

MTM, MTM w. a few errors allowed in the seq.

MTM, MTM, H.C. TM.

TM₀, TM_±, TM_⊥ are descrd in Plan 336-337-20

This section develops ^{MTM} TM₀ :

Lots of general good ideas on TM's.

Part 338-393

Aug 7, 61

T.M.

Dart

338

01: The following will be a development of T.M.
as described in Plan 336.01 - 337.20

T.M.: The operation will be something like this: T.M. first
looks at P₁ input, makes some observations. On P₂ basis of
them, he decides what operators to employ on P₂ input to
produce O₂ output.

When a "new type" of input problem occurs, sometimes
the previously adequate operation method leads to
either (1) meaningless results (operations that are impossi. → they cause
the machine to stop or to get in an inf. loop) or (2) wrong
answer. Then, T.M. tries to find a "signif" difference betw.
the new input and past inputs. This may require a
few more "new type" inputs. ~~the~~ Once successful,
recognized, T.M. tries to find ^{new} operators that will perform
down with them. This is done by ^{try} using old motor ops, or
combinations of old sub-ops.

Under certain conditions of recurrent frustration, T.M.
will try "backtracking" (Plans 333.02 ~~to~~ 334.20).

O.K. now suppose we use a long seq. of div's, a/b, etc.
- and this T.M. does learn it O.K. - then we make
a T.M. ^{def: Plan 336.05} that works O.K. - Now, just what sort of
T.M. would fit into this, and what sort of hills could
he climb? - could we climb all kinds of hills?

Aug 8, 61

TM

Wart.

.01: 339.47

So, we "pass a curve" thro (or near

x, y pts of known. by G . We try to get

for which the mean G on that curve (average

past and future x inputs) will be max. given to a I, D pair whenever repeated variation may do, by making the

.04
.05

N to, 44.3 model is "noiseless" objection to 334.31 ~~objection to 334.31~~ ! that if we

This model, $x = 0$ as input, ~~is~~ ! that if we
make a trial curve by fitting thro G by G pts. only,

we may get the nice form for the curve near around

an x region that is rather insignificant - in G - x axis

that few input x 's occur in that region. Such

an trial would soon be discarded empirically, since

many ~~x~~ inputs resulting in y 's pts. of low G would soon come in.

But it is conceivable that one would continue to

create trials of this poor sort -

How: after a few such bad "trial curves", we would begin to

have lots of ~~the~~ x, y pairs of low G in the most common

x regions. ~~It~~ more It will then be difficult for a poor curve

to "fit" the few x, y pts. of known by G and avoid the more

numerous x, y pts of low G . - A fortiori it's a

"closeness" criteria in the x, y plane is used when

.35

the curve doesn't ~~exactly~~ pass thro a known x, y pt. $\rightarrow 342.05$

I suspect that while this TM with a top. seq. and an ~~ext.~~

external G , can be viewed as an h.c.TM, it is perhaps

best viewed as a special kind of TM i.e. a Reinforced

TM - i.e. r.TM. or R.TM.

One way in which the ~~ext.~~ h.c.TM \neq differs from R.TM,

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Dart

TMX

01:338.40! TM₄: It would seem, that as a type of. nil. in. x, the method of <trying to find out who of the known 'high points' have in common, and a new trial in this <the common mode> not very good - since all one can expect is another trial with about the same average - not a hyper pt.
Have trr - I'm not so sure that this is what I'm really contemplating for TM₁ ~~trr~~ \rightarrow (i.e. ^{plan} 334.31-335.03)

This \rightarrow idea is essentially a special method of h.c. which is useable in situations in which a top. seq. exists.
I.E. say we have a bunch of I, O pairs of known G's. We ~~then~~ pass an operator thru ~~the~~ set of pairs of ~~hy~~ G. This op. then has a hypermax G thru the entire set of I, O pairs (providing the samp. sz. is large enuf and the op. dem is of - low ent cost). If the samp. sz. is fairly smallish, the variance of the prediction of "true mean G" will be largest - which is desirable - in case that we have a noise source of very large true mean G.

30 - A model for visualizing the system:

A 3 dim. will: x, y, and G are coords
x is input, y output, G is the G of the pair...
There is some unknown surface relating x, y and G.
Also the statistics of the inputs x, exist but we shall disregard them for the present.

An "operator" is a correl. relation ^{in the x, y plane} between x and y. 3 in,

Aug 8, 6.
Dart

T 1/8

nl: 34.140 Th. Great Advantage to working on such a RTM, via
 uc TM, is that rtm is close to human, ~~and~~ it is
 to ~~require~~ ~~me~~ for me (a human) to devise hours.
 by starting and perhaps modifying my own hours.
 in objection of 340.05, to Plan 3.34.31, will be partially
 wholly met if we do use a "closeness criterion" and take as many
 of P_1 known (I, O) pts into account as poss. In estimating
 expected mean G of an op., we will give low wt.
 to (I, O) pts "far off P_1 curve" (see 3.39.30 for P_1 model
 referred to). As we get more pts. in the region where P_1
 curve is "bad", we will get more x, y pts. near the curve
 and these pts. will condemn the "apparently good" curves
 discussed in 340.05-35.

Also, a more complete discuss of ~~RTM~~ RTM such
 as 339.30 analogizes ~~with~~ (and is treated in code 273.1.01-
 100311) will elim. this dirty such treatment & takes
 both P_1 & x -statistics and the expected form
 of $G(x, y)$ into account.

Now, a course of action: ~~draw~~ draw up a RTM close
 to intuitive, and close to ^{present plans for} $TM_{1/2}$, and criticize it
 on the basis of P_1 . more complete model (analogized by 339.30
 My impression is that, at first approx., such a TM, would be
 very much like a TM_0 "passing an op. ~~of~~ P_1 ^{known} (I, O)
 pairs— Also, perhaps avoiding certain bad (I, O) pairs.
 This ~~will~~ be done up to evaluate P_1 expected G
 of such ops (before using them to create responses),

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T.M.J.

part

01:342.40 it may be expedient to evaluate them over a range of (I, O) 's using "closeness criteria".

Perhaps another step betw TM_0 and TM_1 :

As the seq. , TM is gn. certain rite I, O pairs but also some ~~is~~ I, O pairs that it must avoid. We must then find ~~some~~ a low cost op. that passes thru the "rite" pairs, but avoids the "wrong" pairs. I'm not so sure that this is much ^{extra} for ~~a~~ TM_0 to do - since in a MTM , there is only 1 rite answer, and all others are wrong.

So in normal TM_0 , if we gave the rite ans. for certain (I, O) pairs, this would be equiv. to saying all others are wrong.

- But maybe we can find problems in which there are > 1 rite ans., and so, giving some "bad" answers would contain info that knowl. of a single rite ans. would not contain.

I think that perhaps the concept of "closeness" is impt. here, in what I want - so we want to get as far as poss. from the "bad" (I, O) 's. Hvr., the fact remains, that the info that TM has may be only that certain (I, O) 's are good, others bad, and he may not know what the good responses are, to certain I 's - tho he may know certain responses ^{to them} are bad. The problem of using this info in an opt. way, to get a low cost of to pass thru some and avoid others (I, O) 's is certainly a fairly well defined problem

Aug 8, 6

part

01:34:40: A possibl. data. of \mathbb{R} . soln. We want
 \rightarrow in \mathbb{R} . futures, th. / mean no. of good (1,0)'s i
pass thro to \mathbb{R} . no. of bad ones. it will pass thro, w
max. We may further stipulate that ~~each~~ ^{every}
to a gen. input is either "good" or "bad".

Another possibl. "study TM" / ^{Bd. TM and TM₁} Several (1,0)'s are gen., and
TM must make his responses is "close" ~~to~~ to them as
possibl.

HW \rightarrow u _____ eventually - i.e. not rite away with tyo. seq

15 - - Impt. Q: Presumably, I will be reinforcing TM₁ ϵ
for speed of response. Then the abs that TM₁ will use
will, presumably, be selected partly with this goal in mind.

u. Q is - how much "choice" does TM₁ have? - i.e. can I
use a fairly simple search procedure, and still have TM₁ end
up with abs. But work well with even that simple search
procedure? Can TM₁ compensate for a poor search routine
and make \mathbb{R} . all over performance (after much tyo), as
good as that of a machine with a better search procedure?

~~It can, we use slower (at first) learning to compensate~~ Can TM₁
"learn" to get around a particularly poor search method?

If so, it ~~will~~ might be easier for me to start out with
a poor (the "adequately above threshold") search method, and
have TM₁ "learn" to use an essentially better search method.

This is what Dick Friedberg ~~with~~ have actually done -
but he didn't know what he was doing so he didn't know
just what to optimize, or whether his search ~~time~~ time was
adequate, or what \mathbb{T} mp. seq. to use and with what speed.

Aug 8, 61

TM

Part

01:344.40: I must do this in more detail, to see just how (if neccy) divide TM into a "TM, part" and an externally controlled by me "search part."

It is poss. to discuss R. above either on a fairly high level, or get a fairly concrete model of a RTM, and discuss R. Q. with resp. to that model. At the present time, my preference is to make the model first, then do the theo. dissn.

Some other things that one might "reward" RTM for:

Good!

- 1) Redundancy - i.e. resistance to random mutation in the computer - the mutation freqs being a true component error rates
- 2) ~~lightness~~ Use of less total memory or less total arith. unit. This would make poss. lighter, cheaper TM's of a given "capacity" - or a greater "capacity" per dollar a/p per pound.
- 3) Total speed - (of course!)

So what must be done now: 1) Some rather detailed descn of TM₀ with emph. on the search method to be used. 2) Descn of TM₁, then giving details of how TM₀ is to be modified to get TM₁, - with emph. on search methods.

Stick close to intuitive descns (i.e. descns of my intuitive processes) - then make more rigorous: Then see what can be done about 344.15ff. -

Aug 10, 69

TM8

Dart

345.40: For a search method, at first just try a simpler
with choice \propto ^{apri} probty. of object. This is slow, but
TM₀ on this basis, and then TM_i. Even this poor
kind of search will probably cost much less on 3 & 4.15
by R. time I get to TM₁.
Hvr, in all cases, make estimates of search time."

346.501

Aug (2, 6)

MTM

Dart



01:34.0.40:

A lot of very good work on MTM

EPM = Theo of MTM = MTM Theo $\beta 768.01$

In particular, $\beta 1084.01$ ff discusses 2 ^{Dart} MTM using ordinary 194704 orders. $\beta 1084.40$

This dvlpmt was discontinued around, because I didn't have a clear theoretical understanding of the problem. I think that I may have one now - certainly enough for MTM.

Using the ^{notational} ideas of EPM $\beta 1084.01$ ff, I think that I will get a TM to learn $A \begin{pmatrix} x \\ + \end{pmatrix} B$. Her., the genl. form (ie. form of these probs. will depend very much on what probs. I intend to give in the future of the type seq.

Some types of probs: $A+B$; $A \times B$, $(A+B) \times C$; $(A+B) \div C$
 $(E + ((A+B) - C) \times D) \div F$; $A + z = B$ (solve for z) ; $A + B + z = (C + D) \div E$

$A+B = C \times z$; $A+z = B+z$; $A+z = B+z$; $A+B+z = (C+D) \div E$
Eqns. in 2 unknowns, then 3 unknowns, then 4, etc.

Differentiation of alg. expressions. (> 1 ans. poss.)

Trigonometric identities (" " ")

Look into the Q of what sort of background TM needs to learn to solve linear eqs in 2 unknowns (say), after having learned about solving eqs in 1, 2, 3 unknowns.

Probably (since TM will start off with my human heuristic observations of "regularities") I should somehow teach TM a concept of what it means to solve an eq.

Perhaps somewhat in the spirit of the book "Lincos".

One way to write such a type seq: Assume (as in "Lincos") that the pupil is very sophisticated mathematically - that he already has the ideas needed 347.01

Aug 13, 61

TMO

Dart

in W

.01: 346.40 Concepts (\equiv defns.) and likely needs to figure out is the notation of the input message

Perhaps start TMs with positive integers only - then introduce neg. integers and zero, rational fractions, irrational nos., complex nos.

.07 SN An old, but very impt. observation! ^{e.g. EPM # 3947, 30} That heur. and "regularities" can be looked upon as the same sorts of things. A "heur." is a prefer. reordering of trials. A "regularity" is a ~~reassignment~~ modification of the prob. of trials due to an observed poss. recording that the "heur." in Q. worked in R. ^{and is expected to work in R. later} past, means that it ~~corresponds~~ ^{is} ~~useful~~ ^{useful} to an "acceptable" means of recording, and is a "regularity".

A kind of ~~heur.~~ impt. heur. would be an ordered list of "regularity types" to look for in a corpus, or in the code of a corpus.

In the case of MTM, say $O_1, O_2, O_3, \dots, O_n$ are successive operators that have been devised to deal with successively larger parts of the typ. seq. Then the corpus for the "heur.", is $O_1 \sim O_2 \sim O_3 \sim O_4 \dots \sim O_n$ - i.e. we must find regularities in this corpus, so we may quickly ~~pick~~ ^{pick} get good trials for

.35 O_{n+1} after we have seen ~~all~~ all of O_1, \dots, O_n .

A very optimistic approach to Dart TM: Plans 81160.20 to ~~approx~~ 8272.40 (~10 pp).

There is some Q. as to just what sort of typ. seq. I really want. Eventually I'd like to have TM learn English, for a simple starter ~~on~~ on "English", or "Baseball" language of the Lincoln Labs group might be good.

Aug 14, 61

Dart



01: 347.40: Note, hrs that there must be some sort of fact for putting "pure data" into TM — This is so we can TM obtain an "education" (\equiv typ. seq.), by reading ^{conventional} books.

One way to feed in "pure data" — just feed in a lit. data as declarative sentences, then ask Q's about R. sentences. ^{correct} Th. (responses to pure data is null. ~~The~~ Hvr., all meaningful Q's must be answered. At first, make Q's follow closely, R. declarative statements that contain their answers. Later, make Q's follow more distantly. This has the effect of making the data sentence that is relevant to a Q, be less certainly determined by temporal closeness alone.

Later, ask Q's whose answs. depend on several data s's, and upon logical manipulation of these s's.

But first, I want to get some sort of TM₀ outlined, so I can then work on TM.

For ~~arith~~ learning — use Polish notation, with parenths.

What I've been looking for, is some sort of uniform format in which to present Q's to TM.

I think there is some confusion about just what is meant by reentry. TM, for "speed." There are 2 speeds involved: TM_i is, at any time an operator. One "speed" is R. mean time for that op. to solve ^{all} probs. in R. typ. seq., we may want to wt. some probs. more than others. Th. idea of "solving" a prob. is a TM₀ idea, hvr., that may or may not be used in TM. A loss of concept 350.01

Aug 15, 61

T.M.J.

Dant

01: 349.40! Involves some function of R_1 soln. times (for R_1 and R_2 & awarded by R_2 external reinforcer.

(b) We want the search for an op. of by G to be fast - in R_1 case of TM_0 , this problem is clearly defined if we use a Mc Search in TM_0 , we take R_1 "first" consi op. that comes along. Thus our search is automatically minimized. \rightarrow Th. Q. is whether we would, in TM_0 , automatically end up with abs. types that minimized R_1 's search time.

In the case of TM_1 , ^{th. goal in} problem (b) ~~is~~ has to be defined for each situation. What we want is that R_1 G's (including computation times) awarded to successive trial ops., ^{effective} as rapidly as poss. Actually, in TM_1 , th. G for an individual response, is gn. on R_1 basis of a response th. external G gn. for that response (2) the response time - This is the sum of R_1 time nacy to find a new op. (if a new one is desired) and th. time of operation of that op. in creating the response.

Not exactly!! We may avoid certain TM_0 op. trials because they take long to test, rather than because they have

.35 low aprip.

Th. probs. of 349.30 - 350.35 will become a lot clearer, once I outline a TM_0 in some detail. e.g. assume any format, then see what differences changes of format make in learning speed.

Part

01:350.40! O.K.: consider R. Alg. format:

We will, in R's typ. seq., want T_M to learn, eventually, equs. of various kinds. Each input to T_M will consist of 10 more equs., some of them containing the letter, Z. After any presentation with Z in it, T_M's response, must be to solve for Z. for each input seq., there will be a special terminal symbol, indicating R's termination of that problem.

so: Inputs might be: $A+B=Z$, $Z=C+D$.

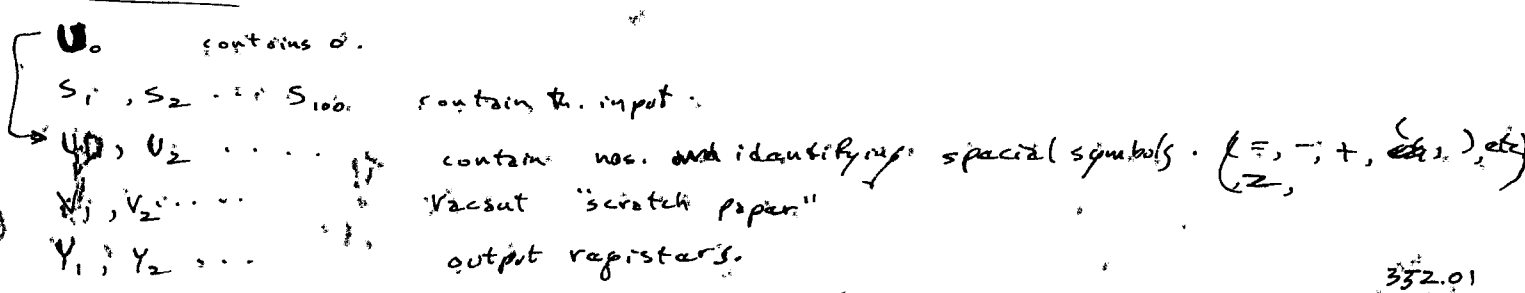
(A, B, C, etc., are random nos., Z is a special symbol, as are +, =, -, x, ÷, etc.)

We have 100 ^{ordered} registers for the input data (Most ideas ~ to EPM 1085:01 ff)

At first, assume that +, -, x, etc., ~~take~~ all take ~~100~~ ^{particular} about R. Some time, use fib. pt., +, x and ÷. This prelim. analysis will not be for any existing computer, but will be simply a study to concretize ideas.

Say our computer orders are: (ERM 1084-35)

- 1) Ad, n +
- 2) sb, n -
- 3) ~~M, n~~ x = (M, n)
- 4) Dy, n ÷ (Acc → $\frac{Ac}{n}$)
- 5) St, n (store Ac in register R_n)
- 6) Tr₀, n If $Ac = 0$, get next inst. " " $\neq 0$; " " " from R_n
- 7) Tr₊, n same as 6) * but inst from R_n if $Ac > 0$.
- 8) Stop.



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T.M.J.

Darts

01:35:40

O.K.

so: first set of probs:

S₁ S₂ S₃ S₄ S₅
A + B = Z

"C_{U_i}" means "th. content of R. reg. whose name is U_i"

~~C_{U_i} = V_i~~ means that Reg. C_{U_i} contains R. name, V_i

C_{U_i} = Z means " " " " " " " " Z ~~XXXXXXXXXX~~
↑ R_{is} will be some number

~~under order (b) Be: Co (≡ compare)~~

~~No! - what I would need have would be a u address to compare with Acc., another for poss. next instruction.~~
~~C_{0, n} means it ~~is~~ C_n = C_{Ac}, go to next order.~~

Use random nos. for each of R. "special symbols"

Constraints on orders.

only (St, Y_i) and (St, V_i) are legal. (we may later want T_M to be able to store special symbols of its own devising e.g. (St, U_z) - for z > a certain no. - but this may be unnecessary).

The "special symbols" can only be used for comparison - e.g.

(Sb, U₃), followed by (Trp, n)

The U₀ (C_{U₀} = 0) is a bit different. ^{originally} I / pot it in, to be used for clearing, but perhaps (CA, n) (clear Ac and Ad, n) would be better to add onto R. ^{set of} orders.

But, I'm getting much more detailed than I want to be yet: At R. present time, I want something like:

Problem: S₁ S₂ S₃ S₄ S₅
A + B = Z



Solu: Add C_{S₁} to C_{S₃} and present as answer.

Remember: I don't expect a computer to do a MC search to find Solns. to R. partly probs. At first, I will solve R. probs. Then

Aug 15, 61

TMJ

Part

or: 35240. from these solns. I will devise special order types, and to work on. set I've solved. With these orders and so it should be possl. to solve new, more diff/lt. probs, by a MS search by a real computer. I think the moral to have the look at the entire (or much of the) tape seq., and then try to get a reasonable op. soln. - Then take statistics on that soln. try very hard to minimize the unintentional coding by defining operations, etc., that are not ad-hock, and since they will be very useful in the future!

Aug 16, 61: It has been taking me several days to get into the spirit of ~~the~~ ideas plus $\beta 1160.20 - 869.05$. I'm not quite sure I'm there yet! ~~mmmm~~ ~~huh~~ ~~huh~~

For input, I can feed in 2 kinds of "data." (1) "facts" (2) "problems." For TM, the "facts" must all be correct, for TM, ^{for TM's} there may be errors in the "facts". The "problems" are sort of short run facts, and must be used to devise an output. The "facts" do not in themselves demand an output. For a gen. problem, the output will be a funct. of all previous facts, and all data ~~input~~ ~~for~~ that problem. In this style of TM, the machine will not have to figure out what practically everything means in its input. Instead, TM will be given much into about the meaning of the input format. This will not make the induction problem easier, but will make the induction problems concentrated on the kinds of regularities in which I'm interested.

Another feature that I'd like the TM to have, 357101

Aug 9, 61

1.10.0,

Part - 3

01: 353.40! would be that ~~it~~ could ~~tell~~ him how to work

The way I could do this in the present mode is to find a soln. to the problem as simply as possible, using ^{those of} TM's ~~existing~~ abss. that are in existence at that time.

In the early MIM type seq. perhaps no "facts" will be needed - any info needed ^{for a given prob.} will be obtainable from that problem's input, and any info needed from other prob. will be carried by the operator that TM has found adequate up to that time. It may be that we will want all "facts" to be somehow incorporated into ~~the operator~~ this operator.

The trouble is, the op. ^{to be} assoc. with a given fact is not always clear - e.g. for the fact $1+3=4$, shall we make

- 1) Problem, input $1+3=2$ yields output 4
- 2) " " $1+3=4$ " " 3 } ? ←
- 3) " " $1+3=4$ " " +
- 4) " " $1+3=4$ " " =

perhaps there is some way to make all of these possible.

In a TM, those of these - responses deemed by TM, to be most useful, would be incorporated into the operator.

We could put all "facts" in the form of problems with their solns. The disadvantage of this is that then we couldn't give TM a book to read for "facts".

O.k. - well, write a type seq. for MIM. Write Solns. in fairly intuitive English.

Aug 3, 61

Part 1)

Problem:

$R_1, R_2, R_3, R_4, R_5, R_6, R_7$ ← input Register nos

1) $A + B = Z, Z$

This is to mean

1 or more examples are given, with ± random nos for A & B
see EPM 1084.01 ff. for some notation conventions.

Soln: Add R_1 to R_3 . program (place soln. in Ac
Action automatically).

2) $A - B = Z, Z$ Soln: ~~illustrate~~

Look at R_2 . If $C_{R_2} = +$, add R_1 to R_3
" " = - , subtr. R_3 from R_1 .

It may be poss. to simplify this prog. much, if we have
a specific set of computer orders. E.g. for ^{8. IBM} 704:

Add R_1 ~~if~~ if C_{R_2} is + Add R_3
" " " - sub R_3

or Add R_1 if R_2 is +, Add R_3
otherwise sub R_3 .

3) $Z = A + B, Z$ ← Register nos.

Soln (this includes soln. to $Z = A - B, Z$)

Look at R_2 (also soln. starting with Look at R_7 is poss.).

If C_{R_2} is = , $r = 2$ also soln. with "If C_{R_2} is +"
" is not = , $r = 0$.

Add R_{1+r} ; look at R_{2+r} ; If $C_{R_{2+r}} = +$, Add R_{2+r}

4) $Z = A - B, Z$ " " $\neq +$, sub R_{2+r}

01. 355.40:

5) We may want to ~~try~~ give

" 2 3 4 5 6 7 8 9 10 " ... so that TM
 $\underbrace{5, 5, 5}_Z = A, B, Z$
 arb. no. of spaces

have to devise perms. under which are "position invariant"
 Such a concept might be useful later.

6) $A \overline{+} B = Z, Z$ } These can be also worked in the
 $Z = A \overline{+} B; Z$ } manner of 3), with little modification

5 One of my goals in this prelim. top seq. is to get out
 good abs (or hours) in TM so that he can perform some
 fairly interesting induction with a random search that is not
 beyond the capacity of ~~existing~~ machines.

Also, to check on my coding proc. of ind. ind. to see
 20 that I've not missed some imp. ~~pts~~ ^{pts}.

2. A list of a few types of probs. that I might want on the
~~next~~ top. seq.:

- 1) $A(B+C) = Z$ 2) $(A \div (B+C)) + D = Z$ 3) various random
 combs. of +, -, *, x with paranth. , =, >, <, >=, <= on either side of =.
 - 4) $Z + A = B$ 5) $ZZ = A$ 6) $Z + ZZ = A$
 - 7) $\overline{+} A Z = V + A, V = 3$ 8) $Z = V + A; V = ZZ + B$
- (Solve 8 by both substitution and subtraction).

We want to continue the top. seq. and solve by hand,
 until we get to a pt. where a significantly diff. prob.
 can be solved by random search on an existing computer.

I think that it might be well to try to solve each new
 problem by recognition operators, that tell how far..

Aug 21
Dart 7

4M

01: 356.40: new prob. differs from the old ones, and
the new operation on the new type of prob. is
in the spirit of plans 67.01-68.40 — In particular of Plans 6740

2:38 AM Aug 21
The problem is of now: to put a top seq. and a soln
the top seq. \Rightarrow one can solve interesting problems, using M.C. methods
on existing machines using the statistic, of this "soln."

The approach I have been using, thus far, is to just try writing a
top seq. of arith. that ^{would} seem to lead to "understanding" of
some fairly complex ideas — i.e. the creation of some very
useful abss.

20 Perhaps a better approach: 1. Find some rather diff't.
probs. which, for a human, would require some cleverness.
Make up a series of abss. \Rightarrow a human who has found these
abss. useful in the past, would tend to use them — successfully —
in solving the diff't probs. of interest. Then work back
to find ^a top seq. that could develop these abss.

in a TM. \rightarrow
Another point of imp't: The idea of 356.39 - 357.03: i.e. the
construction of "obs" ops. to recognize a particular type of prob., and then
devise a method of soln.

\rightarrow Newell-Simon have a set of abss. and heuristics that can
solve many kinds of problems. What I could do, would be
to study their heuristics a bit, then devise a top seq. that
~~will~~ will lead to (at least) those heuristics.

My impression of how a simple man-like MTM might
work: For the inputs, there would be a set of "obs" ops
to characterize the input problem. After characterizing
is decided upon, the input is x'f'd into the output by 358.01

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Dart

-1 M.J.

Tyler G

01:357.40. The suitable op. When a new input occurs for total method up to that time doesn't work, TM to
① to find an obs to characterize this input type
② To find an op. for that input type that gives the rate of

Some of the work of TM consists of simplifying the derivs. of the obs and ops. Certain obs may be merged by $\beta \Sigma$, if one can find a common way to assign them a common op.

New trial ops and obs are formed by combining old obs and ops in standard ways. New obs and ops are trials which, in view of the previous ops and obs. found useful, are of lowest cost - so they can be constructed ~~the~~ Carlo-wise.

20 Suppose we have obtained this set of obs and ops with correspondences betw. them, and we can view the entire system as a simple op that has been created by some stoch. lang.

or more desirably - just each (obs, op) pairs have been created by a single stoch. lang. We then use this same stoch. lang. to create new trial pairs - (the we can create the obs's and ops "separately").

35 Now - ~~we~~ as new important parts of the type, e.g. are learned, ~~we~~ and, in some sense, the type of stoch. lang. is kept constant in ~~the~~ - i.e. the method of going from the corpus (i.e. the system of obs and ops) to the stoch. lang. and its params. remains const. >, is there any tendency for the search time to ~~be~~ through "suitable" obs and ops being selected? ~~probably~~ - see 365.35 We could try to improve the search method and leave the basic obs and ops "invariant".

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-1.M.

Darb

01:35-40: They would effectively change anyway, or we could perhaps
 effects of faster search, by leaving the search method
 invariant and using obs's and ops that were better
 in this respect. It is this latter that we are most interested
 in, and we want to know whether and to what extent
 is possible.

It would be well, e.g., if we took a ~~static~~ system
 of obs and ops, along with the search method and stock log
 and devised an improvement in the search method. Then
 we showed how the effects of this improved search
 method could have been obtained by devising new obs and
 ops to ~~achieve~~ the ~~same~~ effects of the old set.

But this sort of thing is possible is supported by organic evolution,
 with its apparently fairly simple method, of making modifications of
 chromosome strips to make new trial organisms.

Let us at first confine our interests to a T.M.,
 i.e. a classical MTM.

30 What we want, now, is an example: i.e. a task, a set of
 obs, ops, a stock log and a "search method" -
 then an improvement in the search method, then a possible
 set of modifications of the obs and ops, that would be
 about equivalent to this improvement of the search method.

Note: It will probably become clear that certain types of
 improvements of the search method will be inaccessible through changes of
 the obs and ops. e.g. various sampling techniques for dice-throwing
 trials quickly. The part of the search process that

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T.M.D.

3 vt

0.1 359.40 will be accessible, will be the improvement of the first

0.3 header to the beginning of the search. My impression

0.5 all this means is that the prob. values assigned by the system become better and better as time goes on

The ideal prob. eval. meth. would assign prob. close to 1 to one possy and close to 0 to other possys. In the case of Mt. Cor. trays, this would mean usually the correct answ.

1.0 on first try. $\rightarrow 361.02 \rightarrow$

1.1 Say Ob_1, Ob_2, \dots, Ob_n are the set of obs used to characterize the input to decide what op. shall be used on it.

Let Ob_0 be the first obs. that was used to choose (i.e. the identity op. that always says "yes") be the first obs used in the top seq.

Let Ob_1 be the first obs used when Ob_0 alone became inadequate \therefore let Ob_n be the obs. used when $Ob_0, Ob_1, \dots, Ob_{n-1}$

became inadequate for the top seq. \implies then the way we

use these obs. is this: when an input comes in, we first

apply Ob_n . If it is acceptable, then O.K. ^{use Op_n} : If not,

try Ob_{n-1} next, if acceptable use Op_{n-1} , if not,

try Ob_{n-2} , etc. -- finally, all inputs are acceptable

to Ob_0 . When we get a new input, for which our

system gives the wrong answ., we try to create a new

Ob_{n+1} of min. cost, that will characterize this new type of untractable input. I think that Ob_{n+1} must be different

from Ob_0, Ob_1, \dots, Ob_n , or else it could not possibly be adequate.

I don't think this ^{input} system is optimum \neq but leaves 361 of

Dort

01:360.40 : it for th. while

02 : 360.10 → What I'm vaguely thinking of, is some sort of canonical form of th. universal "creative" stochastic process of which th. probs. all approach "correct" values as \mathcal{R}_n sample size

$\rightarrow 362.07$

A sort of heuristic, to obtain closer to minimal coding;

After a typ. seq. (for MFM) has been "solved", ~~we~~ go thru \mathcal{R}_n typ. seq. again, but instead of starting with a prior for \mathcal{R}_n early problems, and slowly changing \mathcal{R}_n search statistics ~~remains~~ in going thru \mathcal{R}_n typ. seq. , use \mathcal{R}_n same constant search statistics for \mathcal{R}_n entire typ. seq. , that were obtained after doing \mathcal{R}_n typ. seq. \mathcal{R}_n first time. After doing \mathcal{R}_n typ. seq. \mathcal{R}_n 2nd time, \mathcal{R}_n resultant statistics can be used again to do \mathcal{R}_n typ. seq. search. I think that \mathcal{R}_n code costs should \downarrow as one codes and recodes again and again in this way. I feel that if \mathcal{R}_n typ. seq. is long, that \mathcal{R}_n statistics at the end of \mathcal{R}_n "first ^{apri} learning" of it, will be far better than \mathcal{R}_n statistics --

There may be some reason to believe that in a search of \mathcal{R}_n type, that \mathcal{R}_n coding method should give some compression. Certainly we expect \mathcal{R}_n 2nd coding to be better at th. beginning of \mathcal{R}_n typ. seq. . Th. Q's, whether th. all ~~other~~ statistics will change significantly in \mathcal{R}_n 2nd or 3rd coding. If we have a T.M.J. with a search method and typ. seq. , then

Part

01:36:40: This idea should be easy to test empirically.

The seq. is moderately long, and mainly "hard data".
(The idea it may work O.K. in this latter case, early part of the two seq. should go rather well by using the known statistics).

07:36:04 → : A ~~part of~~ "example" of a ^{very} simple type of prediction method that is poor for small samples, but may be one very good for large samples, is the gen. method. - i.e. the probty of various digits following a gen. num. As samp. sz. ↑ we can ↑ n. I suspect that this type of probty does not include all possible probty eval. methods, but it may be fairly good for larger sample sz.

20 My impression is that the ^{most} general form of a ^{creative} stochastic grammar would be something like we have at Turing. ~~unrestricted~~ As input, we have AR. A is a fixed string that describes the grammar, R is an arb. random no. (of fixed length) - ^{?!?} ~~the~~ R. ~~is now~~ has uniform density from 0.001, except that somehow R is encouraged to be a terminating decimal. ^{?!?} Then the output of the Turing machine will be a suitable probabilistic set of strings for the stochastic grammar in question. ~~Answer~~ ^{non-creative} defn. of a stochastic grammar is that it assigns a probty to every conceivable string.

Actually, I can get a better feel for the problem in the following way: consider, e.g. the Obi's of 360.11ff. This particular approach really breaks the system down into parts. It may not be the best way to make such a system, but it's close to it.

4/22/2
Part 7

01:362.40 way man does it \rightarrow perhaps not - see .20, but
Anyway, after along / long s.e.
be poss. to get ~~any~~ power conceivable creative stock.
for producing Obj's. Obj's is a ~~simple~~ function
inputs that maps them into "yes" ($\equiv 1$) and "no" ($\equiv 0$)

07 ~~say~~ Say the input is a n digit binary no. Then there
are $2^{(2^n)}$ poss. functs. on this input. The most general poss.
stoch. lang. for creative Pres. functs is a ^{prob} density funct. on $2^{(2^n)}$ points.
It is clear that with a large, ~~enit~~ ^{simple} sampling we can get the ^{2²ⁿ} patterns
of this stoch lang. by ^{simple} freq. counting.

Since 2^{2^n} is usually so large, this is the ridic ad absurdum
of large sample needed - and corresp to the simplest type
of stoch. lang. conceivable. \rightarrow 360.38

19
20 **SN** I sort of doubt that the Obj. system of 360.1) ff. is
at all close to what man does. Man usually makes a standard
Obj. \leftarrow On the basis of the result, he will either make a new
obj. ~~or~~ do an op - both of which would result
depend on the result of the first obj. \rightarrow ~~the~~ ~~man~~ ~~is~~.
On the ~~case~~ the next obj. or ops. ~~are~~ work on, and are
contingent on the results of these 2 ~~id~~ ~~any~~ obj. or ops. and
so on

To fix ideas - consider some of the ops and
ops. used in this ^{arith.} ~~fig.~~ ~~of~~ 358.01 ff. \rightarrow 366.35

37 The probs. as of now, the general MTM system of 360.1) ff. with
the Obj's seems a bit distant from the human ~~method~~ ~~method~~. There
is some Q. as to whether it could be ~~wide~~ ~~to~~ work at all,
and, of course, the more it deviates from human, the more ~~357.01~~

1 2 10
Dart

IMJ



01: 363.40: diff. it is to derive heurs for it ← see 366.35 for what may be
02: Th. very imp. prob. of 359.30 = 360.10 ff sl finding
see it ^{effects of} heurs, ^{being} automatically improved in: Mt. Carl. MTM. Th. idea

04: 359.30 - 360.10 is that the search process devises a creative sto
gram. for th. things searched for, and, with some stock gram. types as so
th. stock gram. becomes very good.

Th. problem, then, was sort of switched to th. ~~is~~ set of Obj;
these are a set of ~~obj~~ functs. Real map th. input into (0, 1)
(= Yes, No). Th. these Obj's may not be a good ideas for MTM,
they are a good study problem for this → Q. 363.07 works a bit with
this idea.

Anyway, say we get a ^{very} large set of strings that are the output
of some (fairly general) ^{creative} stock Gram. How well can we make a
20 model of this stock Gram? 363.07 - 19 gives at least one way, if
th. strings are all of bounded length. - i.e. we just count th. freq.
of occurrence of ~~all poss.~~ strings.

Now, it would seem that this method could be applied to
stock gramms. in which only ^(no fixed) long strings could be produced - the good statistics
on some of the longer ~~str~~ and presumably low probty strings, would, of necessity

25 take over to obtain. 2) Any ^{stoch} sophistication in grammar approximation should
give a better approximate stock gram. sooner (i.e. smaller sample).
There are probably restriction on "Any" - so, I will have to find out just what
gram approx. types are poss.

35: An example of an approx. ^{stoch} gram. that works well for small samples,
and poorly for large. Each symbol is assumed to have a basic ^{probty} and
~~probty~~ (th. sentence termination being such a symbol) and th. probty of
any s. is ^{approx.} product of th. basic probty of its component symbols.
called the naive ^{stoch} Grammar. = NSG.

38: A somewhat better approx. stock gram: One in which sup strings
or sub-string can be given a defin. and have a probty. We may or may not
want to include constraints on what symbols can legally follow which.
Anyway, I suspect that such a gram. starts out better than
365.01

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part

01: 364.40. : NSG (500 364.35 for data. of NSG), and
is at least as good as NSG, since it can do
sanitation and count them if they occur often

I suspect that most ^{such} programs that can at least define
are at least as "good" as NSG in R. sense (in exte

10 - [SN] In addition to being able to define substrings, I think I'd want
to be able to define objects like (a, b) which could combine
with objects like c or (c, d) in various ways, with various
probs - e.g. $(a, b) \times_1 c \equiv a c b$; $(a, b) \times_2 c \equiv (a, b c)$
 $(a, b) \times_3 (c, d) \equiv a c, b d$; $x_4(a, b) \equiv a b$. etc.

In 364.38 - 365.17 we have R. class of some poss. stack programs,
but what is essential in this discn: is the basic (I) R. form of R.
program and (II) the method by which R. sample is processed to
obtain R. params. of R. ^{approx.} program.

We can compare any ^{stack} program with NSG by comparing R. relative
coding compressions achieved by each, for corpi of various sizes.
NSG can be used to "code" by ~~using~~ using a special symbol for every
S that occurs $\geq k$ times. The symbols of S's that have ^{only} occurred once, are coded as themselves and result in a complete Laplace
note for them.

The above discn. sugg. that R. Q of 358.35-38 has an answer of "Yes"
under certain condns. i.e. if the problem of constructing obs. and ops is solved
in a certain way and if this answer is "Yes" only to R. extent
that the answer to the "impression" of 360.031-05 is correct.

This discn. may, conceivably, not apply to R. obj. system of 360.11ff,
since in that system (very probably) it is necessary to describe an entirely
different obj. each time. So R. previous argument about
R. NSG type programs is not "adequate" for

01; 365.40 : large enuf sample. is not valid.

However, the ~~situation~~ situation to which the discn. is a possibly imp. one, and I think the study ~~is~~ is of much import.

SN

In the ~~search~~ search methods being considered, we look for an op. that will operate on the 2 gn. input to produce a gn output. In the simple methods considered, the search does not in any way make use of the knowledge of what the input and output are. Ops are just tried on the basis of a statistics study of what ops have been successful in the past. A Monte Carlo search under such condns. will tend to give a good soln., but will be slow. "Heurs" that do look at the input and required output, ~~can be faster~~ ^{much} give faster success, but will tend to give not such goodly extrapolative solns - i.e. they will tend to be a bit ad-hock. The degree of ad-hockness will depend much upon - just what the heur. is.

I would like to be able to do without ~~se~~ hours of this type. Whether or not this will be poss. may be a purely empirical

Q, and vary considerably with the type of prog.

I am working on a TM_0 - so heurs. of this type are not poss. for a TM_1 , heurs. of "this type" are not poss. in any obvious way (the clearly, a human ($\approx TM_1$) can be shown completed examples and can learn from them.)

35 : 363 37 : while humans don't use 360.11ff for their "action op", they ^{often} do use it at first, to solve new problems. After 360.11ff has obtained a soln., then ^{new} the obs and ops are further processed so as to integrate them with the previous observed system so as to reduce the total cost further. This 367.01

of
ort -

01:366.40 "process of integration" may include the same function as
the same op, etc

Well: To get back to R. prob ② of 369.02: Say we
have a TMs running, and $[O_1, O_2, O_3, \dots, O_n]$ is the temporal
seq. of <action ops> of TMs. (O_{i+1} will usually include mo-
of O_i , and will have a few additions & /o changes). We then
want to get a stock ^{program} to produce O_{n+1} probabilistically.
Now - what sort of stock ^{programs} and "fitting" methods
will have the remarks of 369.04 366.04 apply?

First of all, O_1, O_2, \dots are not a set of ^{indep.} 's' in
the ordinary sense - they are ordered and the ordering
is significant. [This problem was discussed at some length in
Part B 115 ff, but I don't think I knew enough about it
at that time]

For simplicity, assume that the O_i 's are of the obj. of
type ~~obj.~~ outlined in 369.11ff. Then try to get some idea
of what these obs. and ops would look like. Try
to get an example of a heuristic of the kind that
would give better prob. values for the O_{n+1} distribu., and
see if we could expect to get R. effects of that kind
using fairly simple obs, ops and "stock program fitter".

Consider $O_1, O_2, O_3, \dots, O_n$. We have the constraint that
 O_i cannot be identical to O_j if $j \neq i$. - otherwise, just
consider O_{k+1} to be the next object to be predicted
in R. seq. O_1, O_2, \dots, O_n . The non-repetition feature
means that we can never treat R. O_i 's as objects of
which we can make freq. counts.

A simple ^{stock} ~~seq.~~ for approxn. would be one in which

Darts

01:36:40: Th. Obi's wave strings of ^{computer} orders, define upon both compact and articulated upon.

Now - consider any expressible "regularity" in Th. seq. Obi will it (with large enuf samp.) be always expressible by defining a more regu, and their intersymbol constraints, and using symbo. freq. counting? I think that all that the "adequate samp" means is that the regu (ie. Th. defn. of Th. "regularity") may be quite large (ie. by regu).

It may be that ~~the~~ regularities having to do with order in which Th. Obi's occur, cannot be expressed by freq. statements about regu freqs. and inter symbol constraints. - So at first consider regularities in an ^{unordered} set of ~~strings~~ of computer orders' s.

Some strange regularities: say Th. odd numbered (or p-thenumbered) digits of Th. strings tend to be "A". How can we express this fact as an regu? Well, for the "odd" case: for the articulated regu, (A, A, A, A, A, \dots) then combine this with Th.

ntps: $\left(\begin{pmatrix} A \\ B \\ C \end{pmatrix}, \begin{pmatrix} A \\ B \\ C \end{pmatrix}, \begin{pmatrix} A \\ B \\ C \end{pmatrix}, \begin{pmatrix} A \\ B \\ C \end{pmatrix} \right)$ to form Th. regu: $\begin{pmatrix} A \\ B \\ C \end{pmatrix} \begin{pmatrix} A \\ B \\ C \end{pmatrix} \begin{pmatrix} A \\ B \\ C \end{pmatrix} \begin{pmatrix} A \\ B \\ C \end{pmatrix} A$

Here, Th. alphabet of basic symbols is A, B, C.

↑ Here Th. entire method isn't entirely clear, but, anyway...

Another way is to allow freq. counts for each digit position of Th. Obi's.

A fairly complex regularity: If we let A be 1, B be 2 and C be 3 then if we take the product of Th. digits and their sum, these 2 nos will have a difference divisible by sum of 3 modulo 7.

ie. say D_j is integer assoc with Th. digit; then $\sum D_j - \prod D_j \equiv 3 \pmod{7}$.

36:36:19

A fairly general kind of "regularity": say S_i is a "randomly chosen" string in a unordered universe of strings:

art -

or: 368.70. Then If D is a fixed string that decodes R $\rightarrow M(D \rightarrow S_i) =$ with probty P where M is so

This method of decoding a "regularity" is more like a ^{high} ~~low~~
 This statement is only a good meaning if we make some sort
 on the ~~set~~ (usually infinite) set of S_i 's.

A more creative method: say R is a "random no" \rightarrow with a suitable density distrib. (e.g. the probty that it has just n digits is $\propto e^{-kn}$, where k is some const.). Then say D is the

"Decn" of a stock program (see 362.20 for an idea). Then $M(D \rightarrow R)$ is the output of the creative stock program.

(M is a Univ. Turing mach. funct.). ~~is~~ what we do, is ~~select~~ ^{known} find some $D \rightarrow M(D \rightarrow R)$ creates ~~the~~ ^{known} members of R .
 set $\{S_i\}$ as often as poss. We might be able to do this by taking various trial D 's and "reinforcing" them when a random R gets one of the S_i 's.

-- Or, we can make a statistical study of D 's and use a stock program with your data to create characterize the ^{ones with} successful trials

So, we create D 's and using random R 's, we form $M(D \rightarrow R)$ in ~~the~~ attempts to decode R . set $\{S_i\}$. What we want, is a D that can be best used to decode R . set $\{S_i\}$. We are using a Mt. Carlo method to find out (approxty), how well each D decodes $\{S_i\}$:- we must also include in "the goodness of decn", the cost of D .

Now, suppose that instead of using ~~a~~ approx. stock langs. containing n.g.m. data for the D 's, we just used ~~the~~ R . for 370.0,

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Dart

01:369:40: Simpler (the "degenerate") stock lang. that makes freq. successful D trials. (i.e. D's NS of 367.35) How work for very large samples?

Well, just try all ^{poss.} D's in order of complexity, starting with D = 0, 1, 01, 10, 11, 00, 000, etc. Test each D Mt. Carl. - wise by inscribing "Random" R's into $M(D^R)$. Actually, we need more than R. lang. with which produces members of $\{S_i\}$ we need to know ~~how~~ ^{approx.} how freqly D produces each member of $\{S_i\}$, and also know R. pcost of ~~each~~ each of R. S_i 's when produced directly w.o. R. help of R. & D. Using these concepts, we could evaluate th. D's, and eventually we would hit on ~~an~~ some very "good" D's - (since we go thru them exhaustively).

We would like to be able to immediately say that use of a stock program for D's, involving lots of yams, would certainly improve the search rate. However, in R. above \mathbb{P} note that we did not assign probys to R. D's by means of R. analysis - so th. D's did not form a stock ~~lang.~~ lang. in any simple sense. - So if we allowed defs. in our stock D program, we wouldn't know how to assign "probys" to them.

But, even in this \mathbb{P} , we do associate a probly with each D. This probly is R. pcost of derbg. th. set $\{S_i\}$, using that D. ~~So suppose we were to try to construct~~ so ~~we~~ we assoc. with each D, R. pcost $P(D)$.

Now - we would like to can try to devise a creative stock. program that will create D, with probly $P(D)$.

We can use ~~the~~ $\{S_i\}$ R. \mathbb{P} program. hvr.

Thru" \mathbb{P} Mt. Carl. method, we can obtain for each D that

Dart

370.40 We've tried, a $P(D)$ and its variance —
 to convert this variance into an equal equiv. samp.
 for the associated D , and then, we pretend we have
 large ^{corpus} ~~amount~~ of D 's ~~with~~ with ~~repetitious~~ repetitions
 of each D as in no. to the assoc. samp. size. Use
 this corpus, we simply devise a good set of ngm defs. and
 obtain their ^{intersymbol} constraints and freqs. — so we have an approx.
 gram with defs. for the corpus

we then use this approx. gram. to create new trial D 's.
 This gives us a greater sample density near the best hys.
 of the "hill" we're climbing.

If I ever use R. above Mt. Carlo scheme in anything, it
 would be well to do R. analysis in much detail, to give
 quantitative estimates to R. various params. involved Then
 devise a reinforcement scheme for ngm problys to approxi-
 the ideal method.

A brief review of R. idea: I have a bunch of unordered objects,
 $[S_i]$, and I want to make another one that is likely to be
 accepted as one of them. I do this by devising ~~a~~ ^{random no.} D
 D , of stock lumps. $\Rightarrow \Rightarrow M(D \rightarrow R)$ yields S 's in R.
 stock lumps. Each stock lump is a partial defn. of R.
 set $[S_i]$, and so each D can be assigned a $P(D)$
 as the prob. of R. entire defn. of $[S_i]$ using that D .
 We then devise a creative stock gram using defs of ngrams
 to create trial D 's. We use our best D 's to
 devise our next trials for members of $[S_i]$.

Wavy direct prob. pattern of objs :

The above idea sounds very good, because with large and
 sampl. it ^{i.e. in extrapolation,} will become very good. Also, any new heuristic

371.90: devices that I want to add to improve R. upon. defs. > will make it even better - so I plan just "adding-in" any improvement I can think of.

There are a few "loose ends" (1) The business of coming from something close to "G" assignment for D's, to probably assign it on a good theoretical footing (2) my methods for constructing stock programs with ngn (and ntp) defs are not too good (3) My methods of getting predictions or mt carb output from such programs are even poorer. (4) R. idea of inserting a random no. and D into a Turmac will have to be gone over and perhaps put into more practical form.

(5) I don't yet have any ideas of what some good D's might look like

Synthesis : what we've been doing, is using a stock program. (of a ngn defn type) to create / stock programs of most genl type. These secondary programs were then used to create Mt carb-wise final trial elements. Success or failure was used to reinforce the primary ngn defn. program.

Now this cascade of 2 stages of stock program can be viewed as a particular method of using a reinforced stock program. I.e. R.3 cascading of 2 stock programs always yields a stock program. So, we have a ^{single} stock program that we can reinforce directly. This program is in some sense, "universal" and has the capability of becoming very good with large sampl. of Tho. seq.

set
6 6 0
Data

01: 37.2.40: As I ~~was~~ have been using it the
was designed to extrapolate $\{s_i\}$, ~~the~~
set of ~~obj.~~ strings. Perhaps I can extend
system so that it can produce objects that w
be accepted as members of ~~the~~ set $\{s_i\}$
not not of ~~the~~ set $\{V_j\}$ (samples of $\{s_i\}$ &
 $\{V_j\}$ are both variables). The idea here, is
to use the system for hill climbing, by ~~putting~~ assignments
to $\{s_i\}$ / ~~trial~~ ^{known} pts. with G 's \neq above threshold to $\{V_j\}$
if below threshold. Th. threshold can be (R. median
 G obtained up to now) or th G of R. pt. of 10th say
next ~~to~~ G this ~~func.~~ (sa $\{s_i\}$ always has \approx ^{th. top} 10 known
members)

For a fairly emp. H.C. prob. (ie. T.M.), ~~the~~
~~many~~ have the. stack from with new defs. create
(strings of symbols) that are ~~trials~~ trials. For these ^{trial} ops,
one was external input, - then the operator produces
an output and gets its G awarded externally
for that (I, O) pair. For each operator tried, we get
1 or more (I, O) pairs. From these and the post of R. Op.,
we obtain a mean and variance of R. G of that operator.
The mean G ~~is~~ ^{is} some ~~func.~~ of R. G and its variance -
with ~~by~~ rewards for ^{by variance} ~~with long run~~ since this gives a
possy of a truly ^{by mean G} ~~mean~~ G . At each Op. tried, is
computed, and this ~~is~~ ^{is} used to assign that op. to

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Dart



01.373.40: Th. $\{S_i\}$ class set or R $\{V_i\}$ set
There can be 2 stock ^{non defn.} parameters. 1 gives Th. probby
2n op. generated by that param. will be in $\{S_i\}$; Th. stock
of for $\{V_i\}$.

Some genl. simplifications and reconceptualizing of A. Laplace's work.

1) Consider Th. initial probby distribu. on Th. set of random strings, R , as a stock lay (This is certainly true)

Any deriv, D , can define a function, mapping strings into strings, and its mapping Th. strings R into new strings, $M(D \cap R)$ and so Th. stock probby distribu. on R 's maps into a new probby distribu. on R 's, and Th. stock lay. of R , becomes a new stock lay.

This is a general conclusion that any function (e.g. D) on a set (e.g. R), maps any density function on that set (e.g. Th. stock lay. on R) into a new density funct. (i.e. Th. new stock lay.).

30 2) A stock lay. can be used as a deriv. of any certain sets of strings (finite or infinite), if each member of that set is assigned \neq probby by Th. stock lay. Th. extra probst of deriv. Th. set $\{S_i\}$ by means of Th. stock lay. $P(S_i)$ is $\sum_{S_i \text{ or } \text{all membs. of}} P(S_i)$

This was recently analysed in a "recent" (post WJCC) section of "Code" \equiv [Cod 288.01-292.22] \equiv gives idea as to what it is
So Th. total probst of the deriv. of $\{S_i\}$ that employs Th. $D \equiv$ Th. deriv. of Th. stock lay. $M(D \cap R)$ is:
(probst of D) (total probst of $\{S_i\}$ by stock lay. D) (punctation)

001:374 40.

3) As the result of (2), it is clear that D 's is not, directly, a decim. of $\{S_i\}$, it is clear that for D 's, one can, uniquely, derive a decim. of $\{S_i\}$ as the cost of that decim. Furthermore, many of the processes involved can be approximated by M.C. because most D 's do give decims of $\{S_i\}$, one can D 's at random and use the resultant costs of $\{S_i\}$ decim as "G" values to hill climb on, to find a "best" D . Also, the D 's found (in this way), are very good for extrapolation of $\{S_i\}$.

Contrast this random search for D 's (where most D 's fit at least a little bit) with the choice of random decims of $\{S_i\}$, in which hardly any decims fit at all ("at all" means perfectly) The search time is lowered considerably, but, if one uses not random but ordered search (in strict order of cost) for direct decim., the direct decims would usually be better when found.

4) The method used above for search, is for unordered extrapolation of unordered sets, $\{S_i\}$. How can this be generalized to other types of things?

This method of coding a set of objects by first defining a stack lang (e.g. 374:30) is a priority, of code a coding code of code 188:61-292:22. In that case we had an alphabet on the set of decims of all objects $P(O_i)$. We want to decim. the subset $\{Q_i\}$ of objects. In a simple way, its cost would be $\sum P(O_i)$. Here, suppose we have a way of decim. $\{Q_i \in [Q_i]\}$. Then we can view this subset as defining a new stack lang (\equiv probty distribn)

Part

01.375.40 over the set of 2 objects. This. now distribu

is D . h l r v o f o w n r o p o

$\{ \in P(O_i) \} = \{ \in P(O_i) \}$ since $P(O_i)$
 $\{ O_i \in \{Q_i\} \} \quad \{ O_i \in \{Q_i\} \}$

is r u n r e n a l i s c o n s t r a i n t w i t h r e s p o t o $P(O_i)$.

SN) If we have 2 stock logs. $P_1(O_i)$ and $P_2(O_i)$

that be being used in attempts to desc. ~~the~~ a certain set, then P_1 (renormalized) \otimes stock logs $P_1(O_i) P_2(O_i)$

is also a good try as would be (norm) $P_1(O_i) + P_2(O_i)$.

if r u n s o f t h e s e l o g s a r e o f c o u r s e u z i n g t w i c e P_1 o r P_2

$P_1(O_i)$ or $P_2(O_i)$

h. means proposed for extrapolation P_1 - known set of $\{s_i\}$
I would like to think of an example
to P_1 P_2 \rightarrow a ~~study~~ "study problem"

Well, consider P_1 more complex prob. of 373.01-379.04,
with the "Good" set $\{S_i\}$ and the "Bad" set $\{V_i\}$.

Say the S_i 's and V_i 's are computer pps for "searching" for solns. to the
well defined ~~problems~~ "problems," and P_1 is mean search
speed. Or P_1 may be for "well defined hills" in which
the G is a combination of search time and hill hit.
obtained.

OK. - perhaps - but now lets go back to P_1 problems
of 357.04 and .20; and also the prob. of 359.30-367 (= 359.20ff).
want to consider these and this \rightarrow all together, and see if we can
come up with a good idea for a ~~computer~~ P_1 desired features -
ie. most of interesting nts is tied to TM via Top. seq., and
improve P_1 methods automatically with a large top. seq.

01: 379.40:

coding method, is a h.c. method that to anything - also, perhaps, a coding method that λ in \mathbb{R} . no. of probs. worked λ .

In the original prob. of 35709 and 20, I wanted a top seq. that would lead to \mathbb{R} . ability to work a diff. prob. involving "creativity". ~~the~~ \mathbb{R} ^{and thg. seq.} involved could be of type

TM_0 ($\equiv M.TM$), $TM_{\frac{1}{2}}$ ($\equiv N.MTM$) or TM_1 (H.C. TM).

More exactly, for TM_1 , of known input, output, G triplets. $\{I_i, M_i(I_i), G_i\}$ ^{external awarded G - usually includes resp. time} is a set \mathbb{R} is a set. From this, we want to

create a M_{int} (operator) that with tend to xfm I_{int} into a output of G .

The idea we have of 373.01-378.03 relevant to \mathbb{R} is in form like say D_j is a d.m. of a stock and that can produce M_i 's from "random nos." by $M_i(D_j, R_i) = M_{ij}$. Then we treat the M_i 's of \mathbb{R} as data as pts on a "noisy hit" of ut. w_i at pt. (M_i) . We then want to find D_j 's that produce M_i 's of G . We divide the M_i set into 2 subsets - one of n_1 the size of ow, G . We then try to find a D_j that "fits" best \mathbb{R} . n_2 G set or perhaps just \mathbb{R} top fits of that set.

We would then try to find a stock D_j using n_1 data to produce D_j 's that ^{tend to:} internal produce M_i 's of G . Also, we would make a stock ^{gram, L_2 using n_2 models} D_j to produce D_j 's that produce M_i 's of ow, G . ~~again~~ We then will produce trial

Dart

.01: 380.40 M_i 's that use R_i by freq. usage of L_1 and L_2

.03 An idea sugg. by R_i above: We have R_i & L_i of (I_i, O_i, G_i) 's. With D_j we produce M_k . We test M_k by giving it randomly chosen members of known input set $[I_i]$. We do R_i 's until a few outputs of known G are produced. From R_i (eg. R_i of R_i M_k 's) \bar{G} and σ_G^2 are obtained. Several M_k 's are produced by R_i D_j and evaluated this way. We also try other

D_j 's (produced by a uniform def. group) and use them to produce M_k 's, which we evaluate. At any time, if we get new input and need a response, we use the M_j of highest \bar{G} , and use the resultant G obtained, to modify that \bar{G} and σ_G^2 .

The G in. to a D_j is higher, if R_i M_i 's produce \bar{G} result in a fast search for (I, O) pairs of by G . What we do, is try first R_i I_i of highest with a O_i of highest G . Then we try the lowest O_i , then the next highest, etc. The idea is to terminate the search as rapidly as possible by getting a few very high or very low $D_i O$ pairs for the operator.

It may be necessary to use a more randomized input-search. R_i over L_i above R may result in some serious errors - but check on this.

01: 381.40: There is an obvious refinement
 & very imp. one - i.e. the devising of "loosen"
 M₂ don't have to produce ~~even~~ responses exact
 as known responses. A more detailed analysis of the
 (i.e. its quantification) will suggest other refinements

I drew up a 2 dim. analog of this TM, problem
 that is very good for "fixing ideas" 339.30
 Use this model for further design of TM.

So, the way this thing works, is that at any particular time,
 it has an M₂ available which, to its knowledge up to that time,
 is "best". Meanwhile, between 2/0 during / problems, it would
 searches around for M₁'s of higher \bar{G} & σ^2 ^{better} ~~best~~ _{of} σ^2 . It also searches
 for a better L₁ (380.35). During its searches, it presents an output
 of \bar{G} and σ^2 whenever it finds a "significantly better" M₁ that
 the previous "best". This output tells the operator (human / client)
 the next answer to be. The client can tell TM to "meditate"
 until it gets a M₁ of at least a certain goodness - in a work
 the next problem.

In addition to this precy. refinement, it would also be well to
 have TM able to discover the relationship betw.
 (E, O) and \bar{G} , and to be able to conceive and suggest
 new M₁'s of prester expected \bar{G} . than ever before.
 These 2 refinements ^{if realized adequately} could make the resultant - TM very
 powerful!

Altho the idea of 381.03 - 382.37 (including the 2 "refinements")
 might be O.K., it doesn't seem too close to what I feel
 is my own intuitive method of working probs. This diffy
 383.01

Dart -

01: 382.40 may be regarded as a penalty of want; automatically better themselves as more and more worked. Actually, P's TM of 381.03 - 382.37, can be so bad - even w.o. P's 2 "refinements" - Essential TM directs itself to a ~~van~~ problem that is not really P's who problem - so P's 2 "refinements" have to be tacked on ad-hock, rather than appear naturally by themselves.

I think I could work out a TM working on P's same genl. principals that was directed at a problem closer to what I really want, but P's present system ^{would be probly V.G.} ~~is~~ for a "study problem"

Note that this system, while ostensibly designed as a TM, could be used as a T.M. or T.M.₂ in P's follg. way: Say P's G for a trial M_i was P's fraction of problems in P's entire typ. seq. up to that pt. that ~~were~~ it worked correctly. We could reduce P's search time by a great deal by inserting special devices that took advantage of P's fact that we have an MIM typ. seq. This means that P's "closeness criterion" for correct answers is ~~not~~ trivially a valued. Also, P's maximum conceivable G value is 1 ($\equiv 100\%$ correctness). We can use P's ^{sequencialness of P's} typ. seq. in a useful way.

Actually, P's system, with P's 2 ad-hock refinements, really isn't very distant from P's ideal, non-cl. soln. (Use P's 2 dim. model of 339.30 for P's thinking about this). We simply ~~use~~ use P's system, then, separately, we try to devise ① a closeness criterion and ② a functional relation betw (E, Q) and G. Actually, a soln. of ②

4.2.14

01:383.40 makes (1) superfluous, but a partial set is partial or complete soln. of (2)

(2) means that for any M_i , we can get a (I, O) pair we try. The "closeness" criterion way to approx. into the function $G(I, O)$ - i.e. we have a bunch of $G(I, O)$ pts that are known and any new (I_j, O_j) pt. that is not identical to the known ones has its G evaluated by (a) how close the new pt. is to the old known pts. (b) the mean and variance of the G of new pts

(1) The moral is. The system of 38.03-382.37 is useful both as a study problem since it can be used for TM_1 and TM_2

(2) as an approximation to approximate TM_1 ,

(3) as an approximation TM_1 of less character when the 2 refinements of 382.01 and 382.30 are considered. While it is a very serious approach to a very non- TM_1 .

There are several imp. problems:

1) With L_1 and L_2 (see 380.35 for defs) one has the problem of creating S 's that have a max ratio of proby to being in L_1 rather than L_2 . This could define L_3 . i.e. if $P_1(S)$ is the proby assigned to S by L_1 , and v_j by for $P_2(S)$ and $P_3(S)$, then $P_3(S) = k \frac{P_1(S)}{P_2(S)}$, where k_j is a normalizing const: (if $\sum P_3(S)$ converges). The Q. then is - how to make a constructive program for an approx. to L_3 .

Ideally, it would be well if we could reinforce L_3 's program positively or negatively directly whenever it produced a by or low G M_i , resp.

We had some objections to this system - i.e. we wanted 385.01

Part =

01:384.40: something like ~~L3~~ L3 to look at P1

M_i → sub G's — and from all this into a

next trial M_i ~~thing~~ It was somehow that

present system doesn't do P₃. But, it does. The

P₃ past data and trials is summarized to some extent

present form of L₃. The better L₃ is, the better is P₃ summary.

Actually, L₃ doesn't exactly summarize this data. This data is, ⁱⁿ available directly to L₃ (since it is stored). L₃ is an attempt at

using this data to obtain good "next trials" for M_i

—

Let's try to get more specific about — the system:

Say we start with an ~~TM~~ TM₀, and our typ. seq. has 100

examples — each essentially different, so that a new M_i ~~can~~ almost always has to be created to cope with P₃ "next problem". Take

the last 10 successful M_i's (these 100 M_i's have been

constructed "by hand" by me.) Pass a stock program P₃

through them — P₃ is P₃ program of L₃ ~~with~~ ~~the~~ ~~same~~

It is rather imp. that L₃ be able to be the most general kind of stock program possl. About P₃ most complex stock program

that I'm familiar with at all, is P₃ ^{stock} "context dep. PSG".

The ^{stock} non-dm PSG might be better.

Probably the best thing to do would be to look at the set of 10 "good" M_i's, — then find some kinds of stock programs

that could have created them — then try to generalize these types of programs to a more "universal" form.

Of course, one form of univ. ^{stock} program is the one that maps the set of "all random nos." (≡ a density distribn. on the set of all strings)

Part

01:38.54.00 to 01. set of all strings, by use of

maps R into R , new strings

Consider L_3 and R . created M_i 's as constituting a stack operator for responding to R . I_i 's. We would like to be able to consider such a simple stack op. and be able to react directly. Hvr. this does not appear to be the most efficient way to work. When we get an input I_i , we would like to select the "best M_j thus far" to ~~react~~ implement the response. If we just use this simple reactd. op., then L_3 would create a new M_i ~~from~~ which was untested and this new M_i would be used to obtain the response. L_3 then could, legitly be reactd. on R . basis of this response. While L_3 would improve as R . try. seq. continued, the mean G of R . responses, ~~that~~ it ~~is~~ give rise to, would not be v.g. It would, e.g. not need a "closeness criterion" or an internal model of G as a function of (I_i, O) . Essentially what this model does, is assume that R . (I, O) trials are very inexpensive, so it would rather make trials than figure out approximately how good each M_i is before using it to create a response.

Actually, I should be able to take the role of a L_3 ! By looking at the last 10 successful M_i 's, I should be able to see just what kind of program could have created them, and without much extra trouble (if any trouble at all), could create successful trials for the next problem in the try. seq.

After constructing 1 or 2 L_3 's, I should be able to devise a program to construct a whole stack of them.

So what apparently has to be done now: Construct a try. seq. and a set of M_i 's to solve it. Then devise some L_3 's to

386 40. Create M_2 .

Before doing this, think a bit more about R_1 .

easy what I want.

The system certainly does appear to have many desirable features. e.g.

- 1) One can use it for TM_0 , $TM_{\frac{1}{2}}$ or TM_1 .
- 2) One can add the 2 refinements (of "closeness" w/o G as a factor) and get a fairly non-rat. TM . (See 339.30 for a 2 dim. plot of the non-rat system.)
- 3) If L_3 is a "universal" lang then search speed will ↑ as the typ. seq. continues.
- 4) The system can give quick responses by using "the best M_2 to date".
- 5) since it can be a IM , (recently $TM \equiv RTM$)

it can be taught things as an animal is taught, and can be made to work any kinds of probs - including understanding human speech and being asked to improve itself.

Some probs. that must be solved.

- 1) Get a simple typ. seq. and set of M_2 's to solve it.
- 2) Devise a form for L_3 i.e. a good type of types of univ. stoch. program.
- 3) Figure out a way to get L_3 to create M_2 's that have a max. probty. of being in L_1 rather than L_2 .
- 4) What is a good way to modify L_3 as a direct result of the success or failure of M_2 's?

35 → 5) Yes in does not try a guess what the next prob. will be and pattern its trials on this inputs to observing & subsequent (temporal) pattern in the successful trials. See Plan 396.3 for possibly. I think (1) and (2) would be good to work on first. Write some typical M_2 's, and some possible L_3 's that could have created them.

Then, express in intuitive form, various regularities in this (or any hypothetical) set of M_2 's → try to devise a form of L_3 → these regularities are easy to "express."

12N

01:37:40: Say M_i are strips of G_2 ~~or G_3~~

be ^{sub-}strips of orders, nps of orders and asso

Various methods of combining them.

For a goal to work toward - consider th. GPS of 1. Here we have a kind of problem-solving routine that works for variety of problems. Can we make a typ. seq. for which a human would discover their hours of "closeness", "improvement", etc

Then a G_3 must be designed to make such learning not unlikely.

It may be that R_1 Sim-New. hours are sofly complex, so that a TM that could learn them with a reasonable typ. seq. has a high likelihood of being able to learn anything thru a suitable typ. seq. These hours have the advantage of being close to human, so they tend to be easier to work with than most.

20 > Q. that looks like I'm ~~too~~ depending much on recent work:

If I use M_i 's of this form, and make L_3 consist of a single s, that creates M_i 's according to the strictly observed u_{gm} (and h_{tp}) statistics - (these statistics will change with time, and new inputs and outputs will be defined) - could such a TM automatically increase its search speed as was desired and which desire resulted in R_1 . somewhat more complex system.



Well, there is this possibility that we can use simple statistics on M_i 's (e.g. like ZIP 141) to make the new trial M_i 's.

Another thing is to consider "Lisp" as an M_i form ~~man 2/0~~

a G_3 form -

this is pretty much what I had in mind quite some time ago (see. Plans B.160.20ff) > Apr. 2001.

Part - 1

01:388.40: A basic related Q. Suppose we have
 by a set: Suppose we have a set of M_i 's. Now, suppose we have
 ntps to make this set of M_i 's. Does there exist a set of
 "regularity" in these M_i 's. Does there exist a set of
 defs. \rightarrow This "regularity" corresponds to a certain gram / ntp
 If so, is there any reasonable way such a regularity could have
 been derived? - i.e. ~~what would be normally regarded~~
 as a reasonable type seq. for that regularity, for a human, be
 a adequate / for a simple gram, ntp learning mechanism?

A poss. flexibility: We can take this set of M_i 's and
 try parsing it various ways. There will be different sets of defs
 A gram / ntps that give fairly good coding, - yet give essentially
 different extrapolations. We may want to use this flexibility to
 try be able to try new methods of recognizing regularities in the corpus.
 Also, there may be different strings of orders that give the ~~identically~~ some
 M_i 's - but certain ~~strings~~ ^{representations} make certain regularities more apparent
 to the gram / ntp defn. stock. ~~Grammar~~ Grammar: - This is a sort of more
 generalized form of "re-parsing" or "alternate sets of defs".

What I would like would be some way in which all types of
 regularities could eventually be recognized in the M_i set - when the
 M_i set become large enough.

Also, I would like a "creative Grammar" to create the M_i 's
 and characterize that set of M_i 's.

35 Say we have a bunch of ~~the~~ M_i 's. We scan them for
 regularities - out of a certain set of regularities. The regularities
 that occur get defined with the post and are combined to make new
 trial regularities to look for, etc.

Note, here, we would like these "regularities" to be

D₂ A

01:389.40: 14. a form to contribute to 14. str

It would seem that ~~creative~~ regularities in ~~note~~ should be also subject to the same comb. rules for h, ~~as~~ regularities in a directly descriptive grammar.

Note, however, that often, ~~the only way~~ an entirely adequate combination rule for abs is concat. This can be true in creative descriptive grammars. In such a case, I think that definition need be the only type of regularity to be recognized!

I think this last A should be ~~explicitly~~ clarified with some examples from either descriptive or creative "properties".

→ go back a bit (389.35), We want to be able to ~~discover~~ all possib. regularities in this set of h's or M's. Essentially, we have a TM hyper order TM to do this, and we want that hyper order TM to be "universal" and ~~creative~~.

perhaps this idea of combining old regularities by concat. to form new ones can be easily studied in (1) ordinary computer-order setups: (2) LIST operators.

Some simple "descriptive" regularities. e.g. in a sequential corpus: that the symbol "a" occurs with unusual freq. can be symbolized by writing "a". That "bc" occurs with unusual freq. can be symbolized by writing "bc". Note that we will look for the regularity "bc" if b and c are of the same class.

3.5. A more complex type of regularity should for, say, a linear corpus, should be derived intuitively - then have it expressed as a combination of several simpler regularities & combination symbols.

Note say the odd symbols in the corpus tend to be "a" with signify different statistics than the even numbered symbols.

Sort

01. 390.40! This idea can be "factored" into

the rel. freq of the symbol z .

① The idea of the odd symbols as a subset of symbols
concept. ② Also the idea of the rel. freq of z
in any subset of symbols is important. So we can do

our original idea by $F(\text{odd symbols}, (z \text{ freq.}))$.

(odd symbols) is an argument of F . An alternate first
argument of F would be any subset of symbols.

We may want to use Polish notation, or something like
it, here.

Odd symbols: We have the notion here of numbering the
symbols to designate them.

The concept of a subset of these nos designating a subset of
symbols. — The concept of the odd nos. as a subset of
nos. — The concept of "odd nos." — The concept of
integers.

We have listed here a set of concepts that would have
to get by post (and have a suitable type seq. to continue
make their being defined likely) before the idea of
the freq. of the symbol "z" over the odd symbols would be
a reasonable thing to try.

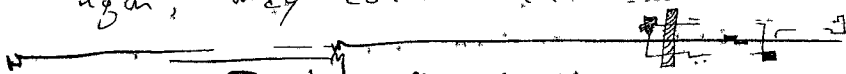
Note that there are other seqs of abs. that could
be defined to make this concept have a principle:

Note that the goal is to find all poss. regularities in a set
of M_i 's — possibly w/o being too economical about sample size.

One idea was to rewrite the corpus, making use of one regularity
to compress coding — then scan this compressed code for regularities.
If we have arranged notation properly, it may well be that

01:39/40: the only regularities we need look for
 like regular nouns & oe ntps. I am sure
 = TB 141, which is a descriptive grammar, but
 analysed to obtain the patterns of the associated
 According to this view, expected routine, any grammar type
 - eg. FSG (strong or weak) or XFL, Gram! (using, hvr, exact
 large sample size. I see so, this would be great for MT work
 in which we do have a very large sample, available. The goal, hv
 of eventually obtaining all regularities (no looping behind with sample),
 seems a bit too ~~ambitious~~ ambitious, and perhaps impossl.

IN = 1B 4: I think that rewriting the corpus in terms of one
 noun, may conceal ~~the~~ ^{unusual} ~~features~~ of another noun.



SN Incidentally, it seems very likely, now that using a nbfsc
 like **ZTB 141** on a large corpus, would result in
 a bunch of noun sets, from which one could "easily" obtain
 nouns (by factoring the set of data ~~like~~ like "The mach. of
 ling. learn"), and then loop rules.

The reasoning is this: say we have a stock FSG.
 then it will produce (substrings that are a pos)'s with
 "unusual" freq. and this "unusual freq" will be recognized
 if the sample size is large enough.

E.g. consider the lang. a^n n^n. Nouns: a, b, and ab
 all have unusual frags. In the case of this large hvr, I don't know
 > just now I would proceed with the discovery of the program,
 after it was in nouns were a word.

There is some point in not having "too good" heur for L3!
 The idea is this: suppose one has a straight MC search for an
 object and one takes the first object that "fits". An "heur" is a
 reordering of the trials by modification of the priord. As such,
 393.01

393

more quickly.

ok: 392.40 : it may lead to the choosing of a obj...

choice will then not have as much likelihood of

aprip. / Note that R₁ heard. modifies R₂. M₂ ...

as R₂ R₁ aprips would different. This apparent ... alters R₁ search order. R₁ actual aprips remain, ... same. This disun, hrr, must be worked out in a b

detail to see just when it applies Some (if not all) new

actually do alter R₁ apripd., since these heard are based on info from previous searches related but not identical to R₁ one in question.

18

In the vein of 388:20 : I think that if Th. + M₁ actually develops from a top. seq. using Mt. C. methods, that R₁ nemy that would appear would actually tend to be very useful in producing new trials by simple concatn. of by post nemy

Now suppose that it turned out that R₁ odd symbols of R₂ M₂'s had signify diffrat statistics than R₁ even symbols. There is no way I can think of to "notice" R₁'s regularity if one is only allowed aprip defs. Perhaps R₁ inter-symbol constraints would help a bit. In fact, intersymbol constraints could do it completely!

e.g. we have 3 basic symbols ^{a, b, c} Then the freqs in the odd positions are O_a, O_b, O_c; in the even positions E_a, E_b, E_c. Then we devise a set of 6 symbols: $\alpha = a^2$, $\beta = ab$, etc, and we express R₁ entire corpus string with these new 6 symbols. This completely takes care of the difference in freqs in even and odd positions.

Well - This still isn't what I want. Suppose R₁ prime ^{ordered} symbols had diffrat statistics than the others - e.g. suppose R₁ prime symbols were always "a".