in a (Scientist's person's) life is Mathematics! — So if he is working on a General Theories for physics, he will use mathematical formulas & formalisms, each, that he is interested in! Math & Physics can be synergistic! So the idea of applying the x-form to physics is much hyped if the x-form has been much used in Math!

.19 → In general cross-fertilizations of distinct fields of interest ("Domains") are very simple in determining PC's, in guiding heuristic search. Perhaps this means that certain concs. that would ordinarily only appear in the Intern. of a sub corpus, could appear in the sub-corpus of a distinct domain.

.24 Also when the Corpus is Algebra (& other Math), there will be definitions of objects occurring in the sub-corpus, that are also useful in the Macro Functions for Math (as well as in other Domains)

.27 So, Math is a particularly good thing to do induction on, for this (.19, .24) reason → 200,32

.28 → While .19 ff seems very true, nonetheless, it seems unreasonable that I shouldn't be able to get more (hyper) SSZ (conformity) of conc's that

.30 are mainly useful in the Macro Functions.

.28-.30 seems like the present main Bottleneck!

I'm not at all clear on what objects .28-.30 is!

.33 Anomorphs related to .28-.30, that seems strange! Say concs A & B are used to define C, & are not used much after that, but C is used a lot. A & B get no more SSZ! — even tho they are part of C! — This may be ok. A & B contribute to the Apriori PC of C, but after C gets large SSZ, that apriori becomes much less important. → (But still A & B are "part of" C!)

Re: .33 ff: if C was definable in several alternative ways would we want C's hyPC "feedback" in a "Bayesian" allocation to it's "definers"? → (193.0)

8·20·00 Bulg

Platinum less wh sheet.
$10/g. = 1 pint
100h = 10kg = 20lbs.

.01: 192.40: On t. otherhand, if we are mainly using these concs in definitions & they don't have a large ssz in that capacity then they get low pc as definitions components. (I may or may not want to peel data from subcorpi — whether I do so, depends on ssz — I can do both & wt. of which error is back, will depend on ssz in both "defns" section & "sub-corpus" section. → .30

Reasons for publishing:
1) Esteem of Colleagues.
2) Need of Money
3) Feedback
4) Enthusiasm by Colleagues!
5) Good to summarize ideas. Put them in a useful form

In t. SAARB TSQ's: I did note that the only concs defined were those asserted that were solns. to problems. I more recently "decided" that this was not a fault of the TSQ, but the method I used for induction implied that this would occur. * — Hrr, it may not be necy!

(1) Often t. Macro Op is in sections so t. soln. of a problem could, essentially, just add a new "section" on to t. Macro Funct.

.14 (2) Secondly: After we have this Macro-Funct; we could partly or completely decompose it & re code it in a distinct way do it its PC in terms of funcs that have been used in other parts of the "Universe" — as parts of Macro Funct a/o parts of sub-corpi.

One point of the re coding in .14: Mainly to find concs. of hy pc that are used frequently/effectively, ←(?) ... (.19 is more to the point!)

.19 → Also: Re coding gives us shorter codes & less expected future error in predictions.

I think I've lost sight of .19 as a Major Goal (or sub-Goal).

I had been focusing on the idea of inventing new concs & verifying their legitimacies — But this is always secondary to the goal of .19. — Getting Max total pc of codes w. t. available etc.

So: what about the difficulty of 192.28-.30? Actually, look at 192.33-.40 If A & B were only used once to define a useful conc, C, then they are, indeed, not very useful for defining new concs, & they should get low pc for that applic. — 193.01-.06 is certainly relevant (& correct). Note .03-.06 on pooling of data from 2 distinct applics of the data of a conc.

.29 →

.30: .06 A Possl Example of such "Pooling": We teach TM a certain conc, by giving it examples. So it can tell if an instance given it is a case of the conc or not. (This uses "conc$_1$" in the sense of computational Irns Theory ("Valiant")) This defn. then is in the sub corpus. We then get TM to use this both either as a predictive conc a/o as part of the data of a new predictive conc. I had planned to use this method of Irning very often in early (at least) TSQ's. The concs Irnd, would be "Terminology" Terms of apparent importance in Math, Algebra (& perhaps Physics)

Def: conc$_1$ is
All concs
some concs
All concs used
not be conc$_1$'s!

So T. CQ. is, in view of T. apparent resoln. of 192.28-.30 : (on 193.19 -.40), can I start to write a complete draft of (T. initial) "Lrng Algm"?
It will be pretty much like T. SAARB TSQ, except that it will be Very User (In English) [redacted] — i.e. More "General"! Also, it will do much [redacted] treating on impt "concs" of Meth, Alg.
(Also perhaps simply concs of Math, Alg.).

Try to give Examples of the Variations of Lrng Meth can be done w. this method — specifically try to get beyond the Saarb TSQ examples. (Do some of the examples not "fleshed-out" on such as Saarb, write be worth going into now. E.g. solns of linear → quad → cubic (& maybe 4th°) [redacted] — & May be hyper, using portions of the Elliptic Funcs)

.100 Write a "review" of [redacted] 163.01ff.
2 parts: ① 163.01 - 184.40 ; ② 187.01 - 193.40 ② includes 176.21 - 177.01 - .40
Part ① is on TSQ in General part ② is on Learning Algm. | Section ① is 1/2 of section ② in pp. scope | ① = 16 ± pp ② = 10 ± pp
While Part ② would seem to be a self consistent "unit" apart from Part 1! This is not so. There is much in part 1, Parts impt. re: Lrng Algm: eg. 182.32 - 184.19 (~2 pp); on "Heuristics", "Categorization"
.117 so Section ② is 176.21 - 177.40 + 182.32 - [redacted] - 184.19 ; + 187.01 - 193.40

176.21 - 177.40 is a good picture of the General operation of the "Learning part" of the TSQ. 181.06 - .19 (Also Top of p. 181 <181.19 R>) is impt.!) Discusses the "Scaling problem" (& 181.19R Gives a kind of Criterion for Soln. — it's a very impt. problem!))))
A long discn. of Heuristic Scaling 181, 182.32 - 183.33

list of "problem" types for T. TSQ.

There are << 10 [redacted] impt. ideas in .117 refs; I should be able to list them & describe them "Briefly". [Later, make more detailed discn as a kind of PAPER! for myself]
1) T. needs Picture of TM operation of 176.21 - 177.40
2) Heuristics & Scaling: 181.06-.19 (Top of P(181)) (181.19R)! 182.32-183.33
.129 3) The diffy of 190.14 → 40, 191.17 → 40 - 192.01 ffff → 192.28-.50, 192.33-193.06, 193.30-194.10
This involves pc's of concs A & B used to define (conc) a very useful conc. (C); C is used a lot
A & B are not used again! Should A & B have hy pc?
.132 4) In T SAARB TSQ's; T. only abss (= concs) discovered, were solns. to problems. How bad is this? Can it be avoided?

5) The idea of Global vs. local references: T. invention of cond, imp formalizm

6)

.01!!94.40 SN In doing TSQ's, The large available HDD's (~ 40 Gby) will make it easy to
.02 store TM's state after each TSQ session. This makes it possible to "Backtrack"
if I suspect that I've led TM down an inappropriate learning path.
.04 I can also try sections of TSQ in various orders.

SN Make lists of types of problems for TM: first a broad list: [Induction of (a) Time series (b) unordered sets; Inv probs; OZ probs.]
Then Alg, Geometry, Physics, Chemistry! .... Then perhaps more specific problems in these areas:

Then: Simple induction; "learning" definitions in Math by example:
Various Math problems: Encouraged by "Hints": (I can study the effects of "Hints" by backtracking (.02 - .04) & redoing the problem.

The "Advice Channel" (MEC artay): Related to "Hints".
"Telling" TM soln. — This is an extreme variation of "Advice" or "Hint". I can either have special "channel" for this, or simply insert the soln. into TM by suitable pgms & reasonable pc's of component concs.

"STUDY" Problems: These are given to familiarize a person w.r.t particular-domain, so he may discover relevant concs. Often this is done w.o. the teacher knowing (consciously) what the "relevant concs" are! The "study problems" may be given as a kind of "Hint" or as a "Hope" by the teacher (who may be ≠ the student), that these probs "may" yield intended insights.
STUDY Probs can be used in 2 extreme ways ① T. teacher knows what concs are needed & knows the study problems contain them ② T. teacher doesn't exactly know what concs are needed, but suspects the study probs "will have" useful concs in them. [Various Mixtures of these 2 extremes]

Parts of this are from "Abs = Abstraction" instead of "Concept" — use concepts for relevant meaning in computational learning theory

SN Would it be possible to do useful TSQ/TM discoveries (at the Beginning) using very small CJS problems (almost exclusively). So it would take < 10^4 for a soln. (may be < 2^{10} for a soln!)? T. Q's: Could I investigate/discover the bulk of the important & useful things using small CJS? Using, say a 1 MHz machine & machine code, these small CJS could be ~ 10^7 or 10^8 bits. — which is not so small!
For SHRDLU TSQ examples, I think each task was "very large" — & I expect that these problems should have been solved much faster using good heuristics — designed to facilitate "scaling".

$433\frac{1}{3} \times 1.25 =$
$5 + \frac{100}{3} \times \frac{5}{4} \times \frac{80}{3}$
$\frac{100 \times 5}{3 \times 4} \frac{125}{3} = 42$
$= 592.$

SN APL has a v. big set of "Functions". — So that combinations of them tend to be "interesting" to Mathians!). Could I use a subset of them as "Primatives" for learning certain areas of MATH? □

8.23.00 Bible

.011196.40 So: Go thru 163.01 ff: See how all of this fits into 8. Learning Elem of 176.21 ff

8.23.00 Bvlg.

.01: 196.40 SN An aspect of t. TSQ that I hadn't recently much considered.

.02 It is t. Soln. of INV probs.: e.g. Proofs of Alg. Theorems.

My Refs about building up good abss from simpler sub abss stem mostly, perhaps from considering problems of this kind. I would use certain soln. techniques many times in different problems. A particular soln technique would acquire a high conditional ps for problems in which it was appropriate.

This seems much different from ordinary induction on an unordered set. — which is what 176.21ff is oriented toward — & t. probs of 194.29, .32 arise in ——→ They may not arise in INV problems, corpos.

.10 In ~~a continuation of~~ t. SAAR TSQ — I could start w. simple induction of defns & simple Alg. notation. The solution of (more or higher order eqns., could be reformulated as induction problems, INV probs or OZ probs.

An impt favorable feature of induction, is that its mainly what an infant (machine or human) does in t. early part of its life.

Hvr, one may also view t. Infant TM as Schmidhuber: that its trying to maximize its mean Gorc — & that it uses induction as a main tool toward this Goal.

Perhaps t. main thrust of t. present work is to get TM "started" — then, using MCT, add various kinds of problems to its Repertoir.

One very early idea I had about TM_2: that it would'n't be of much value until TM had accumulated enuf experience to successfully work on OZ problems of that kind. This idea can be saved: i.e. we can't expect TM to do anything very clever, until it has had adequate (collateral as part of CORPUS) experience in the domain in according "Cleverness"

I think what I want now, is to set it up so I can give it any kinds of problems to TM, & it would have ideas on how to solve them. — Induction in unordered sets seems like a good place to start, but it seems to have characteristics of 194.29, .32 — which are somewhat idiosyncratic!

Hvr, I could just do t. induction now, to start off. It will certainly be an impt. mode of operation. It may be inadequate as a "full TM", but I can add other modes, later. I do want to know if there are some essential deficiencies of t. Pure Induction mode.

T. Pure Ind mode can learn how to work INV & OZ problems — it can even learn to work probs that are not logic. INV or OZ probs! So it should acquire the cones of .02 - .10 often after it has tried to work INV problems.!

199.01

.01 ! 198.40 : [SN] In induction probs: When (at t. beginning), I "give" TM a heuristic,
.02 be sure to put it in t. form that it would have if TM discovered it itself!
.03 E.g. Say it's learning $1+7=8$; $3+2=5$ ... Then $5+3=?$
? is very probably a Number. The operation on "$5+3=$" has to have a number
as output. Then — for t. final funct must be add, sub, mul, div (or whatever
t. set of operations w no. output is at first time). The inputs to t. funct
.07 must be nos. So 5 x 3; 4 poss'l inputs, 4 poss'l functs, so 16 trials.
.08 How we get all this into form in which TM could have discovered it, is unclear!
.09 Assume an adequate corpus for [redacted]. These discoveries,

In .03 - .07 Much of t. "Reasoning" of t. Heurs used, seem [redacted]
"deductive" I will have to put all deduction in inductive form!

Re: .01 - .02 & .08 - .09 : Putting t. heurs into TM in a proper form is very [redacted] important.

.13 An approach: TM looks at t. problem "; "$5+3=?$". It has an "association"
w. t. symbol ? — that it has always been a "number". This "simple" induction
implies that for [redacted] various "features" of t. problem", TM has "associations".
Each "feature" will then have its own "x" property list/set". In this case, t. "feature"
is "?" & its property is "Number".

.13 ff is a very common type of induction — perhaps t commonest (≈ Bacon Sec).

.19 The next step, to realize that t. final funct. in a function must have
numbers in its Range, seems more diff. Each function that TM [redacted]
has been given, or discovers, has a domain (set input) & range (set output) —
COMPLETELY
In general, TM may not know t range & domain of a fun.
So, each funct has several properties; Range & Domain and two: Symmetries
(if any) on inputs & outputs. Numeric functs (domain) can have anti sym.
eg. $x-y=-(y-x)$: division by a "kind" of anti sym unary $x \div y = (y \div x)^{-1}$ (except for zeros)
[this "reasoning" is beginning to sound like Lenat's "AM"!]

On second thot, .19 ff is getting "Too detailed" (no longer "English")

.28 [But I do want to (eventually) find a ways to express (almost) all heurs as part of
induction — as a probistic narrowing down of t. "Such Space"
One BIG reason for .28 is that it will make t. Program much simpler — easier to write/debug.
I do not want to program a special section for heuristics — I want to have only
a simple induction Algm.
So all I have to do is get t. general (ung system) planned. The heuristics
should not modify it in any way! Hur, to test t. (ungrig) system — make sure
that it can acquire & use t. info of [redacted] several impl. types of heurs.
(perhaps including "Quick Abort").

$5 \times \frac{2}{3} \quad \frac{10}{3}$

Laws of Logic: Laws of Algebra:

TM should be able to do both of these rather easily, since these laws are useful in induction — i.e. finding regularities in tasks.

Σ So why did I have so much trouble w. this in SAARS ?]

Perhaps ∀. laws of alg. ∃ solving eqns is only important if our induction has finite c.B. Otherwise, all sizes of all eqns. are dominated by the equation ∃ !. of $pc = 1$. So $cB < \infty$ is necy but for ∀. laws of Alg to become useful in induction.

.06 Re: .06 ! Even if arithmetic does have cc, if it's $\approx \infty$ & the cc of simple reasoning, it would usually be cheaper to do arith than logic, so ∀. laws (really) would not be discovered !

Even w. small ($\neq 0$) cc of arith, if TM trained to solve eqns, ∀. laws of Alg. would prob be useful. — In fact, it may be true, that to learn to solve eqns, TM would have to discover ∃ equiv of "∀. laws of Alg".

If TM uses Lsrch for solving problems, cc would always be an impt. factor in finding ∀. Solns. ! (A what about "Quick/About")

So: Try to find problems for TM, in which some "laws of Alg" are useful.

OR: Just try to get it to learn to solve all kinds of Algebraic Problems: learning discovering.

This is prob "equiv" to learning "Laws of Alg." (!??)

It is equiv. ↔ TM ends up w. ∃ pol's for certain all concs (∃ ∴ ends up w. same cc for all discoveries, as a TM that "knows the Laws of Alg".

ABCdefghi

198.01 — on prooy Thms

However, w. large enough SSZ ∃ equiv cc, TM should be able to discover any derivable conc. So I how do I derive "Laws of Alg" as a conc?

It would seem that ∀. "laws of alg." should be discoverable from a sufficiently large bunch of "Arithmetic Experiments" by TM. — (If it did "Experiments" of this sort.

.28 Another possy: Have it part of TM's "kinds of tasks" was included discover Mathematical Conjectures (∃ perhaps occasional Proof) w. suitable Heurs to help decide if a conjecture were "interesting / useful",

.31 that it would do better in regular learning of Chem, Physics, language etc.

.32 — That [redacted] Cross Domain Inter Domain Lrng. (192.14 → .27) is

.33 can be very impt. — [redacted] .28-.31 we could be an impt example.

I had originally that that Math along could be learned, but perhaps the MaM should be part of as having 2 aspects ① MaM as something to learn: solving problems, proving Thms. ② MaM as an ART form (Making Conjectures, Defining & proving Thms, Making Defns of "potentially interesting" ideas, ect. The direction of MATHART could be partly (or wholely) driven by TM's examination of ∀. Content (& History) of Math. — What has gotten

8.25.00 Bulg.

The interest of Mathematics is what, in the long run, has been found to be useful in Science — By TM's **INDUCTION** on that info.

Going Back to "Laws of Alg". After TM has done a lot of numbers using + − × ÷, it might ask "Are there some relations of interest between these numbers & those functions?"

Try to find problems in which certain "laws of alg." are useful in finding solns. Probably, instead of "laws of Alg" it would find many specific Rules. That is "laws of Alg." would be a way to compress, to express in a simpler form, those rules. —— The value of "a simpler form" is that The laws are then more likely to be extrapolable to new "laws" of interest. The "structures" of the laws could be used for "induction by analogy", .08 .13 which is a very powerful, "creative" Mechanism.

So, in search, I did get something like the laws of Alg, by teaching TM how to work linear Eqns. — It might be possible to go from there to .08 − .13: However, even if + allowed CC & would be larger, I suspect that the size from the existing data, would be too small to allow the "laws of Alg" to be induced —— Tho this is unclear! We could start w/ a very large no. of rules (∴ large size) so the "laws of Alg" could afford some compression.

The large CB needed could be prohibitive! We could just "wire in" the solns to the such. In fact, it probably did take a lot of Search time in the Math community. —— But Math Community also had other heuristics & Pur Goals (like saving time in [illegible] +, −, ×, ÷ & parentheses)

So laws of alg, were discoverable (or either) by

1) Need to speed up + × − ÷
2) Use in solving eqns.
3) A compressed format (= rules of 2),
4) An orientation toward discovering "simple" rules relating Real numbers. Rules relating integers give an entire new area of Math (for many yrs. a Pure Art Form) — Then later found Useful in reducing CC of certain calculations — ~~& Pgs in problems relating to size~~ Also in solving **Diophantine Eqns.**

Re: + & − : they are inverses w. 0 as unity (relation of + to −)
× & ÷ " " " " 1 as unity. (relation of × to ÷)

$$x + 3 = ? + 3$$
$$x + 3 = 7 + ? \qquad (? = x - 4)$$

→ What I want now, is a (somewhat) general lrng algm. that I can easily program. This should be general enuf so I can apply it to several domains. Then I want to try it, first theoretically & then practically, on several TSQ's from somewhat dif'rnt domains.
Also, I want to see if it can detect (in theory) & find (w. available cc) various (impt. necy) heurs.

Superficially, it would seem that the "Laws of Alg." would be like the laws of physics" but over dif'rnt domains, — But maybe (apparently) not!

After being given examples of many eqs & their solns, TM may be able to discover that the numerical value for x satisfies the eq. if its value is substituted therein: What would "motivate" such a discovery?
.13 We mite teach it what $x^2 + 3x + 7 = 0$ means as an "eq. to be solved".
e.g. give examples of eqs paired w. their solns, & see if TM can find the relation (≡ Symbolic Regression).
.16: Well, in looking for regys in the data, TM mite notice that in the problems, solve $f(x)$; 3 say, $f(3)=0$ was always true.
This is a logic. regularity, but it's not immediately clear to TM, that it is useful in predn! — Unless TM knew about approxn, say & could think of solving $f(x)$ as a problem of finding $x \ni |f(x)|$ was min.

Perhaps it would be a good idea to list a bunch of problem types to see how & if TM could learn to work them. By looking at several types, I may get ideas about how to solve them all!
(SN) from $x+x=2x$, $x+x+x=3x$, $2x+x=3x$, etc., could TM induce $ax+bx=(a+b)x$? or, more relevantly, $2x+3x=7$; could it induce $x = 7 \div (2+3)$?
could it induce "counting" i.e. $x+x+x \cdots x = nx$.

Well, we could have TM learn what solve $f(x)=0$ meant as "learning a definition". Learning how to solve $f(x)=0$ is a quite dif'rent problem.
Hrn knowing the meaning of $f(x)=0$ would seem to imply compression in the sub corpus $\{f(x)=0, 3\}$. How TM would deal w. this compression, if at all, is unclear. It may, indeed, be unable to deal w. it, if the only kind of compressions it knows, would have to be one or more functions of the expressn "$f(x)$" that would be possl. solns.

8.25.00 Bulg

.32 **Quick Abort**

.01.202.40 →→ Do I have a ~~clear new~~ idea about induction (ā inserting hours) so I can try that linear → Quad → cubic learning?

.03 That a = b [redacted] implies b = a! There are 2 induction aspects of this idea: (1) How can it be used to compress a corpus (i.e. what corpus? ā what other rules ("facts") would TM have to "know"
(2) From what corpus ā what "primitives" could TM "lrn" .03?
w.o. this info, I'd have to insert .03 into TM in a special way.

.08 →→ It would seem that if .03 were at all useful in predn., that there ~~must~~ must exist a corpus that .03 can help compress. — ā we could ∴ "in theory", use this corpus to discover .03.

.11 To repeat the idea of .08: if .03 is useful (for predn) in any particular induction problem, then a large set of problems of that type could be used as a corpus to "learn" .03.

So .08/.11 mean that any trick that helps in induction can be "learnable" w. a corpus of proper kind.

.16 O.K. then given problem (cond) 1 = x imp x = ?

.17 It would seem that .03 would help solve it. — all suggests that we'd

.18 need a corpus like .16 for TM to learn .03. However, we'd like TM to be able to learn .03 just by observing numbers and their

.20 relation to each other ā to +, −, ×, ÷.

Why .18–.20 is disturbing: I like to think that TM could learn just about everything about R.W. from studying books ā perhaps just the internet. Hvr. .17–.18 suggests that this is not so, that we'd need a specialized corpus to lrn certain things!

.25 On the other hand, if any idea is useful in predn., then that "idea" can be learned w. a suitable (ā commonly occurring) corpus.

So .25 can be regarded as a Theorem about concs useful in predn.

Now, How to "try out" .25? Find some things I want TM to lrn in which the laws of Alg. are useful.

.32 [SN] on "Quick ABORT" heuristic: By speeding up search, it enables us to get higher pc (in Lsrch) for a given CC. Just how TM is supposed to recognize this, is unclear at present.

.35 I guess the thing that disturbs me about "lrng Laws of Alg" — that at first glance, it seemed like it should be easy for TM to lrn such a thing: What is not clear is just when ā when it is not "easy" to write a tsq for a particular lrng. task.

→ 204.01

.01 : 203.40 In t. case of "Laws of Alg" (or any other lrng task), we have to ask: what is t. criterion for TM having lrnd this concept? How can we test TM's "knowledge/understanding" of it?

Clearly, it is not legal to ask TM "what are t. laws of Algebra?" TM doesn't really speak that lang. The only legal tests are inductive problems.

.06 What [illegible] inductions can it make? What new inductions can it lrn rapidly?

↑ (This corresponds closely to a Human Teacher Monitoring a Human Student: Trying to figure out if t. student has "acquired" a certain heur, or whether t. student "can work" a certain type of problem)

Consider "equals": if TM knows about "relations" then "=" is a symmetric relation on 2 things. It is idempotent and transitive.

I could [illegible] have TM lrn defn. of relation: Then ask if "=" was a relation, ask if it were idempotent, sym, transitive.

So: A present possl view: That TM can discover any conc. that is useful for [illegible] t. tasks it has been given (if ssz & cc are adequate).

— But discovery of "Laws of Alg." is probly not essential for TM to be able to work most Alg. problems. When TM studies other kinds of Algebra in which $a \cdot b \neq b \cdot a$ & t. assoc law may not hold, etc. — Then, it may be more likely that TM will try to compress t. set of rules it learned in solving

.19 algebra problems — to give t. laws of Alg. (3 ways to deal w. "laws of Alg" 182.27)

The (203.35–.40, 204.01–.19) is a reasonable way to look at "lrng t. laws of Alg" (also note 203.25), I still don't feel entirely comfortable about t. Question. The overseness of (205.35–.40) still remains!

.24 : 187.17 Look at my initial treatment of TSQ1 (of 181.5) on 187.01 ff. In a realistic situation, TM would be just starting up & would have no heurs that it could have discovered itself. Nevertheless, I could just insert a tsq from to "TSQ1" & see how t. Lcost increases (w.o. use of Heurs in t. TSQ).

T. point of the exerc. would be to see how t. solns. to difrnt problems in t. TSQ, interact (or are indep) in t. solving of a new problem.

.30 → In particular, I'm interested in how Heuristic info is acquired & applied.

At present, I see no particular diffy in .30: TM just has to acquire t. Heur (which is a regular induction problem) & apply t. results to new (post heur acq. original) problems.

Well, in 187.13 ff, (3.7+4.1=7.8 ....) we normally try [illegible] functions [illegible] on t. input expressn. | 3.7 | + | 4.1 | = | ? | (also, perhaps in the "?" the "?" is

we start with t. set of functs that we have: & we continue by making more complex functions on t. input expression, 3.7 | + | 4.1 | = | The "final" function (usually must) map to nos. (This is a heuristic discovery that narrows down t. P.D. for the final function)

The inputs to the (identity) Unary function are 3.7 or 4.1.

205.01

~~49~~ 40

9 x 30 = 270

.01 : 204.40 : (We can "wire in" to Eurisko that the numerical operators identity, +, -, x, ÷ only have numerical inputs so this limits our choices.

This learns 3.7 + 4.1 = 7.8, but this new operator will not work with "3.1 = ?" (a earlier problem I ran on 187.08 AR)

[redacted] So, it now has 2 operators: one for 3.7 = ? (analogy) the other for things like 4.2 + 7.1 = [redacted] ? So each is good for its time, so

.07 they are both useful, & / (or to) comm. progressions obtained. → (16)

.08 TM's motivation to continue searching for a better code! Unclear!

I did write about this recently, hrr. (177.02) The idea was that we needed "serious theory revision", when the system's error rate for the new data was much worse that what we had before (for whatever reason — worse than what we expected). Hmm Hrr., this is a very

.14 **Vague answer!** I'm mainly thinking of Motivation for Theory Revision **in Physics.**

.16 : [redacted] (.07) [redacted] These (for reasons, like (or better than) (.08 - .14)) we look for a way to get

.17 better predn. A common way is to look at the sub corpus in which "add"

.18 was the rule operator v.s. sub corpus in which "identity" was the rule operator.

3 = [?] v.s. 7 + 2 = [?]. We can distinguish them in (several) ways; One way:

.20 ① 2 symbols v.s. 5 symbols! (Hrr can TM "count" at this point?).

.21 ② 2 nos. v.s. 1 no.: [redacted] (presence of) "+" v.s. presence of "+".

[Anyway, this same kind of problem arises when -, x, ÷ are tried.]

The Heuristic of .17 - .18 : How could it have been discovered? ——

what hypothetical corpus could give rise to it? What operators could be combined to discover that regularity? Seems like a very sophisticated heuristic!:

NO → One way: [redacted] (Even w.o. the heuristic .17-.18) TM notices dropping execution of Conds, ... ? .. (working to notice) on n = ?, we try add(n, n); On [redacted] m+n = ? we try Ident[redacted](m); Ident(m).

A (perhaps) reasonable way to get hour .17-.18! It is a way to [redacted] xform the original induction problem into a new inductive problem ("OR" x Plan)

So we end up w. a [redacted] "Times Series" w. Elements [redacted] n = ? → "iden"

m+n = ? → "add". We need a Op that has iden; add as outputs.

I have to investigate the priming of functions. How can they (w. op functs) generate the needed discriminations?

Compare [m + n = ? / n = ?] — 2 subcorpi How to discriminate: The "+" sign is the simplest way. (The "size" is also quite simple).

8-27-00 Bolg:

What I really need to do: Examine OBs & how to use them in induction;
How to use them to disamb. corpi

In general, obs have same inputs as ops. numbers, Boolean, character/string. ≃ basic "selections" of insts

[I may want obs to select from a set of functions, rather than be just True/false]

Mainly, I'm talking about an ob as making a choice among ops. Obs are a subclass of ops

In present case, I want an "ob" that can recognize difrnce betw.
.08 "n=?" & "n+m=?" and use this info to excite the proper "op".

.09 Presumably, I have an ob that can look at an element & give ϕ if a number | if a non-number (see 25)

Some other possl. (ob) funcs of interest: The usual "If a (≤) b" T/F.
This task is "Does such a relation" hold between inputs
A "relation" can be regarded as a kind of "classifier" of inputs "intups"
— Hm, all elements of the relation have the same, so "relations" are not the most general | type of set.

.14 One kind of ob given: If (a) then (b) else (c)
a is an ob; b & c are ops. So, if uses a to choose betw b & c.
Hm, what I want is distinct. from if then else. Those are used usually to control flow of execution: Here do different things in a functional lang.
So, .14 ends up being an op.

The problem I'm working on, is to augment the normal set of funcs used in ordinary Algebra should presumably by by adding to obs.
Hm, other people in GA probably have been using an adequate set of operators. Rozz has (usually) been using LISP.

Returning to Nos Mouton's Say S is the set of non-numerical objects that have appeared in the corpus this far. So S = (=, ?, +) only 3 things.
.26 Assoc w. each is a "ob" that can tell if it occurs at a point in the corpus, by "place"
having a T/F output (Boolean). (see .09)

.28 I could drop this problem for a while, & assume that somehow, somehow I, (say, was able to distinguish betw. "n=?" & "n+m=?"

.30 There is a Q about the form of input: The input is a set of strings of symbols & numbers. If the string is 20 symbols long, I have assumed that the input funct could accomodate 20 inputs, simultly. Evidently long expressions will be hard to accomodate! A possl. way is sequencial input of the symbols. An equiv method; The inputs of the functions have addresses for their inputs, so they can get them from any point in an arbitrarily long input string. — Hm, best to work out this Q when I x/f. from "English" into "Code" 210.02

Perhaps model this language after the way I write pgms in, say "Basic".
Also, if my description of what I do (in English) is simple, I should be able to put it in more exact form — i.e. a formal Lang.

OK: Well "n=?" v.s. "n+m=?" "in English"

.02 So I had Pico 3 approaches on 205.20 — .21 "in English": ① 3 v.s. 5 "symbols"

.03 ② 1 v.s. 2 "nos". ③ presence v.s. absence of "+".

These involve ① counting symbols ② counting number ③ counting non-nos,

④ "counting" "+'s". ⑤ Threshold for no. of counts. ⑥ use of <, ≤, >, ≥, =

.06 for "If" part of statements.

SM Unfortly, having 3 parallel codes doesn't help on normal Lsrch — was just used to single out w. best Lcost. But note: 3 codes has ~ 3 times as much CE, so we'd end up w. ≈ same Lcost of soln! → (There may be something wrong w. this argument, hwr. — it has been my impression that $\frac{pc}{ce}$ would be appreciably better as a stop criterion for Lsrch than is normal Lcost ∴ — so "check this out"! ☺) 8-21-90 TMGen: 1.01 — 15.90 is on this problem.

In .02, I'm keeping discussion "vague" in English: "Continue" in "English". I may not be as good as I once was in shifting from English into Math, but — too bad!

Anyway, in .02-.06, ① ≈ ② how to do w. "size" of expressn: I think this is a very fundamental idea — related to pc (≈ "Alg. Complexity").

— So TM could have Mips "Builtin".

Another "Builtin" skill: in .03 — recogn of "+": This is a "noun(" symbol because it doesn't occur in "n=?" so this, too, is a legit way to recognize distinctns betw. strings.

SN A worry about ENGLISH dcon of tsq.'s soln. and Pico's: That a thing that seems simple in "English" may not be! The main example is "Devry of Laws of Alg".

Back to TSQ.: A possible sequence of problem types:

1) Learn to evaluate alg. expressns. (not in Polish) — Perhaps stay Polish as in tsq — or index to Polish & the both ways!

So 3 + 5 =? 3 × 5 =?

3 × (8+9) =? (7 ÷ (3+9)) + 2

I see if TM can "Get the Grammar idea" of any pb. into tag, — which means it can do evalns. of arby "depth".

Next soln. of eqns x=3; x=y

~~Solve x=y~~; Solve(X, X=1) =? Evaln. of the function "Solve",

X + 1 = 3; 3×X + 3 = 1; X + X = 3; X + 3X + 7 = 2.

("discovering" that X + X = 2*X)

→ ~~Perhaps~~ perhaps my reading that SHARP stuff for a TSQ.

↔ Hwr., note that one of the Great Breakthroughs assoc w. MCT, was that TSQ's should be easy to write! That any needed conc. should be reachable by examples, hints, "telling" or "showing".

So, just by being an "English" TSQ for Algebra! — continue what I've done, including other theories. These theories are usually —209.01

Varieties of knowledge TM should be able to have:

1) [redacted] Condl. P.D. : Input: dem of INV prob: Output: Pd on dems of [redacted] possl. solns.

2) " Input: OZ problem dem: output: P.D. on [redacted] Often Techniques (OT's): Pc = Proby that that OT would give best G for that problem in available cc.

3) Languages: Pd's over sets of finite objects (possibly over infinite objects (as well))

4) Given + "Nature" (= source & assoc. info) of a T.S. & [redacted] a subsequence of it: It has a P.D. on continuns. of that T.S.

5) ——— ———

T. fary: are the kinds of info that I've been mainly concerned w.

~~5)~~ Knowledge of π, $\sqrt{2}$ etc: essentially the ability to calculate them to any precision a/o the memory of a few no. of digits of them. ((6) covers this)

.13 6) Knowledge of how to do various types of math problems like Solve eqns, compute π, $\sqrt{2}$, sin 35; [redacted] multiply, invert matrices. (This includes (5)) ; integrate, differentiate symbolicly

.15 7) Ability to reason logically: Given postulates, to be able to prove theorems, [redacted] conjecture theorems; Give a bunch of data, to be able to deduce info relevant to a particular problem.

.19 8) To be able to plan a/o carry out a research program (like "find cure for cancer" or "devise better CPU hardware" or "find a better theory for physics".)

The "laws of [redacted] ~~the~~ Logic" [redacted] (incl. laws of Algebra) "can be lrnd inductively a/o their equivalent in predictive power can be lrnd inductively

---

Re: .13 (6): These can be taught to TM inductively.

Re: .15 (7): " " " " " " " : An impt. Q is: "can TM learn to apply this reasoning to its own problems"?

Re: .19 (8) I've worked on this problem some: Don't remember if I solved it or not!

.32 One possl. Big Dffty: I thot of starting w. a "simple" TS & Q from Algebra only. But it's likely that to discover or even use [redacted] the needed heuristics, Many more modes of "knowledge" will have to be had by TM. !

8.29.~~90~~ Bdg. 209

$\ln \equiv n!$ — ~~...~~ $\frac{\ln}{\ln k \ \ln (n-k)}$ w?

.01: 207.40. not absly necy at this pt., but put them in, because I want to know ~~..~~ that I know how to do this! : Also note 208.32

I.n particular, note 208.16 & (208.16R) ←

One apparently impt idea is ——— T.M. has to learn to "reason logically" — many heurs involve Logical Reasoning. Since both logic & Algebra will ~~....~~ both have t. same kind of "large SSZ", ~~.....~~ TM's will have about t. same confidence in its predns, in either area. We can perhaps make some kind of Formalism for TM that is equivt. to $SSZ = \infty$. It is hoped that "Truths" obtained w. $SSZ = \infty$ will not conflict — since we would then have to know ratio of sizes of diffrnt ∞'s!

(Difference betw 2 ∞'s might be impt. e.g. $\frac{\infty_1 - \infty_2}{\infty_1}$ $\left(= 1 - \frac{\infty_2}{\infty_1}\right)$)

Th. impt. idea here: That TM has to learn to "reason logically".

— That there is no real difference betw. TM learning Algebra (/Math) & learning (Mathical) logic (as part of Math) & tryng to use ~~....~~ this Mathematical Logic in obtaining Probabilities.

(SN) Many Heurs will have to be lrnd by OSL. We could wait for 1 of "SSZ" to 2, but this would be wasteful of SSZ's Also, it is essential that TM have ~~good~~ facilities for ~~....~~ doing OSL.

.19 So what I have to do, is write a TSQ in English; & Then ~~...~~ figure out t. ~~...~~ needed (or "good to have") Heurs in English. Then figure out what sort of training in logic ~~...~~ a/o Math, that TM needs before it can discover/use each heuristic.

Perhaps Try to find where I've already done much of .19.

Perhaps Reread Original Proposal for Sol 89. It even had estimates of Time Scale(?).

But I ~~..~~ shouldn't be "Too Careful" in constructing TSQ! T. technique is supposed to work w. not-so-perfect TSQ's.

So in writing t. TSQ & Analysing "in English": I can use various amounts of "Heuristic Guidance". To start off, I just want to get ~~..~~ a feeling for just what (& how much) has to be done. Ideally in "Steady State," it should be possl. to train TM using not much more than a TSQ that has been designed "IN English" — That if it doesn't acquire ~~...~~ a certain conc. w. acceptable cc, I will analyse what occurred, &/o just ~~...~~ give a "hint" or give it (or teach it) what I think ~~..~~ is t. relevant heuristic.

.34 I think t. latest TS Q analysis! 287.01 — .17; → 204.24 (dietz 21: 205.08 - .14) Ditto(?)! 205.08 - .14

205.07 ~~...~~ — .40 206..28 — .40, 207.01 — .49; 209.01 ff!

Main points: start w. ~~...~~ ~~.........~~ 3 = ?

Then 3+7 = ? ~~............~~ 3 = ? is solved easily because there are only a few/ (primitive) functions that give nos. as output. That TM should notice that "?" has always been a number, is impt. — ~~...........~~ This seems to

Bdg

be a normal kind of Inductive reasoning. — So Then TM only tries funcs that have nos. as outputs.

> Q about format input — sequential or ll? 206.30–40

.02 [The form of input to TM, sequential a/o parallel will affect encoding lang, but I want to delay seq. v.s. ll until I xlt from English to ← code → "code"]

After it learns 3 = ? we do 7 + 8 = ? [It is able to do this — again, noting output is a no., so only a few funcs have to be tried.

From corpus containing both 3 = ? & 7 + 8 = ? TM gets it rite about 50% of the time — which is v.g. compression.

Motivation to ~~improve~~ ~~comp~~ try to compress further: Well ① it can be given by user, ② for older More "MATURE" TM, it will have some idea as to How Much compression expected in Math, (& 50% accuracy is not v.g. for math.)

To improve compression: TM divides corpus into 2 parts: ① those using identity fn as soln (3 = ?) & ② those using sum( ) as soln (3 + 8 = ?). It must then distinguish betw "3 = ?" & "3 + 8 = ?"

> What is reasoning behind this?

.16 This can be done by ① size of string (3 v.s. 5 symbols), ② no. of nos. 1 v.s. 2

.17 ③ no. of non-nos 2 v.s. 3; ④ Presence of "+" symbol

So at this point TM got a ≈ perfect score. (Assume "Presence of +" is not the criterion used here, because later we will have to revise the other decision methods. ↓ = "Backtracking ... use of (parallel) alternative codes)

> If we use several ll codes ("overscored" W.L.s rep.) then Backtracking is not necy, later.

Next do 3 − 7? This is easily solved, but then we get it rite only ½ of time, since sub( ) & add( ) would both be used here. We have to distinguish betw 3 + 5 = ? & 7 − 2 = ? By "string matching", it's easy to see that "+" & "−" are the only things that distinguish the two.

So at present, we use say ①, ②, or ③ of .16 – .17 to distinguish betw 3 = ? & 4 + 7 = ? This also gives us 3 − 8 = ?; we then use the "+ v.s. −" to further distinguish betw add( ) & sub( ) as solns.

.30 [We could at this point, reformulate the system, & use ~~presence of~~ "absence of + or −" to invoke recogn. of 3 = ? — i.e. the identity operator. Whether we would do this at this pt. is unclear. A "Person" might notice that this reformulation gives a shorter over-all code. Unclear as to how to motivate TM to look for a shorter code when the predn. is ≈ 100% rite!]

Doing 3 × 4 = ? & 7 ÷ 2 = ? involve no new ideas.

8.30.00 Bulp.

Doing 3*(4-5) =? would seem to be quite diff. at this point. TM may associate * w. mult( ) & - w. sub( ), but this is rather vague. Trying mul(5, sub(4,5)) is a reasonable possy.

Similarly, it would learn (4+5)*7 =? and 3*(4+8) =? ect: & 4 contexts 4*(8+3) & (8+2)*4 to be found.

Somewhere along this path, it should begin to do some "hyperorder learning": by noticing that there are symmetries, similarities in its "rules" for various situations. Such hyperorder rules do ↑ pc of comps.

Also, when new kinds of problems are given, they should help solve the new problem immediately or help speed up needed search.

What I'd like, is that at some point TM should discover what "params" "really mean" & induce the general role for evaluating any alg. expression in Eng's notation.

O.K. so that would be an interesting TSQ:

Another TSQ: learns 3=? & 2+3=? 2-3=?

Solve(x, x+3) = ? Solve(x, x+3) = -3, this is the def x+3=0.

Solve(x, x+C) = ? C is a literal; x ? is -C in this case.

If may be easy to get TM to learn mult & div first (?).

After TM learns to Evaluate alg. expressions, try teaching it solving linear eqns. — Some "non-linear" eqns. ← meaning of this isn't quite clear. Well, eqns in which some alg. manipulation is necy, but final eqn is "linear" to be solved.

.24 Perhaps I want to get TM to the point where it's using the logic a/o math it found, to help find models for new problems.

.27 **Long Range Q!** Just how much work will it be to get TM to the point where one can teach it w.o. getting into details about how TM is going to run it! i.e. w/ to teaching a [human] person?

This Q may depend on Domain: i.e. one might be able to do it in some domains, but need more training before it can be done in other domains.

Probly Sub-goal .24 is a large step toward Sub-Goal .27

Anyway, to get .24, I may not need a very large TSQ: Even in simple problems, if one wants to work from w. very small CFS, one must use much "logical reasoning" & "Mathical Analysis"

An impt. Q is how to get this Mathical/logical reasoning to arrise automatically, w.o. any special pgms. on my part. → 216.01

The defn. of Adequacy of an English Soln of a TSQ. .05–.10

12th Sheet o'in pad. I managed to go.

.01:211.40 : A possl. Example of "Much Reasoning" is in the Substitution Heuristic used in soln. of linear → Quadratic → cubic equs.
Also the "Planner" heuristic that breaks probs into "or," "and" nets. — This devises sub-goals.

.05 → SN I have been a bit worried, recently, about mapping from <English desn of the TSQ & its soln.> → to → a program (SW, HW). This shouldn't be too hard.! [If the English soln to the problem is Legitimate then there must exist a {probalistic} practical search method (perhaps ≈ Lsrch) to implement it. So return to the sheep & write the TSQ's! → see 219.19

.10 → So .05 really defines the Adequacy of an English Soln. of a TSQ.

O.k. So let's Go back to the TSQ of 209.34ff: How "complete" is it? — What needs to be added / changed to get it to satisfy .05 → .10?
So start w. 3 = ? : T.M. has a bunch of functions available: at least idem (x → x), add, sub, mul, div.. The input consists of 3 "things", 3, =, ?
a "thing" is something that can be an input to a function. At this point, = & ? are not "things" because there are no [illegible] functions that can use them as inputs.
For each of the "things" (other than nos.) that have appeared in the corpus thus far, we can define a function that recognizes it: Has "T" as output if input is that, "N" otherwise. This SW/HW idea gives some way to estimate pc of various functions (usually primitive functs at first).
We also have a function that recognizes Number; {It is the "not" of the "or" of all other recognition functions.} I may want a function that is T if any its assoc symbol occurs anywhere in the input string. If it does we can look at each position to see if it's there.
Well, O.k. say we have all of these OB functions: for the 3 = ? problem, they could be tried at first level, but they would give "T or F" out — never a no.
Eventually, I'll want a means for TM to recognize that a funct w. numerical output is needed — & thereby ↓ the amount of search needed by a sizable factor. So TM "Notices" that output is always a No., Then
.30 concludes that only functs w. numerical output can be tried as "final function".
.31 Just how do we realize the resultant pc↑ of .30 → 31? What Mechanisms can I use? (I want to desc. the Mechanism "In English" w.o. going into
.34 HW/SW details). → 218.01 see
Perhaps I can skip this "detail" at this point, but make list of these "lacunae". When I have several, see if I can work out a technique that deals w. several (or all) of them.

→ 218.01

2 6.34 spec
.01 2 6.40 : One Approach: One has to predict ?". From pure Uncondl. probly, we find it has always been a "number".

That + − * ÷ have always had nos. as outputs — we can give TM a pseudo corpus that tells it this — since ? as an i/p rarely, it wouldn't have much ss2 of this sort.

The discovery of t. "working backwards heuristic". This can be obtained

.07 2 ways: (1) purely empirically, TM notices that t. Macro function must have a terminal function that has numerical output, i.e., it requires that t. Macrofunct have a numerical output. TM would have to have acquired & recorded cases ("Vocabulary") before this could be "noticed".

.11 (2) Because needing a number for ?" is "sim" always has had numerical output" have large very large ssz, TM would consider using Logic (ssz 200) on these statements. [Also, proper "Vocab" needed before this can be done]

.14 The methods of .07 & .11 need probly t. same cases ("Vocabulary") & probly. some distinct cases.

Tho maybe .14 is it so true! T. main difrnce betw. deduction logic & induction is that in induction, one traces out all possl paths — even those w. low pc. In deductive logic, most paths are of Zero pc., so one doesn't bother tracing them out. In approximate [induction] we will not trace out very low pc paths, so we can approach "Logic". ("Deduction")

(SN) Say we give "teall" TM some laws of logic a/o laws of Algebra. If this info is on "TM Advice Channel", it can be regarded as a "Hint" — in t. sense of reducing (cost of searches for TM.

So: Ok: we have given TM enuf info so that it's lrnd that "?" is usually numeric, that "+ − * ÷" all are "usually" numeric inputs, numeric output. TM always constructs new cands using old cands & useful sub traces. In making final traces, it could "respect" any info that it may have on I/o of t. various operators.

.28 Normally, t. Maps are constructed "Backwards" "Th. final output operator first. ← ? funct

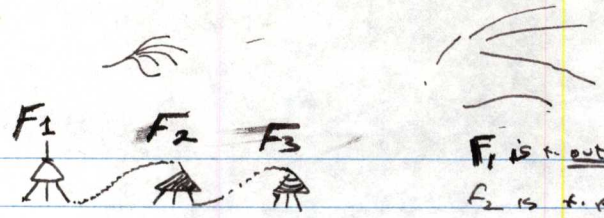
Well we could construct them this way: start by doing all possl. functs w. desired output "Type", connected to all possl. legal input types.

Next, to add one more function to t. "Middle", replace each input to t. function w. all possl. functions, & have t. input to that funct, be all possl. inputs.

To insert 1 map funct, replace all input lines by all possl. functs.

.28 ff seems to get a different result from the method of function generation that I used on (1 2 0.25 — 13.18), which builds up functions starting w. 1 43 20 on symmetries 136.25 on Localization in "Basic", input lines. This method is not so good, if one is certain "Type" for output. On. other hand, it may be that .28 ff is not correct! — that it doesn't get all of t. functions we given output type.

→2.19.01



.01 : ~~218.40~~ 218.40

F_1 is the output function.
f_2 is the possl. additional input to F_1
f_3 is another possl. additional function to (F_1, f_2)

Actually, this analysis need be done for usually no more than 3 or 4 functions, because the pc's became so low, due to the very ~~large~~ large no. of possys, Symmetries
So drop this for while (Note 143.20 : Discussn of Symmetries in Functs, can ↓ no. of possys by a lot!)

The Present Problem: To understand how (1) Neurons are discovrd by TM: what form the discovery takes. (2) How are those discoveries implemented in ↓ ~~[redacted]~~ cc/g for Csrch? I.e. How do we use these ~~[redacted]~~ "discovrd neurons"?

.12 I'd like the neurons to be in some "standard form" to facilitate their applicn. .17

.13 SN In General "Types" of input/outputs of functs could much reduce no of possl. combinns. The formulae of 130.25 ff didn't consider this 143.20 discusses Symmetries; possibly ~~[redacted]~~ a ~ method could be used for "TYPES". I'm thinking of using Mainly 3 types Number, Boolean, String

.17 : .12 → Try to put this Q into as exact terms as possl.: Exactly what is the Main problem Now?

.19 :216.10 One Big problem: (linked to 216.05 ÷ 10): How are reggys (≈ Cones), compressns found? ~~[redacted]~~ After they are found, how are they usd. for predn. How do these new found reggys modify subsequent searches for reggys (≈ compression)?

.23 "Soln" to .19: "Invisibly all of the reggy found thus far" Try all possl. Codings ≈ of the Corpus in order of pc. ~~[redacted]~~ Store all significantly useful Codings. (i.e. is a right compression are obtained)

A possl. Approxn to .23: We are considering only codes that are operators. So we have a conditional pd, betw. input & output. We start w. a set of primitive primitive functs (or perhaps = eprip). We code cond. Operators as functions defined by trees. [I have to work out the details of the coding of these trees — I will try modifns of "Z(4)"]. Hopefully, these trees can be described in ≈ pc order using the "Huffman tree method" (130.19; 142.01 143.05 ff, 144.01 discusses PSG's (phrase Str. Grammars) as models for "tree search" — derived vs. non-derived ~~[redacted]~~ languages.

So, anyway, we list these Conds in ≈ pc order using ≈ Huffman trick. Each function has a pc assigned to it, so we can do the "Huffman listing". Any significantly "short code" ~~[redacted]~~ for the corpus, becomes a newly defined Operator (≡ function).

~~Also~~ After defn of this new operator (we ~~need~~ any defining several of them in "11 codes") we have for each such new Macro Function, 2.20.01

9.3.00 Bug.

.01; 219.40 — A code for t. corpus: This is a "2 part code". Part 1 is the "Precorpus" that consists of the primitives, followed by ~~the~~ all of the definitions of functions (that terminates w. t. defn of t. last Macro funct). This is followed by t. dern of t. corpus in terms of that last Macro Function: This lasts in t. style of Z141, in which t. pc of any element will depend on how many times its been used in t. code thus far. [I haven't yet figured out how to do this ~~for finite string induction~~ US Unordered set induction v.s. TS induction time series)]

Dol — Think of CFG's. — How t. various pc's for choices get updated w. each "use" of them. I do want to write this up in some detail! | Finite string: US induction, TS " "

~~This process was t.~~ My not understanding this process, was t. cause of much confusion about pc's of functions & "sub-functions" in both t. coding of t. corpus & t. coding of new functions & sub functions.

(SN) If I really work well only during Morn (& sometimes before going to bed?) spend much of day preparing problems so they are in form ~~that is~~ most easily worked on in Morning (/ before bed).

Still, much confusion in my mind about difference betw. Stochastic Operator (function) & a stochastic CFG. Well! one big difference: CFG is a stoch. operator for a constant (invariant) input. !

.18 One Try: say we've coded t. corpus (a MTM corpus) using a string of definitions culminating in a certain Macro funct, Fo. We get a new bunch

.20 of I/O pairs that don't seem to fit well. What do we do? : Well, (i) we try our operators in PC order: (?) This would mean starting w. t. primitives & using each of the Macro Funct's that worked in part of t. past TSQ. Actually, none of them could work & they all give a zero pc to t. corpus — (→ they have all been tried on smaller parts of corpus & failed!) so diff't to compare them. Next in order of PC, might be taking Fo & I/O other "Recent" Macro OPS & mutating them, by inserting & substituting different functs for sub operators (sub trees) within Fo and other recent Macro functs. ~~Essentially Mutating~~

.27 Actually, MTM problems are always INV probs. — So maybe try Lsrch. In this case, TM has to "recognize" t. new kind of problem (recognition involves combining concs., just like any other problem.) "Recognition" is a kind of "Categorization" which is an impt. kind of knowledge

.32 that must be acquired.

.27-.32 is a common approach to t. problem of .18-.20. We try to find a minimal code telling how t. "new input" differs from all of t. old inputs. (One ad-hoc way is t. passage of time.) Actually, time may not be that "diff"! The new (intractable) input may be identical to an input in t. more distant past — but "Times change" — so TM has to adapt. One way would be by finding a new soln. for t. recent data, & give wts. to t. ~~old~~ old & new soln. — so we are not

222.01

"MCT: How to do everything Rite, Every time." Title for Talk.

This changes it from a MTM to a NMTM problem!

.08 | 220.40 | 100% sure of t. new soln. T. reasoning wts ~~with bias~~ should bias more strongly toward t. "recent" soln., as time passes.

While 220.27-.32 is, indeed, a common approach to this problem, I do feel it is somewhat A.H. & I'd like a more General approach.

This is in line w. my warning re very simple/common pgm for ALL or TM's problem types & "Modes".

It may be that my considering only MTM Type problems in 220.18-.20 is where the trouble is. In general, TM is supposed to have this very General Condl. PD that contains, stores, all of its "knowledge", including Heuristic "knowledge".

.10 This is MCT → The ● functional Macro Funct. sought in 220.18-.20 doesn't contain enough info to be very useful for Heuristics ← Heuristics are usually "NMTM distributions — i.e. pc's are not always close to 0 & 1)

However, t. MTM d.f. of 220.18-.20 could be (is) a part of the more complete P.D. of .10.

I think that PD of .10 is my simple, unified model for TM Inv. It is t. P.D. that TM2 spends its cc "Updating".

But how does my approach in Sol89 fit into .10?

Sol89 was making new kinds & concs by combining older, useful concs that were found to be useful. How does S89 fit into MCT?

This seems to be t. rite Q! In fact, whenever I get into trouble, I should always ask: "What does MCT have to say about this?" — or/ More specifically what does MCT have to say about t. SAARB TSP's?

So go thru a Saarb-like Tsp. & see what MCT says about it ... how it should "really be done. T. General idea is that t. P.D. helps solve problems, & after each problem is solved (or even just "worked on"), t. P.D. is modified, to "contain" this new info.

.28 So for t. ANL problem, we start w. our primitive functions, & means of combining them. T. funct. Summaries have pc's, as do t. ~~other~~ operations that combine them. This gives rise to a P.D. for all functions. They are searched in w. Cost order (or even just "worked on") that do ANL as far as we t. comes goes. So we find some new functions that are useful (so far).

.33 How do we then Modify P.D.? To deal w. t. new examples that "Don't fit" t. old Macro Functions, we add these functs to t. set of primitives: but what pc's do they get?

Also note: in .28 we used an unconditional P.D. In .33 we need a conditional pd. One of the things a condl. P.D. considers,

.37 is "what to do when we have to revise a theory." Given a "theory Revision Situation" — we want a P.D. on techniques to deal w. it. (denote a)

9.5.00 Bulg.

.31 On t. UNIVERSALITY of a PRIM. Rec. Funct Generator.

.011 222.40 : While 222.37 is close (or just what) we want to do, we may want to "Elementalize" (break into Tentative AND or OR) sub problems, (that may or may not solve t. main problem). { Remember that in breaking a task into sub tasks (AND/OR not) part of one's time is spent trying to (prove) show various of th. tasks are impossl. ! }.

T. cond'l P.D. of 222.33 is t. MCT P.D. It characterizes t. entire corpus thus far, (as well as t. techniques we've used to try to solve probs (which is new part of t. Corpus). Hvr., in this prelim. discussion, I may not want to consider t. Entire Corpus .... but this is unclear just now.

Another point: Storing t. Cond trials as part of t. Corpus, would seem, at first glance, to be pointless, since t. trials contain Zero info: T. trials are completely dscr bd by t. "stated TM" & t. "Lsrch Algm": — Hvr, t. forgo. assumes CB = ∞. For finite CB, it may be worth while to store certain info in t. Lsrch (other than final results). An extreme (tho a bit useful) case would be storing very many of t. short codes of t. corpus, — t. less short ones being good cands for later "Theory revision". [T. Revision Problem"]

12:55 { So T. Big Q is "How is t. Big (MCT) P.D. (up dated / revised) after a (successful) Lsrch? How will TM's subsequent behavior be modified? ← (More General Q).

Consider t. early T.S. Q. 3=? ; 3+4=? : We could start w. a simple P.D. over a (potentially universal) set of functs/functionals. (possibly (.36) a universal or Prim. rec. Universal Algm)

.22 [SN] — It is not clear what a universal Prim. Rec. Algm is: Perhaps if $A(\alpha_A, x)$ is a prim. rec. algm, A (α_A is t. dscrn of a P.R. algm, x is its argument), Then if $U(\alpha_U, x)$ is t. same funct, implemented on Universal Machine, U; Then for all α_U, α_A $|\alpha_U| - |\alpha_A|$ is bounded by a constant that is a function of A & U only. This would be t. case if A were a Universal Algm.

Or, say A^1 & A^2 are 2 algms that can generate all Prim Rec functs, Can A^1 "majorize" all generators of prim. rec. functs & not be a Turing universal function

.31 ⟹ Since t. prim rec. functs are effectively enumerable, A(N, x) could be Prim. Rec. Universal (N is t. index number of t funct. desired), & could dscrb. only Prim Rec. functs. I suspect that one couldn't get signif'tly small "N" 's a Unc !

1:36 It's not clear that my PrimRec. lang. is Universal in t. sense of .31: but Drop this for now

.36 : (.22) consider t. set of successful trials: ① for 3=? find t. number in t. inpt. ? to that no. (Equality function) ② for 3+4=? 3 solns: ① ② find 2 nos. a, b; "?" ← Sum (b, a) ← Sum (a, b). } Same pc for these 2 solns.

1:51 ③ Much hyper pc soln: if "+" occurs, ? ← Sum (a, b) Sum (b, a) else ? ← a (a = any no. in inpt.) ⟹ (224.01)

9.5.00 Bulg

.01: 223.40 [Unclear as to whether I want to include considerations of "?" having to be a number & t. final funct into Macrofunct having numeric output. ("TYPE" info). The "Type" info could be an invariable accompanyment of all symbols., so each item has 2 subparams: Type & Value.]

.05 Or I could let all symbols be same Type, & let TM figure out about "Types"

The .05 would making looking for "+" diff! TM, say remembers all symbols that have occured, so it can decide if they occure again. (i.e. so t. "=" can be xfmd into a Boolean funct. — By using "Types", I ↓ t. no. of Possys Tremendously! This is because t. no. of Numbers is enormous! A way around this is to use only one sample per example, & use 32 bit/random integers. Even so, I'd want to give new random nos. for each new example type. 3:34 & 4:00

Apt #3 #4

But I was mainly interested in running a not-very clever TSQ — since my techniques supposed to work w. all

.16 Speaking in general terms: After a/new subcorpus has been workd on & a soln. for this obtaind: How is Big P.d. updated? (First), t. params of old concs are changed.

.17 Def BPD (Second) new concs added to (B.P.D.) Big P.D. w. assoc. params.

.19 (Next/third) TM₂ can work on entire P.D., TSQ to look for better regys. Possibly a complete Recoding, or partial recoding, or Coding of Code.

.20 Can I use a sort of arby B.P.D. & see how t. new sub TSQ

.21 experience changes it?

.22 [My impression: In a TM working on MTM probs, there are 2 kinds of sub-Pd's: One for Math probs — in which pc ≠ 1 or φ & other PD's used to guide heuristic Search on Math probs & perhaps other

.25 PD's used for NMTM-type Problems.] Noted Before!

To start. .16 – .19 + (.19 – .21)R. maybe O.K. I'm not sure it addresses .22 – .25

. [SN] I had considered a usual Max of 3 or 4 functions to be combind; but this limit could be much exceeded if I had a particular method of combining functions that was often used. (This is a "Meta Pgmg" idea.) One example is "loops", a combination of at least 4 functions: ① Counting function ② Main function ③ threshold of termination function (Boolean) ④ Initialization. (Unclear how ④ is a "function" — it could be an input)

.36 Hvr. Check on Peter's Germ. It may be that ① can always be j = j+1 & initial value can always be 1. So t. loop functional has input n & function ②. Essentially, what we do is take function ② and add another argt. to it, "n". If ② has already got n argts, we need n diffnt functs in

$$\left(\begin{array}{l}\vec{X}_0 = \vec{z}_0 \text{ (initalzn.)} \\ \vec{X}_{i+1} = \vec{f}(\vec{X}_i)\end{array}\right)$$

(2): Essentialy, $\vec{X}_{i+1} = \vec{f}(\vec{X}_i)$ & repeat n times.

.02 A big Q is 224.36: do I need an arby "counting funct" w. an arby form of "stop" Threshold? — This would be useful in telling, say, when an approxn. procedure should stop. — But do we need this for Prim Rec. Functs? — See Kleene "Introdn. to Meta Math".

(unfill'd 225.02 Ex 5.26 cuc ma.)

.02 is'nt actually a central Q: t. Q is: to what extent is 224.16–.19 "adequate"?

{ Re: 224.16–.19: Consider various possl. "Solns" to TM's problems. Can each discovered Regty be put into t. 224.16ff forms?

Can what TM learned from t. Soln. (t. Solving proces) be put into 224.16 form?

In general: 2 parts: ① soln of problem ② Info from t. "Soln. process".

An extreme example: say t. problem is $\sqrt{2} = ?$ One part of soln is 1.414···

T. other part is how this soln was obtaind: t. number 1.414··· mite be stord as a soln to $\sqrt{2} = ?$, since this problem may occur again. Whether or not it is stord depends on cc of storage & cc of computation & expected time before info will be needed.

IN GENERAL in 224.16: This can be done in many ways,

.18 each w. its own "of completeness". In one way, t. B.P.D. consisted of a set of functions, each w. its own pc, & a set of combining rules (functional(s)) each w. its own PC. In 224.17 (first) t. pc of the functs & functionals are changed. In 224.18 (second) New Functs (& perhaps

.22 functionals) are added, each w. its own initial PC. The "New" functs will be Macro Functs used in solns to probs & [Functs. represented by] subnets that are used frequently ("These subnets" — I haven't found TSQ in which they are useful & this

.25 is a source of "Unease"). Re: New functionals: I don't know if this is necy, or if so, how they would be discovered.

In (third) 224.19, the revision of t. entire B.P.D. Code: This also needs much

.28 work: It may involve discovery of "New functionals" (.22), (.25).

Anyhow .18–.28 is only one way to construct & update B.P.D. How well defined, how efficient, how complete it is, is unclear at present.

.18–.28 is an "Eng" desc'n of (2) Induc. System. I could go into more detail, or Generalize it more. — Make it Vaguer: More Englishish!

A MAIN Q: Can All t. induction methods I can think of, be fitted into this (.18–.28) Model?

AN APPROACH: Work some (Human-w.i.c.) problems & see if I can fit t. techniques into (.18–.28).

SN Consider folg. problem: How would TM have ① Motivation ② Ability to work it: Given $f(x)$ from a to b is monotonic ↑, to find $\int_a^b f(x)dx$: Discover Simpson's rule: Bounding $\int$ betw. upper & lower limits.

③ Could TM have motivation & skill to discover higher order versions of this algm when $f'(x)$, $f''(x)$ were monotonic on int. interval? THEN

And: Find order of approx to use v.s. no. of pts evaluated $\ni \varepsilon$ cc was min.

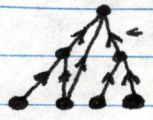
Would TM have to have a kind of "AM" like "motivation" to work on such problems? For the initial problem, the problem definition might be "find $\int_a^b f(x)dx$ within $\pm .001$ & prove this is correct! (or make it very likely that it is correct).

Perhaps study "History of Math" to get ideas of / ordered sequences of concepts.

A possy. is that the function nets I've been considering, ~~do not~~ are not always able to obtain decrb. all solns. to problems. If not, then concs are more general than functions, & TM can make derns of problem solns as "conc nets" (as in Sol(89). The updating of the params, will be as w. function trees)

Woops! There's a big difrnce betw a function (which must be a tree) & a conc, which must be a "net" of concs, but a conc net is somewhat more general than a tree! — The all function trees nets are conc nets

all conc nets are not trees.



This conc. net: it is not a tree.

.22

.23 Woops! A function net need not be a tree either! (.22-.23) Could just as well be a function net!

So it looks like conc nets & function nets are isomorphic.

Neither need be "TREES"

The idea is that any idea (≡ conc) is formed by combining other (earlier) concs. (This is an Eyelid statement / down on the process)

If the concs have simple unconditional pc's, it's easy to update the pc's: They are independent & are obtained exactly as in Z(4) — so pc ∝ no. of times each conc. has been used in the solns to problems that have been accepted.

In the more general case, they are not "indep" but are conditional pc's: The conditions can be arbitrarily complex. I Guess updating indep conc. pc's may be simple (viz Z(4)) ··· but updating cond'l. pc's seems much more diffl.

On second thot, updating (function/conc) nets may not be so trivially covered by Z(4) techniques!

→227.01

9.7.00 Bulp.

.01: 226.40. Consider Functions & Concs. Separately.

First, Functions: To Construct a function: Start at the top (terminal) Function: Using the corpus of successful Macro functs: Each funct has a pc of being chosen as "top funct", — but we will, at first, want to pool this info, because, at first, SSZ's are small. So, say each funct or connection to the input, has a certain pc (initially, say, they are all the same). After the first funct is chosen, Its inputs have to be chosen, & they, too have the same D.F. over all of the functions & (Macro-funct inputs)

So, given any Macro-funct, as well as the assoc pc's of the functs & the Macro inputs,

Def: we can assign a pc to the construction of that M-Funct. (≡ Macrofunct).

These probability assignments can be done exactly like in an advanced Version of Z141, with [redacted] "long" definitions.

Def .13 The Pre-corpus, consists of the M-inputs (≡ inputs to the Macro funct) followed by the primitive functs. We then define our first funct. It can be one of the M-inputs, (in which case, it becomes one of the M-functs) or one of the primitives: If it's a primitive, it is followed by its inputs, which must

.15 be M-inputs or The next function can have as inputs, any previous function output, or any m-input (Loop Recursion) to itself. The P.C. of each choice made = (no of possi. identical choices)/(no. of possi. legal choices)

.17 So. if an input was F_3 & F_3 occurred N_1 times in the previous code, & there were N_2 legal choices in the past code that could have been made,

.20 the pc is N_1/N_2: This is same as normal Z141 coding. (also ≡ "Laplace's Rule")

.13–.20 Tells how to code/get pc of any M-Funct. That's pc will be indep of the pc's of any M-Functs that this M-funct does not "refer to" (That does not "refer to" in this defn of that M=function — There can be parallel defns of this M-funct, that use entirely different M-functs in their defns).

The isomorphism of concs & functs can be used to assign pc's to concs in a formally identical way to M-funct assignment of (.13–.20)

In a Conc. net, each node in the net is a "combination rule" w. several inputs: the inputs are "lower order concs."

.30 In funct. nets: Conditional pc's can be obtained empirically by case counting, but I suspect that this will not give much info usually, since SSZ's are

.32 too small: Probably condl pcs will arise from "reasoning", analogy, etc. → (228.01)

Re: (.13–.20) (The method of): That gave the mathical value of the pc. To actually get it, each Funct would have assoc. w. it, its "slots" ≡ "property list", a N_1 & N_2 value (.17–.20) So we could rapidly compute (& update), its pc.

(.13–.20) seems fine for functions & fine for Concs. Since funct. analysis is simpler than Concs., Do function problems first. — But the analysis will apply exactly the same to conc nets. One big primary problem is Conditional PCs (.30)

9.8.00 Bulg:

.07 Present Branch of Activity: But this on the STACK

.01: 227.40: Re: Analogy (227.32) This is common structural properties of concs. Usually a conc will have several possl. derivs. Only using certain of these derivs, will "structural similarity" w. another conc be recognized. So holding "parallel codes" can be useful in this way. Also, if we suspect that 2 concs may be analogous, we could examine alternative codings of both of them.

.07 → So: Maybe do this: Do Alg. Impl. using only 227.13-.20 (& possibly t. simple condl pc's of 227.30) See how far I can go w.o. needing more complex concs, or more carefully constructed condl. pc's. (than "Functions")

.10 Some Q's: will 227.13-.20 work for only MTM funcls, or will it also work for NMTM funcls (Stochastic Operators)? Note that B.P.D. (224.18 ff) is a stoch operator

.10 is an interesting Q! → Because it deals w. a confusion that I may have had betw. t. pc of a function (which can be deterministic or Stochastic) & t. params t. derive t. condl. pc. of a Stoch. Operator. → See 247.19 to at least 249.30 for discn: I'm not sure this was complete discn. to t. stochastic problem!

I usually Consider 2 kinds of Probilistic operators/situations:

.16 1) T.S. predn. (regression) (symbols a/o numbers)

.17 2) F.S. predn. (Languages) (called "Symbolic Regression" by Koza)

Give Examples of stoch operators I may be able to discover using models like 227.13-.20. Well, there is, for T.S. predn, Z-cfl, w. Wolff's Breakthru term (of "parsing every once in a while"). Note that this is a coding method, & doesn't directly give a prediction operator.

The T.S. predn method I have of defining all post fixes of t. corpus, & giving a wtd sum of their predn., is closer to a stoch. operator.

It would seem that t. method of Z-cfl could be applied w.o. much modifn, to a corpus of unordered finite strings (a "language") — (Also I have all these ideas on PSG diary)

Avery BIG Q in all of t. large. To what extent will it be able to deal w. t. "Scaling Problem": The ↑ in no. & variety of available concs: t. necessity of having heurs (condl. pc's) for t. concs to narrow their pc's down to t. application (& which may need to be (progressively) used.

So: Thrust: To make a TSQ "Acceptable": I have to show how each new needed conc, is derived from older, useful, concs. Also, I have to have condl. pc's for these sub-concs, to show why they were likely to be tried as components in t. (present, relevant) Sidra [illegible] 187.33 ff

.36 I think our idea of t. "ob-op" approach was that four good condl. pc's: One way (not t. most general) to get Reason is by building up a set of "ob" concs that build on one another hierarchically, to be "categories". These "categories" are then "correlated" w. t. op operators

9.8.00 Bdk.

Avg 3# (kcat → 2936 136
kcwd 3860 179
Pushing busts of Lithography 3854

.01 So we can know under what conds (its more likely) that we should try these operators.
In general, t. rate of Growth of t. "Ob" vocabulary, (conc set) must be at least as rapid as t. growth of t. "Op" conc. set. — These rates can be "Served" by TM_2 toward approx equality. One signal that there are "too many OPS per OB" is that t. cc of search for acceptable ops is too hy. — (But there could be other causes of this last diffy!).

Re: 228.07–.17: This problem needs clarifn.! I don't have a clear picture of how discovery of stoch ops is done: how "categorization" (condl. pcs for search) fits in, etc.

I may want to do a good review: A more detaild descn of just how I expect this Prelimy TM to work.
Probly best: outline how non-stoch ops found; first: Random w. a few kinds of stoch ops: (228.16–.17)

"Similar subtrees" in many cases — coded by similar strings.

On t. coding of a seq. of MTM problems: (solns. are functions)
The TSQ is A, B, C, D, (each is a set of examples of a function).
Given A, the (greater) soln is F_A.
" A,B " " " " $F_A \frown F_B$ where F_A is a descn of an operator in terms of prim. funcns.
F_B is a descn of an operator in terms of prim. ops & F_A. F_B solves (A,B)
.22 (NB) Even w. only 2 functions, F_A & F_B can have common sub-functions
.23 (≡ sub-trees) that will ↑ pc of t. code.
Given A,B,C, $F_A \frown F_B \frown F_C$; F_C is a descn. of an operator
.25 in terms of prim ops & F_A & F_B. Also, t. remark of .22–.23 holds even more strongly for t. code $F_A \frown F_B \frown F_C$ than for $F_A \frown F_B$. (More likelyhood of (above-threshold-frequency sub-trees) in a larger corpus).
When common subtrees are found, they are added to t. function list (w. proper
.29 assoc. pc's) for t. construction of new trial functions.

{.22–.23; .25–.29} seems to clarify a lot of my confusion and "unease" about this lack in my SAARB TSQ.

I suspect that t. no. of ~~examples~~ new example types in the SAARB TSQs was far too small for t. effect of .22, .23 & .25–.29 to be ~~observable~~ observable!

Also Note: It is possl. to "not need" the ability to detect sub-trees, of .22–.29 if one puts examples in t. TSQ's that use those subtrees as solns. (No! In t present system, this would not work! Even if I gave TM a problem in which a particular sub-tree was as soln. (say t. sub-tree was f_0). Then f_0 would never be a M-funct. — we would (230.01

.01: 229.40 have to have an Ob recognize when Fo was relevant, & have Rel (Ob, Fo) object, & joined (added on) to [redacted] M-Function. — Or have a general reformulation (total code revision) to fit Fo onto M-Funct. — But in general, Fo could never (by itself), be an M-Funct — (unless it was fr. soln. to the first [redacted] set of probs. in the TSQ.

In fact: I had Rel idea of teaching TM definitions (in Algebra) as always an important part of the TSQ. The Rel could, indeed, simplify TM's learning of simple concs: I think the mechanism of finding "common traces" [or much more difficult: OSL of such concs!] would be necy. before TM could use those concs in solving new problems.

T. Main TROUBLE w. OSL, is that it requires either an enormous amount of searching (for functions), or some kind of INFO-Retrieval — (≈ "associative" Memy).

One approach to the OSL problem would be to look at cases of subtraces that I'd want TM to learn, & try to find ways to access such — either either by ↑ of pc a/o by some hardware method to cheapen associative such — or associative search whatever

Case Based Lrng is OSL, but I think it models the Conscious aspects of the process only. There are probly many strong, idiosyncratic, (individualistic) techniques of categorizing, association, that are not "conscious" — perhaps many kinds of "Analogy" (≡ Structural Similarity).

.22 One kind of OSL that may be common: Say I give TM examples of a named conc. to learn ("Definition Irng"). The rest of the concs that are learned in this way, occupy a special "named" kinds of "positions" in the resultant M-Funct. When new problems arrise, TM looks at those positions as being good Cands for OSL.

Another approach (a mild Generln. of .22) to OSL: That the M-Functs are normally "pickled" to some extent, so we do have Obs & Ops & correlns between the two — So TM could do a search over all Ops that had a highly correlated Ob — in the OSL srch. i.e. "which trick (Ops) have been useful in the past".)

.33 An elementary TSQ that could be very revealing!: MTM tsq, small amount of simple Irng (say ANL). Then we teach it a defn: Then we give it a problem in which that defn. would help solve it. This is OSL, & I am concerned w. the mechanics of how TM would deal w. it. I think the problem is easier than looking for common sub-traces (— because there are many, many sub-traces to look at — in this case Based Irng, OSL, we only (search & examine) a much smaller part of the corpus for relevant cases (I think!)

2 31.01

.01: 230.40 In general, [redacted] Lrng. definitions involve creating an OB. (funct. w. Boolean Output). Normally, just lrng a definition is not very useful in itself. It can be/become a useful OB, only after it has been correlated w. a corresponding OPS. (or a corresponding op.)

SN I have been considering only HTM problems that were "instantiations" — as opposed to problems involving a sequence of operations: usually obs followed by ops: The obs told "what to do"; the OPS "did them". Or, we can think of a "ob-op" combination as a single OP, & just have a sequence of ops.

Re: Lrng "Definitions" &, & OSL: If an OB has ever been "useful", it is a Cand. for OSL. Hvr, certain "utilities" are more useful than others. Useful in a taught "Definition" will have a hyer probty of being a Cand. for OSL. Hvr, in general, when a problem is presented to TM, all of the obs that TM has lrnd, are applied to it to understand it (the problem) in as many ways as possl. Also the no. of OBs that TM has will ↑ (perhaps linearly) w. time; & (amount of time) needed to find a match, will probably ↑ much more slowly.

.19 I'm thinking of OB recognition" as simple sub-string recognition! That an OB is represented by a string or set of strings. To recognize or not one of a set of Cand. strings in a longer string, is a well defined problem. The Cand. strings can be in lexical order. — So recognition is like my "soln" of the DNA sequencing problem. (use of B trees): Other related methods: Zato Coding, Hash coding (see Knuth's Volumes 3 on this last — or perhaps he has refs to other work on directly-related to the problem of .19) Maybe ask Simon!. Ask Alex.

.19 occurs in Z(4) in reversing the corpus.

Since .19 appears to be an impt., often occurring operation: perhaps special H.W. can be used. For Decryption such string matching H.W. is available — but I don't know it. H.W. is always revised w.r.t. "Moore's law": A large amount of Parallelism can be used, & of course, any very large computer will be highly ∥ (say Beowulf).

Hvr. Obs can be Boolean combns. like AND, OR, NOT. Still, recognition of the components of a Boolean Expression can be done via .19.

Also, there are Hierarchical means for Recognizing Obs. — so a part of an ob is recognized, so a large set of Cands are now narrowed "super Cands" — which narrows down the problem.

Zato Coding
Hash coding
See Knuth for

232.01

.01: 23/40: Another technique for Ob recognition is that used in LZ coding. (very fast & very "cheap" very little RAM needed!)
— Here, in contrast to LZ, I would also store info on how many times each substring occurred.

SO! the OSL problem may not be so "computationally intractable"

This [illegible] is a bit simplified! In many cases, an OB is more complex than just sub-string matching. I.e. to recognize that a certain substring is an equation, or that it's a linear ODE, etc. is certainly more complex than string matching! A More Realistic Model would be Pandemonium — in which each OB looks (in II) for strings that are "a case of it". A Q would be the CC involved in this. Also, whether one might not be able to save CC by doing sequential testing for Categorizing.

Do a bunch of Examples of Definitions and a perhaps other forms of Concs. that have to be learned by OSL. Go Further by the learning of the Conc. & the [illegible] later application of the Conc.

Normally, we give TM a new chunk of corpus (= "sub-corpus"). Say it learns to watch probs of the corpus; even w.o. recognizing the new sub-corpus, this probably will much of pc of coding. But next step is to "recognize" the new section of corpus. This is an "Ob Job". — presumably made much easier if some of the many Concs have been (made as "definitions".

How to use Neg exs: We just have to categorize the input into 1 of 2 categories. TM assigns pc's to each category: the Conc is a max value of $\prod p_i$; p_i being the probs assigned to the "Rita" category.

O.K. so: Defining problem is given:
α, examples, Def. Say Rita is pursued for learning that example is a case of α. — TM is then given queries of form ?, example, Def. or perhaps Def, α, examples, (Y)"; then Query is "Def, α, example, ?".
Say TM finds Ob_α that gives right answers for the [illegible] examples sub-corpus.
We also have to tie Ob_α to "α" and "Def": Assoc. w. α & Def are Boolean operators that recognize them, B_α, & B_{Def}: so the final operator is Ob_α And B_α And B_{Def}. (Note I've ommited some details here). Say OP_M is the previous M-Function — so we want the new M-Funct to be Ob_α of B_α and B_{Def}; else OP_M.

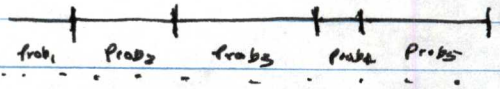
In general, the form of the M-functs (at first) will be elementary:

.37 1 If ob_1 Then OP_1
2 If ob_2 then OP_2
⋮
n If ob_n then OP_n.

See 235.16 for the comparison

Bug.

.01 : 232.40 : The form of Soln. of 232.37 – .40 is o.k. for a certain type of corpus: One that consists of a sequence of "separate, clearly defined, problem types.

.02 

I think that each of these "problem types" & gives TM_2 some info about where to spend much time on a "single problem", vs. how many of the entire past corpus & expected future corpus.

This form of corpus is quite "restricted" but many, tho less, simple, difficult problems can be put into this sequence. A big Q is: Can I put the Understanding of English text into this formalism?

.08 Superficially, Obj_1 can be a device that counts the no. of probs that have occurred this far — but I can outlaw this by making TM regard the problems as "Unordered". — by giving TM no functions that could detect this order. On the other hand, I may want TM to be able to detect "Recency": So if I teach it a defn., then in the near future of that defn, it will expect problems using it. (This is a rather broad "Hint" — perhaps an "unmotivated urge.")

I probably can arrange so that TM within cannot do .08 : — but I'd like to know just what other restrictions on TM's capabilities this would imply!

Actually [redacted] if TM did .08, it would be unable to solve new problems! In fact, the soln. form of 232.37 – .40 is also very limiting! If I give it a new problem that is not [redacted] any of its training problems, it will not know what to do! — Yet, after .22 it has solved a seq. of diff problems, it should in some sense, be very smart!

.24 Well, Look at it this way! Say TM has worked the TSQ up to a certain point. "We then" given it a problem: "string ? ". How should it best code a P.D. for "?" ? Well, presumably it tries to get short codes for "string of", — α being various possible continuations of "string".

.28 Actually the situation at .24 isn't so different from the situation when .28 TM is given its first problem examples of a new problem sort.

In .27 – .28 TM has to get a pgm. to produce $\alpha^{\frown}\beta$ (where "β" will be essentially replaced by "?" in the query as part of the prog.) ....

.33 Well; 2 comments: (1) If a person were given a totally new problem under .34 those circumstances, he wouldn't know what to do. (2) The [redacted] TSQ given to TM was really of the form $[Q_i, A_i]$ — but it was broken into distinguishable subsets $[Q_{ij}, A_{ij}]$ "j" is the name of one subset.

Consider the TSQ: Solve linear eqns, Solve quad eqns, Solve cubic eqns. We can consider this seq as 3 subseqs or 1 big seq.

235.01

235.10

9.13.00
Bolg.
b

235

9.13.00 Joyce 783 8785

.05 :234.40 i.e. we can use t. info about t. 3 divisions or not. In t. [redacted] latter case, we mite regard them all as "equations" to solve.

Also, even when [illegible] TM clearly uses t. structure (234.02L) and t. recognition Ob, 232.37-.40, we can regard TM as solving 1 category of problem, with 232.37-.40 being an analysis of t. problem that tells TM what to do. — Hvr, I guess "232.37-.40" is a poor analysis of t. problem: it doesn't tell how to extrapolate itself into new domains: It is rather A.H. (even tho it has a lot of "learning in it" (234.22).

.10 :234.34 [230.33-.34 may be t. rite idea! — That unless TM has been specifically taught to generalize input analysis, it will not be able to do it — & shouldn't be expected to be able to do it!

So, one Q is: 234.02L is a desirable structure for lrng: & it does get into init (234.22) [illegible] n w. t. soln. format 232.37-.40: Can we somehow use that form to get more useful lrng?

.16 [A Very Serious Criticism of t. 232.37-.40 soln: T. search for OP & t. srch for Ob are separate & not related: Normally, I'd like t. input to create a condl p.d. for t. search of OP's: This should ↓ cc of srch tremendously! It's a really broad area for creation of heurs! So: while 232.37-.40 mite work for a relatively small corpus, it would not have enuf heurs in it to scale up to a large corpus. — As t. corpus Grows, t. no. of cases available grows, — w/o any narrowing down

.23 of t. search that would be obtained by "looking at t. problem". → (236.16

.24 To t. extent that 232.37-.40 doesn't "tie t. lines" on MCT, it is deficient. Ideally, TM should use info on past (corrlns/relations) betw. Ob & assoc'd OPs,

.26 to help in each new search for an OP.

.27 [One way to do .24-.26: After TM has (using 234.02L & 232.37) done a reasonably large corpus of example probs., TM should try to find useful, (predictive) (probliste) relations, betw. t. OPs & their corresponding obs. This "scales"

.29 o.k. — since t. corpus is small when one just uses (232.37, 234.02L) Later, one really needs t. relations of .27-.29 to ↓ cost of srch.

Unfortly, for a infant TM, discovery relations of t. sort needed in

.33 .27-.29, may be too diflt. — It may be necy to put this in ≈ AH. Another way to deal w. this: Use t. strict 232.37, 234.02L TSQ form, until TM has acquired enuf "smarts" to do t. induction required in .27-.29. — Hvr, unless one has very great IPC (maybe 10^15 bits/sec?) it may not be practical to do this, because before TM is smart enuf to do .27-.29, it will run into "scaling problems" — I could do t. partly AH thing of .33 to help — to avoid need of super hy IPC. equiv. to A.H., would be different degrees of BROADNESS of HINTS.

236.01

.01 : 235.40 : So, OK. Say I have this set of [problem$_i$, ob-op soln$_i$]
.02 Def corpus of soln's of t. "STSQ" (Simplified TSQ) (of 232.37 - 40, 234.02L)
.03 for this set of pairs (of .01R) I try to find a probabilistic operator (as in 233.27-29) to ↓ t. p.c. of that set of pairs. I could do it in an operator form of solving $[Q_i, A_i]$, in which I assume "no info" in t. $[Q_i]$ set.

So .03 ff sounds fine. Note t. stuff on by IRC a "Hints" of 235.33-40

T. original problem set (TSQ) "STSQ" was a $\Sigma [Q_i, A_i]$ form, w. "clumps" $[Q_{ij}, A_{ij}]$ ("j" is a "clump name"). So we ended up w. $\{Q_i, OP_i\}_{i=1}^{n}$ & we want to extrapolate this sort will work w. Q_{n+1}
.11 t. problem of ill is just like any other $\Sigma [Q_i, A_i]$ prob, except that up until now, I've been doing MTM probs & this one is MTM(problistic). OB
In .11, I should be able to use OP_i w.o. to associated "recognizer" (π) for t. "j" group.

.16 This gen'l ("M_2-like") method is also used for unsupervised parsing heuristics into t. Lstrch, & to deal w. "scaling" — it deals w. 235.16-23
What I'm doing in .02 is a "recoding" of t. "soln" of (.01R).
STSQ of .02 does give a logic code for it's corpus: Th. code is one
.20 single final M-funct, along w. t. recognizers Obtd 232.37-40.
Nvr, this code can be much compressed! .01 is one way to do it.
The corpus + t. code of .20, gives can create [the data string of (.01R)].
From (.01R), any stoch operator that can get a good pd of soln's from problems as input, could ↑ t. p.c. of t. M-funct of .20. One way this works: say we are able to ↑ t. pc of one of the prob: soln pairs of .01R: then the pc of t. final M-funct is also ↑, since t. pc of t. final M-funct depends on current pc of each previous M-funct. If all of t. component M-functs have their pc ↑, t. final M-funct has it pc ↑ even more (perhaps n cases final M-funct is $\approx \Sigma$ #'s of all component M-functs.

How the stoch operator ↑ pc of component M-functs: It looks at t. problems & makes a kind of pd on possible solns. This p.d. can be very direct, or it can work by making narrower conditional pd's for t. functions that are components of soln's. (These "conditionals" can be on t. partly-constructed function for solns.) — in either case, λ.
.35 pc of soln's would ↑.

Next, I want to go into more detail on t. "entire system":
.37 First, we have this set of sets of $[Q_{ij}, A_{ij}]$. "j" is name of subset of Q,A's of common "soln". At this prob we do $[Q_{i_1}, A_{i_1}]$ first: we do an Lsrch
.39 over primitive functs to find $\exists f_1 \ni f_1(Q_{i_1}) \approx A_{i_1}$. (Since this is at first 237.01

90,72. B

.01: 236.40 : a MTM TSQ; "a" "13" = "in 236.39)

When we solve problem 1, we go to problem 2. We solve it by Lsrch, using primitives + F_1, to obtain F_2. For the entire corpus $[Q_{ij}, A_{ij}]$ $j=1,2$

.04 F_1 works about ½ the time; F_2 works about ½ the time. This gives a stochastic Operator: which we try to improve, by finding obs that can look at a Q_{ij} & tell if j=1 or 2, so it can tell if F_1 or F_2

.07 is applicable.

We then do $[Q_{ij}, A_{ij}]$ for j=3 and integrate it into a single Grand Funct (≡ M-Funct) as in .04. .... we continue this with higher values of j, until the cc of solns is too big: They are too big, because we have too many functions to construct our new trial function from — & these functs all have about same pc.

.12

.13 .... Next we have to go into Phase of (236.01 – .06): We want to find regularities in the code for the operator that does $[Q_{ij}, A_{ij}]$

.14

.15 SN Perhaps better we go to .13: In the soln. process of 236.37 – 237.12 I had been thinking of just using functions that were primitives or were F_j type of functs, or Obj type that identified j of $[Q_{ij}]$. I think I will need more: I will need simple OSL. Also, OSL (& MSL (Multiple shot learning)) & [redacted]

Dat .18 learning from subnets (≡ subfunctions) ....

SN: TM could easily lrn meaning/criterion for "Soln to eq" & Soln.(X, X+7) = −7, T True: How would it go about solving "Soln(X, X+7) = ?, T"? IN FACT, any INV problem can be put to TM in this form, & the best way to solve it is Lsrch (if TM has proper hints!)

SN Perhaps Make "Graph of My/Pgm in Recent TM work, so I know where I am, & what needs be done.

I think the critical work now is .15 — .18 & then .13 – .14.

.28 In 236.37 – 237.12 we can have just one "i" value (≡ 1, say) for all j. This will work if we use long random nos. for examples. — This tech will prob'ly not

.30 work in non-numerical domains (& prob'ly wouldn't work in "Number Theory" either!)

236.37 – 237.12 is a kind of Gross Elementalizn of the Task. It is of some interest in that it appears able to do some lrng. — (perhaps at a serious level?) (.13 – .14) & (.15 – .18) are improvements. (.15 – .18) is conceptually simple, but the details of implementation — could be very diff'lt to get it to work w. reasonable acceptable cc.

.36 (.13 – .14) = (236.01 – .06): Using the idea of (.28 – .30) (i = 1 only), we have this set of $[Q_j, Op_j]$ pairs (op_j = Mfunct$_j$), & we want to find a good stoch operator ∋ $F(Q_j) \approx$ Mfunct$_j$

.39 or ∋ $\prod_j P(\text{Mfunct}_j \mid Q_j)$ is max (Mult by pc of F.) 238.01 .13 max

Non-Conventional TM could do analysis of Wholistic Medicine (it may be that there is almost always no good data to conclude much — but TM could read a lot!)

Bulg

.01 ! 237.40 : Actually, we should save much cc by doing 237.36-.39 After every "j" problem is solved; since it should make soln. of next "j+1" problem much easier (See 236.16-.35 for discussion of why i now) Using the i=1 only idea of 237.28-.30, we can think of TM's work as "linear regressn" "Sequential prodn" in the following way:

In linear prodn; we use info upto j=n to produce get a pd. for x_{n+1}. After we are given the true x_{n+1}, we use this to get the pd. for x_{n+2}, etc.

In TM, In all cases, the pd for X_j is based on the knowledge of the true X_j values for $i=1/j-1$.

.10 α In TM, similarly — knowing the MFunct$_i$'s for $i=1/j$ and the Q_i's for $i=1/j$; we get a d.f. for MFunct$_{j+1}$. We use this d.f. over MFunct$_{j+1}$ to do an search to find a "actual" MFunct$_{j+1}$.

.13 we then do $j \leftarrow j+1$ if loop to α (.10)^5.

.10 - .13 is very interesting! Somehow we are able to combine this d.f. for MFunct$_{j+1}$, that incorporates all of our previous knowledge of $\{Q_i, \text{MFunct}_i\}$ for $i=1/j$!!

.17 Doz MF = Macro Funct

It would seem that After I have given for the MF$_{j+1}$, & finding MF$_{j+1}$) there is no need to find a way to find j 's a funct of Q_j. — i.e. finding an OB that identifies j from Q_j. — we don't

.20 need this in .10 - .13!

Well, what happens after we find MF$_j$, and we are then given a problem Q_5 ($j>5$ & Q_5 is one of the previously done problems). Then, presumably, TM has the pair (Q_5, MF$_5$) available to it, so it will give P(MF$_5$ | Q_5) a very high pc.

.25 → My impression is that MF$_j$ does not need to include info on how to work problems Q_i for $i<j$! MF$_j$ need only solve Q_j (N.B. .32-.34)

.25 has to be looked into: Along w. .17-.20, this gives a Quite different picture of TM than 236.37 - 237.12! It looks

.29 less efficient, more capable, but I'm not sure it's right! One fear is that my "stochastic operator" model of a cond'l d.f., is flawed — the

.295 it may assume that there is No Info in the [Q_i,] set. → see 241.01 - 242.05 for good understanding of the "stoch operator" model.

.31

.32 Actually, in .25, the d.funct that finds i as a funct of Q_i need Not

.33 be very SMART (as will not need it to be scaled upto) — All it has to do is find out how Q_j is diff from all $Q_{i'}$ for $i'<i$ (!) — so perhaps (.10 - .13) seems much better, more direct than (236.37 - 237.12)

[However N.B. 236.37-237.12 was meant as a preliminary step of TM: A kind of "study problem": In .10 - .13 getting the d.f. for MF$_{j+1}$ from interim [Q_i, MF$_i$] $i=1/j$ is not so easy! — But it may be eventually cascading always eventually, necessary. [.295 - .31] → is a possible worry.

239.01

9.17.00 Bulb → ..... 3.60 − (.44) = 3.16

so 3.16 V. 150 k. 3.16

22 Ω at 3 volts

$\frac{3}{22} = \frac{1}{7} = 140$ ma. 600 ma hrs → ≈ 4 hrs.

.01 : 238.40 : I'm really not sure how impt. t. Q at 238.295 − .31 is. (on info in [Q_i] set) — see 241.01 – 242.05

Perhaps drop it for a while & see if I can continue w.o. answering it.

.02 O.K. Say we have this [Q_i, MF_i] i: 1/n ! We use it to obtain [redacted] $P(MF_{n+1} | Q_{n+1})$. for v.g. version

[redacted] I think I'll have to see how I would work such a problem; to get some feel for how TM would do it.

.06 A Perhaps Relevant Idea: for a given sequential corpus, "a", we can have a "Ideal Summarizing Machine, M", so that Random inputs to M, give [crossed out] outputs w. pc's ≡ pc's of possl. contins. of "a". M is a complete summary of all regys in "a".

.09 Normally, "Summary Machines" are approximations of this ideal summarizing Machine: → 242.07

An interesting Q is: [redacted] if we concat. a string α to "a", how does t. Summarizing machine M, get modified? Well, if α is not very long, we simply put lots of random inputs into [redacted] M, until they produce α, & continuations of α. These continuations of α will have t. pc of continuations of a^α.

There is a corresponding "Summary Machine" assoc. w. t. [redacted] "Prodn of an unordered set of strings". I don't immediately remember how this works: See that so 199: "T. Computer Journal": "Two kinds of Prob. induction" → 241.01 – 242.05 13 v.g. on Summary Machines. also 242.07

Anyway an approx. (summary machine) for a t. [crossed out] bag of strings! would be a stoch lang. [crossed out] for PSG's, (stochastic) Each such lang is defined by a set of production rules & prob choice probys. T. prodn. rules correspond to NT's (which correspond to sub-languages — or sub-grammars)

.21 When we add new data to t. Bag: we usually modify t. Grammar by (1) changing t. xfn. probys (2) Modify t. Grammar by adding new NTs & assoc choice pc's.

.23 (3) More diff & less likely: Deleting certain NT's & adding others. (Hard to do).

In any case, we end up w. a new Summarizing Machine (≡ P.D.)

.02 is t. problem of updating a "Summarizing Mach.", Hvr., it's dif. from normal updating of a Stoch Gramm, because it's a Conditional P.D. that has to be updated. O.K., but ≈ t. same things in (.21 – .23) hold — we want to change pc's (if any), add new NT's or functions & perhaps revise t. [redacted] structure of t. MF. ⋮

More specifically: When a new Q_i OP_i is given, we must devise a new operator and/or try to combine t. new Op_i into t. old one in a compressive way; and/or revise t. entire system in a compressive way.

A GOOD APPROACH: Watch how I work a specific problem of this sort: Express my soln. in t. forgg. Language (concs), then perhaps Genz.

.35 Hvr: My impression is that Demonstration of t. (kind of behavior I'm interested in) can only occur w. a fairly smart T.M., or (i.e. a rather long TSQ). T. regys required are rather complex & invisible (via M) a small corpus. Only by A.H. insertion could t. needed concs. be visible in a small corpus. ← see 240.03 for what this means!

240.01

5.18.00 Bulg

I guess Q_i^1 is the symbolic part & Q_i^2 is the numeric part

.01: 239.40 : 239.35 is another example of a kind of TM behavior that cannot occur until TM has reached a certain "sophistication."

.03 Specifically, the behavior "looked for" in 239.35 : After a fair no. of $\vec{Q}_i, A_i$ pairs < in which Q_i is of form Q_i^1, Q_i^2, and TM is able to work problems w. input Q_i^1, $\tilde{Q}_i^2$: here Q_i^1 is an input that it has trained on: $\tilde{Q}_i^2$ is part of the input that is New. & it is able to get proper $\tilde{A}_i$ → TM is given an $\tilde{\tilde{Q}}_{n+1}$ in which both Q_{n+1}^1 & Q_{n+1}^2 are entirely new to it, & it is able to get a

.08 "reasonable" pd. for A_{n+1}.

5.21.00 Looking at 241.01 → .05 ! In particular, 241.20 → 40 and [proof of adequacy ≈ 242.01 – 05],
It is clear that the [Q, A_i] TSQ. is adequate, & that making stochastic functions P(A|Q) using ALP should be adequate. — That the only reason it didn't seem to work, was that the P(A|Q) functional forms I was considering, were too limited — Not "Universal" — or not able to express the needed regularity, (Universality is usually unnecy)

(238.10 – 13) & (241.01 – 242.05) seem to tell how to do TSQ's !

I had 3 kinds of Induction that I wanted to demonstrate in TSQs:

1) Simple acquisition of concs & stat[redacted] soln (MFoncts) as in the Search TSQ's
2) Finding subfuncts / sub[redacted] nets (not necy trees) & using them for new induction.
2.5) doing 2 for both OSL & Multiple MSC.
3) A more elaborate kind: say (.03 – .08) (≡ 239.35 – 40) This involves NMTM — truly probilistic predn. (.03 – .08) seems to be a v.g. example

.22 → Look at (.03 → .08) say $\vec{Q}_i = S_i, \vec{N}_i$ N_i is Numeric (random) part. S_i is symbolic part

Say F_i is the [redacted] functional soln. to $\vec{Q}_i$; so $F_i(\vec{N}_i) = A_i$.

we've after solving several (n) problems, we have this set of pairs:

[redacted] $[S_i, F_i]$ $i = 1/n$. My Earlier, simpler ideas ~~allowed F_{n+1} to~~ gave a pd. for F_{n+1} based only on the unordered set $[F_i]$ $i = 1/n$.

Now I want the Pd. for F_{n+1} to include all of the info in $[S_i, F_i]$

Ideally, this can be done by a (set of) function(s) that map $S_i \to F_i$ ($i = 1/n$).

.29 The map is from a kind of string, to a function.

Superficially .29 would seem to apply only to ANL, but in theory, it could cover solns of any kinds of equations. — so, in theory, TM should be able to learn essentially diff't task via this tsq. type.

.34 Bookmark (where I was) ! 1) Trying to do simple TSQ 2) Trying to devise

.35 simplest possl. Learning machine that would work on any TSQ's. → 243.09 244.37

The recent (241.01 – 242.05) stuff makes it clear that a pure $[Q_i, A_i]$ TSQ would be always "solvable".

The idea of (.22 – .29) of my S→F type tsq as a special kind of Q, A tsq is very nice

243.01

(.01) CONDITIONAL PROBTY!
2 forms of Data: [redacted]
Convergence Thm 242.01

On Conditional (complexity) (Probty). It should be poss. to devise a Thm like S78T3 for conditional probability! This would seem to be closely related to MCT.

Anyway, say some corpus has been generated by some condl. Pd: $P(y|x)$. T. corpus itself is $[x_i, y_i]$. If we assume that (x_i, y_i) pairs in t. corpus are unordered, then this is an example of Bag prediction (= unordered set of finite objects)

Actually, we'd need $P(x)$ as well as $P(y|x)$ to generate t. corpus. Will t. error bound in [redacted] $E\,(P^M(y,x) - P(y,x))^2$ be indep of t. x d.f.?

$P(y) = P(y|x) \cdot P(x)$ If $P(y|x)$ & $P(x)$ are finitely describable then $P(y)$ is also "finitely describable".

Hvr, there is also an approxn of $P(y|x)$ that is "indip" of $P(x)$ — that does not use info in t. x d.f. It tries to find a $P^R(y|x) \ni$ $E\,(P^R(y_i|x_i) - P(y_i|x_i))^2$ is min?

More exactly, it trys to find P^R's w. short desns $\ni$

.16 $\prod_i P^R(y_i|x_i) = \max$ or, more $\approx$ exactly $2^{-|R|} \cdot \prod_i P^R(y_i|x_i) = \max.$

.16 is the form that doesn't use $P(x)$ info. (?)

.17 Finding a $P(x,y)$ assoc. w. t. corpus $[x_i, y_i]$ is an (apparently)

.18 diffrnt way (better way?) to get t. approxn to $P(y|x)$

Q: does $P(x,y)$ imply a $P(x|y)$? : (Yes!)

Note that $P(x,y)$ (gives / implies) both $P(x)$ & $P(y)$.

.21 9.21.00 We are dealing w. 2 diffrnt Goals for our model! In one, we want $P^R(y|x) \ni$ .16 is true i.e. $2^{-|R|} \cdot \prod_i P^R(y_i|x_i) = \max$

in .17–.18 " $\ni$ $2^{-|R|} \prod_i P(x_i)\, P^R(y_i|x_i) = \max.$ ≡

~~So, taking logs in both cases; The first way gives wt. to all x_i. The second wts them in accord w.~~ or .17–.18 ≡ $\prod_i P(x_i) \cdot 2^{-|R|} \prod_i P^R(y_i|x_i) = \max$

It differs only by t. factor $\prod_i P(x_i)$ from .16

So in .17–.18 we have t. same conds. as .16, except that t. model must also give a good Pd. for $[P(x_i)]$. When the $[x_i]$ are from my TSQ, the $P(x_i)$ d.f. is somewhat "A.H." & is related in a weird way to $P(y_i|x_i)$, so I may not want a TM to try to model $P(x_i)$ — particularly at a cost of a less good $P(y_i|x_i)$

So: for t. present, use .16 : It also has t. advantage that $P(y_i|x_i)$, (which is what I really want) is obtained in a direct way: in .17–.18, $P(y_i|x_i)$ is obtained from $P(x_i, y_i)$, the joint distrbn, and involves division by $P(x)$. $P(y|x) = P(x,y)/P(x)$: $P(x)$ is obtained from $P(x,y)$ by integration or summation
$P(x) = \int P(x,y)\,dy$ (or equivt Σ form).

There are impt. situations in which .17–.18 is better than .16 — in which t. $[x_i, y_i]$ corpus is from R.W. entirely & not modulated by my "Heavy (A.H.) Hand"! (242.01)

"Summarizing Machines" (.07)

.01: 241.40: The convergence theorem for $P(x,y)$ (or $P(x)\cdot P(y|x)$) is about the same as that for Bag of strings. In fact the "Bag of strings" convergence theorem follows as a corrolary of .01, when X = constant or X = A null

.05 Probably best look at present proof of Bag of strings convergence theorem (in MCT files) to discuss proper proof for .01. (.15)

.07: 239.09 "Summarizing Machines" that are theoretically complete, need never backtrack.

The only reason we ever backtrack is that our "summary" is imperfect.

On convergence of $\sum_i \left(\ln \frac{\delta_i'}{\delta_i}\right)^2 \leq E \sum_i \left(\ln \frac{\delta_i}{\delta_i'}\right)^2$:

For S78T3 and for Bag of strings:

we do have constraint that $\frac{P_i}{P_{i-1}} > C$ and the convexity of $\ln$.

— Best limited excursions of δ_i from δ_i'.

.15: (.05) On 241.01-.40: In the case of 241.16: what is the default (A.H.) down?

We need this to tell if we are making progress & how much progress.

Also, how does QA system work, when there is a body of data to use for reference (e.g. an Encyclopedia)?

MCT would seem to hold the key!

In much of my thinking in 241.01 — .40: I had been thinking in terms of "MTM", in which there was one formula that successfully did all predictions, & the next best formula was far behind. For a small corpus, perhaps this is not true.... But in such a case, a more "soft" probabilistic down would have higher PC. — This is true in MTM w. small corpus. There are 2 aspects of "Corpus Size". One is the Nonsense point: easy to get it to be effectively large by very long random nos. Second is the symbolic part: No easy way to assess.

If $Q_i \to A_j$ is 1 to 1 then there is a single code for the corpus.

If $Q_i \to A_j$ is 1 to many, there will be a code then there will be many codes — of which doesn't corpus exactly.

The normal way to think about this is eq. 241.16. Each pass there is a Prob for each. For each Q_i, there is a P.D. of poss'l. A_j's. In a MTM corpus, of much length, these P.D.'s are very narrow.

Almost all wt on one A_j, but all other A_j's have non-zero wt. since they have poss'l codes.

.36 So, in this in a NMTM corpus, say A_{10} has many $\geq$ poss'l. Q_{10}'s: Q_{10}': Q_{10}'' Then it may are of = P.C. Then both will have codes of about same length, for Ref Q_{10} input. Essentially we

.39 have a Prob. machine w. 2 inputs. One input is Q_i, the other 2 43.01 2 don't

9.22.00 Bdy.

.01 : 242.40 is "random" input, or just all possl. inputs. It's this Random input [redacted]

.02 whose code length we're interested in [redacted]

I think [242.36 : 40 ; 243.01 – .08] is t. rite way to understand this stuff. → Actually, there are 3 inputs for machine: ① an input that defines t. machine behavior: its length is |R| of 241.16 ② The Q_i input of 242.39, ③ t. "random" input of 244.01 : 02.

"Ep'stly" speaking we want an R & bunch of "③" (random) codes

.08 ∋ ≈{|R| + [redacted] $\sum_i$ | code for A_i |} is min.

.09 ① To what extent have we solved (mode progression) 240.34 – .35? → 244.37

② Just what was t. QA_i TSQ I was most optimistic about? 240.03 – 08; .22 – .29

What we have is our T.S.Q [Q_i, A_i], we start w. a few Q_i, A_i pairs ($i = 1/n$)

We find a R pgm and a set of codes for t. A_i (= a_i) w. smallest $|R| + \sum |a_i|$. Using that R, when we put in Q_{n+1}, we get, using random input on #2 input, a d.f. for [redacted] A_{n+1} & we then ← ~~If that's~~ what we want.

[redacted] do trials (Lsrch) ~~...~~ until we get t. rite (given) A_i as output. This, of course, assumes that R doesn't change. In fact, as we get more & more $Q_i A_i$ pairs, we will want to Revise R. — This has been all very theoretical thus far. —

[redacted] I want to see what these operations correspond to in a real TSQ.

Say we are doing MTM w. $\vec{Q}_i$ ≡ [redacted] s_i, N_i [like 240.22 – 29]

[redacted] We end up w. a set of solns F_i ∋ $F_i(N_i) = A_i$.

Consider that t [s_i, F_i] correspond to a xform of t. original [$\vec{Q}_i$, A_i] (set.)

Given the [s_i, F_i] $i = 1/n$ How do we get D.R. of F_{n+1} from s_{n+1}? ↑ Not "Bags" in this case

.25 One Gross simplifn: we have ^set of^ primative ~~set~~ functs, we then want to find a set of functs (w. wts), so that w. t. primatives, we can use this ~~set~~ ^Wts.^ set of functs to generate [F_i] w. max pc. [we haven't yet included [s_i] info] ← Includes OSL

.28 These new ~~set~~ functs will be subgraphs of the [F_i] $i = 1/n$. functs.

~~After~~ Next, we look at t. inputs [s_i] — try to ~~find~~ generate them from a set of primatives, & find larger abss ~~...~~ so that t. ~~...~~ set [s_i] ^small^ ~~large~~ total dern. } Similar to .25 – .28 on [F_i]

Next, we try to find "parts of the [s_i]" that correlate w. that corresponding F_i or parts

.32 of t. corresponding F_i.. ~~...~~

.33 We do all of .25 – .32 so as to minz (.08)

So .25 – .33 is a rather gross dern of t. kind of activities one wants to do to minz .08.

It is a very "elemental" way to do it: to break down parts into sub-parts & correlate them.

[redacted] There are certainly ~~...~~ more "global" ways to Minz .08!

.25 – .33 are some simple induction methods. To ~~...~~ get t. mechanics of them & to find more methods, Do run thru a TSQ & see just what

.39 techniques I use.

9.23.00
Bulg.

.01 : In 243.25-.33 : Make sure I understand just how each operation ↑ pc (≡ ↓ 243.08).

N.B. The ANL TSG I did in most detail in Saarb was somewhat non-ed, in that it simply tried for a start overall psm. in 243.08: Perhaps look at that psm in some detail — find out just how it did this.

.06 I think the way it would work is: We started out w. [redacted] start w. primitive function set. S_1, found a f_1 by Lsrch: Then in Adding f_1 to the function set, we find a f_2 ∋ it works for both S_1 & S_2. ... Adding F_n to the set of funcs we may use Lsrch to find $F_{n+1} \rightarrow \{F_{n+1}(S_i) \rightarrow A_i \ (i = 1|n+1)\}$

.10 Then $n \leftarrow n+1$ & loop to α

.06 - .10 suggest an improvement to itself, that might make it better than 243.25-.39! We do .06-.10 as stated, but the functions we use, to generate trials are/ much more general than those used in .06-.10

A more Global (but still a bit ad hoc) approach to .06-.10 : In hunting for

.15 F_{n+1}, We are interested in "modifying" F_n, because f_n is already a short code. Similarly $\{f_i\}(i|n)$ are all short codes. Also, any common sub-graphing functions (subgraphs) in $\{f_i\}(i=1|n)$ would be useful

.18 in ↑ pc of all of f_i $(i=1|n+1)$ [I think I may have discussed .15-.18 not so long ago!]

.19 Anyway .15-.18 is a justification for using $\sum f_i \ni i=1|n$ & parts of those funcs to construct trials for F_{n+1}, & it tells us (I think) what pc's to assoc w. them: In particular, it enables us to quantify

.22 **OSL** — which is quite important in this problem (I think).

.23 It may be possible to augment the techniques of .15-.22 by those of 243.25-.33 : looking for good sub functions (acquired)

.25 in both the $\{Q_i\}$ & $\{F_i\}$ set and correlations between them: Any such copies would be found would ↑ pc of the F_{n+1} in .15-.22. (The F_i & Q_i [in .15-.22] mean somewhat different things than the S_i & f_i in 243.25-.32.)

.28 So: .15-.28 is essentially the method used in ANL in Saarb, with (I hope) a better understanding of the theory (Theory = 243.08 and 244.06 and 244.01-242.05 <242.01-.05 is of particular import> Also .19-.22 gives theoretical basis for pc's of component functions in the Lsrch trials for F_{n+1})

Also, because of MCT, I have a greater variety of data types I can put in the corpus — which make TSG writing easier than it was in SAARB!

T. Corpus Stuff is all for OATM. It may be that the general way I do the TSG in the Corp. .15-.28 can also be used for OZ problems.

.37 Also Note (Book Mark) the sub-goals of 240.54-.55: How General (i.e. (very. technical of .15-.28? Would it be best to work reasonable TSG's, and deal w. the Scaling problem? T. "correlations" of .25 may help a lot.

Bols

65
16
81

5 PM Sunday / Som.

The Analysis methods of 244.23-.28 may not all apply so nicely to the QA problems! While [Q_i] sets & [S_i] sets can be identified & "understood" (mainly coded) The things corresponding to the [F_i] sets of 243.25-.33 are not as clear!

.05 Hvr: 244.15-.24 should be regarded as a guide in trng TSQ's: That I will write 6.TSQ: decide on what types of things TM needs to learn, what kinds of types it needs to know about, & see if I can fit it into 6. schemas of 244.15-.24! This schema may need "Augmentation"! → .17

.09 One main Q was: "How far could I get" w. a given schema for trng? Presumably, at some pt. of complexity, a schema would become "Universally Extensible" so w. suitable TSQ, it could learn to do the equivalent of any needed Modifn of its schema

.15 (.09) → SN Another ("Very English") approach to TM: Write TSQ: (heuristics) Write down all the heuristics, tricks, comes, that I seem to need to "Trn" test TSQ's. Then try to devise a (formal) device to encode, using ideas in [redacted] MCT. 244.15-.28 would be a good first approach to such a model! (≈ .05-.09)

In .15 ff: [redacted] After I write TSQ & "solns" in English, Make up a compact lang. that expresses easily & naturally, concepts, tricks, used in solns.

SN A common "Meta heuristic": when a problem is solved, or when something [redacted] unexpected occurs: Try to see how this new (conc) info fits into & past — to predict PC. Opn in a more advanced TM — into & expected future — So, what kinds of probs would it be useful for? — This puts "hooks" in "condl pc's" on & new conc., so we know (under what circumstances) when to try it in & future.

Ok. So go back to those TSQ's! 181.5, 182 have some stuff, but I think I need more extensive listings of things to put into TSQs.

180.24; 167.24-.34 : 173.10-.16 ; 180.34

173.06-.08 Gives a big scope TSQ refs — Also possible future Extensions

If I get stuck: review ~160 ff (or 167 ff), lots of good ideas, if

→ Q: How did I "get stuck" in 160 ff? — What were diffs?

Perhaps go thru history since 160 & look at each of diffs that occurred & how each was solved.

As I see it: The way it's Done: I start w. a stochastic language that is Universal 2 ways (1) it can express anything (2) Its expressions are at most a constant "longer" (or pc < by > a constant factor) than any other stoch lang.) [cond (2) implies cond (1)]

.37 < Words! in .37-.38 stochastic lang → stochastic Operator >

9.24.00

Bulg.

.01: ~~Re~~ 245.40: **Next**: ~~...~~ using the stochastic lang, we input Q_0 — we put in our "random" codes (Lsrch) until A_0 comes out.

— We then make a "summarizing machine" & do Q_1, A_1 t. same way.

245.37 ff is a bit "too general". More Narrowly,

— I'm thinking of a stochastic function-language. (Prim-rec funct lang — not quite Turing Universal — can't express Ackerman funct)

.07 Anyway I use this lang to get A_0 from input Q_0, using Lsrch.

Next, I modify t. lang so that $Q_0 \to A_0$ is more likely, but I keep t. lang "Universal" (prim. recursive universal). Then I repeat w. $Q_1 \to A_1$.

ABC

.10 In .07, if we get all codes for $Q_0 \to A_0$, we have a "complete summary machine." & we should be able to "reuse" our "random inputs" using Q_1 as input, & eventually get A_1 as output.

(SN) N.B. our "universal" Machine probly has to be able to make "Definitions". (at least!).

.15 In .10, After "input #2" creates A_0, Input #2 is concat on to input #1 (Input #1 "defines t. current Operator").

Woops! .15 wouldn't work (I think)! After creating A_0, with the concat'd code on input#1, I don't think t. system would be more likely to give A_0 as output (w. input Q_0).

.20 ⟶ I have to work out t. "Summarizing Machines" for stoch Operators! → (.25)

Perhaps in .15, ~~After~~ Before creating A_0, t. machine put its "state" on t. stack. After creating A_0, it goes back to start state but for new input, it remembers how A_0 was created, & is ~~...~~ more likely to try it w. whatever new (Q_2) input is.

.25: (.26) Look at "2 kind of ~~...~~ Prob. induction" for summarizing machine for Bays.

I did have a "summarizing Mach" for Bays but it used a less convenient Model:

A better summarizing Machine for Bays:

.28 **We list** all stochastic ~~...~~ operators $S_i(x \to y)$ in aprio order (i is order, $P(i)$ is assoc. pc.

The summarizing machine ~~...~~ is equivt to: Chose a S_i with probly $\propto \; P(i) \cdot \prod_{j=1}^{n} S_i(Q_j, A_j)$ & use it to evaluate $S_i(Q_{n+1}, A_{n+1}$ (all possl))

{ In practice, this means "store the S_i's w. t. hyest

.33 $P(i) \cdot \prod_j S_i(Q_j, A_j)$" The fewer we store, t. more "Backtracking"

.34 we have to do.

The next problem is how to do .28 [redacted] - .34 in a practical way

— **Heuristic Methods to ↓ cc.**

eq .33 induces a pd $P'(i)$ on all stochastic operators, $S_i(Q_\#, A_\#)$

When a new Q_{n+1} (or Q_{n+1}, A_{n+1}) comes in, we'd like to try them in $\frac{P'(i)}{cc(i)}$ order, but it's not so easy to do, if we just have, say 1000

.01 1246.40 S_i's w. $\hat{y}$ $P'(\hat{y})$ If we had all S_i's in $P'(\hat{y})$ order, then this would be fine & we'd get a v.g. Pd for $Q_{n+1} \to A_{n+1}$. [redacted] What I want is a way to approximate $P'(\hat{y})$ in a way that gives a useful, shareable distribn on $\hat{y}$, for [redacted] Lsrch.

.05 One way might be to look at $\vec{S}_n$ ← 10 dom. vectors, which is the 10 best S_i's that I've found, for $[Q_j, A_j]$ $j=1/n$. (So: 10·n S_i's.) & "pass a grammar thru them."

.08 Or possibly "piss Grammar thru" the 1000 best S_i's for the current value of n.

[SN] in Sol 89 (Isaac): §5 "How Updating is done": We use Lsrch to solve several probs/pgms. We take the set of solns & pass a grammar thru them. Then gives us a pd. to use Lsrch for to solve the next problem.

When described this way, the Lsrch is used "conditionally", so that its P.D. is indep of the nature of the problem. In .05–.08 we do a similar "passing a grammar thru", but we have a conditional probability dist., $S_i(Q, A)$. [Note: $S_i(Q, A)$ is usually not a joint Pd on Q & A; it is the pd. on A, as a function of Q.]

So the P.D. on S_i is supposed to be "problem-independent."

.19 One Q is: Just how do I express a cond'l p.d.? Can I express it like those "function/concept 'nets'"? Well, a p.d. is a type of function, but it is a very special kind of function — So we'd want to construct a representation method so it would generate (almost) only Pd.-type functions (compact on $[0,1]$)

One sure method is to simply use an approx of ALP!

to get $P(Q, A)$, we have many functions that map from Q to A, & we put in Q & then the Probability of A is a function of the no. of funcs from having that A as output.

[redacted] We may want to weight these funcs, for their a priori probability.

It would be neat, however, if we could just put in Q & A & get Pd. or, put in Q & get a list of the most likely A's w. their associated pc's.

We can get a P.D. on ordinary funcs, by considering the Net that represents their generation. Each guided choice in the net (each edge) will have a certain pc.

Perhaps a better way: Consider the set of (non-probabilistic) functions $[F_i]$

.33 that generate [redacted] $Q_j \to A_j$ $(j=1/n)$. Say P_i is the a priori pd of F_i, the set of functions. Then we pass a grammar thru the P_i-wtd. set of functions F_i. The funcs are represented by nets, and we can look for common sub-nets in the set of funcs. There is a finite set of such funcs, so we may be able to narrow the P_i $\leftarrow \frac{P_i}{\sum P_i}$.

However, [redacted] the meaning of "passing a gramm. thru the set of funcs" is not clear, because SSZ is not clear, and SSZ has to be clear before : 248.01

#vlg.

.01: 247.fo We can have a "corpus" whose code we want to minz. This enables us to tell if a code element (a "regularity") is log.t. in derbys. w. "corpus"

.03 In the present case, we know the true stoch grammar & for the set F_i. If it is the zpriped on $[F_i]$ with zero wt. for all rules that [illegible] aren't 100% compatible w $\sum [Q_j, A_j]$ $j=1/n$. We want to approximate the true P.F. P.F. from this

.06 ~~the~~ with. sample, $\sum f_i$. Is this a "surface reconstruction" problem? Probly not. I think I've been here [.03-.06] before! It may have just solved it. It may have been in SGA, in which I wanted to approximate $G(\vec{x})$, given $[G(\vec{x}_i)]$ $i=1/m$ — an interpoln. function. [i.e. a set of pairs $[\vec{x}_i, G(x_i)]$] I normalized G, so that $G(\vec{x}_i) \leftarrow \frac{G(\vec{x}_i)}{\sum_j G(\vec{x}_j)}$

Then I wanted to pass a grammar thru these $G(\vec{x}_i)$'s — which were like a P.d. Lots of trouble!

There is an A.H. soln, that assigns P_i to F_i for $i=1/n$ & zero to all other funcs.

.14 Perhaps try for a shortest run of the set of pairs (P_i, F_i) $i=1/n$. or (F_i, P_i) in the sense that we want an operator in which we input F_i & get P_i (or close to P_i) for $i=1/n$. We want the operator to have

.17 a min. d cm.

.18 Did I ever find a way to deal w. the pc of "correcting" P_i when $F_i \to P_i$ wasn't exactly right? Well, one way is to have $F_i \to P_i$ be a pd rather than a deterministic function. We could have a deterministic func, then a Gaussn D.F. w. common σ^2 for

.22 $i=1/n$. This seems Not BAD!

So (.14-.17) & (.18-.22) may solve the problem!

It looks like a non-linear regression problem ("How many params. to use?") Well, this "n.l. regression problem" may not be so easy! In fact I did try to model G_i as a function of $\vec{k}_i$ in my recent work on (GA/GP). — But I didn't think of it as a n.l. regression problem!

Still, it looks like a rather hard problem! n.l regression or not!

One hooble! I do know a grammar that obtains the P_i from the F_i. (in 247.33) P_i is the comp of the F_i & is usually easy to find — in that F_i are generated by some (known) stochastic grammar & P_i is the pc assigned to F_i by that (known) grammar.

.33 So the question is: is there a cheaper way to derive the set $[F_i, P_i]$?

Well, we may add more info — say on negative (vs positive) cases. — in which case GOOD!

the old grammar that generated $[F_i]$ would give wrong pc's to the funcs that should be given near zero pc. I will probably have a lot of F_i's that I know (empirically) do not fit.

Hrr, see 252.29 for discussn!

~~The poor~~ Derby for positive cases only is easy; its pc = 1; it's just the zpriped grammar (pc=1) generating the P_i's from the F_i — total pc = 1.

22 × 16¾ → 27.65" diag ←

> 22 × 16½ → 27½" pretty close ← 3,4,5

If we list the [F_i] that satisfy i constraints, this will be a complete dscn. of a set of F_i's that excludes all other F_i's. It's total PC

.03 is $\prod_{i=1}^{n} p_i$ This is the A_H. dscn of the [redacted] set of positive & all known negative [redacted] cases.

So .03 is PC of the default (analogous to "random") case. Any shorter dscn is of (some) interest. — dscns of smaller pc are usually not of interest unless we have a way to add up a lot of them. ABCDE

But perhaps all we really want is a way to get functions that are "close" to F_n. — or to [F_i] i=1/n. Actually F_n is the only funct. that satisfys all of the constraints — only it of [F_i] will do [Q_i, A_i] i=1/n. We use the rest of the F_i to "expand the sample" in a "soft", approximate way.

(No!) On second thot, p_n is the pc of F_n, & F_n is enuf to be a complete Dscn of [Q_i, A_i] i=1/n : [redacted] positive & negative cases.

I want to dscb. [F_i, p_i] i=1/n plus [F_j, ϕ] for f_j that I've tested that did not work.

I guess F_i (i<n) are all impt also, because they are all attempts to find F_∞ (the F that works w. all Q_i in an optimum way.) ... But I've forgotten about the problistic nature of F mapping Q→A.

But still F_n is "Uniformly Better" than F_{n-1}.

Should I try to find a grammar that just gives F_n hy pc, & very low pc for all other F_i that I have tested that faild for one (or more) Q, A?

I could often just have one positive instance of F_n ∴ small ssz (possibly more if I "Oversearch")

Re: use of data of F_i; (i<n)! These can be considered to be "approxs of F_n"! Each has a pc of "correction" to bring it to F_n accuracy. Usually, I imagine, this pc of correction is quite small so these [redacted] $F_{i<n}$ mite not be worth considering (?). — Maybe not so small! In the first place, $F_{i<n}$ have aprops >> that of F_n. "Correcting" F_{n-1} could involve just "recognizing" Q_n & then giving A_n or a suitable function for $Q_n \to A_n$ — Total pc could be not much worse than that of F_n.

.30 much worse than that of F_n. → [See 249.33–.40; 252.01–.28 for ending and a kind of resolution of this problem]

[redacted]

CRITICAL

.33 247.19–249.30 is an attempt to solve a very impt problem: How to develop a good stochastic function/operator that maps $Q_i \to A_i$ [redacted] [Q_i, A_i] i=1/n.

.35 There is a standard ALP soln. to this problem: Say [F_j] is a set of functs of aprop [p_j] ∋

.36 ∀ i,j $F_j(Q_i) = A_i$. Then $\sum_j p_j F_j(Q_{n+1})$ is a wtd.

.37 d.f. over possl. "values" of A_{n+1}. — i.e. a d.f. for A_{n+1}.

So just what is 247.33 ff. trying to do?? Well consider .33–.37: If we only have a few F_j's that do [Q_i, A_i] i=1/n, then we will only have (at most) a few

252.01

9.28.00 Bulk

Meta Problem of TM (.34)

.01: 249.40: I want to do a good review of ~160 to 249: T. present idea is to
.02 write a TSQ. i write out detail'd heurs for humans: Then develop a method
.03 of lrng. along with this: 160–249 discusses various problems: solves some:
I should be familiar w. these (probs/solns) while I'm doing .02 – .03.

It should be possi. to do .02 w.o. (.03) (i.e. developing lrng. method). I would list all needed heurs, then look into Q of how each heur could have been devd. — which heurs (if any!) were needed for each heur discovery, etc··· down to
.08 a Basic set of heurs or a primitive set of functs/functionals.

Some impt. probs/ideas in TSQ design:

1) The idea of local v.s. Global (symbols/nos/vars): So if I ask TM.; If $x+3=7$ what's x? It must realize that x i $x+3$ etc are to be usd in this problem only, but that "Most" other cases (unless specifically told to be "local") are Global i apply to All problems

2) Is [Q, A] format an adequate / good format for an initial TSQ? — for all TSQ's? Conceivably all problems could be "usefully" put into this form. — E.g. If it's a TS (1), bag lrng (2), inv (3), OZ (4), problem, we could have Q w. an additional component that tells which of these it is so, (1)(TS, begng of TS): "A" would be next element in Time Series.
(2)(Bag lrng:, Finite bag : Part of Quel): A is next symbol in Quel or rest of Quel.
(3)(INV, Function, F, $F(x)=y$; y is Given): A is to find X.
(4)(OZ, G(·); cc : A is to find $x \ni G(x)$ is Max in time ≤ cc.): A

[Note, hvr, Most INV probs are solvd by converting them to OZ probs

TM could learn a lot by watching an O.T. at work: The O.T. would find a sequence of $\vec{x}$'s in attempts to Max $M(\vec{x})$: TM would try to find a p.d. that characterizes a $\vec{x}$'s G distribution. TM would not have to know $M(\vec{x})$ in order to do this, but TM could regard this as a kind of Cond'l p.d, w. $M(x)$ being part of the "Condition" (≈ Question).

By watching many O.T. for many kinds of OZ problems, TM mite get a good idea as to how to estimate $G(\vec{x})$ very fast i thereby do OZ probs very well.

Note that part of the OZ problem is t. cc of a (triad): This would have to be appreciated by TM!

60
12
4:45
Migr at 3:30
4:
4:20
5
1 30
1
20 00

.34 Work on "Meta Problem" of How I should work on Designing TM! Taking into account my ↓ power of memory. Sounds like t. "unsolvable" OZ problems where $G(\vec{x})$. F(T)(max of soln) is to be maximized. F(t) is my failing Memory!

1) Some ramblings: What was bottleneck after Sol 89? — Why did I
.38 not finish TM @ Saarb? Reasons: (1) T. TSQ's were very hard to write (2) They solns (251.20)
3) I have made several impt. discoverys since Sarb: (251.01)

(b)

"WON"

Very imp. problem ("Bottleneck")

How much progress on problem of interaction of tasks? "AND" "OR" "NOT"

Meta TM prob:

.01: 250.40: 3) (cont.) They are: (a) Sort of basis of (what to work on next) for complete AND/OR net.

.02 (b) M&I proofs understanding (c) Understanding of TM's Gen.

→ The main Subprob is Understanding English Text.

→ A BIG Q is: How much info do I have to introduce (≈ A.H.) into TM (by TSQ's user/designer) before it can learn much from less carefully designed Tests? How will this Threshold depend on IPC of Machine Used? How well can we do w. 10^{10} ops/sec, 10^{11}, 10^{12} ... 10^{15} ops/sec? Or, more exactly, how well can we do w. 10^n ops? (≡ CJS)

For a given t.s.q. I will have to put in fewer steps ("less" "shinnerian") if I allow large CJS. [redacted] Say $> 10^{10}$ [redacted] or n" steps

If I have smaller CJS, I'll need more examples. So t. Q is: What is (no. examples) × (mean CJS of the examples) = total ops needed for TSQ:

N × C: Can we minimize N × C (for a TSQ that goes up to "understanding text".) by varying C (Then N is a function of C: N(C)) — so minz C · N(C) by varying C. C · N(C) could be small for small C, but we would then have a TM that had no experience (& hence no Hours) for problems w. large CJS.

Large CJS's use certain hours that are not practical for small CJS: — e.g. large initial overhead.

So, In General, we may want to use as large a CJS as we can afford; Its H(C)·C. (Recall t. total cc of t. [redacted] TSQ.)

↑ ≡ Total cc of Corpus

.20: 250.30: (a) Only used CJSs that were complete solns of previous problems. (b) There was no use of sub nets (sub functions) of previous solns. I recently (since ~1960) wrote about this — why it occurred. I think it is intimately involved w. how "Big", how complex t. TSQ is, & t. exact nature of t. induction system used: What t. form of t. inducting (inducting?) is, e.g. Int. TSQ's ideas of ... ? what did I have in Mind?

(c) T. Soln. method didn't "scale" well — as TSQ continued, too many functions in "storage": (see 253.12 on Big "scaling", see 253.02 for dern of SAARB

Scaling

.27

SN In "inserting Hours" we do this so that TM could have learned to Hour by observing enuf exs ~ cases in t. past. What SSZ should be used? It may depend on just what t. hour is: but try OSL: just one recent t. past.

9.30.00

Bulg

Olymp's fund (is US)

STRI or S2RI (.35) (.38)

String to Real Induction

.01: 249.40 : possl. values of Anti : If we are looking for a particular known Anti, then we may well be unable to find it in the small set of possl. Anti's!
— so we would really completely "lose" on this search.

247.33 ff tries to deal w. this by extrapolating a set of functions, [F_j] of 249.35-36, by "passing a stoch Grammar" thru [F_j]. The spirit of this is that of Z141 — (passing a stoch grammar thru a set of strings).

A major difty is that of SSZ. We only have a few "strings" (w. wts, p_j) to pass our grammar thru, & I don't have a clear idea as to how to do this...

It may, indeed, be a not viable way to solve this problem.

.10 I had the idea of extrapolating a set of pairs, [F_j, P_j], w. (P_j = 0 for F_j trials

.11 that did not pass thru [Q_i, A_i] i=1/n.) 247.33 — 249.30 tries to deal w. this problem.

Note that because of the large no. of negative cases, the SSZ for the problem of .10-.11 is quite large.

A somewhat difrnt approach to 247.19 "How to express a condl pd":
Stochastic Grammars (like in Z141) are p.d.'s; They are uncondl pd's:
If the Grammar were a function of Q_i, and A_i were given P.D. by that Grammar, we'd have what we wanted! Hvr, it's not clear as to how to do this — the deterministic (or even probilistic) mapping from Q_i to (a) Grammar(s).

In General, it may well be that I will get a better "feel" for what kinds of condl. P.D.'s I want, when I try to express my "English" total (including hours) solutions to TSQ's in the format condl. P.D.'s.
At the present time, my ideas on this would have to be too general.

So this is a good motive to go to the point [250.01 — .08] ← ≈ [163.16 / 163.01] "early work" (6)

.26 in which I write a TSQ in English & write solns & all hours in English, then try to devise a language & induction rules to acquire the TSQ

.28 viz the solns & hours in (.26)

.28 N.B. the idea of 248.33 of including negative cases, may not help much!

A v.g. fit could still be obtained by deriving the [F_j, C_j] set exactly & giving $C_j = \phi$ to all other cases: (particularly if there are only a few $C_j > \phi$). Hvr, still it is possl. that we may get a grammar that gives approximate values of [C_j] that has a whole greater PC.

.35 Anyway: Consider the problem of "converting" [F_j, C_j] to be "not well solved"

2 impt. cases: Thus fan F_j is set of funct, C_j are known zeros & zeros for neg cases.

(2) F_j are Functions or strings denoting objects in [GA / GP] & C_j are their known scores. → Much work on this

.38 DEF Call it the String to Real Induction "STRI" STRI.

N.B. Stochastic Langs is one kind of STRI ~~form that I have "solved" soln.~~ functional form

> Parkinson's Law: "Work expands to fill time"
> To get work done fast, allocate less time to it!
> e.g. TM. → 255.01

.01: 252.40 That a STRI problem soln. can have. Hvr T. heurs of Grammar discovery have not proved to be very useful in solving STRI!

.02 (SN) I wanted to use Y STRI soln. for SGA! My "soln" of Koza's "3+2" 11 input Muxer problem was simply Lsrch over a simple function space; then add to t. set of functions, any new func. found that were much closer to t. Goal. — This seems close to what I did in t. SAARB TSQ's. This suggests a "Back Door" soln. to t. STRI problem!

— So if I have F_1 & F_2 & F_3 w. hy C_i values, I just add them to t. "function pool" w. wts, perhaps a their C_i values. The functional combinations can be at most of depth 2 or 3, because of rapid combinatorial explosion (w.o. heuristics). We also have rapid incr. as no. of functions to be used in t. combinations ↑.

.12: One trouble w. this: By itself, it does not scale to larger probs — t. no. of functs available grows (≈ "linearly") w. no. of probs solved — so it eventually becomes impractical — unless one devises heurs to "narrow down" which functs to try. ← One aspect of t. "Scaling" problem

STRI

.16 → Some Common Mappings from strings to reals: ① strings represent ② equations or ① Systems that have evaluable parameters. Example of ① would be a string representing an electrical ckt: & an ω; the real (or complex no.) is t. impedance at that frequency. Or the string des. a network & its input; t. real (or complex) output is t. that output. The representation of SPICE ckts by Koza is a good case in point. — The "output" is t. "fitness function" of that ckt.

A common way to discover t. string → real relation is to experiment w. t. numerical params in t. "string".

A string can represent a physical system (say in QM) — t. "real" can be some computed property of t. system (Energy, velocity ...) Often, t. "text" will set up t. eqns & t. "numbers" will furnish "parameters"; Occasionally Numbers (integers, say) can help set up eqns., but this is usually either a specialization of a continuous param or t. nos. are descriptors that are assigned "symbolic" meanings in a rather A.H. way. — So they have to be treated via largest SSZ.

This is true in Koza's systems: "Reals" & "Symbols" are dealt w. in different ways in GP.

In assigning pc to strings (stock lang.) t. strings usually have no numbers in them. This is ALP's way of Bringing "Quantity" from "Quality". It is a very special kind of functional relation. What it does is related to "counting" t. no. of times a certain feature occurs. We can get a pc-like number by suitable normzn. — or by multiplying such normzd nos together.

Considering this ↑ way, it seems clear, that usually, STRI

Bulg.

depends very much on [illegible] TM has about what to symbolize "part "means",
I had decided that I couldn't really work on this STRI problem until I had a less "abstract" idea of what to expect in the relevant functional relationships.

Well; now I have a somewhat better idea of what to expect!

ANA is one simple example of STRI. (253.16)

Another type would be recognizing certain "substrings" or "structures", country Dance is having Alg. functs. of f. [redacted] various "counts". [illegible] Passing a Grammar thru a set of strings is one way to get a number (≡ pc) for each string.

Also: "Speed" is an extreme case: Looks at symbolic (abstr. descr.) & has (param values) & calculates its path in time: [redacted] (≡ seq. of numbers) or any other param of resulting system.

Or to invert "speed" Given params to design (Symbolic) chart & params (nos) (This is what Nozzi's filter design pgm did).

The dcom of functions in terms of prim funcs, involve mainly Symbolic (not Graph) [illegible] dcom + prim funcs + 1, 2, π, e, γ [illegible] & some small integers as "params". e.g. $\sin x = (e^{ix} - e^{-ix})$ / (divided by 2 because numerator 2 funcs in numerator ... (from "Country" operation).)

.01 Admin: Parkinson's law: "Work Expands to fill Time":
Early in TM rsch, I felt that TM could easily take rest of my life: It looks like I've been arranging it to do so!

A (possl) way out: Make outline of what needs to be done on TM:
(A sort of PERT (?) diagram), w. indications of interaction,
a AND/OR net of Tasks:

Describe each task clearly & tell how much eff't I've done, & how much more time I [redacted] expect I need to "finish" it. (dif'nt def's of "finish" may be possl.)

For most tasks, Give them "Names" along w. clear definitions of them.

→ Also Arrange Journals so I know where stuff is; Also so I can refer to a page or set of pp.

Stuff since Saarb; since S89; since S86 are of most import.
Work on S 89 was probly very impt: (Also much on S86).

• Main Subgoal is "Good enuf" Understanding of English, to begin to read & understand English text, so that % of understanding ↑ w. time.

.17 • Subgoal before that: Understanding of Algebra or some part of Math. Enuf to work probs of some difty.

• Sort of Subgoal: To be able to do NMTM problems — "Soft" induction in the area of Math & eventually in other domains.

.21 • Devise TSQ for Algebra of .17

.22 • Devise Lrng. strategy for .17 & for general [Q_i, A_i] problem — 15 & gen. of (.25), i.e. STRI problem

• PSG discovery: Probably very useful for various kinds of Grammar models... Grammatical induction.

.25 • STRI problem: Given [String $\vec{X}_i$, real r_i] to induce $P(r \mid \vec{X})$ for new data.
String, Real, induction:
This is important in learning OT's & understanding "fixing", GA. | or to induce $P(r, \vec{X})$.
This is a special case of .22 R [Q_i, A_i] induction

---

tailored

Superficially, it would seem that Q,A lrng would be a rather general soln. to .22 & .25 — both impt problem areas.

.31 I think I felt that I wasn't doing so well in thinking about .22 "in the abstract" —
.32 that a better way to deal w. it was to write a TSQ for Alg I along w. "solns" in English, w. all hints I could think of. — Then express this "soln." as a lrng. strategy i.e. (.22)
While the lrng method would be somewhat tailored to this TSQ, I hoped I could generalize the technique in a useful way.

Then I would continue the TSQ & solve the probs "completely" in English & see if the previous techniques for lrng. was adequate — or whether new lrng. techniques had to be added.

0.6.00 Bulg.

m/k

m/m! m ~/= m | !⊕ ABCD

.01 : 255.40 ! So there are 2 ways:

1) Work on 255.22 ~ [Q_i, A_i] lrng problem directly, abstractly, in rather general case. [Q, A can be all or partly symbolic ; numeric

2) Do 255.31 –.40 — which seems very promising.

In 255.22 I had t. updating problem! i.e. say I had a good induction model or d.f.

.07 for [Q_i; A_i] i=1/n ; How do I update my model when

.08 Q_{n+1}, A_{n+1} comes in? I did do work on this & t. soln. was a "Complete Summary machine" & then there was no diffy: If t. summary was incomplete then

.10 Backtracking is warranted/necc.

.11 Another tack would be: If $P(Q, A)$ works upto Q_n, A_n,

.12 what is t. a priori for t. functional form, P — so we can search on it for a fit? → See 267.29–.35 for possl. relevant remarks! maybe soln!

Can (.07–.10) & (.11–.12) be unified? — It would seem that they should be!

Re: .07 –.10 & Summary Machines: T. stuff about "Backtracking" is

.15 correct if "summarize" by only including t. shortest codes of t. corpus. — in this case, "Backtracking" retrieves some longer codes for t. corpus, that are now relevant. BUT there are ways to "summarize" ∌ .15 :

— Just what are they? Consider an approximate code plus a correction code. Is that a case? Also what is done in "Theory Revision" in physics?

.20 In Physics — (say introduction of Relativistic phenomena) we first decide that we are working in a new area (hy velocities/energys), so we "label" t. area, so we can distinguish it from t. old area ("old corpus"). We then make models for t. "new area" specifically. When we get these models, we then try to get a "grand theory" that unifies t. models for t. old & new corpi —

.25 A "Grand Reformulation" that ↓ overall code length (This is ≡ 257.34–.40)

.26 This would amount to finding a good model for (Q_{n+1}, A_{n+1}) (or a sequence

.27 of [Q_i, A_i] i=n+1/n+m) and a way to identify [Q_i] i=n+1/n+m.

.28 — Then attempts to "unify" t. new model into t. old.

.26 –.28 is easily applied to learning "3+7" then "5*9".

.30 There is t. problem of deciding This deciding how big "m" should be in .26 –.27. This means — deciding how much of t. corpus should be covered by t. new model. Well, we just fit t. model to small m, then we see how large an m we can take & still have t. new model fit!

T. lrng assumes t. corpus is very nice about introducing new concepts sequentially: Say t. new corpus was 3+7 ; 5*8 ; 4+2 , 4+7 , 3*7 ··· Here "3" & "7" etc are just 3 & 7, they are not long random nos. —

.38 this means we cannot simply code 3+7 , then 5*8', because t. model is to big to be able to afford any code compression w.r.t. 1/or 2 examples. (257.01)

ABCdetg.

~ "BAD"

'01: 256.40 Well, this 256.38 may well be an example of a "BAD" ("imperfect", "corrupted") TSQ. {I wrote this report "Perfect TSQs & the cost of corruption" — on this problem} — I had the idea that it could very much ↑ cc of search, & sometimes make it prohibitively large.

My more recent ideas, hrr, were that we shouldn't depend on a TSQ being near "perfect". Maybe for infant TM, we may use perfect TSQ's — but later on, I expect TM to be able to deal w. very imperfectly constructed TSQ's!

.09 One way would A.H. deal w. 256.38 (& with a somewhat more general difty of this kind): TM finds an operator that works 3+7, then looks for other Q.A.'s in which this operator works, then

.12 trys to find a way to identify the QA's in which that operator works.

Actually (.09 – .12) is not so A.H. It amounts to a way of [redacted] "partitioning" the corpus in a useful way!

(N.B.) the problem of the 256.38 & its "soln" (.09 – .12) is an example of an unordered "Batch" corpus — Harder for TM to learn, but much less likely to be "imperfect": That it can be a realistic corpus for TM, & very good, in the sense that TM has some idea as to what its "future" problems will be. This enables it to deal w. what would ordinarily be done by TM2 — i.e. to have a feeling for what "choices" are "interesting", in the sense that they seem likely to be relevant to "advanced" parts of the "Batch" corpus. — I also vaguely remember this "Batch corpus" somehow enabling TM to deal w. cc [redacted] minimization in a fairly non-ad-hoc way: — I'm not sure just how this was supposed to work! — I would seem to

.26 make the writing of TSQ's a lot easier! → 258.06

Note, hvr, that my initial goal is to get info into TM! — That the "sequentiality" of the corpus can contain implicit info, & I start by giving TM that info "free". Later, I can wean TM from this by giving TM augmentations of the corpus in larger & larger "chunks", until I am adding chunks that contain info that TM will not be able to deal w. at the time. So he will learn to recognize these corpus sections &

.33 [redacted] realize that he is not yet able to deal w. them. — "Textbook" → 258.06

.34 So this ⊃ (256.26 – 28) ⊃ (256.20 – .25) is [redacted] standard, simple, method of theory revision used in physics: Old model (M) doesn't work on Q_{n+1}, A_n. Find new (usually narrow) model (M') that will work on Q_{n+1}, A_{n+1}; find way to recognize Q_{n+1} as diffnt from

.28 Q_i $(i < n+1)$: (Integrate / Unify) M, M' → $\tilde{M}$, which is a shortening of code for M, M'. → (note some aux. problems! 258.15 – .17). → 258.05

.39 ↳ Note one non-ad-hoc way to do (.34 – .40) → 271.34 – .40

.01: 257.40 While 257.34 is "not bad" it is not optimum. It is ≠ the most natural way would, when Q_{n+1}, A_{n+1} remain, do a direct search for a good model for $[Q_i, A_i]_{i=1}^{n+1}$ directly. For this to be even mildly interesting, we must have a reasonable description length search space of $A + f(Q)$ functions. (≡256.11)

.06 257.33 257.26 A Commonly Occurring "Batch corpus" of this kind, would be the problem of getting TM's to (understand) compress a large English text. A v.g. way for TM to work on this would be to find parts it could compress easily, then find more difficult parts, etc. I expected that TM would learn to "read & understand" English text in this general way, but I had not thot of it as a "Batch Corpus" w. TM going back (later to try to code or recode a section that it didn't "understand" the first (or 100) time around. I really want to get TM to the pt at which it can do "Batch corpi" of this kind! It makes writing TSQ's very easy!

One recurrent problem: TM has to decide when to (temporarily) quit work on a part of the corpus & work on some other part (— (hoping to eventually return to that part later (& wiser)). This judgement also occurs in e.g. 257.34-.40!

.15 — (Is How Good a Code is "Good Enuf"?)

We found M' for (Q_{n+1}, A_{n+1}). How good a model must M' be before we go on to

.17 the "Q_{n+1} recognition problem" & the "M, M' "unification" problem"?

We might do one M' then try to recognize the cases that part didn't work well, go back & try for a better M'. (or even a better M! Super back track!)

Could I get TM to learn Algebra in "Batch Mode"? I would start out (perhaps) w. a "(Sequential) Linear" TSQ & switch to a "Batch Mode" instruction as soon as feasible. The "sequential" TSQ has more info in it — the explicit ordering — really makes it a lot easier for TM.

Best way, start out in Sequential Mode, then see if I can get TM to work in "Batch Mode".

[One of the Good pts about "Batch Mode" is that it might be easier to define TM's long term Goal in terms of this batch. If the Batch is of infinite size! We want TM to understand, work as many probs as possible in a given time. This means we have to define TM's "Horizon"!

.31 Also the trade-off between Good compression for a small part of the corpus,

.32 vis. weaker compression for a larger part of the corpus. I think the Horizon & the trade-off (.31-.32) Must Be User selected.

One possible Good aspect of "Batch mode" TM could be like a Grad. Student & Ask Questions of User occasionally.

Another aspect of TM that TSQ's that I've sort of neglected is the "ADVICE CHANNEL".

.38 Another idea I've neglected: That one of the big reasons of my wanting TM to work on probs of large CTS, is that only in such problems is

259.01

.01:258.40 t. "overhead" of / development of / various search enhancement techniques feasible.
How much Time to spend on such developements is a TM_2 problem —
— Having to do w. "Horizon" as well as expected cc of soln.

Hvr., in early TM, I will be giving TM problems that will ~~help find, or reduce CJS search enhancement techniques.~~ or be ≈ search enhancement methods, or will be close to such methods — in t. sense of reducing CJS of finding such methods.

.08 Still, t. "English TSQ" approach of 255.32–.40 is very attractive, & I'll prob'ly do it.
— Hvr, I want to have as much Theoretical Understanding as possl. Befor I do it, to guide my creation of what I hope to be a very general lrng
.11 algm.

.12 [SN] My functional lang. didn't use Recursion because I felt it could be faster in execution w.o. it. Hvr., it seems clear that use of recursion can vastly simplify t. expressn. of many algms. — Even t. simplest recursion in which a function calls on itself, can make a lot of "simplifn." Landt may express all recursive funcs in 2 parts: ① the final simple value of t. funct for a certain (usually low) value ② the self-referential definition part.

So I can use recursive defns of funcs, then have a compiler that translates them into loops (or some other fast, low cc, form).

So one slot can be Recursive pgm. Another "slot" could be a low cc version of t. pgm. We "Mutate/modify" t. recursive pgm, because it is a very ss by ps expressn of t. function is derbs.

O.K. so now, T. main problem is my state of understanding of t. Theo. Soln of t. lrng. problem: 257.34 is reasonable approach:

for t. Theoretical optimum! 256.07–.12 (Summary Machines) 256.11–.12 (problem of apripd for L srch)

.27 for "Summarizing Machines" I want a more General model other than t. one that simply takes a bunch of t. best codes.

For t. "problems of apripd" of the (Q,A) d.f. — Perhaps all I really want/need is an updating algm for when we add a new Qn & Ans. This may be tied up w. t. form of t. Summarizing Machine of .27.

One diff w. Summarizing machines — I had no idea of other models than
.34 a set of short codes: In fact any f.d. is an approximate "Summarizing machine". So Stochastic CFG's, linear regress w. stated σ²
.36 are all cases of Summarizing Machines.

T. method of updating, will depend on t. Model (= Summarizing Machine approxn). Most Models don't do OSL. Stoch langs update by ~~modifying prods, defining new NT's & by general Modifn of Grammar.~~ 261.01
defining new NT's & by General Modifn of Grammar (new loops, etc. undefining old NT's...)

0.7.00 Bulg. # HARDWARE

To run Many computers in ‖ for Lsearch { worker computers.

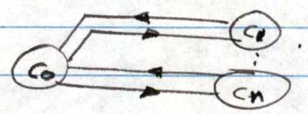
One (control / king / leader) computer := C_0 C_0 gives search ranges to each C_i ($i=1/n$).

C_0 has an input register for each C_i ($i\neq 0$).

" " " outpt " " " C_i.

So C_0 has 2-way communication w. each C_i.

When C_i gets its range of problems, it starts work on them.. If it solves one, it tells C_0. C_0 (scans its inputs for statement of soln), C_0 then tells all C_i that are working on solved problem to stop, and rcv. new (problem/search range.) C_0 also tells C_i's that have solns. to send them to C_0; C_0 receives them & stores them.



Whenever C_i finishes a particular trial, it looks at its input from C_0 to see if this problem has been solved yet.

This arrangement would involve many only ordinary I/o ports.
— Hvr for C_0 to give C_i a new problem/search range
and " C_0 to receive soln. from C_i, will require a serial port, (or ‖ port)

I really don't know how "ports" work — how to address them, & how the communication cables are to be set up. The ‖ port is ordinary a way.

As I see it the C_i machines are all normal ordinary machines but C_0 has n I/o ports: These are actual HW connectors, — perhaps each has a latch.

Alex felt that I could probly do what I wanted w. a central ethernet "server" & a bunch of computers connected to it. — So all computers would be "ordinary".. Is there much overhead time for this ethernet system?

I was thinking C_i need not have a hard disk: Maybe 64k ram & a floppy to start it up.

David G. Willis
Googel

.01: 259.40: Another kind of P.d. is that used for T.S. [redacted] regressn. Here we update by modifn. of params, ↑ no. of coifs, use of non-linear coifs:

In both Seqn loops & T.S. regressn, I'd like a seq. of updating operations that are "Universal", in t. sense that ALP is universal, & always finds any finite describable regy, at ssz & cc over large amt. ← (like S78T3)

.06 In regressn, by employing many linear & n.linear coifs, one can approach
.07 a kind of "Universality", since any funct can be approxd this way.

Similarly, feedforward ANN's can approx any function — in some ways, better than .06 – .07

Tho, I think even ANN is not really "Turing Universal" in t. sense of S78T3

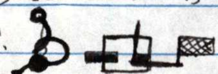
Rissanen's work may be relevant, but I find him very hard to read!

I do have a paper of his, somewhere on Generalized Regressn. Paper given at that conf I attended, ISIS conf. Abstract on p4 of ISIS conf proceedings.

A purely numeric form of [Q_i, A_i] is: Q_i is a numeric vector & A_i is a scalar or vector. This is t. general/N.L. curvefitting problem.

.15 Again, using many N.L. coifs. or ANN can do t. job. → .22

A nice way to approximate ALP is ≈ what Ries does: He lists a large (perhaps infinite but enumerable) set of Pems, & he uses t. lose system in S64a (T. one I describe as "Moonyplaces").



So, in General Induction, we can [redacted] use a wtd mean of predn. methods, wts. oc [redacted] (apriori method) * (pc of corpus [redacted] wrt. that method); This should be applicable to all kinds of QA problems; Symbolic, Numeric & mixtures.

.22: (.15) Another interesting possy: Say we have k numeric inputs and a set of functions of various types. We can make all functional combinations of ... using h functions & k inputs.

.25 N.B. In .22 we also have to include constant inputs) These are continuous params in addition to t. normal k inputs. This also applies to my earlier analysis of "Functional Composition" of 130.25 — 131.18

.28 Anyway, one of t. ideas of my functional lang, is that any (Prim. Recursive) funct. can be describd by a finite n composition of primitives. (This is much shortened if definitions (as in Z141 notation) are used.) T. notation is a prescripts corpus string as in Z141 & pc's are automatically assigned, as in Z141. → see 262.12 for recursive defns

→ There is, hvr, t. problem of listing t. functions in ≈ pc order

If we allow t. set of primitive funcs to be both symbolic & numerical, we can get a grand P.d. over all Q→A functions

Some such primitives: AMSaxM (string) → True else False.

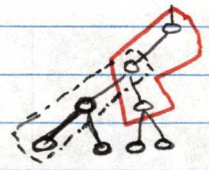
Boolean functs of T, F → T, F. ; T. various Numerical functions + − × ÷, etc.

Various Control functions.

Actually, t. set of funcs used in "BASIC" would be about adequate. — It does both strings & numbers. 262.01

.01: 261.40 : The development of 261.28 - .40 is fine, but its pretty much
t. same as A1P: except that I use a kind of extension of t. TRMC,
.02 and [I automatically sum over many of the codes, to get a pair of various functions. Thru t. Z141 method,

.05 In coding Function nets, is there much of a parsing problem? ——
I think there might be: I could code this net using either the read
macro, or the [illegible] macro.



So, if I have a certain set of Macro functions —
each w. its assoc. pc, I will have to try
reparsing it, every time a marker a new node ■

"Macro function" definitions (this is Wolff's idea) → .18

.12: It might be possl. to write regular/Recursive Definitions using t. idea of 261.28 - .40.
T. recursive defns, that would be evoked would be of t. form:
Defn. of F(■): F(1) = 1 : F(X) = X*(F(X-1)). [defined R(x) for positive integers only
I'd have to find a way to express this in t. Z141 formalism.
First Go thru t. formalism of 261.28 - .40 in some detail.

"REFORMULATION"

.18 : (.12) Actually, this "Reparsing" could amount to an important "Reformulation" Theory of t. code! One time it may occur is when a new defn is made
or t. pc of old defns has changed much — Both occur w. Growing Corpus
While this may be a common type of Reformulation, it is probably not the most General kind.

One "kind of" "reformulation": Tell (ring) "1+2; 3*7, 4-8, 6÷3;"
as individual classes! Then it looks at Pro(soln) expressns for all of them
& creates the simplest expressn that covers all cases. —— A T. "reparsing" of .05
does not cover this. In this case perhaps "Integration" is a better term than "Reformulation"

The reparsing of .05 could result in an entirely difrnt looking code
for t. corpus!

Anyway: T. problem I've been working on, is to devb. a less Cl. way of
doing Q,A induction than 257.34-.40, 257.38-.40 is pretty "standard" in science,
I imagine. On t. other hand the Non-el method of 261.28-.40; 262.01-.02 is
close to A1P, but w. t. advantage of (.01-.02)". As is, it is not ordinarily practical, for
a large corpus, but it could be used for a small augmentation of t. corpus
(like Qual., Anal.)

The 261.28 ff is probably usually used to evaluate a parsing of a corpus.
The actual functions defined, are normally found by other means (as in
t. old Z-141 ... for finding sums by noting tracks of occurrences.)

Bolg.

.01 : 262.40 Another track on this problem: stock CFG's give a p.d. on finite strings.
— So it's a sort of soln. to the Bag induction problem. It works by assigning pc's to various rewrite rules (or Matrix Multiplication rules.), Z 141 does this also. —
for BoW & grammar & the (proper) corpus.

Hvr, the function described in 261.28-.40 are deterministic: Could I do a similar thing & get probabilistic ones? Well, as a matter of fact, I do! The functions are constructed probabilistically, so what 261.28-.40 does, is give a p.d. on

.08 Functions. These acting (//) on Q_{n+1} to give a p.d. on A_{n+1} (!)

.09 It may well be that "Backtracking" is the only non-el. way to get better models in a "Universal" way. [In the Polly discussion, $(\exists A_i)[\exists\, Q_i, A_i]$ is meant to be a large chunk of the corpus w. a common soln. (common Macro function)]
So, say I get a soln f_n for $[Q A_i \to A_i : i = 1|n]$. F_n doesn't work on QA_{n+1}, hvr. So I backtrack: When F_n was found as a soln for ... we (also) had other solns that were not as good (if we did any "overscorching") — perhaps many of them. In "backtracking" we try these solns on QA_{n+1}. If none of them work, we go back to $Q A_n$ & look for solns that work for both QA_n & QA_{n+1}. If we can't find any we may backtrack to QA_{n-1}. To try to find a soln that works for QA_{n-1}, QA_n & QA_{n+1}.

Hvr, Normally in inductive coding, we don't use such Non-El. Methods of Backtracking. We use heuristics that narrow the search down considerably & in doing so we find solns faster, but we tend to miss certain (V.G. empl.) solns that a Non-El technique would find.

.23 In General, when I work problems using 257.34 a/o other El. methods, I will try to see how I can make them less el.. Also in 277.34, the (Unify/integrate) phase could sometimes employ the Backtracking of (.09)ff.

I want to write a review of recent ideas on Non-el'ness of 257.34. Also, perhaps expansion of the idea of 261.28 — 262.02 (A. Non-El soln. of T. problem) plus the notion P.D.'s of 263.01-.08

.20 T. Fusion This is imp.! One of big advantages of ALP over other inductive formulations is that given an induction system, it suggests how to improve it. I really want a system that is "fully open ended" I know that w. finite CB, I cannot expect to get best possl. induction. T. Q is, given extra CC, how best to spend it.

One way to spend it is on hy pc nols that haven't yet converged.

.25 I think a part of the soln. to .20 - .23 is: We start out w. a general UMC — biased by whatever ideas we can think of. We use this UMC w. Lsrch to find solns. to simple problems. We then bias the UMC in view of the solved problems. We then continue Lsrch — used to biased (than still Universal) UMC.

Hvr, the trouble w. .25 (if it is, indeed the trouble) is that the

.01 : 263.40 "Summarizing Machine" used to incorporate & solved problems, is NIX a complete Summarizing machine, so we have to use "backtracking" like in 263.09 -.23. This is rather non-el backtracking, & my impression is that it is very expensive.

.05 I'd like, perhaps some kind of Mix of the eliminated of 257.34-.40

.06 and a Non-el. method (like maybe backtracking, but not exactly 263.09-.23.)

So .05-.06 seems to be where I am now.

I want to be able to use Heuristics, yet I want to be sure the system is Universal. It may well be that "Scientific Method" as practiced by the Sci Community, is not universal, that it depends too heavily on certain heuristics — that it does tend to "Paint itself into a corner" so much, that the amt. of backtracking (= Recoding) that would be necy, is quite impractical! E.G. drop all of Modern Physics & start over!

However, Modern Physics summarizes an enormous amount of data — so any new formulation would have to track its models closely! — This is what Witt. focuses on & Standard Relativity & QM & "Standard model" do.

Unfortunately, the Sci. Community doesn't much tolerate dissidents!

.18 Another track: I think one of the original ideas, was that one starts (say the TSM using a complete seq. of instructions: that the initial problems are such (A Matching between initial probs & initial inst. set)

.20 that the best fractioning set of makes Levin find solns rapidly, & After solving a few new problems, the PD is Modified in various ways, new data are made & the various mistakes are different. PD's) — Making it easier

.22 to solve what were before, excessively difficult problems, loop to .31

Note that a full MCT TM can use practically any kind of info to improve its P.D. A very imp. source of info is the traces of its own attempts to solve problems. This is the source of Many heuristics.

.31 During the loop of .20-.22, the Machine/PD remains universal!

.32 All searches are always made over a universal (inst. set/PD).

I have to Reconcile .18-.32 w. the idea of approximate "Summarizing Machines".

.18-.32 Does seem to be a very (GOOD desirable) way to run a TM!

A kind of "Summarizing Mchn" like More like the spirit of .18-.32, is what one does in linear regress! After each new data pt comes in, one updates all of the coefs — Of course this induction system is not universal.

.33 261.28-.40 is a complete (Machine/P.D.) formalism. The Remark of 262.01-02 Makes this a system more than just a ~ UMC for ALP. It does induce a pd. The changes Changes automatically as we code a corpus. — we get variations in pc's of various concs, & we introduce (define) new concs as we code the corpus. A hour would seem be (probably) a further biasing of the pc's — perhaps in a way that depends on previous codes.

Hints noted; 267.01 -.03

0.10.00 Bulg.

Sears
Silvertone Guitar ~ 1960
Sell to that Guy who wanted to buy it?

.01: 264.40: 264.33-.40 suggests that parallel coding trials can influence one another!
The way this works: We try a certain code for the corpus but it fails.
A MCT-TM could learn from this one failure (OSL) and perhaps influence the pc of a code ~ to the "failed" one!

So how does 264.33-.40 relate to "Approx. Summarizing Machns"?
(I probly want to consider various induction systems & see how they can be made Universal & how they do "summarizing") e.g. linear/nonlinear regress, ANN, > tech layers (cf. a C.D. Grammar)
The P.D.. of 261.28-.40 and others that I may think of. Does this suggest any amusing classes of Predictive functions?

Def Sumes Consider summarizing Machns (Sumes): After a Umc has coded part of a sequential corpus, the ratio of 0's to 1's in the shortest code, could be a kind of very weak summary: But in general, this ratio will be ≈ 1 because the codes are "random"
We can, tho, consider non minimal, ll codes: They need not be "random":
Consider a pure Bern Seq: T. pc's of its symbols are a complete summary of the P.D. of all possible strings in any of its corpi. generated. — So these pc's are a complete summary of the P.d.
In the case of 261.28-.40 The probbys are a complete summary of the Bernoulli-like regularities in the corpus: These probbys also include several "derived Concs".
Considerable Effort to find a way to express Conditional probbys. (like the probby vector of symbols that follow α.)

Presumably, We use probabilities, the probbys of creating the actual corpus, is maximized.

.23 A Big Q is: Can any Considerable P.D. on all possible functions, be expressed by something like 261.28-.40? This formalism expresses or can be modified to express, all computable functions. — But all p.d.'s on Ream?

.26 An analogous Q about P.D.'s on Strings: Can all possible P.D.'s on Ream be expressed by a Bern-like seq in which various Universal strings can be defined as given probbys.? I would say that this is extremely unlikely! — But consider it anyway. Well in .26 not all, say we consider only pd's on finite strings that are derivable from summing over the pc's of all infinite strings that they are the prefix of. So the pc of a string must be monotonic as one moves along it.

.29 .31 Well, I suspect that whether or not 261.28-.40 is "complete" in the sense of .23, # — that that approach is pretty much what our "Scientific Method" is — and possibly it is as well as can be done by humans (??)

The Q of .26-.30 is interesting! Surely we have more "Universal" methods of assigning pc's to strings! — Probly we can apply Ream (analogously) to 261.28-.40
Are Hidden Markov Models more powerful than that ≈ Bernoulli Model? I think so! The Bern model is a HMM w. 1 state! So 261.28 could be augmented into a HMM. 266.01

0.10.00 Bug.

CSL equiv. to Linear Bounded Automaton ≡ TMMC

→ See Hopcroft–Ullman "Intro to Automata Theory... 1979" p 225

No: CSL lang is equivt. to a finite TMMC w. 1 tape containing input, only the cells occupied by input used for work tape.

.01: 265.40 Well, HMM's correspond to finite State Machines! Which only approximate full Umcs! Hvr, CF langs & CS langs are more powerful.

There is some belief that CS Langs are Turing Universal. See to what extent this is true — It would have to allow some shortening of strings. — i.e. rules like ACBD → AD otherwise, it would always be possl. to tell if a string were in the lang. (It's Not!)

Since strings can be made to represent all functs in a universal way, a strong problistic lang on strings should give us a good ensemble of pd's on functions. **GOOD!** Hvr, I don't want the (constraints on P.D.'s on strings) derived on (265.29–.31). They are not what we want. We want an arby pd on finite strings — so like L. Viti's Discrete D.F.: Umc's that stop at end of string. T. P.D. assoc. w. a stochastic Lang, rather than sequence extrapolation.

There are 2 problems here: First (the easiest) to get the "language" of 261.28.. to be actually universal! (or Prim. rec. funct. "universal") — say to model Lisp or another functional lang.. Second (hardest); to get t. P.d. on these functions to be universal. I have been using Bernoulli Models, F.S.M models, CFG's, CSG's, Umc's (the "Chomsky heirarchy") Stack Machines

.18 One approach: See what kinds of models are needed to express various heuristics that I will need. [for a list of some of them see my "How to Solve Problems" sheet.]

One impt. Heur: "Express problem in a difnt lang." (A Necessary meta heur is to keep one's eyes open for new "Languages"). Also, t. process of preparing, (modifying) t. problem to express it nicely in t. new lang, may involve (inventing/discovering) new useful concs that could be used in other ways for "compression" — or to simply help solve t. problem in a more direct way.

OSL seems to be outside t. "2 code" model — outside the Chomsky heirarchy.

Hvr OSL can be dealt w. by auxilliary Mechanisms (Presumably!)

.28 A vague Model of learning that stays universal: We start w. a set of universal concs, so that by suitable composition, we can create any function (a universal inst set in t. UMC sense). The set has initially all = apriPs. After we solve some problems, we have new functs & we have modified pc's of primitives, but our system is still Universal. ≡ (264.33–.40) which was a Big "Accumulation Pt." in my recent thot.

.33

.34 As for the "Summarizing Machine" that idea may not be so good in present context. For practical induction, T. Summarizing Mchn. is little more than an approxn to full ALP. The model of 264.33–.40 is also an approxn to ALP. It is my impression, hvr, that since 264.33... retains its universality at all times, then at all times a sufficient ↑ of CB must solve t. problem (if it is solvable!)

6.11.00 Bulg. Rev 162. impromptu review

.01 A possl. problem. 264.33 -.40, is that while it's universal, in the sense that it can code any function, it is not (ordinarily) universal in being able to express any possl. P.d. on all functions.

.03

.04 However, when one begins w. TSQ, (QA tsQ), the system can and does do an almost-short

.05 search over all possl. functions. It does so & it ends up w. a simple distribution over all possl. functions, T./wh. will be over only a few functions.

If the corpus is large, we will have a wh. over all functions $F_i: (Q_i) \to A_i$.

wh. will be [illegible]

I'm confused! .04 seems to cause no problem, but but but ... !

After the search of .04 - .05 (i.e. it oversearches enough), it will have several functions that fit [QA_2] i=1/n exactly

Perhaps a summary of the last wk's work: (since 155.01)

(1) Mainly work on QH Irng.

(2) 256.07-.10 T. updating problem for QAn+1! No problem if Complete (CB=∞)

.15 ALP is used! Otherwise, Partial ALP is, in general, impractical, (tho occasionally it can be used w. Backtracking as an effective updating method).

[In .15 when I spoke of "Backtracking" I meant "BLIND" backtracking:

Here, Fn doesn't fit QAn+1, so one simply goes back & finds for Fn some that fit & upto QAn & also upto QAn+1. One should try to find lots of them, because QAn+2 will cut down their number again — perhaps to zero! (From the pt. of view, no of O.a. pairs is exponentially w. n, so it gets to <1 fast)]

There are 2 Moddings of/ Backtracking Algt. that make it more practical,

(1) When Fn does work on QAn+1, we don't just do Blind search on QA data up to QAn: We do Heuristic Search — Being aware that the known Solns must fit QAn+1 as well We do this, say, in Physics, we know QAn+1 & we know what QAn+2 who have to have to fit up to QAn. This latter is characterized by equations that summarize the data & any new trials must closely approximate these equations.

.29 (2) When I In the updating Discussn of 256.07-.10, I was considering discrete codes, a finite number of them by "Oversearch". In fact it may be often possl. to obtain ∞ codes that fit. e.g. linear regressn: for every set of corrs, we have a slope & σ^2 & ∴ a code for it. corrs. The pc's assoc w. each code will depend on its σ^2 & on how many pairs it uses. As new data comes in, we always have codes in the set that will fit it.

.35 .29 ff may be an answer to 256.11-.12

(3) The Common Method of updating in Physics, (& possibly in the TSQ's of Szeeb: AIML)

256.20 -.28 is one only dim of this: Other dim in 257.34-.40 256-267.

268.01

162
267

.01:267.40 (4) Batch Processing: 256.30 — 257.33 ; 258.06 — ~~.40~~ :

(5) English TSQ 255.32 - .40 ; 259.08 - .11 ; 266.18

(6) On Recursion in algebraic T. (any TM uses) 259.12 ff.

(7) Why we want TM to work on B.I.G. problems! 258.38

(8) On "Summarizing Machines" 259.34 - .36 ; 266.34 - .40 (these are not v.g. refs!)

.05 (9) T. General (supposedly Universal) model (or TM, Post's also proposed). recursive

266.28 - .33 ; 264.19 - .32 ; 264.33 - .40 ; (Use of functional language: possibly recursive: → 262.12
261.28 - 40 then gives more details on how this is supposed to work)

0.11.00 Bulg. Cosma Shalizi

TELNET

Visiting for >yr. Madrona Edea.

[redacted]

.01! 268.40 Defin. bound of Universal Lrng [redacted] Algorithm Algor. tion

ISN It would seem that A_L(x) should be computable by simple crossed loops!

1) for QA lrng:

Given a corpus $[Q_i, A_i]_{i=1/n}$. T. Algm gives, as cc↑, successive approxns of $P(\bullet A|Q)$. [redacted] which we call $'p_n^{t}(A|Q)$ (for cc or assumed $'p_n^{t}(A|Q_{n+1})$. we want $\lim_{n\to\infty}$ to be within a constant factor of t true $P(A|Q)$ whenever such a P exists & is derivable by a finite string (vary terms when P is partly d cr bd by continuous params)

Furthermore, if t true $P(A|Q)$ is indep of n, then R is "constant factor" will be indep of n for $'p_n^{\infty}$. (or another way of looking at it as $n\to\infty$, R is an upper & lower bound for R is constant factor).

Essentially, all I'm saying is that t. successive approxns. to $P(A|Q)$ should have t conditions of Sol 78 T3 and ∴ will converge nicely to t. right values, at the proper rate.

{N.11.00 Partial rec. funcs are effectively enumerable. Better to use them!

One way to do R is: (↔ to ALP): Consider only Prim. Rec. funcs —

.15 Let [redacted] $P_i(A|Q)$ be a enumeration of t. funcs of interest. We can assign them $p_i = 2^{-i}$ and R is a complete prefix set since $\sum 2^{-i} = 1$.

This is not much diff from ALP & has all same disadvantages.

.18 Ideally, we'd like to list all functions $P_i(A|Q)$ in order of $\left(\prod_{j=1}^{n} P_i(A_j|Q_j)\right) * 2^{R_i} = \phi_i$.

2^{R_i} is t. prop of a $P_i(\cdot)$. List them in order of ϕ_i

[redacted]

.21 268.06 Gives 266.28–.33 refs to 268.06 Gives other forms of a lrng system that seems to be universal.

At any time, if we use another cc, it will eventually try t. correct function. — so it will however be worse than "t. correct function" by a constant.

But this constant is based on t. props of various fixed terms, & these change as n↑ as we progress in t. seq. If t. lrng system is any good, t. constant factor, will ↑ w.n (I assume!) ☺ However, if t. right funct is included, any of t. lrng systems I consider will probly give it a high wt. (No! — they will give it a high [redacted] aps! part but t. apri part need not be so hi — in fact,

.29 if I don't have a adequate soln to t. "scaling" problem, t. apri part will be

.30 "swamped out" by many, many new conc. distributions!

In .15 ff I wrote about a NTM system for QA probs. The "functions" I try are not probabilistic. But deterministic. are accepted only for a perfect fit over $i = 1/n$. For $i > n$ they give a not infinitely sharp probability distrib. [Actually false. In .15 I used $P_i(A|Q)$, a "soft" probabilistic funct.]

This is true only of $P_i(A|Q)$ are not probabilistic, but deterministic functions

However, if, in t. problem of .15, I regarded it as a NMTM, I might want to use probabilistic functions for trials. One simple way is to use deterministic functions w. an "error cost" function" for correctness thereof. Formally, this is t. same as t. deterministic treatment of .15 ff, but it seems to make search easier. Because we get some measure for every function we try — Also we get (usual) gray tuned back to steer t. search over functions.

270.01

0.12.00 Bulg

Try to find TELNET on Mac's copy.
(2) Int explorer.

.01: 269.40 : There are 2 common ways to get corrections: (1) If outputs of function is binary seq., the error would depend on no. of output bits & no. of error bits. I found 2 reasonable functions that did this O.K.

(2) If output is real (or complex) m.s. error can give amount of info needed for correction.

.05

A more general form of function (of 269.15, .18) is $p_i(A|Q)$, in which $p_i(A|Q)$ can take any of forms. In 269.15 I used Prim. Rec. functions so they were enumerable, and I could assign $pc = 2^{-i}$ to the i^{th} one.

But perhaps that is not the pt.. We can (perhaps) easily find a set of functs to each of which Pr is "Universal", — that will work for a simple problem that are "matched" to the set of universal functions. The Q is: when we "learn" we modify wts on the sequence of functions [redacted] was each of in: Do we retain universality?

.13 If so, what happens to the "constant factor" < does it expand (desirable) — or
.14 get smaller (Undesirable) ?>

Probably it's easy to retain "Universality": The big Q's are .13 - .14.

.16 269.21 - .30 discusses the "Constant factor" problem.

.17 SA Going Back to the "Summarizing Machine Problem": w. a Complete

.18 (ec = ∞) Summarizing Machine, there is no problem (except, perhaps OSC (?)).

.19 For a summarizing machine that uses only a small no. of codes, the no. will ↓ exponentially

.20 w. corpus length & eventually we have no codes & have to backtrack.

One way to sort of deal w. this is to have a very large no. of codes — i.e. every trial increment is an acceptable code, but in most cases, the pc of correction (.01 - .05) is very small & the code gets very small pc.

However, we can hope to [same large no. of codes for the corpus as n ↑], because every time n ↑, all of the old codes are still valid, but they have different amounts of "correction pc". This [redacted] seems to deal w. the exponential ↓ in the no. of valid codes in .19-.20.

.27

However, [redacted] the .18 - .27 will give lots of codes, I suspect that they will not be good codes, because they are all the same initial function, but w. various amounts of correction for each corpus ↑. Think in terms of a coding Q, A problem plus sequence. We start out w. 1000 functs that map Q into A. For each of the functions can do the entire corpus, & each will have corrections for each $Q_i A_i$. So, essentially, we try each of the 1000 functs on the entire corpus & we add the needed corrections for each to get it total bit cost & → proc. I don't see much "learning" taking place.

Could I do ≈ .18 - .27 with new definitions, changes of wts, & "reformulation"? (1) would it run? (2) would it be universal? (3) would the "constant factor" remain large enough to be practical?

Using a "Gray" fit criterion, it would be possible (perhaps) to use conventional GA (or GP?) to find short codes! The "corrections" need not be "part of the code"

27 Nov 01

0143.00 Bulg.

.01 Z70.40: as for as t. GA GOF (goodness of fit) is concerned: It is simply part of t. info that computes the GOF scalar.

.03 Hvr, I think t. Main problem: My expected Method for TM is 257.34-. Q: Is it universal? If so, does t. "concept" c remain large — or does it stay large enuf to be practical?

.06 Well, in 257.39.., We normal (at first) denote a TSQ so that "$c^{\#i}$" (which is essentially ≈ t C.J.S.) is reasonably low.

Perhaps t. real "Bottleneck" of ≈ "Universality" is "Good heuristics." TM could read Math Literature, try to verify proofs, etc., w. t. goal of finding heuristics in this Corpus. Perhaps TM could look at Notebooks of GAUSS & Ramanujan, ect: General books on heuristics might be of some value. T. Book "Numerical Recipes" & some Books by known may have many heurs. Also try Carr book (that Rsm.. cut his teeth from) for heurs. My "General Problem Solving" list.

So, is 257.34.. "THE Method"? (Many details need to be worked out).

SN There is t. general Q of "how short a Theory code is "adequate"?. In MTM, this is easier to answer, but even then, If we have a very long Proof, would there sometimes shorter!

.20 On t adequacy of 257.34.. ; This seems to be t. way t. Sci Community works. There is t. general Q. of whether "T. true models of t Universe" are accessable via Past perf. It may be that by starting out on a different path, one would end up w. considerably different models; some of which might be the best for certain Domains, than current "Science".

.25 So: I think that t. Problem is not "Is 257.34 Universal", but "Is it likely to give good models for various Domains" — i.e. Does it have "reasonable" C.J.S for t. "True Models" of those Domains. — A system could be universal, yet have excessively big CJS for simple problems, for t. TSQ that we would use for it.

As is, I am assuming that I can give TM an education close to what a human would have, plus its ability to read much much more than a Human. The Selection of "what to read" by TM will probly be Much diff'rnt from that which a Human would choose! So I might want to tell TM what areas to read.

.34: 257.39: A way to Non-cl=. 257.34 a bit! when TM is given (Q_{n+1}, A_{n+1}), It tries first to find a function that will solve $[Q A_i]\ i=1/n+1$. ← (The non-cl problem)

This can be done by direct Search, using a suitable guiding PC function. Hvr, One heuristic way to do this is 257.34-. do i.e. Try first find a funct to do (Q_{n+1}, A_{n+1}) only, then (Magnify) unify this funct into t. old one for $[Q A_i]\ i=1/n$.

(Actually, this is not a bad notion!)

272.01

.01:27140: 257.34··· can further be genzd. so its more "MCT TM". The Q's can be inv. probs, OZ probs, ~~Induction of Time Series or Bag induction~~, and t. A's can be the answers to these problems. [6 would work in an obvious way for inv. problems. For OZ problems, t. Q would be t. problem dcrn: M(·) & CB. ; t. "A" would be a reasonable result. [This doesn't seem reasonable!]

Anyway Bag induction could be done w. null Q's & various A's.

How to do Time Series induction is unclear.

Anyway: A better way would be to do as MCT says: Have 4 (or 5 if we want Q,A as a special mode) modes: (1) Inv, (2) OZ, (3) T.S., (4) Bag induction, (5) QA.

.09 All info from any soln. or trace of a soln. is put into t. Grand PD

All searches for solns. are simple Lsrch's (except for OZ problems — in which case we have Lsrch over O.T.'s (which are non-simple objects)

T. "Grand P.D" also contains info about O.T.'s. Given an OZ prob dcrn as input, the "Grand PD" gives a P.d. over appropriate O.T.'s (≡ opzn. techniques).

Def GPD There seems to be an inconsis in my ideas about t. GPD (Grand Proby Distribn). In an OZ problem, t. GPD would give a direct ordering of trials. (a cond'l pb., w. t. prob dcrn as "condition". On t. other hand, in heuristic search, one has sub-goals & t. GPD tells one how to achieve these subgoals. — which is not compatable w. GPD order t. INV trials directly. → Note 273.27

.20 One way out: Given a INV problem: 2 ways to solve it: (1) Use Lsrch on GPD's directly obtaind conds (2) Try to break down problem

.22 into AND/OR Net & solve it. ↑ This part is unnecy: (2) Does it anyway/whenever it is reasonable.

SN In all INV probs: A proof that there is no soln. is also an acceptable Soln. So usually one tries to find soln and prove there is none (unless one can prove soln. exists, which disproves non-existence of soln).

Actually, .20–.22 is pretty much a standard Heuristic. At t. Begining TM will not be able to do much in t. direction of finding AND or OR Breakdowns of problems but later, these will become very common. My [redacted] Genl. Soln. of t. AND/OR net may become impt. — (Tho I have a general MeThodology, I've mainly

.30 worked on "OR" problems and even in these I'm not so sure of t. soln.) ← 273.02

Most Math probs of any difty at all, are broken down into AND/OR nets of problems. Xfmg a problem into an AND/OR net can involve Great Intelligence!

Both AND & OR skills will have to be lrnd.

So: TM would do Lsrch (for INV problems) only if they were "atomic" I.e. not breakdownable into AND/OR nets. : Unless any of t. other 4 types of (.09) could be "decomposed". Sometimes Time Series & Bag induction are decomposable, ~~but usually~~ "OR" by xfmg t. original data in an invertible way. (This may be done in the linear → quad → cubic soln. of eqns!)

A time series or Bag, mite be better viewd as t AND of several T.S. or 273.01

.01: t. Bags might be considered as the AND of several Bags.

OZ problems can be formally xformed by a monotonic function of the form.

In MzR problem $x \to \ln x$ or $x \to e^x$ often simplifies an optrn.

.03: 272.30 One can imagine this heuristic: start out w. Lsrch on a problem: if it isn't solved by a "certain time", then try breaking it into AND/OR nets.

Then "certain time" would have to be found by TM.

In Solving probs (all sorts) by Lsrch, the Lsrch is not always to find a

.07 direct soln to the problem. — The Lsrch is on "What to do next" → .27

Def Info on the tasks of this sort is contained in GPD (Grand PD)

.11 SN On Updating GPD: In linear regression after n data pts, one has the vector of coefs, $\vec{a}_n$, that "fits best": When data pt. n+1 comes in, the search for the update on $\vec{a}_n$ can be in a "small region" around $\vec{a}_n$. If $\vec{a}_n$ is a k dim. vector, this search is over a k dim space, say over a small Gaussian hyper space region:

In ordinary (linear regression), the data in this space is uniform isotropic; but

.17 in non-linear curve fitting, it need not be isotropic. • Perhaps the easiest way to do this: The $\vec{a}_{n+1}$ that gives zero error for the n+1st data pt., is a k-1 dim space ($\vec{a}_n$ will normally not be in this space). Draw the shortest line from $\vec{a}_n$ to this k-1 dim subspace (it will be orthogonal space). We can then compute just where on that line the "best fit" $\vec{a}_{n+1}$ should be. The tradeoff is between error in the n+1st data pt. v.s. error in all of the first of the

.33 data pts.

To what extent is the General problem of updating GPD (for arbitrary (numeric/symbolic) functions, similar to .11-.17 & in particular .17-.23 ?

.27 Perhaps an easier way to get heuristic info into GPD!

A "soln" to a problem is not just the final soln, but the trace of the soln — or perhaps emphasis of the "correct" decisions path" in the trace.

This is analogous to: $\sqrt{81}$ "is not just "9" but the process by which "9" was discovered. A simple case would be the soln. of

.32 (non-linear) linear equs. see 273.13 & 275.03-.14 for relevant "Comments" on this →

.33 (.27-.32) looks like a very simple idea that I've forgotten! → 275.01

I tend to think of $\sqrt{10}$ as an OE problem (inverse of x^2 — finding $x \ni x^2$ is close as you want to 10)

But " " " " " solving linear equs as inv. problems, tho they can be solved

>> OZ [redacted] probs.

One kind of "Hint" for solving linear equs would be to give an example trace. The "degree of Hint" would depend on how close the actual problem was to the example.

274.01

~ REV

revs 162 267 268

5 types of probs:
Inv, Oz, T.S. induction, Bay induction, Q.A.
1 2 3 4 5

268.40 ⊂ Reviews

.01: 273.40 Impt I deas since 268

① (269.01) Dfn. of "Universality" of a learning Algm.

② Idea that $\vec{0}$ is ≈ equiv to small CJS for all of the problems in the TSQ. 271.06

③ à 269.29 that we certainly have to deal w. the "Scaling" problem, if we want "universality"

.05 ④ For the Model of lrng of 257.34.. : ⓐ One kind of very impt. heuristic 271.34 (≡ Solve problem "locally" then unify "globally") — can be any of 5 types.

ⓑ Generalizing 257.34 ... : so "Q" is desc of problem (INV, OZ, induction, QA, ...)
"A" is the correct Answer. (272.01 is in this direction, but not as good!)

ⓒ Use of standard AND/OR net for soln (as heuristic) (272.20)

ⓓ Dfn. of most problems (like solving eqns) is that "A" must be the
.10 "Trace" of the soln. (not the soln.) [Note 275.03: TRACE contains many properties ("slots")

What (beyond .05–.10) must be added to 257.34 to make it a "complete" lrng. Algm? Well several aspects of it haven't been worked out.. A big one is (unification/integration) ⊂ (257.28) — it is a "special case" of the more general

Def LTTM₂ "Limited TM₂ problem" of making a hyper pc model for the known $[Q_i, A_i]$ set.
Woops! — There is the original corpus $[Q_i A_i]$ set, but more important is the

Def discovered $[Q_i A_i^{TC}]$ set: A_i^{TC} is the Trace soln(s) for A_i.. A_i^{TC}

Do we want TM₂ to work on a hyper pc model for $[Q_i, A_i]$ or $[Q_i, A_i^{TC}]$?
Perhaps BOTH! from a LTTM₂ pt. of view A_i^{TC} implies A_i (≡ can be derived from)
So they would be equivalent.

Hvr TM₂ (unlimited) is mainly interested in near future & somewhat more distant) Depending on "Horizon" as
future averages of cc — (perhaps weighted cc — wts. ∝ "importance" of problems) | Given by USER

274.13
a kind of
Review

.01 : [crossed out] : (Spec 273:33) : 273: 27–33, on using t. TRACE of TM's soln. to a problem as t. TM's remembrance of t. soln. of that problem!

.03 Well, t. "Soln" can have many "slots" properties: The "Trace" is t. most general one: The actual final "value" is another (kind of "summary"), The cc is another, Perhaps other impt. propertys of t. trace. (was it recursive? Did it have loops? What were its I/O "Types"; Range, Domain . . .

Anyway, I had previously been considering using (only) the function trace of t. soln. as t. "soln". The "function trace" is t. same as t. "trace", in cases where funct traces are possl. The "trace" is more general; can be used when t. lang. has loops & / or recursions. The trace also has cc info. & t. Funct. trace — well it does also, since it is = t. Trace for t. problems in which it is usable.

.13 So 273. 27–33 is really only a (mild) genzn. of what I've been doing
.14 w. my "Functions as Answers" approach.

Another possl. Genzn. The Trace functions I've been using correspond to Deterministic Automata. Hvr, a more general lang is non-deterministic automata (in which a state may go to 2 or more other states), & its problistic variant, in which t. state transitions have pc (Hidden Markov Models)

N.B. Tho P. HMM allows loops of states, it is still much weaker than cf languages (Phrase Structure / More generally Trace langs ⊇ PSLs)

In MTM we want > 1 "soln." to t. problem if possl. Do I want to express this set of solns as a (Non-Determ.) lang, or a stochastic lang?

The Trace of a soln could involve exploration of t. paths of a Non-Deterministic Lang.

Given a INV problem, Giving a stoch lang. as a "soln" is a way of narrowing down t. pc for Lsrch.

○→ SN The First GHT says that pc/cc order is best for minz Σ cc of soln. Hvr, this may not be our Goal! We may want to be getting info: in which case, we may broaden our search! (Perhaps the Monte Carlo Srch of GA?) In fact this may partly justify t. otherwise very-wasteful Mt. Carlo Srches of GA. Hvr, GA usualy (or always) doesn't use t. info very well — i.e. it doesn't use info from "poor fit" trials very well. It does get more "Diversity" than Purely honest Lsrch.

T. main idea of t. TSQ: In t. initial TSQ, I will be very aware of all t. concs. usd & what TM's cjs is for all probs. I will use Lsrch for solns. I think that is the only system that will work! To have acceptable cjs for a problem the / very concs have to have been acquired by TM. After TM has aquired a reasonable no. of Good concs, I want to be able to devise TSQ's wo. my detaild understanding of their Conceptual Structures. After TM has done enuf TSQ's I expect it should be able to continue w. less & less care on my part in writing TSQ's.

.01: 275.40 T. most immediate Goal is to acquire a fair understanding of Algebra: (possibly enuf to work somewhat hard probs.). Next, learning to understand English Text about Algebra.

T. (string) methods of 274.05–.10 should be unnecy to specify. T. entire lrng. techeque should be expressed in t. "English soln." of t. problem. Hvr, 274.05–.10 may help me express T. English soln. in (clear) unambiguous form.

.07 Also in t. English TSQ & soln: When a heuristic is found or we just want to incorporate t. info "discovered" in a soln. of a prob. in t. TSQ, — First Tell, in English, just how this new info is supposd to influence subsequent searches for prob. solns. Then, find a formal lang. in which modns can be inserted so as to modify t. P.D. expressd by t. modify lang. so it expresses t. heur or t. knowledge lrnd from t. problem soln.

.13 → So t. English TSQ & solns should guide language design. 66 $8\times 8 = 2$ $8\times\frac{?}{?} = 5'\times 5'$

By "language", I mean TRMC or "P.D. assoc. w. a TRMC".

At times I don't know just how I want t. info in t. newly solvd problem to influence future searches. — I often have a qualitative idea, but no key as to a quantitative idea. T. quantitative part usually comes from ALP analysis.

→ Hvr, I do have to express what I'm doing in some exact lang, & as soon as I do so, ALP is able to give me pc's. Imped a lang that can express heurs as well as problem solns".

[SN] T. no. of Bits in a heur or rule doesn't nessly measure its pc! Any "OR" statements mean the rule has really several parallel codes so pc ≈ pc×2 for each "OR"! "And" statements ≈ divide pc by 2.

So, say to start, I have a lang. to express functions. A set of primitive fncts & composition rules [?]

.25 Consider $3 \frac{?}{?} 4 = (?)$ type problems: Say TM recognizes "nos" & t. symbols (= nouns.)

.26 add, sum, div, Mul operators are available; they go from 2 nos. to 1 no.

.27 So it's "natural" to try them on t. 2 nos. in t. problem to get t. soln.

I don't have a lang. for T.M. to do such "reasoning".

The fact that "sum", say maps no. pairs into nos., is an "empirical fact" found by TM thru statistical analysis.

[In general, all of Math, including axioms, can be regarded as "statistical info" obtained w. very large GSz & t. very likely to have pc close to 1 or ϕ.]

So: I want to show how t. "reasoning" of .26–.27 is t. outgrowth of statistical observations on t. corpus. It should be possl to assign ϕ pc's = ϕ.1 to natural ideas as I like. — i.e. the generating P.D. should be at least that flexible!

There is t. general Q of how to express various constraints obtained by "discovery" into t. P.D. in a way ∋ Lsrch could still be used effectively. (see 277.29 & .32 for good ideas on this.)

I had this idea that PD's could be expressed in many different ways & that (277.40)

0.18.00 Bulg.

.01: 276.40 One big trick in problem Analysis was to view a P.D. in various ways — that enabling Soln. of a problem. ~~Various~~ "Viewing in various ways", is a Major Screening Heuristic. (eg. change of coord system; Time domain ⇔ Freq. domain; Minz ⇔ Maxz; Gorc ⇒ Monotonic function of Gorc;)

One (apparently "IMP") idea is that ∃ Trmc formulation can express any derbable P.D., & in this form it seems easy to implement Lsrch.

One apparently simple way to create a Trmc from a P.D. is to list the strings in pc order, then assign pc's via Huffman Coding. Every finite string has a P.C., so one starts w. null, 0, 1, 00, 01, 10, 11 .... & assigns P.C. to each via ∃ known P.D. We then somehow assign Huffman Codes to them. ← (No! This is wrong! We want to assign Huffman codes to all strings of same length (since Pcorr ∝ ε pc = 1).

So first assign codes to 0 & 1, Then assign codes to 00, 01 & 10, 11: The code for 00 say, will be ∃ code for 0, w. extra bits.

→ This will not work! A code for an n bit string will be n bits long!

Perhaps read Cover-Cheung Paper. (But actually, I don't think this is impt. int. person problem: ? See .17

.17 So an impt. Q is: Is it possible to put all P.D. into Trmc form & for Lsrch? Can I do ≈ Lsrch w. a less constrained form of P.D. ? All I really need is sands (or strings) in ≈ PC order. (Perhaps knowing ε c at each pf. in derivation could help relax need for exact knowledge of pc ordering)

Anyway, I want to look at ∃ P.D.'s that I will be using, see how various Heuristics affect their pc ordering: See how they affect Lsrch.

As I see it: ∃ main Q is, how problem scales affect ∃ PD (since Heurs can be expressed as "Prob. solns" w. suitable sizes), in terms of ordering of pc (& $\frac{pc}{cc}$)

We don't want ∃ approx. ordering of trials to leave out any very promising ones — Tho it could include many low pc trials. — This is an analysis of ∃ relative badness of different kinds of errors in ordering Trials.

.29 One way to get reordering suggested by a Heur! To look at ∃ reordering that occurs in ∃ "Hypothetical" Pseudo Corpus that disregard ∃ Heur in Quest. — Perhaps make an "analogous" reordering on ∃ problem that ∃ Heur is to be applied to.

.32 (SN perhaps) → Every time a problem is Solved, a "demon" is created to look for probs of that type "I" in ∃ future. (Or for "Analogous Problems"!). Periodically, various demons will be (unified interpreted), perhaps w. some Compression as well.] Mindful of Smithspeit pandemonium.

An explanation .32: It can deal w. OSL. It can be an Impt way to deal w. "Scaling" (280.34)

0.19.00 Bulg.

.01: 274.40! OK: Back to the ANL TSQ of 276.25 [redacted] & f. Hour of 276.26-27.

It is possl. that TM has studied f. 4 arith. operators statistically & "knows" much about their properties, "slots"

[With random nos as test objects, it's unlikely that TM would discover that division by zero is imposed.] Perhaps the best way to get this info to TM would be by "hint" or some kind of "Telling". Another way: use [redacted] $\frac{3}{4} = .75$ but $\frac{3}{0} = ?$, $\frac{4}{0} = ?$ or $\frac{3}{0} = 3$, $\frac{4}{0} = 4$ etc. Or $\frac{\pi}{\phi}$ = largest possible float. $\frac{-3}{\phi}$ = most negative possl. float.

Another way: T. special nos, ϕ, 1, π, e, 2, 3, [redacted] -1, -2, -3, have special properties. Some integers have special properties. Another Lenat Heuristic

One nice heuristic that Lenat used: When a new object type was discovered/invented, AM would investigate its properties in a systematic way: There was a certain fixed set of "slots" & it would fill each of them (to varying degrees).

But Back to 276.25 "in English": Perhaps stay in English more: Not too much formalizn. at this pt. Yet! The f. "details of formlzn." of 277.29, .32 are v.p.!

T. [redacted] hour 276.26-27 can also be in Eng., & As a separate problem, we must ask how TM acquired that info.

So do 276.25 w. at least 2 hours! ① T. operators maping 2 nos. into a no.

② The simplicity of the structures of 3+7, 3×7, even & of add 3,7; Mult 3,7 ... The general idea of ANL (recursive interpretation of expressions) would consist in a recursive say a/c over heavy a stack.

A Main line of rsch: That we may want to concentrate on 'ordering of trials"

.20 So, in English, figure out rules for ordering trials, Then find way to assign pc to those trials. These 2 operations may interact a bit, but see if it's possl to do .20.

So: We start "knowing" T. function recorded will be a combn. of f. 4 binary alg. fcts & perhaps other functions. Since final funct. must have numeric output, the reasoning about "Types" (numeric v.s. Symbolic) can be easily formalized w.r.t. "Type" language/algebra. The 4 arit binary arith operations are all f. same wrt Types on I/O. — So this is a big simplifin. of f. problem. For complexity level of 1 fcts, there are only 4 possl type maps. $n_1, n_2 \to n_3$: for nos 1,1; 2,2; 1,2; 2,1:

[Also, there is f. possy of ① functions w. no input ② Unary fcts of inputs
The only unary fcts we have are x (Identity), & $x+x$; $x-x$; x^2; $\frac{x}{x}$ ← unless $x=0$

.30 .31 Also "unary" fcts: $x+2$ (arbitrary const.) $x-2$; $x \cdot 2$; $x \div 2$. — So sort of, 2 fcts, but 2 can have different typ in $x \div 2$, $x \cdot 2$.

Re: .30-.31 drop it for a while! Just concentrate on f. 4 arith operations & this particular "Typo" heuristic.

.33 .34 One trouble is that TM has had no "previous experience" w. function formation, so it couldn't have derived this heuristic. Actually, it doesn't really need that heur just now, because f. problem is so simple. In fact, Much (or all) of TM's early lrng will be w.o. heurs. It will use "large" KB's for solving these early probs, but, in general, f. probs will be "easy". After this phase of lrning, various heurs can be discovered, because there is a large enuf corpus of data to base on which to base TM's discovery of heurs. (Apart from OSL discovery, that needs

281.06

279.01

Oct 5, 00 Nature: 2 arts on Concerning Research, other Comments

.01: 278.40: only 1 or 2 cases ... actually "1½" cases)

.02 So just try to do a Paramt. of t. TSQ w.o. "Heuristics". — Even if (CB CJS) is quite hy. When TM has enuf experience" I may then start by adding sequences of data for TM's discovery of hours! See what CJS is needed! A large CJS may be tolerable, since it could take many yrs, for t. Sci. community to discover 2 new heur.

T. "w.o. Heurs" of .02 may be a bit extreme! take: $3+7=10$ $x_1 \cdots x_5$ 5 arguments.

4 binary functions: 5×5 possl. input configns for each:

So we quickly find Add $x_1, x_3 \to \Sigma \to x_5$ & $\text{Add}\, x_3, x_1 \to x_5$.

We get $+, -, \times, \div$ in this way: now w. x_1 opi $x_2 \to x_5$ when we use random opi = random $(+ - \times \div)$ we still get good corpus compression, but not

.11 "perfect" (whatever that means!) (.19)

SN In "Batch Mode", TM might develope "curiousity" & various other behavior patterns appropriate to a "crazed Scientist" rather than a "Pure Engineer".

— How this works: TM finds that "experiments", "investigations into the properties of functions" — inventing & using functions" all do result on often useful abbs for future problems. However, unless TM has a v.g. criterion for "interestingness", v. long-run yield of this behavior will be small.

So, v. Sc. Community (& Organic Evoln) develope "Criteria for interestingness".

.19 : (.11) Some useful functs $f(x_i, x_j) = T$ if $x_i = x_j$; else F

T, F is somehow used to "control" other functions.

$G(T, \alpha) = \alpha$ $G(F, \alpha) =$ undefined.

This way to do it "correctly". The exact Mechanics of how to have the functs (primitive or non-primitive) controlled by "Ops" is not yet clear. In t. "Ob-of" algebras, correctns were used.

SN On 250.34-.40; 251.01-.02; 251.20-.27 I discussed just what t. Bottlenecks were, after the Saarb TSQ's, & how I have since developed methods to deal w. these Bottlenecks. It would be well to expand (a) just what were t. Saarb TSQ's (b) More detail on t. bottlenecks & just how I would now be able to deal w. them.

First, t. TSQ's themselves: (1) The ANC TSQ ended w. recursive evaln. of Alg. expressions. Expressions were in Polish notation, & Pushdown stacks was used in the "working lang" of t. Model: W. Paul may have suggested that part. (2) A TSQ to learn to solve linear alg eqns & this by discovering t. equiv of t. "Laws of Algebra".

(3) A given (substitution & "desubstitution") heuristic, used for solving linear eqns, then when $\sqrt{x}$ was added as primitive/soln of quad eqns then when $\sqrt[3]{x}$ was added as primitive, soln of cubic eqns.

280.01

.01 ! 279.40! Bottlenecks ① Dfcty of writing TSQ's ② The soln. obtand only usd
.02 complete solns. of previous problems (no sub-trees of functions used). ③ The solns.
didn't "scale" well — as t. TSQ continued, t. no. of concs in many tinearly, so
t. pc of t. concs ∝ ✱ ∴ it became more & more dfclt (more cc) to solve
problems. Problems w. "soln depth" of k required cc ~ n^k. In human prob → (.34)
.06 Solving, cc ~ constant or a slowly ↑ funct of n — more like β^k, where β is about indep of n.

→ Idea of task has k params: AND/OR of 2 tasks gives a new task with params: Also try to prove any tasks impossl.... Try conserve given cc

Since Saarb! Several impt. developments; ① Sort of understanding of
.08 Soln. of AND/OR nets (WOA problem) ② Soln of MCT ③ Better
.09 understanding of TM_2's Gorc (281.01). ④ Some general Heurs for solns of TSQ's:
e.g. Find Fn ∋ An = Fn(Qn), then (Unify / Integrate) Fn into t. previous soln upto n-1.
This last was done using Ob-Op algebras & observing "correlations" betw.
certain Ob. ouputs & t. successful use of certain Ops.

Re: .01: ① Dfcty of writing TSQ's: Seemd much eased by MCT — since a much
larger scope of "problems" could be put into t. TSQ. In particular, t. use of
"Learning Definitions Inductively."! (Definitions are usually "Told" to a human
.18 student). One effect of this is that t. need to discover useful sub-trees
in useful functions, would be much relaxed. { A Mixed Blessing! —
since we do want TM to have this skill — But in t. mean time, omitting this it
may be!} [This is because t. "useful subtrees" (like x^2
or maybe sin x?) would be lrnd as "definitions" & would ∴ be ≈ "final solns
.25 to problems" (.01–.02)↑.

Also, I had t. idea that I should be able to use practically
any TSQ usable by humans, — just as long as it didn't contain need for
R+W. specific info: (That I might be able to insert such info in TM,
or modify TSQ so it's not needed. So: use TSQ from Human
Text Books. ② Re: sub tree discovery: see (.18–.25) But also it may be
that t. TSQ's I was giving, simply didn't have t. complexity of concs, such
that sub-tree discovery was needed. {Tho it would seem that in eq. solving
.34 sub tree discovery would be useful!} ③ on "Scaling" (see .02–.06): The ~β^k
is obtaind by finding suitable heurs, for each conc., so TM "looks at t. situation"
& this suggests certain concs should be usd; thus ↑ t. pc's of these concs enormously!
Also, see 277.32 for a possl. way ("Demons") to realize heurs of this kind: Also
away to discover them.
→ (see 283.32)

.01:28.40 : (4) TM_2's Gone; (28.08-.09) Main breakthru: that Much of TM_2's Gone must be "USR supplied" e.g. t. Horizon" & perhaps t. read of TM_2 cc to TM_1 cc (Properties this is part of t. "Horizon".). "Values" like relative cc to spend on "Cancer cure," V.S. "Theory of Everything" must be USR supplied.

→ Could I then just tune the Soar ANL system & modify it to fix its deficiencies? deficiencies?

.05: 28.34: Re: this objection that TM has no history at this initial pt., so "really hard hand crud hours" : I should be able to "bend t. rules" a bit; to enable hours at this point. I want to understand this heuristic (trap/insertion) as soon as possl.

Well, overgross T. Lang for just trap. E.g. a alg operator notations is quite simple — no where near universal. It only has num, num → num functions! It has no Boolean functs, no way to implement recognition of strings, etc.

I was thinking of a simple functional Lang w. definitions implemented via the

Z141 "precorpus" notation. It assigns pc's to codings of the corpus in perhaps a better way than simple Trmcs would — since it automatically summed over macros & codes to get pc's. At present, I can start w. any set of basis functions, & it can generate all compositions of them, & make definitions & use definitions of any functions create. But it can't yet, do loops or recursions.

One way to introduce loops & / or recursion, was my use of functionals. I think I can do both Loops & (simple) recursions using functionals. I have written about how to do loops recently, using this formalism:

```
z = 3
for i = 1 to n
z = (z+3)(z/2 - 3)
next.
```

The input to loop are "3", n, $(z+3)(\frac{z}{2}-3)$ real Integer function (real → real)

The output is real no ("z"); I. input is (z_0 (= "3")), n, f (real → real).

```
x = x0 , y = y0
for i = no to n.
y = f1(x,y) : x = f2(x,y)
next.
```

input, x_0, y_0 real real ; no, n : integers
fonts f1 real, real → real, f2 real, real → real
output, x, real; y real.

11.64 13.27 (3.27)^11 / 11.64 = 4.23

.32 (L)

```
F(0) = a
for i = 1 to n
f(i) = G(f(i-1))
next.
```

← This is basically to look like t. loop form of a recursion! (R)

recursion: (1) f(0) = a
(2) f(n) = G(f(n-1))
Defines f(n) for ⇒ n integer ≥ 0.
Which is also done by

So (32 L), t. loop form is equiv to (32 R), t. loop form, but is of lower cc (usually) but (32 L) seems to have a more complex depth. I think Peano stack it doesn't not,

262.01

0.23,00

Bilg.

ABcde ABCDE
ABcde ABCDE

.01: 281.40: In Post 1982 Lanet Paper (why AM/Gurisco à Gurisco seemed to work), he gave an example of a recursive depth of "2 strings being of = length." Could I simply extend that into a loop? Perhaps it is possible (even easy!) for "simple" recursions (is Ris p rim. rec. funct?).

.04 It looks like I could use Z141 "program" formalism to define loops à recursive functions, in terms of constants à previously defined functions, and get suitable pc's for the functions defined. .35

.06

So what I need now, is some more primative functs. to enable TM to recognize conditions (ops) à control functions ops.

What I want, is a nice way for Boolean ops to control numeric ops.

A possl way: $F(T_{rue}, f(x)) = f(x)$, $F(false, f(x)) =$ undefined.

I don't like Pc's because it could lead to inconsistant expressions.

or $F(T, f(x)) = f(x)$ $F(F, f(x)) = \phi$.

We want a funct to be $f(x)$ if $\alpha = T$, but $G(x)$ if $\alpha = F$.

So $S(x) = F(\alpha, f(x)) + F(\neg\alpha, G(x))$.

We would define a new funct: $S(\overset{\text{real}}{F(x)}, \overset{\text{real}}{G(x)}, \overset{\text{Boolean}}{\alpha})$

Analogous functs can be defined if (F & G are Bool → Boolean) or (Bool → real) or (real → Boolean)

Functs of this kind would be useful (& natural) to map 3+5 into add(3,5), à recognize "+" → "add".

Maybe Types of vars: real, integer, Boolean, String.

Note! Recursive funcs are normally defined for integers only. The extension to reals can be done by say $\sin'(x,n) = [n \cdot \sin(x)]$ so we can get arbly close approxns to $\sin(x)$. This is not, htr., the way I had recently been considering defns of real recursive funcs. The usual X! to obtain example of a recursive funct is Oh, as long as it is clear that x must be an integer in this type of defn. It could be used to define X! for all reals if one defines all values on say (0,1) as a boundary cond. — But in the case of X! this would be difficult.!

.35: .06 Anyway, this Z141 technique is fairly general, so I could end up w. a universal density function on all finitely derivable functions. "Universal" in the sense that every derivable function has pc > 0.

This is not yet (closer to) the universality of the Universal Pd's assoc. w. UM's UM's. 283.01

6.23.00 Bulg

Abcd.

.01: 282.40! Each (integer / finite string) (usually, if not always), defines a function: Usually many integers will define the same function.

SN T. function is defined by a sequence of integers. The upper bound of each int. in the sequence is uniquely determined by the integers that occurred before it.

Can I somehow map all (Q one) integers into functions? Well, yes: Since I have a p.d. on functions, I can assign integer code words to them, such that form a prefix set.

However, since $\sum_i p_i < 1$, all integers ~~without exception~~ strings will not nearly represent strings of functions.

.10 Anyway, I guess the Q is: Given a finitely describable, probly density function on all finitely describable functions, will the density function on functions of 282.35 multiplicatively "Majorize" it. (> it for all functions (within const. factor))?

Since each function is represented by a finite set of finite strings,

— Then the universal p.d. on finite strings (the "Discrete" d.f.) will induce a universal d.f. of functions. (may be!) .10 Is where I have been before in discussing

.16 Universality. <269.01 ff>

5.5 → 7

SN It looks like, recently, I have been "playing my arms about," w.o. good direction! Perhaps go over where I am, what has been done & what needs to be done! Perhaps looks at previous reviews of this Q.

1.5%
≥10%
≥150

SN In OOPS™, the Certain Sets of OOPS are in Universal Sets (Bags), The "Sequentiality" The "what comes before what" in the TSQ is of vital importance. It may be well to think of any t.s.q. as being essentially a (sequential) T.S. problem! (Except for "Baton mode" Thg)

On the other hand, it would seem easy to write TSQ's for TM! If I know the concs needed & their hierarchical structure (The "concept net"), I should be able to make up examples, & TM should not have much trouble learning the concs, using Languages like 282.35. The main problem (perhaps) is to have adequate levers, so we can "Scale up". In general

.32 In General, a lever must be good enough to narrow down the choices so the info needed for choices does not ↑ (much, if at all) as concs ↑. (Is 277.32 adequate?)

[See 280.02–.06; 280.34–.40.]

I think in my solely analysis of TSQ's & "Conc nets", I did not treat the "scaling" problem. I.e. I must treat a "conc net" when obtained, would be enuf to insure TM could devr the needed concs. NO SO! If levers of .32 are needed, — if the CPU is to be kept below some upper Bnd (as required in .32.

284.01

0.23.00
0.23.00

Bulg

284

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.01:283.40: So, it would seem that what I should do is devise a conc. net for t. set of probs I want to solve, then find heurs that suitably reduce t. CJS's.

If necy, any conc. should be learnable by TM thru induction. Perhaps, for "practice" see if I can do this for arby concs.

T. simplest kind of conc. is that defined in "Computational Lrng Theory" (Valiant et al.)

Some examples: (1) an example of correct addition: 3+7=10; T. "example of addition" is t. conc. Neg examples: 7, 7+ (meaningless); 3+7=2; 3×7=21

Perhaps learn "addition" example then "correct addition" example

My mind is rather vague on t. forgg. concs. — just what I want them to be — just how they are to be combined, usd. "correct" & "addition" could be 2 combining concs.

[SN] Suppose I simply took certain English words that referred to concepts, & gave TM examples of them. This kind of info, would later be of use to TM in lrng to understand English text.

---

122 [SN] A (perhaps Novel) approach to TSQ Design: List problem types I want TM to solve. Give equs./functs/strings that solve them: Give several

.24 alternative solns when poss/relevant. The functs are closer to a conc. net than t. problems. Hvr, for humans, the English (vague) descns of t. problems have vital heuristic associations.

So, perhaps in line w. Lenat's AM & Eurisko: Each conc. will have various "slots": Some will relate to equs, functions: Pgms: Recursive, non-recursive forms,

Bot also Associational Baggage [Lenat said that Humans found this more useful (usually) for devising heurs, than t. pgm stuff] seems very reasonable: — This would include t. "name" of t. conc. (In English — mnemonic, meaningful for human debugging & following operations of TM). When conc. was devd/introduced. T. context of t. conc. [where it occurs: what kinds of probs it occurs in; generalizns of conc; specializns of conc.] Also, t. "subject area" (Algebra, Chemistry, ect)

We may devise a "special Language" for TM to deal w. t. "Assoc slots" info.

In (.24) "Soln" can mean several things! 3 most impt. are:
(1) single number (2) function, giving "(1)" as funct of problem descn (3) The Trace (285.01)

Bulg.

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ABcdefg $\int_{-\infty}^{+\infty} e^{-x^2} dx$ —
ABcde

*(3) "TRACE" of the soln(s).

.01: 284.90! Of the soln. : for very simple probs (& some not so simple), T. Trace is the same as the function (soln.) definition: It is a pgm (or equiv.) that defines the function.
In other cases, the trace will be a sequence of steps that result in the soln. final soln.. As a simple example, the final function might be very simple because of cancelations, or substitution of a very very simple expression for a complex one. The trace shows these cancelations a/o substitutions being done; T. "final soln." does not.

One (sometimes) easy way to teach, is to give examples of (the correct) working of a specific problem (This was the usual way they used to derb gen'l. solns in the very old days of Math.)
Anyway; TM must then Genz. this specific soln. so it is able to work similar problems.
e.g. show how to solve linear eqns: 2 linear eqns in 2 unks, 3 linear eqns 3 unks....
Then can TM genz. to N eqns in N unks. <N eqns in N unks.>

So: Try out 284.22 ff : Some possl. elements of the T.S. Q,

.20 3+4 = [?] problem! soln., sum 3, 4.; desired trace (at first) sum.3,4 ;
This trace does not include unsuccessful trials. I'm not sure if this will always be true! An unsuccessful trial can have input in the that leads to correct soln. —
— If one has suitable Heurs. Otherwise, not — which is the case in (.21)

Next set of probs. (or single prob. w. long random real): 3*4 = ? .
First soln is mul 3,4. This compresses the corpus of 3+4 = 7, 3*4 = 12,
where "3" & "4" are long random nos: If the random nos. are long enough, we
.29 can compress further. (w.o. this further compression, we get ≈ 50%
compression). If further compression is ≈ 100% (length of random nos. ↑.
(corresponds to SSZ if we have > 1 example but many small
random nos. — log of 1 big random no. is equiv. to sum of logs of
.35 many small random nos) → 286.16 → 286.16
.36 In fact we can use (.29) ff for MTM to figure out when
it has solved many (if not all?) Math problems!

286.01

0.25.00

Bulg.

286

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.01: 285.40 Say TM knows what params in the problems are "random nos." (corresponding to ~~numerables~~ general real variables.). As lengths of reals → ∞, compression should → 100%, or for given length of random vars, Length of Code of Corpus should be Konstant + f(L), where $\lim_{L\to\infty} \frac{f(L)}{L} = 0$

f(L) could be a constant (or perhaps ln(L))

However, in MeM, 0, 1, ~~*~~ were integers, positive integers, can be exceptions that would not be caught by Random nos. The most common exception (by far) is division by Zero (and associated exceptions.)

.16: 285.35 (Dropping 285.36 for the moment): to continue 285.35:

We need to distinguish between "+" & "*". "In English"; + → sum; * → mul;

But we want to find + & *; Then decide on a functional formalism (op, Ob algebra?). (The "finding" * + occurs when 3+7 =? alone has to be found.

So; 2 problems: ① Deciding where the nos & symbols (= strings) are.

② Deciding on Formalism for functional form of + → sum, ect.

Actually, I don't have to decide at this pt. I have a good idea as to how much "info" is involved, & that's all I need.

I can make a (recursive) decision at this pt.

When the problem 3+7 =? comes in: There are 5 positions.

The possl. inputs to "sum" are 3, +, ?, =, ? : Only 3 & ? are defined — "sum" immediately rejects +, =, ? w/o computing output. This saves a lot of CC, because sum, 3, =, say, doesn't have to be opened w. 10.

In fact, for sum, =, 3, we don't even try 3 before rejecting when = is put into "sum" operator (? ?).

Next, How do we recognize "+" & use it to "exite" "sum"?

287.01

O:25:00 Bulg. 287

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3+7=?

.01:286.40: Well, try Identity function: id(α, β): α, β, can be strings or numbers: output is T or F. (Boole).

We also need a way for Boolean variables to control numeric variables:

One suggestion was cn(α, n) [α Boolean, n numeric] = n if α = T, = 0 if α = F.

So to choose sum, * → mul.

.06 sum (cn (id(α, +), sum(n₁, n₂)), cn (id(α, *), mul(n₁, n₂)))

A rather complex expression! But much symmetry.

Estimate rcost: $R_{cn} \cdot R_{id} \cdot 5 \cdot 5 \cdot R_{sum} \cdot 2 \cdot 2 \cdot R_{sum}$

5·5·2·2 is 100 already! The symmetries will be extremely hard to exploit!

Say we have just sum, sub, mul, div, cn, id, Sa [Sa ≡ same: Sa(x) = x].

7 functions. So $R_{cn} = R_{id} = R_{sum} = 7$ so $7^4 \approx 2500$

.20 2500 × 100 = 250,000 = M/4 already! If we didn't have sub & div its

.21 only $5^4 \cdot 10^2 = 25 \cdot 100^2 = 62{,}500$ only. A factor of 4, better.

W.o. invoking symmetry, better both halves of the sum. The entire expression would be $\left(\frac{M}{4}\right)^2 = \frac{10^{12}}{16} \approx 6 \times 10^{10}$.

Reachable but bad. Since id & sum are sym, we get an additional factor of $2^4 = 16$ so $\frac{10^{12}}{256} = 4 \times 10^9$. Also, symm for mul sum ⇒ 2×10^9.

If we did this not using "sub & div" as in .20 & .21, we get an additional factor of 4×4 so 2.5×10^8.

However, while these symmetries ↑ pcost they don't influence kcost or lcost — which is the real problem.

.30 If we could implement Random choice w.o. replacement, we'd get pc! Pseudo random trials would do this. Perhaps HASH coding would be more general.

← Not always! This present case Appears to be a special case in which kcost >> bcost. — whether it is or not, the main Q would be, how much C to spend on that trial? Hvr, when this is not relevant, we can get ≈ rcost to control search! So: When is it not relevant?

288.05

Bolz

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.05: 287.40 While 287.20 may be of great eventual importance I may want to return to main track: How to implement the symmetry of the 2 halves of 287.06 — & how to get more heuristics in, in general.

A different track: Can parts of the function of .06 have ≠ compressions ≥ ±2? hm

.10 More reasonable: Try to Implement the idea of finding which 2 functs work for each problem, then "correlate" the functions w. the problem domains.

Suppose these we had found equiv to 287.06 as solns. to many different probs: I have indicated in Red on 287.06 the invariant & constrained parts of eq(s). The variables are d, t, sum, *, mul, n1, n2 | 7 7 7 7 7 2 2 | 7 positions

recost:

So recost: 50 × 50 × 28 2500 × 28 = 70000 = 70k

This would be the recost of using the function 287.06: ×a mult by the recost of the function rscale (which is probably hy).

.19 So at present .10 seems like a more reasonable approach!

.20 What we do, is first get the set of pairs: $\{3+7=?\}$; $\{3*7=?\}$ ect. $\{\text{add}(3,7)\}$ $\{\text{mul}(3,7)\}$

So this becomes a new QA problem. To a human, it seems trivial, but what tools are needed/used? "=?" are constant! 3, 7 correspond to 3, 7 (long random nos)

.25 + → add ; * → mul. But how to implement this? .30

.26 The ob "+ was seen" correlates w. the op "create sum" ✓

.27 .19-.26 may be an instance of a more General (heur.): Parallels: "look at various

.29 Parallels: Mirror, Crossover (combine) Plan for total final Solns. Mutations (crossover) Should be part of normal Lisren. They are usually "OSL" cases. "OSL" cases.

.30: .25 In .20 we want to find what 3 is a funct of; what 7 is a funct of, what "add" is a funct of. Clearly 3→3 ; 7→7 ; I. mapping could be made by "correln". Another way to look at the "functionality" In .20 3+7 .. = 3*7 differ in only 1 symbol add 3,7 = mul 3,7 .. .. " " "

So the functionality must be betw. those symbols.

.35 On "correlation" ("+", "*") is a predictor of ("add", "mul") w. "100%" accuracy so pc = 1.

In .20, we look for test various symbols in top row for efficiency in predicting symbols of bottom row. This is good because it works in "Grey" cases → 289.05

0.26.00 Bulg. 289

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.05 : 028.40 ! as well as "Black & White". This "correln" is a standard heur. for doctecting "causality".

.06 One main Reason I'm not 100% happy w. 288.20 - .25, .30 - .40 ! I don't see it in a "standard form" as coming from a real or ~~[illegible]~~ hypothetical TSQ.

Pol perhaps drop. objection .06 for a while. Later, I ~~[illegible]~~ will (I hope) have experience w. Discovery of heurs, & I will ~~[illegible]~~ have ideas about how to correct deal w. .06.

.11 So: I've more or less gotten TM to do the TSQ $3 \overset{+}{*} 4 = ?$ T. cjs's ~~[illegible]~~ are all very small.

first each $3 \overset{out}{\underset{+}{*}} 4 =$ is learned, individually, then 288.20 is used, followed by the "correlation" Study of 288.35. This is all pretty much in "English" so the details have yet to be worked out.

Next: to learn evaln. of arby alg. expressn. Best have all notations in Polish. This is simplest. Later, a prob. for TM will be ~~[illegible]~~ betw. Polish, & rev. Polish ⟷ infix ⟷ Polish w. parens, etc.

Polish

Perhaps a fundamental Operation for TM, is substitution of expressn for a variable or expressn → expressn or expressn → variable. Another possible, is use of push down stack. (fund. operation)

+ (3, 4) → sum (3, 4) ← it means : to get ~~[illegible]~~ $\frac{3}{7} =$

+ 3, * 8, 9 → sum (3, mul. 8, 9). This is not very interesting! TM just does "+" → "sum", "*" → "mul", etc. The stack is not explicitly used.

(lang. of CPU.)

perhaps have TM trn infix → Polish

Just try this out! See if any interesting things are found. [Perhaps try Polish ⟷ rev. Polish.]

If may be easier, more "consistent".

~~[illegible]~~ ~~[illegible]~~

3, 4, + Rpolish	3 *, + , 4 infix
sum, 3, 4 polish	sum, 3, 4 polish.

In general, the ease of discovery of anything will depend critically on the notation.

Also, I could try parens w. polish & r. polish. : Any of these variations don't make much difference in learning + − * ÷ ~~[illegible]~~ individually. ~~[illegible]~~ Look for differences (if any) in more complicated Combinations of + − * ÷

Def pol, rpol. T. problem now is evaln. of arby alg. expressn. I will try r.pol (= rev. polish) → polish notation.

3, 4, + ⟶ Rpol	3, 4, + , 7 ~~[illegible]~~ *
sum, 3, 4 pol.	mul, sum 3, 4, 7.

Write out some possible "final solns" But I'd like : see if I can lead TM to it (them)

.37 A desirable soln. : replace 3, 4 + by (3+4) obtaining 7, 7 * , & evaluate this. Perhaps a recursive program! I 3, 4 $\overset{+}{*}$ occurs, rewrite expressn w. its evaln. Repeat until → 290.01

290

.01 : 289.40 no longer possl. If result is number that's t. result.
The idea of "rewriting" is of "Recursion" could be primitive or "heavily hinted" concs ("heavily hinted" means teach w. many big hints — like examples of worked problems.

SN Use of example traces of solns. might be an easier way to do early training than giving simply probs & solns. I may use a Mixture: Use t. "Trace" example only for probs too hard for TM to len other way.

Actually, 2 09.37 is n't such an unlikely thing to try. Say TM knows about "substitution" (& Rewriting). This substitution always ↓ no. of elements by 2, so it does "move toward a soln."

2 09.37 works for any of 3 diff. x/em problems! — It xfms prob. into a new prob. that seems simpler than t. original! The "OR" net. — which I considered to be a major "primitive" planning heuristic. → So This Looks Pretty Good! It is in English - so it's still quite flexible

.15 Int "OR" plan: "Any way I can "simplify" t. problem?" Well, in t. past, substitutions (int. form of evalns.) always ↓ T. length of an expression — (One possl. "GPS" measure of vector till ht.), so it's a good thing to try. → see .35 for OBJECTION!

SN T. routine of 289.37 is not really "recursion". It is More like a "Do-loop w. "Until" or "While". As such, it has less cc than "normal Recursion". Most simple recursions can be easily represented as loops of this sort, & t. ca pc of dcrn is identical to that of recursion, because it has t. same defining "parameters".

SN Solving Eqns of all kinds by progressive "simplifn" (.15) is a commonly .24 used technique. The use of an invertable substitution is a "special case" (as in soln. of linear → quadratic → cubic (& possibly quartic & possibly quintic, since in all cases, one adds a new kind of function that enables t. soln. to be found (for linear quad cubic, it is ~~the~~ linear inversion $\sqrt{x}$ & $\sqrt[3]{x}$; for quartic & quintic, I don't know yet — perhaps look at soln. of quintic ( )

Substitution (& perhaps inversion of substitution? (.24)) should be primitives in t. language or very easily defined from Primitives.

I want to Estimate pc of this "soln" of .37: Also estimate its cc, since I need to know $\frac{pc}{cc}$ (or $\frac{cc}{pc} \equiv cjs$)

.35 Using t. "OR" Heur of (.15) Woops! in .15, TM doesn't yet know that "doing a legit substitution leaves t. value of t. expression invariant." ← These are all concepts foreign to TM at this point. Would it be feasible to teach TM these ideas before having it ~~acquire~~ acquire the heur of .15?

291.05

Bulg.

.05 ? 290.40! I mi. to teach TM About equality: Func. is invariant under equal substitutions & other trans. unary, binary, ect. x fns: e.g. if a = b then f(a) = f(b) for any f, a, b.
Maybe not exactly true if F is floating pt. function! Then, TM works probabilistically;
These relations are usually true enough (so 99% accuracy): later-TM will try to find
how to get closer to 100% accuracy by looking at error cases;
Perhaps it could learn about the "division by zero anomaly" this way!
First, to teach what equality "is": Examples, pos & neg.

SN pos & neg examples: There are usually far fewer models that need be considered
if neg. as well as pos. info examples are included ... ~~But still~~ One can have
very A.H. models (w. + & - cases) but v. A.H. models look very A.H. — they have very
small pc.

So 3=3 3=4(no): 3+4=7 < This is rather complex! TM has to find
~~So~~ sum operator in its internal lang. An input Q., perhaps is "In what order to
introduce concepts to TM?"
Perhaps first teach "identity", then equality. Identity is between relation
betw. strings & strings or nos. & nos.
So: Is 3+7 a string or a no.? In Lisp, maybe, it's a list. Its eval'n
is a number. So maybe identity should relate strings to strings only.
3+7 is not "identical" to 10 tho "3+7 equals 10" because eval'n. is implied and "equals".
"Identity" can be a primitive conc (string α) = (string β) means (eval'n of α) = (eval'n of β)
If they want equality of 2 nos x & y to be primitive: It ~~means~~ is tested by ~~testing~~.
evaluating x − y : if result is zero, then x = y. TM will learn that if
x ≠ y then y ≠ x.

TM will learn to evaluate the function Solve( ):
.35 e.g. Solve (x, x+3 [redacted]) = −3
.36 eval (3+7) = 10
string.

N2.06 Bugs

.01 : 291.40 ! The present approach seems v.g.. Perhaps review recent work! Have I been "drifting about?".

Superficially, The "OR" plan of 290.15 is an "English" descn. of how to solve the problem; Trouble is, it does seem to use several concs that seem distant from "primitives".

So it suggests 291.06 ff i.e. To teach TM what "equality" means:

SN This somehow seems closely related to teaching TM what "quantity" meant — But seems more "exact"!

291.06 ff is beginning to sound like LISP! Distinction betw. strings & evaln. of strings, etc. Perhaps all of the concerns that I had about what I meant

.11 by various expressions: (e.g. distinguishing betw. occurrence of a string & the string itself) would be automatically taken care of by LISP. .18

The slowness of Lisp can be taken care of in various ways:

1) Lisp → Forth 2) People in Fortran (Fanatic Pgmg group) translated some Kozo's Lisp stuff into Machine lang. & a/o CC, C++.

Perhaps have TM learn Lisp (as a TSQ!). Another possy is to learn MAPLE.

.18 : (.11) Another thing I was confused about, that Lisp deals w. in a possibly o.k. way: the idea of Binding temporarily: An assignment statement is only true, until the next modifn. of the variable.

Also, the idea of Local v.s. Global variables: In Basic, we use local variables in subrtns. All other variables are global.

Also: I have a file on "Functional Langs": It includes a long review Art. on "functional langs"

At 10:06

SN A possl. way to deal w. large (or ∞) cc of trials in GA. Use it Lsrch: i.e. For each GA run, set each has a k value, & truncate a trial as soon as cc.pc > k. For the next GA run, do k ← 2k. Pick some reasonable value of k to start the series of "GA runs".

.30 SN On the possy that the 2141 d.f. of Rec(x) Lisp functions, is a universal d.f. on all functions. To show this: Consider these functions:

F_1 maps integers into Lisp functions: it is able to map all possl. functs from integers.

F_2 maps integers into integers (essentially what a UMC does): Picking (Discrete (skipping)) is universal D.F.

.34 $F_1(F_2(\text{integer}))$ maps integers into functs w.t. universal D.f.on functs.

$F_1 F_2$ is a function realizable by LISP — it has > ∅ pc. (finite cost).

So if we give Lisp random inputs, this D.f. will ("Memorize") $F_2 F_1$

.34 functions — so it, too, will be a universal D.f. on Functions.

Q: Is this a non-trivial result?

(This way finitely describable function $F_2(F_1(\cdot))$)

263.06

N 10.00
Bulg

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.06 : 292.40 Well, what is defn. of a "universal d.f. on functions?"

Say G is a D.F. on all functs G(F) = pc of F. ; G(F) has a finite domn.

If $G_u(F)$ is a Universal d.f. on functs then $G_u(F) \geq k\,G(F)$ for all F, for some k > 0. (k is indep of F: is same for all F)

If 292.38 is true, how "picky" do I have to be in evaluating pc's from Z141? How bad is MDL?

Note: Tho t. F all have finite domns, they need not be effectively enumerable. (?)

[redacted] Does 292.30ff prove Universality for t. Z141 → Lisp d.f. on functions?

.16 **Two Previous Discussns of "Universality" of D.f. on Functions!**

282.04 – .06, 282.35 – 283.16 || 269.01 to at least 270.16 (270.18 – 271.02 maybe relevant)

271.03 – .40 seems to be very relevant.

.19 Back to 292.30 – .38 ; F_1(F_2 (integers)) maps integers to functs, & gives a "universal" d.f. on functions. "Augmented Z141" is a way to map integers onto functions .... it is a possl. "F_1"

The "functions" that F_1 derbs, are from integers to integers. ..... → 302.16

271.20 switches to t. problem (perhaps more relevant!) of t. adequacy of 257.34 ... (which seems el. — tho 271.34 – .40 perhaps makes it less so.)

.235 Since 257.34 is prbly what I will use, & it will guide t. intellectual path of TM, 271.25 suggests that t. Universality of 257.34 (or of augmented Z141) — w. low

.25 is not as important as: "will it give solns on probs that we & not it enuf cjs?" T. idea was that I would lead TM along a path that a human could follow, using human heurs; that TM could (w a large tho possl CB) learn itself. — After going along this path enuf, it would have enuf heuristic power to continue w.o. much guidance on my part.

T. role of "Universality" of TM's model, in t. prgg. is unclear.

One simple (tho possibly impt.) way: When TM searches for a soln, (that is a function) it really is able to try all possl. functs. That t. "soln" is beyond its CB, is a necy restriction (debility) of all "Real" mechanisms.

.38 Perhaps a very impt idea is 284.22 ff ← (Algorithms) Novel way of TSQ design. → 294.01

→ This Makes t. TSQ more impt. than Universality. Even w. a non-universal lang, an adequate TSQ can bring us closer and closer to solving a problem — But universality w.o. a good T.S.Q. — one can (in practice) solve very little. Hvr. even w. a G. TSQ., we need a lang. that is at least strong enuf to express all t. concs. we need. Prbly, it might as well be Universal. → 304.01

N1700
Bulg

294

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.01:

†. Main goal of TM is "compression". Ideally, this also makes the future work of compression easier. One part that would make it easier is if "significant" ("Significant" / "insignificantly") longer codes are included. These are codes one would normally "Backtrack" to in Theory Revision.

Would "indexing" the corpus be of use here? Perhaps list all things one might do to make the tasks of the future, easier.

At least 4 kinds of future tasks: ① Augmentation of Corpus — we have to code the new stuff. Prediction of the future is perhaps best way to do this (also see ④)

② Modifn. of tasks/goals of system ("S.W. Maintenance"). If system is maximally modularized (so each part of system might also have alternative S.W. codes ... & ways of doing it)), this makes Modifn. easier.

③ System should have many "adjustable params" to deal w changes of applicn. of system.

④ Theory Revision: when the prodn. system of ① no longer seems to work adequately.

N19.00

295

Marcus:

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This is a reply to Marcus Hutter and parts of Jurgen Schmidhuber's "Learning how to learn".

Consider the problem to be RTM, w. h not specified but CB specified. Each [redacted] "trial" by TM is somewhat expensive, — so cost of the response (by the envt.) will, in general, be computed distinctly from TM's internal comptn.

As first approxn, consider RTM & w. h = 1. ("horizon" = no of trials to be used.) This is a somewhat less "straight OL problem". For h ≥ 2 the problem is to make 2 trials s.t. best of the 2 is max. This allows "experiments" (which h = 1 does not allow). (I think Jurgen was concerned w. h = 1, only).

If TM has intelligence ≥ that of user, this can be a very dangerous activity! TM will try to manipulate the USER in to giving it a high score..

For h > 1, I think the problem may be a "Dynamic Prog" problem. There are also aspects of the (unsolved) G = G(t, x) problem, in which G decreases w. delay.

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.06: 1.40 :

Int. [redacted] Problem of 1.18-30 At each pt in t. process, we have a sequence of Painting, Goodness pairs: $[pn_i, G_i]$ i=1/n.

.09 → We could regard it as a simple OZ problem; We want a painting pn_j → G_j is max, & we want to find it/within time T.

We would/have to use a rather sophisticated ~~OZM~~ OT ~~■~~ if we wanted anything of much value in

.12 a reasonable time. → (Note 4.12!). An OT is any thing that derbs a discovery strategy.

T. Method of 1.18-30 is perhaps a greedy method, in which each choice of painting to present, is t. "Best possi." that TM can do at that time.

More General (& Less Greedy) ~~■~~ OT's allow for "Experiments".

By allowing t. most General kind of OT's (Those describable to a UMC) we get t. best possi soln by Lsrch.

.20 So .09-.12 really is a "theoretical adequate" soln. of t. problem!

I was thinking that the "Cure Cancer" - (research direction) problem, was close to the "Paint an accordingly good painting" problem.

In t. painting problem, feedback to each "Trial" is a scalar.

.24 (It could be more info, ~~like~~ Like ~~■~~ a more detaild dscn. by t. "patron"

.25 of just why he liked/disliked t. particular Trial.)

.26 Anyway, in "Cure Cancer" t. feed back is in a very general form.

Superficially, if we Generalize t. idea of O.T., it could deal w. any ~~■~~ possi. form of feedback; so it could deal w. t. feed back of .24-.25 ~~■~~ and even t. very general type of F.B. in .26.

Finding a Good OT ~~■~~ of this more general kind is t. same kind of problem (w. what looks like an identical type of soln) as my older, simpler OT's w. only scalar F.B.

.33 Furthermore GPS w. i.e. vector Gorc, could be implemented by ~~■~~ this more General OT.

My impression of t. OT's used in "Lsrch": That one would best use only 1 OT — ever! Using >1 O.T. means loss of info between O.T.'s — so ~~■~~ OT's can't use info that other OT's have requested — ~~■~~ They may even duplicate trials

1) Most INV probs are solved by OZ or OT. T. invention of a good FF for a INV prob to convert it to a OZ prob, is an impt skill that TM would acquire early.

N22.00 Hutter

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.06 : 2.40 Made by other OT's.

If t. trials are very Expensive, this would be very inefficient!

Selecting t. [redacted] "Most promising" OT Thus far considered by TM could be quite good — if t. OT selection was like I envisioned it in MCT.

In MCT, we have this "Grand PD" That looks like a problem (≡ t. "conditional") à outputs a pd. on all possl. soln. trials. In t. older styles of t. input were a OE problem,

.12 it would output a pd on all OT's wrt. that problem — A possibility is that as TM

.13 matured, it would end up w. almost all wt. in one P.D. — Seems likely!

[I was concerned that in t. "Cure Cancer" problem of 2.20, t. results of experiments would be "probilistic" — But in fact, all results of any action are "probilistic" as far as TM is concerned!]

SN It would be well to integrate logical/mathical reasoning into ≡ "Lsrch" or ≡ MCT — since much development of OT's involves reasoning of this sort.

I did write some about this in last 6 mo. ... try to find it!

But do Go into t. "Cure Cancer" problem in some more detail! I think I did Not Solve it, last time around.

SN perhaps t. ideas on non-Lsrch suggested by .12-.13 would back. back Way to deal w. t. G = G(x,T) type of OE problem!

SN for "Any Time" problems, use Levin's Time-shared Version of Lsrch for OE probs

In General, if G(-) is expensive, Levin Normal Lsrch may be "rather poor".

In "Cure Cancer" prob à other OE probs w. expensive Gorc! ① cc of Gorc may be larger à less constant à known by TM. ② cc of G may be known in advance of each trial — by a formula $f(x^{(\text{trial dem})})$ Given to TM. ③ cc of Gorc is only known by TM as its own induced p.d. from previous observations of G, cc pairs.

This last seems close to t. "Chaos" problem. In Classic ALP, one can't ever get a d.f. for t. cc of a trial — but in Resource Bounded ALP, one might legitimately make an approx. guess.

If "Cure Cancer" is an ≡ INV problem, (so one knows when it's been done — or a certain threshold of goodness of soln is assumed) then t. ideas of t. First Gambling house Thm are relevant — one wants to mine E of Σ cc before soln is found.

4.06

N23.00 Hutter

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.06 3.40: In General, t. Methods of MCT can be usd. w.o. using Lsrch. Tho, in many cases, Lsrch is ≈ OPTIMUM.

W.R.T. OZ problems, Lsrch seems O.K. for induction problems. — either sequential or unordered — say w. fixd corpus. Lsrch is v.g. for finding t. unnormed PC of a corpus.

.12 T. "Method of 2.09-.12" automatically tries to "simulate" t. expensive Gore — an OZ problem: Hvr. induction is an OZ problem; for each OZ problem, TM finds a P.D. over appropriate OT's! → see 302.27

Normally in induction problems, one does not use Lsrch (in t. usual sense) for trying cand. codes. (in which any trial is or cannot succeed for)

Usually we arrange so that every trial results in an acceptable code: we do this by comparing t. amount of "correcting" that would be needed to xform a "non-code" string into a fitting code. More Generally, since induction is an OZ problem, TM recognizes it as a "special kind" of OZ problem (all OZ probs are "special" since t. dcrn of t. problem determines what kind(s) of OT's will be tried), and uses special OT's to solve it.

Hvr, there is a common type of induction problem in which we do use Lsrch: These are t. induction problems that I had in early Algebra learning — like ANL.

We don't use Lsrch for diffl INV probs (we usually re formulate them as OZ probs, w. suitable FR .... or as GPS probs w. vector Gore (2.33),.....

Yet, INV probs should be ≈ optimumly soluable by Lsrch: "If all needed theorems can be put into t. "Grand PD"."

For OZ probs, if they used Lsrch & "all theorems into working "Grand PD"" then it would, indeed, be ≈ optimum! — for same Reasons as for INV probs. → 301.15

.30 I want to look at Several Issues: to what extent can they be expressed as Modifns of t. "Grand PD."?

Take t. AND/OR Net "Plan" heuristic

Remember: In most INV probs, we're searching not for just a soln., but for a strategy that will solve t. problem. Lsrch is over Progs "strategies".

In "Cure Cancer" problem: In view of past info, what is strategy w. max Probby of solving problem? — one w. min expected CC of solving problem? ← First G-H thrm. First G-H. Theorem.

A Heuristic would be part of t. dcrn of any strategy. This idea of a strategy is one of

3.06

N23.00 Hutter

300

N.B. .18 – .40 should
Be in Bulg file
Perhaps make Hutter
pp 1 thru 5 → Bulg 296 thru 300.
Bulg 295 is on Hutter.

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.06: 4.90! To Mm. Lcost — but can
.07 it deal w. "Quick & dirt"? [Actually Quick Abort is perhaps hours like it. Couldn't TM speed by very much! The Expected CS is always ≈ $\frac{CCI}{PCJ}$ ≈ CJS. → 301.11

Re: Use of Hours: Ordinarily, only 1 or 2 will apply in a given case; so not much of an "exponential explosion": Once a hour is chosen, it could closely guide the soln. of the problem (not much ↓ in PC).

.11 GPS is and AND/OR planning are fairly General Hours, applicable to almost all problems. GPS for ENV. problems: AND/OR Nets for (I think) both ENV & CNV.

.13 Another very General Hour: "Use analogy to successful soln. of known problem"

Using Hours in this manner ("When a problem occurs, find associated hours to guide soln") Makes it sound like AM, Eurisko, Cyc — Lenat — But it certainly seems much different in spirit from Lenat! — Certainly NO "Teaching by Brain Surgery" I guess I by Difference: My TSO's are much less "shamanic": Much larger CJS than Lenat's. — Also, they are always probabilistic.

.18 SN In/retrospect: One of the most important thorny problems that I ran into was teaching TM Definitions of various concs. My mind was very vague as to what relationship was better: many concs & examples of that conc. I had hoped that my Reading about details of "LISP" would help here. → .24

.21

.22 T. 3 Hours of .11–.13 are really quite impt Hours! Used almost always (GPS & AND or Nets). "Analogy" [Nets or successful soln.] used maybe less.

.24 (.21) In .18 → .21: I think my own internal ideas of what concs. mean, were too vague: E.g. the "ideas of" quantity". — Yet vague as they may be, concs of this sort can be very impt. in Irng! I think the problem was that I did not recognize that the conc. I had in mind was really quite vague!

So, T. Q is: how do I implement this Vague Conc? — More Generally how are vague Concs. conveyed ("implemented") to TM?

T. eval / quote notation of LISP is probly very useful for teaching concs. — but still vague concs will occur, & I need them in that form — just how to deal w. this & when to deal w. it.

Wrt. explicitly vague concs: Perhaps just give myself lots of examples & give a no. (maybe PC) saying how much the conc. is represented by each example. Thinking Vaguely is an impt. method of inquiry/learning.

Perhaps "Fuzzy Reasoning" is ≈ what I want.

30[illegible].06

N25.00

Bulg.

301 6

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:06 :5.40

The hours like 5.11-.13 [redacted] do seem expressable as modifns of t. pd. What Raysay is: "t.i.s hour has a pc of .3 of success" & (T. hour itself will be a sequence of instructions (some of them probabilistic) on how to solve problems) ← T. hour is expressable as a modfn. of t. GPd (rand.)

.11 ; 300 :07 : [ON Quick About]: Perhaps it would be possl. for TM to discover t. "Quick about" hour by "Logical Reasoning", rather than [usual methods of heuristics]. I haven't really gone into t. role of Logical reasoning in problem solving & hour invention.

.15 ; 299.30 : ((Lsrch for OZ probs ..., + optimality of): T. idea in OZ probs is to use Lsrch (Lsrch over OZ probs) over t. P.D. of OT's. So, say a person/α (non TM) has a better way to work a particular OZ problem than TM. Then α must have a reason (way) to chose a particular good OT for that problem, & that reason must be based on past experience &/or Logical reasoning. If those reasons were available built in to TM's GPD, it should be able to get about t. same results.

.20 The args. on .15-.20 would seem to hold for "practically any" OZ problem type: — including those w. "very expensive evaln. funcs"

So t. weak pt. (if any) of .15-.20 (suboptimality of Lsrch for INV. probs as well) is t. "Logical reasoning" part: (also in .12). "Logical reasoning" is a special case of probabilistic reasoning [redacted] — but the pc's close to ϕ & 1.

.26 So t. Q is: **How can TM learn to reason "Logically"?**

Well, suppose we first teach TM Math & formal logic in a purely [theoretical] way, so it can solve problems but knows no relatnsh betw. (math & Logic) & other problems math is given. We then give TM problems (B) in which its math/logic knowledge is relevant, & it could find there is an "analogy" betw. math/logic solns, & t. solns. of problems, B.

It might also be possl. to teach TM "Practical Applicns" of math/logic along w. t. Reas. understanding of Math/logic. "Practical Applicns" are work TM's own normal problems.

→ Try to find a "practical problem" for TM, in which math/logic would be useful. Wall: TM2's work. — initially, TM2 is mainly Lsrch, but it could be made aware of t. more general goal of min E(cc) for total soln. or for "T. horizon Time".

→ Another approach I had to ANL did use Logical reasoning!: I didn't know how to do it, however. → 305.06

N28-00

Bulg 8

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.06: On Summarizing Machines & Prediction.

Say One has a corpus, C, & one has found "several poems" that are not bad for this corpus. Predictions can be made in 2 ways:

.10 1) W/ a mean of the several poems. ← "THEORY REVISION":

.11 2) Devising new poems: These use consensus of the "several poems", w. appropriate wts.

Hvr., T. Mechanics of .11 is not clear: It may be that 10 is almost always used — unless it seems to not work anymore; In which case one goes to ".11" in "Theory Revision" mode ← .35

to not work anymore: In which case one goes to ".11" in "Theory Revision" mode ← pilot.

D3.00: Z 141 using a complete lang. (like lisp, say) generates a Universal P.D. on strings.

.16 One (Apparently) big Q is: Can one generate a Universal d.f. on functions this way? ?? ← .23

Some recent writings (within last 10 or so yrs of Bulg) suggest that this is not very important — that what is more important is a good TSQ. Even if the D.f. on functs was Universal, it could assign a very small pc to the "correct" function. A good TSQ assigns by PC to the known soln., even if the D.f. on functs is not Universal. (See 293.16 for biblio. of refs on "Universality"

see 293.25, 293.18, 27h.25

.23 .16 If we had a nice way to map strings into functions, that would do it! (Say we need funcs of single strings only). We might be able to do this very easily in "Lisp" by restricting the set of symbols in the Alphabet that generates "Lisp expressions". (Note 293.19 on this)

.25 .16-.25 May have a very simple soln.! I'll just have to think about it in more exact detail!

.27 Anyway: In MGT, I was thinking of the 2 kinds of induction problems being "special kinds of problems" — the other 2 kinds are OZ: Env. probs. In fact, Induction probs are treated by MGT as "Just another OZ problem": The GPD looks at the induction problem, then because it is an induction problem, it uses special OT's for it. The nature of the OT will depend further, on whether it's seq. extrapol. or more general Bayes induction, & whether the params are continuous s/o discrete.... Also the source of the problem will help determine the OT. OT.

(Seq vs Bayes induction)

Well, "Theory Revision" may be a standard kind of OZ problem: but as induction, if OSL is possible, the OZ version of the problem becomes more complex (?)

303.06

Bulp
TSQ's

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.06 : For TM to learn
(1) good set of OT's & ~~[illegible]~~ (2) when each is to be applied to a given Problem.
(1) → Getting set of ~~good~~ OT's: If I insert a set of OT's that I think are good:
I should factor them as much as possl., into a compact code. This makes their form to be more likely one learned ("by itself") by TM.

(2) The assignment of an OT to a problem can be made on basis of (a) source of problem [Does it come from a Math, physics or chem" Domain? i.e. is it a Math or a physics or chem problem] (b) Type of problem; Is it an inductive problem? Is it continuous or discrete? (c) "Logical Reasoning" (301.26)

What are Main Problems (Bottlenecks) in TM rite now?

.19 1) TSQ design..... (see .26 for a perhaps novel approach).

2) Clear understanding of what t. Lrng Algm. is! 257.34 ... (Augmented by 21.41) & its non-el version 271.34 - .40. 271.25 ~~suggestion~~ suggests that 257.34's "universality" is not as impt. as "will it enable low cjs solns of probs?" ← (see 293.25 ff).

.26 3) Look at 284.22 ff. for an approach to TSQ design. --- see .35 also

4) T. mechanics of putting OSL into all induction systems. (Probly an IR system is needed)

5) How TM should operate so that its future tasks will be easier (294.01).

6) How TM can use logical reasoning; How TM learns to do Logical reasoning: What TSQ to use?
.30 See 301.26 ff. One new idea: Simply have TM do logical reasoning. Then figure out how TM could have
.31 lrnd this Technique. → 305.06

.35:.36 **TSQ TRACK** (not nearly Now): Take bunch of Math probs. Solvable by humans. Write down how I seem to Solve them. — Then how TM can ~~[illegible]~~ lrn to solve them — At first, not nearly using "a known codes", but eventually reformat it into this form, & devise a TSQ for them. I "sort of" did this for linear, Quad, cubic eqns. — derv of Solns. was at a "relatively hy level (English)". This seems V.G. How does it differ from 284.22? I think 284.22 ff is more into details. — But in fact, both (.35) & 284.22 ff can (& should) be used. I think (.35) is more general. → (304.2f)

.01: (293.235–.38)R : "Expressing a [redacted] Conc" means that we can "define it" so it becomes a single word, w. [redacted] an associated pc. This enables us to "compress" by using that Conc.

284.22 is a simpler way to design TSQ's. It's not as complete as a Conc. not, but closer to a TSQ. → [See Also 303.35 ff]

.05 So, in view of (293.235R)ff, all I need is a good TSQ, i a lang. adequate to express its needed Concs. "Z141" Along w. a "LISP" will perhaps be adequate for obtaining pc's of concs.

So how does t. elaborate treatment of OZ problems in MCT fit into .05 ff? Well; MCT is about GPD updating. How is .05 related to GPD updating? — Presumably, GPD updating is just another OZ problem. Like Most OZ problems, it has special techniques for doing it.

→ [SN Like OZ probs, Each INV prob. can have(a) particularly appropriate way(s) to solve it. GPD looks at an INV prob. i gives a PD over soln techniques for that particular INV problem.]

I had assumed that t GPD would simply give a pd for t Lsrch of any INV prob. presented to it: (Probably) not so! As I noted in t past — Most INV probs are solved as OZ (or GPS) probs, not via t Lsrch.

→ It looks like I am gradually Abandoning Lsrch as a Universal prime problem Solving Method! — Tho in many circumstances, it does seem to be optimum.

Also, it may be that many ostensibly non-Lsrch methods can be (usefully) expressed as modifn. of t pd. i (usually if not always) amenable to Lsrch. (There is, of course T. "Quick Abort" problem!)

.24: 303.40 Consider all sorts of probs, at all levels: e.g. "Cure Cancer", Devise Good unified theory for physics"; Devise V.G. painting from stored set of paintings (do it for poems, music, etc.). Make big list of interesting problems that I'd like TM to be able to solve. Chess, Checkers. (Good evaln. functs; Cheap, effective)

.29 On Prob. Svng. in General: First time one solves a particular kind of problem, it's pure [redacted] Lsrch. $\frac{cc}{pc}$ depends on t. pc of t primitive cines used to der b.f. soln.; Only "simple" probs (simple wrt t primitives) are practically solvable.

Next, one solves probs by "OSL" find problem in past ~ to present problem i use "~" soln. techniques.

Next, when one has 2 or more examples of a problem type, it is easier to decide on pts. of similarity of t. prob. i soln. so ccot/lsrch for soln is much reduced.

It is possl. to solve probs w.o. t "OSL" phase: The first i second times one solves t. problem it is by pure lsch in both cases. After t. 2nd soln., one recognizes it is ~ to {prob. i soln.} of previous problem. By eliminating OSL as possl. soln. method, one is probably at a very big disadvantage, but it is still

.37 possl. to do TSQ's i acquire "fly intelligence" ← (perhaps) —

In t. SAARB work on TSQ's, I ran into t "SCALING" problem: → SPEC 306.01 → 305.06

D8.00 Bulg

Logical / Mathematical Reasoning:
How TM can do this,
" " " learn to do this.

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301.12
303.31
.06 : 301.40
(See 303.30-.31)

.08 Some areas where TM uses Math to do induction:

1) Linear Regression: we want a set of coeffs to minz. Σ sq. error — an OZ problem. In General, each OZ prob may have its own Soln. mechanism. Consider a normal analyzer OZ problem: To find ƒ: max of $\sin^2 x + 3x^2$. People learn how to do this. We could teach TM to find extremum this way. It would first have to know diff. calculus. (which is one way: Probly Archemedes could have worked it w.o. dif. calc!)

2) There has been much work done on Continuous optzn: "Info Based Complexity": Joe Traub et al. Also, on discrete optzn. (to some extent). TM could be "taught" all of this.

Bulg.

Int. SAARE tsq's,

spec
.01: 304.90 : I put solving of all problems in "Memy" w.o. indexing them, so pc of memy use → $\frac{N}{N}$ (N = no. of probs solvd) — A clearly non-viable method of prob. soln.
This might not be too bad if one accessed memy only once in the course of one problem: But for 2 or 3 memy accesses, the process becomes quickly of too low pc.

Some simple methods of indexing: (1) Recency of occurrance (2) Source of problem (Math, Physics, Chem, etc). (3) Similarity of structure of problem description to that of previously solved problem ("Analogy"). (3) is perhaps most important.

So, Given a problem: Subproblem is: What is/are the previously solved
.10 problems? (or previous unsolved probs?!) ← To avoid unsuccessful techniques of the past.

.11: 304.37 : Actually, how much trouble is OSL (≈ ϕ!) — w.o. OSL, when a new problem comes in, we look for "induction systems" ← An induction system is a problem soln. structure that is common to ≥ 2 problems in the past experience corpus
that have been obtained w. SSZ ≥ 2. This does involve some searching.
Here, the searching is easier than in OSL, for 2 reasons: (1) There are many more "solved problems" to search over than "induction systems" (perhaps not quite so!
— we normally search over the total of solved probs that have not been put into induction systems + induction systems: the cardinality of the latter is usually << the set of all solved probs; à perhaps not much larger than the set of all "induction systems"

(2) In an induction system, the relevant features of the problems have been "distilled out" (= isolated) — This makes it much easier to tell if it is relevant to the new problem. On the other hand, comparing an unsolved problem to a solved
.24 problem — it may be difficult to decide on the relevant pts. of similarity.

On the Basis of (.11 - .24), it is not always clear that one might save any cc at all (even w. comparority a con) by not doing OSL.

.29: : Re 304.29 ff: (our Prob Solving primarily by "Case Based Reasoning")
Often we will not find previous problems or "induction systems" that will solve the present problem or be relevant to it. Pure Srch is not often the way to go.
We'd like (at least) to be able to do Srch over combns of cons (larger than primitives) — But the system of 304.29 ff only puts prob solns. & "Induction Systems" in memy (i.e. "defines" Recog). There is no mechanism to obtain intermediate size combs! (only primitive cons & "Induction Systems")
One (not completely A.I.) method: Factor the "Induction Systems" to find the pc of the entire set. The "factors" are sub cons of all sizes.
This "factoring" is part of a fairly normal part of prob. solving: i.e. compromise
the all-over-pgm found thus far. We then do Srch over the parts (= cons)
to solve problems. It may be that our factoring using 307.01

D.11.00 Bulg.

.01:306.40 Solns of previous problems (could) would be a special case of Lsrch, — So wd have an integrated system.

I have been considering 2 kinds of General Prob. Solving Methods:

.05 T. Older Method: All probs solvd by Lsrch: T. Pd for each Lsrch is a cond'l. P.D.; T. "condition" being t. problem dcrn. The unproved assumption is → that all heurs are expressable as modulus of t. P.D. ("G.P.D.") (It's known that "Quicksort" is not included) A (perhaps) associated assumption is that this probliztc form of th. heurs can always be obtaind from t. Data that t. "Human form" heur was obtaind from. — This "Data" includes (manual logical reasoning)

.11 T. newer method: 304.29 — .40, 306.01 — .10; .29 — .40: Idea is that new problems are always either solvd by raw Lsch or by finding probs. (or sets of probs) that they are ~ to, & using ~ soln. methods.

T. new method seems closer to t. way I solve probs, but I'm not sure about its "completeness". Could I do a TSQ from Elementary Algebra, using this method?

D.12.00

.15 Part of t. "older" approach was to try to do TSQ ~ to that of Humans doing, say, Elementary Algebra. How this fits into t. older Model of .05, is not immediately clear! It seems to be like t. newer method of .11. — Perhaps I can integrate t. .05 & .11!

One idea is that I would do .15: That .05 would enable me to express all prob soln. techniques & heurs: Thus perhaps .11 would be a major heuristic.

A Main idea of (older) .11: That whenever I (or anyone else) solves a problem, it is by putting together concs. that derb. t. soln (method)., so an "adequate" sub problem becomes t. acquisition of t. subconcs — usually by solving simpler problems.

A (recent) idea on writing TSQ's: → 303.35 Amass a large set of problems, w. one or more "solns" (= method of soln = "trace"). From this, see if one can make a Conc. net & order t. problems suitably.

303.35 & 284.22 are on this ~

Another (perhaps related idea): To write out how I solve probs on a very hi

.30 level (of English). E.g. "How to evaluate Alg. Expressions": & Find a part that I can evaluate. Evaluate it & substitute it, obtaining a new expressn. Recurse to & until I have a pure no. or until I can't find any sub expressn. to evaluate.

I could write a Lisp pgm to do this: T. Lisp pgm would be "factored" into concs. that would be lrnable from other problems.

As a Heur for above; A subsn. of an evaln. "Simplifies" t. expressn — thereby bringing it closer to goal. This would be an "Adequate" heur to solve t. problem. It is an example of an "OR" decomposition of a problem. (AND/OR nets)

Re "Simplicity" a shorter expressn is usually "Simpler", Hvr. in t. present case, 308.01 309.01 Spec

D.12.00 Bulg.

The Horizon problem: Rather unclear as to how to define final goals. Say one wants to ~~goal~~ (maximize) yield per unit time to be max over "long time". This is equiv. to max total yield for "long time." So one would do meta problems first. (always first) — So one would never do non-meta problems.

One way to fix or. thy is to spend say 10% on meta probs, 50% on non-meta probs. Also, it may be necy to put time scale on meta probs: how far into future is it oriented?

So at least 2 params needed ① f. % on meta probs. ② f. time scale.

An automatic time scale might be found in terms of ability to predict f. nature of future problems to more than (say) 10 yrs into f. future. (This 10 yrs would be as ~~beyond~~ part of TM & beyond that of SC community)

So, anyway, since I will be (assisting/giving) TM lots of params, I have no qualms about giving it those params. People who want Rotair machines to be "truely autonomous" don't like Rotig, but I certainly don't want a "truly autonomous machine"!

D·14 00

Bugs.

·01:307.40 getting monotonicly shorter "equivalent probs" would usually be adequate to solve t. prob.

One big Complaint about t. SAARB ARL soln. was that it was not a way a human would solve the problem, so it was hard to identify useful comps in t. soln. In 307.30 ff we are close to t. way a Human could solve it. We Do seem to need t. concept of "evaluation" — which may be a subtly important idea so we can make it "primitive".

One Big direction of TSQ work was to start at t. Beginning & work up to a very Smart Machine. A Perhaps Better Approach would simply work on "patches" of TMS training at various levels, then "patch" t. patches together. Working on patches of various levels will give me a better idea of what t. real problems of TSQ design & TMS learning Alg.s are.

.10 .10 certainly isn't a new idea, but I haven't really been doing much in that direction. An extreme case of .10 ff would be to take an existent "Expert System" (such as a Symbolic integration Pgm in Maple or Macsyma or ...) and slowly convert it to a learning system. — Try to find out how each of its routines could have been "discovered".

16.9.00

.01: ON MUTATIONS, (Recombination) CROSSOVERS : a) SSZ=1, SSZ=2 respr.

On mutations as SSZ=1 ($L = SSZ+1 = 2$). This is fine. We can use any rep'y we can think of that (seem to be) appropriate to the "Domain." It as a basis for generating a mutated form of the parent.

In the case of Recomb, SSZ=2, ($L = \Sigma: SSZ+1 = 3$), we learn about the same situation as SSZ=1; we are much controlled by rep'ys observed in the General Domain of the problem. For each such domain, we must learn how best to design recomb. techniques.

For Recomb, the main idea that ALP contributes is that child should have as many common parts of parents as poss'l. (but having part from one parent is also good (a Mutation))

As Before, my impression is that in certain domains, a well designed mutation scheme can be much better than an inappropriate Recomb. alg'm.

One could evaluate a given Mut. Alg'm using the "present population". We could get its mean & var. in predicted G of child.

Similarly we could Evaluate a recomb. Alg. (for 2, 3, 4 or any no. of "parents").

.17 **A META Problem:** To devise new (Mut/Recomb) Alg'ms, & select best ones (A GA-type problem). What languages what Mut, Recomb. to use for this "Meta problem"?

Also, the Core for a (Mut/recomb) Alg'm isn't so obvious! — we want an ideal M/R alg'm that Maximized "rate of ↑ of Gmax".

↳ An alternative View: To get a good, cheap pd out of the existant Corpus of Cands. ($\equiv p(X, G)$). — This is in the Direction of "PM_2"

Another View of .17, would be that we mutate or recombine a (small?) population of (Mut/recomb) alg'ms. (eventually, this scheme might become "Recursive" so this mode of mut/rec. could also be (mut/rec)ed.

— Re: Human Creativity: An experienced Scientist will solve diff't problems rapidly: his pd's for trial solns are narrow, but often correct. — His searches are short. Often, however, they will be not interesting solns.

The novice Scientist will have a less narrow search alg'm. It will be much slower in finding a soln to a problem, but because of diversity of trials his solns to problems are less ordinary, and he may solve problems unreachable by the "experienced Scientist" a/o he may find/use unexpected things along the way.

A very successful experienced Scientist will have a somewhat diffrnt direction of search than most. Also, if he spends lots of time, he will eventually go to strange unusual places and find great discoveries.

Very poor Scientists & Engrs: Use Excessively Elite Srch. →186.01

8.12.00 Bulg:

Mutations, Recomb, Search, "Creativity".

.01: 185.40: Also, poor researchers do not use good search methods: Many quit after trying only the hyest pc trials. They don't properly consider hypoths in Lsch order. e.g. 2 cncs. w. pc's of .1 & .01 should be given no less wt. than 3 cncs of wts .1, .1, & .1. This latter is spuriously Elitist in spirit (consider only hy pc. ideas — severely hy pc ideas - rather than one low pc or a mix of hy & low.).

A merit of much of Koza's work may be that he has (because of not so good prob gen'n methods) a very broad, very diverse set of cands that he trials. It takes him a long time, but he is able to solve very diff't probs. in "Creative" ways

I have this (Mild) fear that doing a such in Lsch order may be "Non-creative" due to low diversity. — But I think that on the average Lsearch (using all available info) will get me the best solns. in available time. If we give it enuf time, it will be "creative" & (on the average) more so than non-Lsch based searches

8.21.00 Bulg.

195

160 ff names Review Recclaim
outline of and TM — 163
Princeton / 66

Genido / 76.21
Mary McPeak. 172.11,34

IPC of Human!

Gary Kasparov
The IPC of (≈) Deep Blue, t. machine that beat Human world chess champ.
Said to be ~ 10^{12} flops. very vague in terms of bits/sec, hvr.
Got some idea of speed by [redacted] no. of Board evalns/sec. ← which may give.
Hvr, no. of "Bits" "flops" in a "Bd. evaln" is very unclear; well from ratio, I can get no. of flops in a bd. evaln.

→ Perhaps read up on just how Deep Blue actually worked

The speed of "Blue Gene" is to be 10^{15} flops (or 2×10^{15} flops)

Check on this by other comparisons, other estimates of speed of Blue Gene.
"Blue Gene" speed is also given as multiple of Deep Blue.
But much of this can be vague & inconsi.

What Human Computing does well ① fast parallel optical & perhaps acoustical processing. ~ 10^{11} bits/sec.
② Fast ∥ info retrieval. This could be, by far, the most massive amount of computation done. The amt. of HW (dendrites) used, could be enormous.
→ This last could be impt. for OSL: to find situation in past, ~ to present problem or "relevant" to it. "Associational" lrning: Perhaps done by ≈ Zatocoding.
So large "Content Addressable Memys" would be a very cheap way to get what appears to be "enormous IPC"! — Hvr, at present time there is not much market for "CAM", so it (perhaps) not cheaper than what is now, t. conventional way to do information retrieval.

The fast optical & acoustical (& maybe touch — tho this can be slower) need not be used in a TM of great intelligence.

Latest civilian super computer cost/performance. \$ 45M for 6 T flops:
\$ 7.5 /flop = \$7.5 per M flop, \$7.5 M per Gflop seems hy!

An upper bnd on human IPC: cost per flip in ergs is $E_n \cdot$ Time $\geq h$
so $E_n = \frac{h}{t}$; $\frac{1}{t} = \frac{E_n}{h}$

$\frac{1}{t}$ is bits/sec for 1 bit. ; for N bits w. $\frac{1}{t}$ sec. each: $\frac{NE}{h}$ Joules used.

$\frac{E}{h}$ $E = \frac{h}{t}$; Power $= \frac{E}{t} = \frac{h}{t^2}$; $\frac{1}{t^2} = \frac{E}{ht}$ = Power for $\frac{1}{t}$ bits/sec.

so $\frac{1}{t}$ = bits/sec = $\sqrt{\frac{E}{ht}} = \sqrt{\frac{watt}{h}} = \frac{\sqrt{watts}}{\sqrt{6.626\cdot10^{-34}}} = \sqrt{watt}\cdot 3.885\times10^{16}$ Bits/sec

$h = 6.626\times10^{-34}$ Joule · sec

$\sqrt{watts}\cdot 3.885\times16$ bits/sec. for 25 watts (human brain)

this is only 1.942×10^{17} bits/sec. ~ 2×10^{17} bits/sec.

How much of t. 25 watts is used for computing, is unclear.
There may be a bug in this reasoning, hvr. ; ∥ procsr. w. bits rate ∝ √power may give ~ ∮ power for much ∥ procg!

Also, t. Q of how much power (if any) is needed for Memory access

8.31.00 Bolg:

EBL

A possl. model for EBL: A problem occurs in a particular domain, in which one has much "knowledge", both domain specific & logic/skill. There is enuf to solve (or almost solve) the problem deductively, using that "knowledge": but there is a fair amt. of cc involved in the particular process of deduction.

So: Next time a "similar" problem occurs (presumably TM has good enuf, relevant enuf categorizing skills to realize this particular relevant "similarity") it reduces the cc of deduction considerably by performing after the previous case. This is a bit like CBR, (& OSL)

Bdg.

G(x) · f(t) : "+ Super OZ problem".

on the "Super OZ problem" : G(x), F(t) : G is Gore for candidate, x,
[illegible] F(t) is monotonic f funct of time: Both G & F are known to TM.
find x [illegible] in time t ∋ G(x)·F(t) is Max.
[More generally Given G(x,T) w. G(x, T_1) ≥ G(x, T_2) if $T_1 < T_2$.]

Use technique of MCT to deal w. this:

TM has many cases of (OT_i (G_j(•,•) , G_{ijk} , T_{ijk}) ← omm; x, T

.07 OT_i (G_j (·, ·), G_{ijk}) ← each ij will usually have several (often Many) "k" trials giving different
.09 x_{ijk}, T_{ijk} & G_{jk} values.

[OT_i is the optza. technique. This is true, tho each OT_i will have a single stopping criterion that may or may not be met.]

From data on many OT_i's & G_j's (TM's Life experience)
.11 TM is able to obtain a condl. p.d. giving P_{rob_y} (OT_i, G_j(·,·), G)
It is the p.d. of OT_i giving the Gore G for problem G_j(·,·).
.13 From this and (.11) cond. p.d., we obtain P_{rob_y} (OT_i, G_j(·,·))
which is the probly the OT_i will have the best (over all OT's) Gore,
for problem (G_j (·,·)).

.16 NO! (Using .13, one does an Lsearch over all OT's. [illegible]
{ The usual Q comes up. "When does one stop" — Each OT_i tells one
(as part of its defn. "when to stop")

.19 .16 is wrong! One simply picks the simple OT_i that is most likely to
.20 give max G for problem G_j(·,·). The data used for the PD
of .11 and .13 (which is directly derived from .11 w.o. any new data)
say include (a probly will) include info for several Trial T_{ijk}'s for each
OT_i (.07–.09 & ((.07–.09 R)) The Data is (.07–.09) The
Resultant PD's are .11 & .13.

.26 O.K. Now Consider regular OZ problems. I was not (I believe)
ever able to show that Lsearch is (practically) near optimum for OZ probs
(tho it's second to be for INV problems). — Could it be that for
Normal OZ problems as well, this is the case (i.e. F(T) = constant)
The soln. of .19–.20 is best? i.e. just pick the "Best OT_i that is
most likely to be best & stick with it.

In both the G(x)·T(x) & the classical OZ problem; As we work on the
problem, the pd's of .11 & .13 may change, but we cannot usefully use that
info directly. What we want to know is the PD over the OT_i's involved
in a jump to those OT_i's. In the case of many OT_i's this involves a "startup cost" —
one can often use data from trials thus far on an OT. for info on
.37 prospective new OT_i's. → 214.05

→ In much of the [illegible]: One must spend time computing probabilities
that various OT_i's will be best. There is a nasty tradeoff betw (214.01

8.31.00

Bulg.

.01: 2.13.40 that time & t. time used to make t. TRIALS themselves. In t. Forge discussion, I have assumed that cost of a trial is >> usual cost of making pc estimates. We note, hvr, that t. cc spent on pc estimates can be arbitrarily large: it's not clear how much cc to spend on them!

"On use of Forth" for Cand Prmg.

.05: 2.13.37 In general, for even "regular" OZ probs. It would be best to start out w. t. OT that is most likely to be "Best". stay w. that OT unless/untill t. data on cands tried thus far (or any other info), makes it clear that (including cost of switching & startup costs), the Best OT is no longer OT_1 but OT_2 — so we switch to OT_2, Then to OT3 or maybe back to OT_1 — depending on how things look.

The Success of this method will depend much on how Good one's PD is. — but t. same can be said about normal Lsch for OZ or INV probs

.14 A Big Advantage of (in Normal OZ probs) of picking t. "Best" OT & staying with it untill it no longer looks Best!

.16 If there are several OT's all about equally Good — all "at top" of t. Goodness Range — Normal Lsrch will spend about t. same amount of time on each of them, rather than just pick one and work on it (usually to t. end) — This last is usually t. most efficient way.

One serious disadvantage of this last approach is that we will get much less data on "non-top" OT's, so we will tend to not find anything Better! It is an "Elitist self-confirming Hypothesis"

→ Is this Criticism also true for the G(x)·F(T) Gen?

① $\frac{pc}{cc}$ v.s $\frac{2^{-keff}}{cc}$?

② Failures in Many (([redacted] procs.

③ MONTE CARLO TM (MCTM)

In 1999, world production of 32 bit cpu's for computers was ~ [redacted] 100 M units. M units at 10^9 bits/sec is 10^{15} bits/sec! ≥ IFC of "Blue Gene".

10^5 may be > 1 human, so 100 M units may be ~ cap of Sci community — so "over threshold". In 10 yrs, ICP → ×100.

One problem w. ((mechanics of this number, is failure of individual uproc [IBM C. Blue Gene] is supposed to steer away of dealing w. this.

Another way might be Monte Carlo pgms. In 1990 (or 1991) I did a study of MC choice of candy to buy. The probty of buying a particular candy was ∝ its $pc_i^{\epsilon-1}$ (or was it $(1-pc_i)^{\epsilon-1}$.?) Anyway, the goal of that pt. was to get the effective cost of $\frac{pc}{cc}$ as ordering of trials rather than $\frac{2^{-keff}}{cc}$.

If it worked ☺, it might also be a good way of dealing w. occasional failure of individual uprocs.

In general, my work/interest in Monte Carlo methods of such has been motivated by the possible use of "Crummy Components".

Correction of Errors in Languages / (from Grammars): Several "Takes" on My:

.02 1) Recent Stuff I wrote Wolff, & my Corrections Ref. — I had 2 Somewhat distinct "Sols". — Hwr, .06 Ref is the nicest way to do this!

2) Method of correction used in linear/n.l. regression; Rec model gives a prediction & was based P.d. in with "correction".

.06 This method might be adapted to Language Problem, use ideas in my version of of S78T3 for Languages! — So that for languages, each bit in the corpus is given a Pd.

.09 In fact, .06 is a v.g. way, since we get from the error Recovery, an expected RMS error (& Likelihood error) for the predictions.

Call the analog of S78T3 for unordered finite strings "I. Grammar Error Thrm." V.S. S78T3 "The Time Series Error Thrm" or "Regression Error Theorem" So is "TSeries Form" = S78T3 or its Corollary.

Gram Err Thm. GE Thm. TSerr Thm. TSE Thm

Hwr, "Symbolic Regression" usually may refer to (& usually does) unordered finite strings.

Hwr, I have to work out more details of .06. Just how is applied to all sorts of langs. — I think the corrections are to be applied to each bit of the corpus as it is generated. This Area Needs More Work!

ANN & Genzel Hill Climbing.

In hill climbing, momentum is often used to guess at direction (& size) of next step & i.e. Say $\vec{x}$ is the confign. vector, & we are trying to find $\vec{x}$ s.t. $G(\vec{x})$ is max. ($G \equiv$ Gene).

"Momentum" $(\vec{x}_t - \vec{x}_{t-1}) \cdot (G(\vec{x}_t) - G(\vec{x}_{t-1})) = k \cdot \vec{m}_t$.

Our next trial, $\vec{x}_{t+1}$ would then be $\vec{x}_{t+1} = \vec{x}_t + k \cdot \vec{m}_t$, where k is a factor to be inserted by USR.

Actually, k depends on our estimate of the second derivative of G w.r.t. $\vec{x}$, & we can estimate it empirically from past (perhaps x window) data.

.10 Say $\Delta G = m \Delta x + \frac{1}{2} s (\Delta x)^2$ ($\dot{s} \equiv$ second derivative)

we want Δx s.t. ΔG max $\quad m + s \Delta x = 0$ so $\Delta x = -\frac{m}{s}$

We find m & s by doing a curve fit over the last r set of data

.13. pts. (or use an x window of width, r).

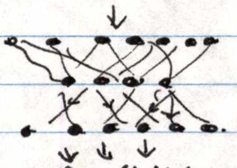
We can only do .10 & .13 once, however, since it keeps $\vec{x}$ on a simplex s.t. line in $\vec{x}$ space. After we do .10 - .13 once, we might do a trial in a random direction. — However, size of step is unclear. Size of step could be correlated to "Noise level", which we could have obtained in the previous .10 - .13 calculations.

9.12.00 Bulg

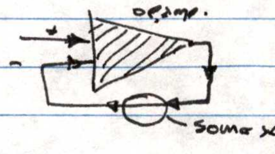
ANN, Fractals, Fractal Compression, CFG discovery

01:221.40 Discussn. of Simon (Grad Stud. w. J. Pollac at Brandeis.)

3 layer ANN



Output is fed back to Hidden layer: This is analogous to



This performs inverse of xfmn. A.

Say "A" is a Parsing Algm. — Say its a Matrix Parser w. Real (not Boolian) wts. — Suits for a stochastic CFG (or CDG)

The params of A a/o t. wts in t. ANN are adjusted using standard ANN a/o GA rules.

I Guess that we want to adjust wts. of ANN & Params of A so that t. input & output of t. ANN is ≈ ~~~~ its input.

Doing Matrix Mult. w. Reals, is much more expensive than using Boolian values! So one mite use stochastic ~~~~ Boolian values in t. Matrices! — This would then be like IFS (Iterated Functional Systems). Instead of a proby value of P in a matrix element, T. element would be 0 or 1; ("1" w. proby P, hvr.)

Consider a CFG:

$\Sigma \xrightarrow{p_1} a\Sigma$ $\quad$ $C \xrightarrow{r_1} Cd$

$\xrightarrow{p_2} a$ $\quad$ $\xrightarrow{r_2} \Sigma$

$\xrightarrow{p_3} bC$

Here, we have 2 N.T.'s: Σ & C: They correspond to Matrices. The p_1, p_2, p_3 correspond to wtd. addition rules on Matrices. There are 3 terminals: a, b, d.

We could start w. a very large no. of possl. production rules for each N.T. So Σ would have 10 proby values assoc. w. it.

As we continue t. process, hvr, t. we discover that only a few proby values are ≫0.

We mite start w. t. Grammar

Σ → abdamr } These are t. elements of t. Corpus!
→ adbdrm } This is t. A.H. Grammar.
→ } We start w. it & mutate it toward a u.g. "Goodness of fit" criterion.

additional rules
C → aΣ
→ Σa
→ bz
→ Σb