

Best

~~base~~

IDSIA Talk 1

Introduction: General Goals, Method of Project

A Progress report on a system for Machine Learning that I've been working on for many years.

System for ML: Able to learn to solve practically any kind of problem, after having a reasonable "training sequence" to teach it to solve problems of that kind. As the talk continues, it will become clearer as to just what I mean by "Learning", "Kinds of problems", "Solveable" and "Training Sequence."

I will describe the system briefly, to give a general picture of how it works: then I will go over the system in more detail, telling just how the parts work.

This will be followed by a discussion of extensions of the system, present limitations of it and weak points in the present state of the system.

The system starts out pretty much like a newborn baby - with very little specific knowledge about its environment, but with some simple, very general, *learning* algorithms and *problem solving* algorithms. We give the machine some simple problems, which it solves easily. The solution techniques for these problems are then integrated into its problem solving Algorithms, so it can solve the next set of more difficult problems. After it solves them, it again integrates the solution methods into its problem solving Algorithms - So it can work even harder problems - as we continue our training sequence of problems of increasing difficulty, the system becomes capable of solving problems of greater and greater difficulty.

A bit more detail: What kinds of problems does it solve: At first:

Two kinds - Inversion problems and Optimization problems.

Inversion problems. Given a function f defined by a specific computer program. The program has input x and output $f(x)=y$. The Inversion problem is given some "y" value, to find x such that $f(x)=y$. Here x and y may be digital strings of symbols a/o numbers.

A simple example is $f(x)=x^2$: If $f(x)=9$, to find x , or $f(x) = \sin x + x^2 + \ln x$; $y=3$.

A more complex example: given a theorem in Math, we are required to find a proof of the theorem. We are also given an algorithm that is able to look at any theorem and any string of symbols that represents a proof of that theorem, the algorithm then says whether the string is a valid proof for the theorem:

Alg(Theorem, Cand) 1 or 0 (Yes or No).

INVersion problems correspond to the P and NP problems of Comp. Complexity Theory.

Another kind of problem, is the Time Limited Optimization problem. The formal form of this problem: given a function whose input(s) are strings and/or numbers. It's output is real numbers $G(X)*Y$. To find in 10 seconds, say, an x such that $G(X)$ is as big as possible. If we were given 20 seconds, we could probably find an x with larger y . An example would be to design in 6 months an automobile with certain characteristics,

Second hour

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This is different from the optimization problems in which time is not specified, and we want the value x such that $M(X)=\max$. In general for most of the optimization problems we will work on, there will be no knowable value of x for which $f(x)$ is max - that in the specified time we will have to do as well as we can, but expending more time usually means better optimization.

That talks about the problem class: As we continue, you will hear more about the problem class.

The next question is "How does the system solve problems"? The simplest version of the system uses a variety of "Blind Search" called "Levin Search" or "Universal Search".

The way it works: say we are solving INVersion problems. Any finite binary string (which represents a computer program) is a "candidate for solution". Lsearch uses a probability distribution to guide its search over these candidates. Candidates with high probability are tried first.
- But since we are interested in solving problems as soon as possible, we don't spend much time testing a particular candidate, unless it has high probability.

Roughly speaking we test candidates in P/T order, P being probability of the candidate, and T being time needed to test it. Using Lsearch, if a candidate has been given probability P and it takes time T to test it, Lsearch will find the solution in total search time T/P . If there is > 1 solution, it will find solution of $\text{*min* } T/P$.

Note: the probability distribution guiding the search, is a **Conditional Probability Distribution**. So input to P.D. is problem description. Output is list of candidates and associated probabilities of each candidate.

P (problem description) (cand sub i , p sub i) in rough p sub i order.
The function P contains all of the problem solving skill of the system. P will change as the system learns more and more.

Suppose the system has solved a few new problems. How do we modify the system so that it is able to solve more difficult problems? We modify the Conditional Probability Distribution so that usually the solutions to the new problems are given higher probabilities than before. Since solution time = T/P and T remains invariant for that particular solution, increasing P will decrease solution time.

Our method of modification of the Guiding Conditional Probability Distribution will be dealt with more fully, after we discuss the estimation of probabilities.

The particular method of *probability estimation* we use is called Algorithmic Probability. Suppose we are doing sequential prediction: we have a long binary sequence, S , and we want to know the relative probability of the next symbol being 1 rather than 0. -

Let us assume that we know an a priori distribution on all possible Binary strings. $P(R)$ assigns a probability to each such string. Then the relative probabilities of the continuations 1 or 0 will be $P(S1)/P(S0)$. This would solve the sequential prediction problem, if we know $P(R)$.

HOW TO FIND $P(R)$

For a heuristic understanding of what follows, consider Ockam's idea: That simpler hypotheses are more likely than complex ones. We will quantify this idea and describe some results on the accuracy of the resultant system.

To quantify Ockam: Suppose B is a finite binary sequence. - Then D sub B is "description" of B if M sub $r(D$ sub $B) = B$. Here M sub r is an algorithm or computing machine that is able to take the binary string D sub B as input, for which it gives B as output. D sub B can be regarded as description of B in terms of Machine or function M sub r . I use the subscript r , because M sub r is to be regarded as a *reference machine*. Clearly, by changing reference machines, the needed descriptions for B will change. The shortest description will be the one with fewest bits. - containing least information. We could approximate the probability of B by 2^{-D sub B . For more accurate estimation, we use $\sum_i 2^{-D$ sub B_i . (Also mention *theory revision*). Here we are getting the total probability due to all possible descriptions of B . We will call $\sum_i 2^{-D$ sub B_i the Algorithmic probability assigned to B by M sub $r = p$ to the M sub $r(B)$.

There are a couple of loose ends here: first, what reference machine to use for our *probability calculation*? There is a group of machines called "Universal Machines", that are particularly good for describing things. They have the property that if we compare 2 of them, there will always be a constant factor that tells how much they differ from one another.

Two machines M_1, M_2 then P to the $M_1(B)$ and P to the $M_2(B)$ are always with a constant factor $C_{1,2}$ of one another. This constant factor depends on M_1 and M_2 , but is independent of B . The constant factor can be quite large - so it usually makes a lot of difference as to what reference machine one uses.

In our system for Machine Learning, we will be periodically changing

distribution that guides searches for solutions to problems *is* an induction problem. How is this so?

Consider all of the (problem description i , solution program i , time i) triplets that have occurred this far. What we want is a bunch of relatively short codes to describe all of this data. From a set of codes of this sort, we can then extrapolate the data to any new GPD sub 2 (problem, solution) \rightarrow probability distribution on Time to solution. From GPD, it is possible by a process of integration, to create GPD sub 2 (problem, solution) \rightarrow probability that this is the fastest solution to the problem.

This last distribution is the guiding distribution for our search for solutions to new problems.

I will now discuss Levin's universal search procedure: suppose I'm in a Gambling house and there is a kind of lottery with a single large prize. I am the only customer. Each lottery ticket has a certain probability of winning - which is printed on the ticket. In the first kind of lottery all tickets cost \$1. The best ticket to get would be one with maximum win probability. If it doesn't win chose next largest - and so on.

In the next case, each ticket costs a different amount of money, (P_i, M_i) are the probability cost associated with the i th ticket type. The best choice to make is the ticket with maximum P_i over M_i . You get the maximum probability of winning, per dollar spent. If you continue to buy tickets in P_i/M_i order you are certain to have least expected money spent before winning.

The gambling house can be translated into a problem solving environment. Instead of money for a trial, we spend TIME. The best trial to choose is one with maximum p_i/t_i .

Normally, in trying to solve problems, one may know p_i , but one doesn't know t_i . In this case, a very good strategy is Levin's time share strategy - (which I think is likely to be the best possible). The way it works: you work on all trials simultaneously, but you work harder on trials with large p_i values. The rate at which you work on trial i is proportional to p_i . Suppose we use this strategy and after a while the j th trial gives a solution after spending a total time t_j on it. Since other trials have amounts of time spent on them proportional to their p_i s the i th trial will have $P_i/P_j, t_j$ time spent on it.

The total time spent on all trials will be $\sum_i P_i/P_j t_j = t_j/p_j \sum_i p_i$. $\sum_i p_i$ will in general be 1. So the total time needed to solve the problem this way is t_j/p_j .

So: this is Levin's time shared search and I think it's as fast as is possible for the kinds of problems in which it is used - i.e. INVersion problems.

This technique can also be used for time limited optimization problems. In this case our trials are not programs that attempt to solve inversion problems: They are *optimization techniques* each one takes as input, the problem description, $G(X)$, the function to be maximized, and the time available for solution. As before, we have a conditional probability distribution, that looks at the problem description and assigns probabilities to various optimization techniques to be applied to that problem.

The way we do the search: say our time limit is T for some small τ we spend time $\tau \pi_i$ in the i th O.T. and slowly increase τ . We keep track of the O.T. that has best G value of all the O.T's when $\tau \pi_i = T$ for any OT, we stop spending time on that O.T. When $\tau \pi_i = T$ for any OT, we stop spending time on that O.T: When $\tau \pi_i = t$ for the best O.T., we stop. This is our best value. Total time $= \sum \tau \pi_i = \tau \sum \pi_i = \tau$: $\tau = T/\pi_i$, so time to obtain best solution in time T , is T/π_i , so we have efficiency factor of i/π_i - as we did for INV problems.

.04-.05: T. idea that in Lsrch, the main problem is to Modify P.D. for the problem \Rightarrow in Lsrch to factor 2^2 is not large. — ideally ~ 2 or 3 or even one! \rightarrow buskato: (15)!

Q: Is .04-.05 literally correct? One doesn't ever do Lsrch w. large CJS? Just exactly what did that "small 2^2 " mean?? I think the idea was that Lsrch was the optimum if "all the info was in the P.D.". — If a long Lsrch (w. large CJS) is the best way, then 2^2 will be large.

I think 2^2 is usually not very large — that an imp. part of TM's work is to modify the P.D., so that 2^2 usually is small (but not nearly ^{as small as} 2 or 10!).

Another Q: I had the idea that Skinnerian TSO's (very small CJS) were non-v.g. — that a long Lsrch was more "educational" for the student. This would be true if the student tried "interesting" trials. — if the soln. was really "educational". To what extent is a normal, large CJS Lsrch "educational"? — How is it "educational" ~~what~~ in what specific ways?

This seems like an imp. Q! Perhaps CRITICAL!

An Approach: Consider 2 ways of working a large CJS problem:

1) straight Lsrch 2) Normal method(s). Do they really differ (if so) how does the "normal method" give more "training" to TM than straight Lsrch?

E.g. Consider 2 students.

In "straight Lsrch, w. a "Big Problem": T. soln. is pretty much as Envisaged in SB9, using a "conc. net". A big problem is solved no differently than a "small" problem, but the CC is larger for a large problem. After solving the large problem, TM is not nearly any better off than if it had been solved in a Skinnerian TSO. — in this latter case, TM would have used fewer concs. used, but if the problem is solved as a single large prob., then the concs used don't nearly have $CC > C$ & need not be recognized as "useful". So normally a Skinnerian TSO with 6. ~~prob~~ concs in it will normally be better than having TM solve the final problem at very large CJS (w.o. a TSO, leading to the 20th, as in Skinnerian case). $\rightarrow 201$

from "Expt 1.40"

01: On a non-optimal situation: Consider very large problem: One starts out by doing much "Meta-long": ¹⁰ ~~10~~ Ayer level (ing. a/o learning ingenuity) related to final problem.

E.g. Gang & Collopy before trying to solve Expert Problems.

["Meta (ing)" is not a nice term "Meta data (ing)" is about how to / in; about how long works (i.e. fails). For a large problem, one might do some Meta long in preparation for it, but one would mainly do (range in T. field of study & related fields, ^{which is what I'd like a formula}) T. main idea here is that this study is Non-precise — not doing trying to directly solve problem.

TM₂ could do this "pre-education" as a kind to "Self taught TSO"

10 → A perhaps better tack: Say to P.D. for Lsrb is conditional on ① + initial problem down ② T. trace thus far of + work on this problem. This certainly seems like a reasonable way to do prob solving ② take advantage of previous experience. So to P.D. is on "what to do next" in view of ① (.10) & ② (.11).

10 seems like a more general method of prob solving than a "Simple Minded" Lsrb.

15 While "Simple Minded Lsrb" does prove a factor of 2² slower than "Best Method", th. ①: Does it give us any way to make 2 "small"? Could 10-11 be regarded as a possible way to do this? Is th. P.D. of 10-11 can be regarded as a P.D. on prob solving process? So, w. a N.G. TM₂ Smith would design a P.D. for Simple Minded Lsrb, that was really to solve as 10-11.

In 15 I think th. P.D. on prob soln. would ~~not~~ ^{be} conditional out "trace buffer" —

So perhaps TM₂ could use that info for a "Simple Minded Lsrb"

— In 10-11 one has both + original p. solving cond. proc. and a modified it proc. as it evolves (≡ runs a trial coin). — So prob of soln. could drop down to trace —

This could be a way of implementing "QUICK ABORT" !!

25 Hvr. it would seem to be out of th. "spirit of Lsrb" because, doing th. trace of a trial's pc would not be monotonic, it could ↑ a lot if it began to look like a soln. was "Near". Would 10-11 have this difficulty?

One way to deal w. 25: For Lsrb control, ^{ie. change next cond.} use only initial (pre-execution, pre-trace) estimate of p.c. of cond. For discriminating a trial, use recent ~~the~~ (trace) info. [I'm not sure th. ② will work ① will defend / (concentrated) !]

Another point: P.D. & Cond. themselves are ~~probabilistic~~ "probabilistic" — meaning, what?

32 Stochastic — Markov-ish? (I may have once used an idea like this to deal w. th. defense between pc & bc in Lsrb) → see 35 hvr.!

[5N] Define th. Main problems Clearly, Exactly!

One attempt at one problem! Be: 10-11? In order to express many ~~th. 10-11~~ ^{pc} we need to use trace info from each trial. Hvr. P.D. screws up Lsrb, because ~~it~~ ^{pc} not-monotonic during th. trial, ~~it~~ ^{it} can ↑ as well as ↓, & ~~it~~ ^{it} increases up!

35 On 32 we discuss M.b. code: Hvr. by use of Lsrb we can make th. deterministic, if we like!



Remember that the "soln." of a problem is a "trace" (i.e. method) rather than a simple final answer.

Still, a soln (trace) to a problem can be completely described by a finite, describable string.

Another way to describe a soln. is by a probabilistic string, in which the prob. of the string w. lang. R. I think this ends up w. a probabilistic d.f. on strings that has "t. sequential property" — i.e. $P(a_i) = P(a_{i+1}|a_i)$ a is a string & Σ is the set of all symbols that can follow a.

So it looks like .05 is essentially, t. demands using LSEARCH on a P.D. of solns. Best to look at 1 or more prob solns. to see how LSEARCH fits (if, indeed, it does)

How to solve linear eqns? Move all terms to LHS. of eq. Then simplify as much as poss. Try to get eq. into form $ax + b = 0$ then $ax = -b$ then $x = -b/a$.

"Simplify" is a rather complex operation, it has to be learned from examples:

Perhaps learn what term "simplify" means by examples —

< "In many cases, "simplify" may have special meanings. >

[SN] One of the main things I had against Skinner's lang, was that we might not include certain very imp. concs. that were needed for non-Skinnerian lang. Sk. lang is a lot like Loran's teaching cyc by "brain surgery": by carefully giving it "facts" that he thinks it may need. I can give examples of such a conc. — "normal" "Expert Systems", t. "brittleness" is due to lack of those concs.

[RB] In OZ probs (which are most problems) TM spends very little time in "LSEARCH". Each OT is "its own thing" & may or may not use LSEARCH internally. I think the main advice was ordered via LSEARCH, i.e. "vector P.D." has a job of matching an OT to a given OZ problem (or to find a P.D. over t. OT's for a given problem).

Inv. probs are often (maybe usually) converted to OZ probs via "G.P.S." method of a Vector Conc. .12 could be viewed as a kind of "OZ" soln to an INV. prob.






"Simplify" is a kind of OZ problem, but t. Conc often has several components. Certain components are more "inv" (maybe essential) in certain probs.

Getting back to 1.15: An assoc. problem is: If Loran is t. machine, then most/all hours have to be inserted into t. P.D. I need some good examples that are easily parsed.

If a hour, reduces/compresses to some space, this would seem to be expressible as a mod. of G. LSEARCH Grand P.D. It would seem that many hours would be of this form. —> See 4.13 for poss. serious objection/diffy.

So there seem to be several general schemas for prob solving. Trying 34, which assumes all needed hours are compressed to G.P.D. modules. (This G.P.D. is conditional on natural problem, only)

There is another (perhaps different) method of prob-solving in which one solves a

UNKS on IBM TP:
1) Rtsider
Bos 3 AM
Jacks
2) Navata
00/00 sol!
a) 
b) 
So 5 items.
3) Back of the
EOT cover:
3 litos. battery


When nothing
"on".

Not talk
bottom, but
both changing.

1-3-00 Bulg.
#.....

Abcdefghijklmno
Abcdefghijklmnopqrstu . ABCDEFGHIJKLMNOPQ

A

01:34. Problem by Lisch, using a Z141 type P.D. we can look for ways to compress G.P.D. — using the latest problem soln, but not necessarily using it: One could simply try to compress whatever parts of cards are seen copies in: but using the ~~latest~~ latest prob soln. as part of a compression is better because its more likely to be relevant to future (i.e. never before) problems.

• **T: Main Prob** I'm working on now is ~~the~~ "What is the over-all (true) Alg of TM?"

How is 3.34-37 in 3.39 safe.

Perhaps related to 3.39: T. idea of my working a problem, then trying to break down my soln into a "conc net" — that would implement its soln. via Lisch.

RE: T. Alg of 3.34-36 on why more hours can be implemented as modifs of a G.P.D. : Some poss. Objections:

1) Most hours ^{are} ~~not~~ ^{main} for ~~main~~ problems but for sub-problems, so they quit (immediately) relevant to Lisch for "Main" probs.

1 □ $\sum_{i=1}^{\infty} e^{-i}$ ~~□ □~~

2) Many of the Modifs of P.D's demanded by 3.34-36 may not be easy to implement for Lisch: i.e. the modifs may give a P.D. in which it is very diff. to put cards in a PC order.

Looks VERY IMP!

[L.S.N.] RE: 3.13 P.D's come in various forms of device of Perm. Some in some forms, certain operations are easy, other operations diff.

Xfng. from one form to another is sometimes a way to enable certain operations (like putting cards in a PC order for Lisch).

Abcdefg
 $\sum_{i=1}^{\infty} e^{-i}$

A few types of P.D. devns: 1) Intermittent & VIO machine w. random input: Could be a Mt Carlo type.

2) Input is string, output is PC of string. 3) Input is N (numbers) output is N's most likely string.

(This last is useful for Lisch). 4) The Z141 Model: Is this a variety of ~~3~~? or CT [Z141 Model] ^{Abc} ~~defgh~~

5) General Mt. Carlo: P of output = P of raw string. While the Model of 1 does P's, we need a conditional Mt Carlo: e.g. Given 6. first 6 data of a T.S. (condition), to put a 6th card d.t. for the next data pt. (This is what a "Summary Machine" ~~does~~ ^{does}).

Can I put some (or All) of the hours implied by 3.12 (Hours Solvilinear Equ) in form of Modifs.

of the G.P.D? (Note all 3rd hvr.)

* 1-4-01 Perhaps look at older A.I. literature for a system to do Alg. probs: Look at hours used, & see how they can be put into form of modifs. of G.P.D.

Perhaps look in "Encyc of A.I."

31 The idea is: All A.I. probs involve search: Any hour must narrow the search: If the search is characterized by a P.D., then any given P.D. narrow the search, can be interpreted as a modifs. of the P.D. used to identify search.

From 31 f. Q. Then seems so far **(13)** ... ordering the P.D's for Lisch.

see with 99 TM 9.05 (1.5.99) for a different proof that "All hours can be expressed as modifs of the P.D."

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Maybe good idea to write a few "Internal Reports" (or "Papers")!

1) On ~~optimal~~ Practical Optimality of Lsearch. This is an explanation of Why Lsearch, properly done, may be close to optimal. The best possible achievable with given informational and Computational constraints. Computational constraints are cc . [Give (many) Examples!]

~~Informational~~ Informational constraints are: t. sp., "heuristics", memorized techniques, general knowledge of the domain of interest

Lsearch guided by t.
a) That, by Lsearch, I mean to Guid. PC, $P(\alpha)$: Given a problem ccn, α (a string), it is an output P.d. on strings that are poss. solns. of the problem, α .

Give examples of (kinds of probs); (T.S. extrema, bsp extrema, OZ probs, JUV probs) and what $P(\alpha)$ is a d.f. over.

b) Heuristics can be defined as techniques to speedup (more or less, make less costly) the search for solns.

c) Because of b), heuristics can perhaps always be expressed as modifiers of first $P(\alpha)$. no. not always (only) see 11.3, 10.5, 12 for why not!

They are in classical A.I. modifiers of a search; instructions to look in certain regions of the search space. In Lsearch, the ordering of the search trials is governed by priority to form of $P(\alpha)$:

So (we can implement heuristics by modifying $P(\alpha)$). So, if we use $P(\alpha)$ to guide our Lsearch.

d) Though a particular heuristic may, indeed, be implemented by modification of $P(\alpha)$, those modifications could take a form in which it is very difficult to order the ~~trials~~ $P(\alpha)$ trials, in pc order, to implement Lsearch. How can we make modifiers in $P(\alpha)$ that a ① will implement + heur ② will make it easy to order the trials in pc order.?

One poss. way? Any particular heuristic must have some reason for existence:

① Either thru the empirical history of its discovery (i.e. what solutions to what problems suggested this heuristic), or the heuristic might have been obtained by analytical/mathematical reasoning.

If ② is true, then we can take the set of past examples (or invent a suitable

set of "~~trials~~" pseudo past examples") and use our normal induction algorithm to modify $P(\alpha)$ in view of the past of the "~~trials~~" pseudo T.S. → 6.05

If ① is true, then a reason (but heuristics not proved) that a similar modification of the P.d. $P(\alpha)$ can be made — that analytical/mathematical reasoning is a kind of probabilistic ~~reasoning~~ reasoning, but with probabilistic of α only. ... Then why consider it to be analytical/mathematical reasoning? — if reasoning is — "So we can apply the modifiers of ② to it to modify the P.d., $P(\alpha)$."

This "normal induction algorithm": this is so-called our first. See GOF on this

17.01 1 Bulg Expo.



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.05: 5.30 On "T. normal induction Alg.": TH is Given a P.d., $P(\alpha)$ and a set of examples (\rightarrow BAG) assoc. w. a particular value of α (i.e. problem data). Or choose $P(\alpha)$ and a set of α_i in a BAG for each α_i . The α are: prob. data, the BAGs are the problem solns.

[Guess to elements of BAGs and traces of prob-solns]

.10 The problem is to modify $P(\alpha)$ so that $P(\alpha) \rightarrow P'(\alpha)$ so that $P'(\alpha)$ best expresses
.11 +. Post corpus plus the new augmentation to the corpus of $[c_i, b_{ij}]$ set. \rightarrow see .15
The problem down for $P(\alpha)$ is: $\{ P(\alpha) \text{ and } \dots \text{ plus } \dots \text{ augmentation of } \dots \text{ corpus.} \}$ And the standard request to do
From this, $P(\alpha)$ outputs a ^{p-wtd} ordered sequence of ~~subproblems~~ DT's (optimal length request) (Note DT's change order if p's are given)
.10-.11 is a problem of modifying $P(\alpha) \rightarrow P'(\alpha)$ so that $P(\alpha)$ (P'deen) \rightarrow $P(\alpha)$ \rightarrow BAGs is May

.15



I guess the expert problem is to get a good set of DT's for probs like .05-.15

Or perhaps \rightarrow simple good DT's

A possible Approach! Do some simple problems in Algebra: See how the techniques of .05-.15 map into the way I solve the problems. Well, maybe: but the kinds of problems described in .05-.15 seem more difficult.

Maybe also discuss F.N. Part Sol 89 (26)

An attempt to outline a country's debt, what I will say at Lugano (IDSSIA)

Some imp. items:

- .02 1) Done but + system is within factor of 10^2 or 10^4 of optimum for PC is intermittent constraints. Give several examples. Give 3 ways to show approx. of hours to modulus in P.D. Give counterexamples, then.
- .03 2) ~~PC~~ Eyring's detailed descr. of meaning of 3 GHT's. (GHT's is .02)

- .05 3) "Quality" of a P.D. — what it means to "compare P.D." — maybe include better than P.D.'s are partially ordered (maybe > 2 dims) — maybe include better PC of coops is time to complete coops. we want $\frac{P_i}{T}$ to be max (at least!)
- 4) Discuss varieties of P.D.'s, how they are often whole or partially applicable to an actor. Our next formal to output in PC order (w.r.c's): Max PC first.

In Sol 86 I discussed "Comparing P.D.'s" There are only one (or two) aspects of "importance to P.D."

STU QUICK ALERT "Quick Alert" may be simulated by assigning a (nominal) low PC to a bursty msg. (My mind is not clear on just what this may mean!)

- .23 5) ~~PC~~ Say I have 1 second to work on "comparing P.D." If I had a well defined Genc for this, it would be a regular OZ problem. Now, I do not — Genc is partially ordered. Is such a meaning for in this context? — do I have to convert to a scalar Genc? Also, "Horizon" is relevant here: one can "improve P.D." in various ways, which are more or less complex & how much more cost? But .05 & .13 are separate problem areas involving TOLZ. A ~~MAIN~~ MAIN

Q: it is to solve .02-.03. The "normal" varieties of δ system ($\approx TM_1$).

One way to do this: make \approx concept net, as in Sol 89. Some off. "conc" can be hours. I have in mind 2 different ways of solving problems: One way, one person/assembly various coops (sequentially), to form δ soln to δ problem. As an example, perhaps δ prob here in δ early search pems (ANL), deriving pems, using a grammar of "introductions" or of "questions".

.24 Another way to solve prob, is sequentially: One is in sequential situation, what to do next? one does it's own in a how-to-situation, what to do next? — loop to δ

.26 ISN Another Aspect .02-.03: To show that "blind search" can \rightarrow heuristic search, if P.D. is suitably modified. (perhaps this is what .24 is about!) (This is the F.N. of Sol 89, 15) — so at 4 aspects.

I should be able to take any prob solving method that I can think of & show (via the 4 aspects) how TM would go about discovering that method.

.25 O.K.: Consider "solving most known prob" (w. 1 unit): working backward: try to get eq. into form $a \cdot x = b$, then $x = \frac{b}{a}$. Do reform the prob system in a "simpler" form of a form that seems easier to solve. TM must have some concept of "simplicity" or idea of what forms of eqs. are "closer" to being in $a \cdot x = b$ form. So this looks like an AND/OR net problem. The "parts" of this "heuristic" are of by PC... They are often used in solving, & other probs. Also, ISN is in the direction of GPS a fairly general problem solving method.

Some weaknesses of T.50's -> 1) How did TM get to be that successive "legal" refus out. instead of. Is it way to go?

b) In what sense can we say 7.35-40 is essentially defined by a P.D.?

c) What is f. L used on? In 7.35 ... ~~... to~~ "cond", f. active 7.55-90 "ppm"?

SN Is blind search ~~... of~~ ^{FNPS} of S89 discussed in Banc? ^{Near final} ^{Now P.D. and!} } Section 4.2 on "Learn".
also 4.3

Well, first look at 7.35: In what way is it a "search" problem? Well, we get ~~...~~ searching for a seq. of operations that ends in $x = \frac{1}{2}$.

All INV. probs can be regarded as "search for string that satisfies conditions"

In 7.35, we can consider the "closeness" $x = \frac{1}{2}$ "heuristic" (as "simpler" form, etc.).

We seem to be learning "during" the search, because when we find a candidate closer to " $x = \frac{1}{2}$,"

we try modifying it (i.e. GA. w. Mutation only). "Learning during the search" makes it not blind!

So how does this square w. the Sol89 footnote?! Learn is (normally) for "Blind Search" only.

What did the PN say? - That any non-blind search could become blind by x paid into an overall blind search of our task and params. It's not clear how this is supposed to work: Maybe need f.N. (By "parameters" I may have meant "x" or "traces").

Unless ~~...~~ f. P.D.'s modified during search. - In which case, each cond is chosen "blindly", but ~~...~~ successive conds may have quite different P.D.'s.

If f. P.D. is different for every cond, when is Learn possible? - or more exactly,

Do we get expected cc for each of $\left[\frac{T_0}{P_0} \right]$? (or $2 \cdot \frac{T_0}{P_0}$).

Also, what about ll (times traced) Learn? - one doesn't really have ^{completely} ~~...~~ set of general fossils in ll search. (Perhaps look at Li-Vitanyi Book for alternative search techniques.)

21 It is easy to generalize 7.35 to a ^{very} common search heuristic: We start to solve prob. into to be pattern via a "closeness" criterion, then use ~~...~~ proper Modification - only in GA.

26 Try this: Say the Cond (a trace) is characterized by f. traces.

In a) formulation as 21! The trace, is a sequence of x terms \in f. pc of each x term depends on f. pc assigned to previous x terms, so the cond. traces are strings

in a stochastic lang. - so Learn is appropriate!

So conds are generated as follows: One has a vocab set of x terms plus. At each step in the generation of the trace, each x term has its own pc, i.e. a format of "f. entire trace thus far" for that cond. This is certainly an ordinary P.D. to which Learn can be applied. ~~...~~ PC's may have been P.D. idea of the Sol89 footnote!

Whoops! The pc's obtained as one generates a trace are ^{not} monotonic! - So perhaps not a real P.D. (could one still search anyway? Usually f. pc's has one going by a trace, as opposed to normal generation of conds by stochastic Grammars)

Student

Looking at FN. §4, 5489: "It is not difficult to prove that" (1): any hour α can be considered in Blind search, in a space of sufficiently powerful cores. A good α must be permitted to have more depth than supply to present problem down: T steps may include any info obtained as result of a single iteration, or previous core α 's.

In 01-02, I guess its f. p. data cond. as a part of (1) prob. data (2) other things. If f. p. d. is diff. for each cond: ~~then~~ then search may not work! \Rightarrow previous trial history. Nontheless, the p. d. part is updated after each trial does seem to be what many hours do!

The seq. of trials is a progressive long experience.

In 7.35 ~~also~~ again when is/with a trial at all — α one doesn't expect it to be a soln. More α 's when executed, can be regarded as "Experiments": Things done to gather information.

Another Question from ~~1/2/85~~ 2/2/85!

"Any Hour can or any other info should be insertable into f. p. d., by assigning it to pc or an operator (w. a pc) that tells it when it is to be used."

6/13/89
6/13/89

1:30"
40"
33"
43"
40"

15

Well: T. "Soln." In 7.35 ~~the~~ thing we are searching for is f. pgm that solves

f. eq. This is always what IMV search looks for. We were not looking for f. "trace" of f. soln. (Re f. "trace" is priority of f. soln: it goes into a list of f. soln. pgm.)

Finding is applying suitable hours to get this pgm. & then f. final present problem. It may involve logical/mathematical reasoning ("non-probability").

For knowledge devising heuristics, any experience of f. past iterations would be used by humans. Hvr, if we are to search, hours can be derived only implicitly from "previous problems": If we also use from f. present problem, f. P.D. will change during f. search — which invalidates our CJS estimate.

well, if we use the T-2T mode of search, any changes made between T-2T operations would not screw up f. search; It would only \uparrow f. PC of f. soln. — which is fine! So if we find several or discover several modifs in f. P.D. during a T-2T "round", we do not apply these P.D. modifs, until the next round.

Hvr, since the T-2T doesn't occur many times during a problem solution — one could be rather limited in one's source of data for hours.

One help is to use T-2T w. $\alpha < 2$. Already $\alpha = 2$ is suboptimal but only slightly so: $\alpha = 3$ is, I think, optimum. Some how much worse $\alpha = 1.5$ is. (Making α smaller means the T-2T operation occurs more often (approx. no. of times, per soln. of problem))

The analysis for optimum α comes about when I use using the T-2T method as a way to combine H.W. errors in large-scale parallel computer implementation of TM. (The references within f. list of me (I think).)

Another possy is to restart f. search ~~at~~ ^{Whenever} some u.g. (looking through new hours (modifs of f. P.D.) are found, (the this might be too useful!), Another possy Another possy: that it's useful for many (or even any) new hours to be used during f. problem soln. (?) — multiple case with no troubles! \rightarrow See 2.2.26 for more on changing α during f. search.

Self (I think) knownable to find f. number of hours out. proposed Hatched 1989. Search box: when.

Of interest: At 7.35, I was interested in the implementation of Hours in Lisch: yes 7.35 is not exactly a Hour! It is a problem. But solves many kinds of problems.

So: finding out how 7.35 could have been derived, is now a **BIG PROBLEM**

Actually 7.35 is a sort of Hour: it is a problem. But can be easily modified to solve a great variety of prob.

Main Immediate Goal: To demonstrate (to Paul) 7.02: first: that for GNU prob, Lisch is within a factor of 2 of true optimum if all info is in P.D.

Specifically, that any hour can be put into P.D. is any hour is "learnable" by f.s. system, if it is learnable by human: a fact which hour has to be inserted, or TM has to be helped in finding it via "Hints".

The 3rd says that "any hour" can be expressed as a P.D. method.

1) Hours are ways to speed up search - by reordering trials: This can be done by changing P.D. on trials via Lisch. (5.15-19)

2) We can find out how to modify b.p.d. by hypothesizing how a Hour could be found, deriving a f.s. q. (or a taken T.F.2) to get a result that it would change a f.s. P.D. (5.25-30)

Also a hour can be obtained by logical/mathematical reasoning: (5.32-37... Hard to be worked out)
~~Since 3.05.06 (5.12.00) for some development.~~

3) "All info is in P.D." If a person has a better way than TM to work the problem, then that person has a hour that is not in TM's P.D. This assumes all Hours are expressible as P.D. methods.

4) T.F.M. of 34, Sol(89); [9(70) - .02]: Blame I ^{still} don't understand this. Can't work

(Is correct:)

SN on: For ^{SOME} hours to be "learnable": TM should be able to "match" problems being solved.

Good!

So that it can observe Ropyz in f. solns. Perhaps this observation gave rise to that FN in Sol 89: In general, it would seem that human learn based Hours could be based on any ^{kinds} repetitive observed in problem solving - so all of this kind of data must be available to TM if it is to be able to derive (almost) all hours.

I think FN of Sol 89 is imp., but one must state 6.20 to understand why!

~~Blame I~~

Quick ABORT - a Generalisation.

for **SEARCH** of paths to INV. Paths
 Heuristics / est. funcs (cores), methods of such paths. Fast method than (less expensive) in time &/o memory.

They are of 3 kinds

1) (same Cond. paths of search invariant, but vary order of trials) - This is in Recursive.
 Not doing heuristics mean e.g. doing trial on least order in fixed time cutoff for all trials, or Random & trials w. fixed time cutoff. (find more interesting Examples: random has the

13. only type of hour I can prove for "x4" effect for.

2) Least order of trials to solve but speedup some or all trials. Simplest examples:

a) Get faster machines. b) Quick abort: [trials terminative are modified because of order of trials & invariants to be less cc.

Example of Q.A: Say problem is to find shortest codes for strings, S, using VLM.

Not using Q.A. wait till either output machine stops before computing outputs & BS.

using Q.A. reject trial as soon as ^{out} bit decreases than from that of S.

[This is not a necessary flow: one could run machine on till a certain threshold or output bits were wrong: If \leq that no. more in error, than abandon make code by adding in bit string fully the where errors were: This can be done in (at least) 2 ways:

The newer decision will \downarrow of error but for "Best" results for search.

See test (1 recent) work on Zif: Rec: Garry Wolf:

Q.A examples would check change parts of output to error (they go to be in error, & reject error if they had more error nearby correction than to Best Best Cond.

Flow for

3) Method of 11 & 23: Examples for OZ problems, bits. We want to find (linear regression) coeffs $a, b \in \mathbb{R} \rightarrow$

$a x_{t-1} + b x_{t-2}$ closest to result to x_t ; for $t=1, \dots, n$. $t=0$ to n (using least sq. error criteria: Non-linear way:

Search Do xhaustive branches a, b plane: start w. low resolution search

$\mathbb{R}^2 \subset \left(\frac{b}{a}\right) \subset \mathbb{R}$: for successive trials use larger R & higher resolution

(e.g. double R), but to fixed size limit (this multi by 16 no. of submesh round!)

Heuristic of type 2: for each round, find best code & its error.

For next round limit region in a, b space to best approx $\pm \Delta R$ radius value \pm best approx.

A type 3 has: find a, b by solving linear eq. involving correlation matrix.

Its not clear if flow is type 3: It goes directly to the optimization, so usually sorted to order of trials best as in to non-homothetic order but it eliminates all but one trial.

to In all these methods Flow is extra surf. of cc involved in

Computation - particularly in the (linear eq. soln. "heuristic".

The method of 18 can be used for non-linear regression, but to narrow down search in a, b space. Then when we are close enough, use (linear) approx to get successively closer.

Another type 3 example: Using Machines M1: Flow usually changes order of trials as well as speeding up all over process.

This is an OZ problem!

It is also an OZ problem!
 using linear & problem minimized error flow pk.

Acct no. 2581096745 personal 9792

2-2-01 Bulb

Quick About → Generalization

So far, I'm only saying that a hour is a part of modula. off P.D. if it hour is of type 1.

It may be possible to show for some type 2 or 3 hours as well: I don't know.

SN on Lsch: Consider non-Lsch of 11.02-05: How bad is it? Say we do have w. time limits $T \leftarrow 2T$ for successive rounds, but get Lsch: Say we want shortest code for string s. We take all trials of length $|s|$ to start. As soon as we find a code of length $\leq (s)$, $P_{1/2}$ becomes our new limit. If t shortest code is of length $(s) - (0)$, (it would be a v.g. SM code) ~~that~~ $P_{1/2}$ method would \downarrow cc a lot but still cc would be enormous.

Q: Is this method of 11.10-14 a "horrible"? It would seem not to be a INV problem horrible, because ϵ goal is unchanged. ~~Essentially~~ Essentially, it's an OZ problem, anyway! In fact, most of the examples have been \approx OZ problems

The problems in 11.27 could be made soln. of 2 equs. in 2 vars.

Q: if $f_1(x,y)$ & $f_2(x,y)$ are both continuous, & both change sign in region R — Most there to be a pt. where they are both 0? No.



For linear set of equs, ~~the~~ if there is a soln, it

is unique, so goal $(f_1(x)) = 0$. One can find regions w. low ϵ out. Sort of,

typical search (11.02) for INV problems Given $f(x)$, to find $x \ni f(x) = z$.

If $f(x)$ is a unimodal Lsch ~~can~~ using P.D. on x ($P.D.$ is a funct. of $(f(x), z)$.)

If $f(x)$ is diffrat for each a , then whenever $f(x) \geq a$ from that point x , it will not reach a for x to diffrat a (!).

Say one is looking for a from $f(x) \geq a \ni \forall a, f(x) = z$ ~~when~~ when $f(x) = z$ has a soln.

Trouble is, one can't verify that $f(x) \geq a$ for all a (!).

See p. 504 in 1997 Liv'ing. (P503(bid) lists a few interesting INV problems. ~~40~~ 40

Using Liv' notation, we want to invert $\phi(x)$ given x we want $y \ni \phi(y) = x$

~~If we use~~ Look for only (Least is a finite ϕ only) Say A is an ~~an~~ inversion Alg

Then if we use a P.D. on A , as a function of ϕ alone, the ~~time~~ time to find y will

be $\leq \frac{TA}{Pcost(A)}$ This will be true for any x .

if ϵ our P.D. is on A as a funct of ϕ and x , and A exists, then

$\frac{TA}{Pcost(A)}$ will not be constant, but $Pcost(A)$ will be a funct of ϕ and x .

If we ~~do~~ invest in $x = 3, 100, 150, 250$ v.s. q., then ~~Pcost(A)~~ $Pcost(A)$ will be, presumably quite large for $x = 200$ — based on the similarity (if any) of the solns. for $x = 3, 100, 150$.

2.2.01

180y:

Quint A Bort

13

What is a good set of examples for each of \leq cases of 11, 02, 05, -18.

2.3.01 Bulp

HORIZON problem

Partial v.s. linear ordering of P.D.'s → (2, 1)

The way is to not solve a horizon problem:

For a horizon of H , at time t , TM acts as if he had to ~~finish~~ ^{max} total R_t (over document) ~~over~~ ^{acts} $(t, t+H)$. "h" is horizon. At time $t+1$, it acts to optimize $(t+1, t+1+H)$

At time $t+x$ it acts so as to optimize y over $(t+x, t+x+H)$

If it did they would not work on Self Improvement. It would never work on

S.I.
Self Improvement

actual problems. It would always do S.I. first in any interval of length H , since this would help more in solving other problems later. → (2.2)

A possible way to get a finite horizon is to spend a fraction f of one's time on

Self improvement. The effective horizon is then $(\text{exp } hf)$ or $\frac{1}{1-f}$.

But "or" ("proportional") What to what? How do I get from box to "out of this"? ^{"diminishing"}

formalising

Anyway, (in general), it might seem diff. to distinguish between normal problems & "self improvement" problems. However, in my own formalism, we have logic for finite present problems using existing P.D.'s; "Self improvement" consists of improving to P.D. - of finding shorter

Coarse for it (mainly) w/o other means of "P.D." improvement.

Since τ Gödel's P.D. is not only partially ordered, this is not really a well-

-defined problem. **ABCDEFGH**

In .09, one does not get a "horizon" what is obtained is a "slowdown factor" of $1-f$ in work on the present problem. If it would take time T_0 to solve w/o

self improvement, it will take time $\frac{T_0}{1-f}$ to solve it w/ self improvement

fraction, f . " $<$ " because self improvement will tend to \downarrow T_0 needed for the problem.

Any sup v.s. that of (S.I.): Working on main problem is a bit better S.I. could help "direct" S.I. - since the P.D. is a "big" object. One can improve "parts" of it. Working on main problem would tell what "parts" to work on.

But even doing it this way, TM would never finish working on a problem!

ABCDEFGH
= bcdefg

On f partial ordering of P.D.'s with ("Quality" "Goodness"): D.I. is good for

f P.D. to solve by PC (to f. corpus) & it is good that it ever solve P.D. PC rapidly.

So \therefore partial ordering because of 2 Goals: (2) now useful to Göc by using

PC (corpus) / Time (to evaluate corpus) ? Essentially Min Cost for corpus. (N.K.T.)

well, S.I. doesn't seem degenerate to P.D. is a fund. P.D. Perhaps min total K_T for all of the problems thus far? MIN total K_T for all probs R_{ix} for? If corpus was only ENTR probs.

Perhaps look at stuff on TM's Göc. Map from Set Substructure Box.

See 2.0, 2.1 for Review of P.D. problem w. a sort of Soln.

Major Immediate problems:

1) A good set of examples for each of the "Q. About" ^{hours} 11.02, .05, .18 ^{3 types of hour}

(None are yet usually Q. About probs, but they do make clear, t. kinds of hours that Occur in R.W.)

2) In particular, 11.02 is, I expect, I. work kind of hour (change of ordly of trials, only).
So I want lots of examples to work on - to see how TM uses these hours - how it xfers them into Modulus of t. P.D.

3) ~ "TM2's Goro": In view of MCT, what is a good Set for t. P.D. ^{main cond.}?

for large P.C. for its data. This for, but also quite evaln. is imp. (4.27 has similar ideas.)

First consider EXEMMA corpus w. only DNU probs. Say P_i, T_i are t. PC & solve time for i. q. problem; then total solve time is $\sum \frac{T_i}{P_i}$. If t. T_i 's were independent P.D. \leftarrow NO!

then we want to minimize \sum and so with $\frac{T_i}{P_i}$. We could modify WAPU any of P_i individually or in subsets. Perhaps we would get most mileage by working on P_i 's of probs w. large T_i 's (??) \rightarrow (7(-01))

Some PA's are very slow (like ALP is slow but u.g. P.C. is good)

Next, consider a corpus of OZ problems only.

4) L case v.s. T/PC (6(-01). 16.08 - 11 ^{gives} t. dirty) IT

5) Each of these problems has 2 aspects: ① a bottleneck in getting TM running

② Tutorial: explaining how system works to people.
issues #1 (100) & #2 (02) are most imp. bottlenecks

① How does SOL, etc. etc. AAE problem? AA P.D. into A/D of P.C. These are probly related to TM2's Goro. T. TM2's problem.

Item 3) is of most import: The very dirty solns. may be available & adapte to.

Also of much import is applic. of L such to OZ problems.

1. Most imp.

priority is to understand how any all hours of type 1 (various) can be used to modify P.D. for L such. — for DNU probs at first; then for OZ probs.

I want to understand this & I want to find lots of examples.

2. Next in import: to understand TM2's Goro.

K_T v.s. p_c/T

$T \cdot 2R$ v.s. T/p_c

-01

The problem involves the first Gumbel House Term: That if P_i is the probability of success:
 T_i is ~~the~~ T_i cost of a trial, then best order (least $E(T)$) is
 $\frac{T_i}{P_i}$ order / ^{smaller first} In general ~~this~~ this will not be the same as $T \cdot 2R$ order, but if P_i is 2
if $P_i \approx 2^{-d}$, it is easily correct. The discrepancy occurs when $P_i \approx 2^{-d}$ is a
poor approx. P_i will be sum of codes, which could differ much ($E(P_i)$) from 2^{-d} .
(Normalization is not relevant, since we're only interested in ordering ~~to~~ — Hvr. can do it.
probys are not normal, GHTI gives a result that assumes they are normalized,
so $E(T)$ will be " \ll " $T \cdot 2R$ because $(2^d)^{-1}$ should have been normalized.

.08

In general, the errors are small!
First, is the order of trials about right? A factor of 2 in the sum. term (for the $T \cdot 2R$ method)
is acceptable.

.11

Second, is $T \cdot 2R$ a good estimator for cc of sum. (\ll Expected value)?

Actually, it would seem that one should be able to get expected cc of Lenk
w.o. considering "probability" of GHTI. In fact Lenk does ~~seem~~ to be ~~the~~ fine
w $2 \cdot \frac{1}{2^{-d}}$ to find its sum.

I've written a comment on this problem, but I really don't know where any of it is.

Perhaps after making a good list of critical ~~un~~ unsolved problems: Go thru my notes looking
for relevant work.

So your problem is TSP for Inv & Oz prob: Assoc. w. this is understanding just
how Lenk works to solve the problems — using hours to modify P.D.

Perhaps the dirty is this: Lenk really has cc but at $2 \frac{cc_i}{P_i} 2^{2i}$.

But optimum such would perhaps (like GHTI) be bounded by $\frac{cc_i}{P_i}$ which is
sometimes much smaller (forgetting the factor of 2, $\frac{1}{P_i}$ is usually ~~much~~ $\ll 2^{2i}$ (?))
(is " \ll " the right word?), ~~is~~

The "Proof" that Lenk is optimum with factor of 2 expects Lenk to be cc
to be bounded by $\frac{cc_i}{P_i}$ & this is (perhaps) $\ll cc_i \cdot 2^{2i}$

Also note the dirty of 17.12 ~ 40. It has the same implication. A human, say, would notice
that a bunch of codes were above the "sum" & use this idea to \downarrow cc of search (a lot!).

Meta Heuristic

Could a Hybrid hour deal w. both. (6.01) ff & (17.12-40)?
Perhaps 2.4.30 & 3.25 are examples of this!

01 The main goal here is to "improve P.D." Because a RD has both CC & PC in it & because certain parts of E-P.D. are more important than others, T. problem is not "simple".
 E.g. a Nerve soln, would choose a P.D. for which the PC off-computer was huge.
 T. General ALP p.d. does this, but takes too long.

03 Another thing I may have mentioned about the OZ aspect of the P.D. - that the MIN P.D. does NOT give the PROB that a given OT will be best for a given problem.

05 It gives the PROB that any particular ~~OT~~ O.T. α_i will have $G_{\alpha_i} = G$, w. $CC = \beta$, for problem P.

06 From this info, plus work on SOY, AAB, SMA, etc, we get the P.D. criteria for
 07 being best for pattern P is $CC = \beta$.

Going back to (1998?) work on MCT, we know a p.d. for various agents!
 One kind of arg. is that of 05. It may have been p.d. for arguments in 06-07 but that part is "Mathematically redundant" since it can be derived from the P.D. aspect of 05.

AB, all is not lost for OZ & ENV. P.D.'s in the case we want D.F. for the best p.d. see 37

12 A poss. dirty in using Lurch for OZ probs: Say draw over a bunch of OT's that are very similar (so that success using one is highly correlated w. using another on the same problem). Then it would seem best to spend all one's time on one of them rather than divide it up (about equally) among the set of OT's.

I ran into a similar problem w. INV problems! My impression that it would not be a problem! I don't remember why hvr! It may have been called the "First D.F. problem" I think I went thru the Mechanics of an Lurch, & it ended up making very little difference!! (Hvr, I don't see just how this can be!)

24, 30 suggests what looks like a partial, (short) soln.

21 Oh! I may see it! On the final round of $T \leq 2T$ we spend our time on the first candidate that works, & we don't do the others!

22 What about the time wasted in previous rounds? Also, if there were 8 candidates w. $\frac{1}{8}$ PC. If there are n candidates, $\frac{1}{n}$ PC each of $P_{\alpha_i}, T_{\alpha_i}$; we save time $(n-1)T_{\alpha_i}$.
 $T_{\alpha_i} \rightarrow T_{\alpha_i} - (n-1)T_{\alpha_i} = T_{\alpha_i}(\frac{1}{n} - (n-1))$; but usually $\frac{1}{n} \gg n-1$ so it's not so important!
 23 So it would be found 3 rounds earlier & $CJS \leftarrow \frac{CJS}{8}$ } This cannot improve anything! 2 problems

25 So it may well be that my old "PROOF" of insignificance was wrong! but it would be well to find it, & make sure!

Anyway, the "PROOF" is CORRECT would apply to OZ as well as INV. probs.

30 If one could somehow recognize that the candidates were very similar - this would perhaps help.

31 The fog. Seems similar to ~~the problem~~ "PC vis. a trust" problem (G.(01)).

32 A way of a PC that is not taken.

33 Another relevant term of investigation that we often have PD's that don't sum to 1 - they could sum to $\gg 1$ - like the PC that one of the soln with some PC's will solve a certain problem! Perhaps all of them will! I did finally find a way to deal w. P.D. that someone's RA was always deduced so that it was a true P.D.

37 ? (e.g. In INV probs, the PC (x) is the only one that is the best (single best soln to a problem) - 37 is probably (?) true of the P.D. for ENV probs. Also in OZ probs. set of candidates "exclusively".

Some General pts:

(Grand PD)

Re TM2G problem: Using = wts for all parts of GPD they learn a deeper default soln.

There remains: problem of partial ordering betw. PC & CC: Sum of the Least ($\frac{CC}{PC}$)

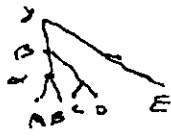
.01 ordering is adequate. Look at examples.

.02 Another big unsolved problem is TM's use of math/logic in (discovering) heuristics

Good! One way I considered was that TM first learns lots of math; then finds analogy betw. problem it's "RW" (i.e. Heuristic construction) & mathematical worlds!

.07 (01) Re: TM2G: In addition to cc v.s. pc problem, for pc problem has at least levels: One can work on a particular epistemic area (narrowing up) or one can try to find all problems. This can be done at various levels. We may think of problems as being leaves of a tree.

tree



Say A, B, C, D, E are problems. By working at "level x" - we find abcs. Common to all pc's of A & B. At "level z" we look for abcs. common to A, B, & C, etc.

We write 2 approach by directly at z by regarding A, B as a single problem "only" and C, D as another "single problem" - so z has to unify (to summarize)

from 2 "macro" problems. [So improving "pc" can have many levels: Also, one can work on another branch, if one suspends it will be used in the future.] 19.05

On the "Cure Cancer" problem: 2 parallel methods:

.20 Good! 1) Dynamic Programming: Consider all possibl. sequences of (Experiment, result) pairs.

This may be considered to be a highly branched "tree" (i.e. many branches at each node (exactly continuous))

At each node, there is an expected yield (E Goal) for the branches from that node.

(This Goal is hard to compute, but probably "dynamic programming" is relevant.)

From each node, one chooses the next node that has Max Goal.

2) Simplified Epistemic Reasoning: First, I want to understand Cancer, to initiate Experiments will be on that subgoal. (How this is an approx way to do it is unclear!)

In .20 the problem would be much simplified if the pc of each (experiment, result) was indep of what occurred in the past. This independence may be the "usual" case in Dynamic Prog.

In present case, each (exp, result) is very much a function of what has occurred previously in that sub of (temporal) (exp, result) pairs.

Actually, .20 is like feeding random Seqs into a UIM: The output string is the seq. of (exp, result) pairs. How large is a big difference!

In .20 the seq. is (exp, res), (exp, res), etc. While the pc's of all seqs are indep of all previous seqs or res's; the exp's are not indep of pc's assoc. in them.

My impression is that this is a very complicated problem: Maybe like chess - i.e. Q12, is there an alpha-beta pruning method?

2.6.01 1st: "Dynamic prog": T "cure cancer" problem: solution 26-40
 There are 2 solns. to "cure cancer" but I wrote 1. This is not a "short" soln.
 This is not a "short" soln. This may be to "short" soln.

01: 18.40: Unlike chess, but eval depends on past as well as distrib. of poss. futures. t. opponent's not mistake.
 Like chess (in our version of problem) + length of program is on team.

02: Way to simplify problem! To Goal: to have, after a fixed k moves, max prob of cure or some other N cure, or Max % curable for $C \leq 20$ day or whatever.

04: reading Marcus's paper (on 2 papers) will help. — probably: given it's approach would be wrong! (16) ← Looks like an adequate "formal soln" to a "find length" problem. (16)

05: 18.18 (TM26): Int figure of (18.134): hours of effect both A & B. $\approx \frac{T_A}{P_{cA}} + \frac{T_B}{P_{cB}}$ is \downarrow

06: My or level hours at β effect A, B, C, D, so $\frac{T_A}{P_{cA}} + \frac{T_B}{P_{cB}} + \frac{T_C}{P_{cC}} + \frac{T_D}{P_{cD}}$ is \downarrow
 So as to keep prob of hyper order, it's more likely to \uparrow to allow over cure, a lot!

Thus, hyper level hours are harder to find than low level ones: 1. My level/ones may not be applicable as often.

General Conclusion: That "Improving P.D." is a rather complicated thing; that one can do it in many different ways. However, since OZ probs. become like repeated INV probs (\equiv Induction probs) ~~via~~ via MCT (see 17.03-05 (+06,07))
 There is only one P.D. a parameter $\approx \frac{T_i}{P_{c_i}}$ criterion is o.k.

15: The: A long time series w. lots of data, could get 2/3 of wt. by way! \rightarrow (20(-01))

16 04 Consider 1 step case, then 2 step case.

1 step: e_t, r_t ; (expt, result) pair. we choose e_t w. best "expected" r_t .

We have, in this case, some simple way of evaluating r_t : T prob of $e_t \rightarrow r_t$ will be depend on ~~entire~~ previous history of e, r pairs. \rightarrow as well as "expected value" of r_t .

r_t is not a number, but it's way here nos. in it; $E \neq T$; "expected (utility) of r_t ".

r_t is defined to be t 's measurable rate for cancer patients after the result e_t from previous ~~seq.~~ seq. of e, r pairs]

So, for one step t 's soln. is clear: a each "situation" has an expected utility (Assuming one made optimum choice)
 for 2 steps, each pair. e_t, r_t has \hat{e}_t d.f. on r_t . After this r_t , ~~we have a~~ \hat{e}_{t+1} .

1 step situation, we know its expected utility — so

26 (500) Goods

Starting over: "A history" is a seq. of e, r pairs starting from beginning.

Each history has a "1 step utility". This is t 's utility of r_t if all t 's e 's part could follow it. (or t 's U of t 's e w. max U that followed that history)

On 2 step utility: Each history can be followed by an e_t, r_t .

for each chosen e_t , we have a r_t d.f. of histories, (as r_t is generated)

Each of these histories has a 1 step utility, so for e chosen e has an expected U .

We choose e_t e w. max expected U ; this is t 's 2 step utility of t 's initial history.

In a similar way, we can define n 's 1 step utility from ~~known~~ knowledge of n step utility. Th. n step utility is assignable to assoc. w. any seq. of e, r pairs.

n step utility is U of a history that has ended. it mainly depends on t 's last r .

So .33-.35 define it recall n (recursively).

It may be better heuristically to define n step utility using "h-n" where h is t 's finis no. of e, r pairs ("horizon")

(N.B.) n -step soln. to t 's cure cancer problem is for fixed horizon: not complete soln, since horizon's usually open

the reference machine as a way to accommodate new things that the system has learned.

Second: The method of computing probabilities of strings may seem a bit arbitrary, - and for several years after I thought of this method, I wasn't sure it would work - wasn't sure about the details of the method. However, I finally did work out a proof that the system was very accurate is estimating probabilities from empirical data.

Third: It turns out that if we use a universal reference machine, $\sum_i 2^{-D} B_i$ is not computable in a finite amount of time. We can't even be sure that we've found the shortest code for B! The reason is *not* so much that we can't try all possible codes: - Very long codes don't contribute much to the sum. The main problem is that there are certain short strings that one put into the machine, and the machine runs and runs, and after a long time, we still don't know if it will output B and stop - and there is no sure way to tell. That Algorithmic Probability is incomputable may seem like a strong argument against it - but it is *not* - this incomputability is an essential part of probability - of science itself.

The short codes that contribute most to the $\sum_i 2^{-D} B_i$ correspond to string regularities in the data. Consider the string $(01)^{1024}$. We can easily program a machine to write that sequence by simply telling it to write 01, 1024 times - which takes 2 bits to say "01" and 10 bits to say 1024 plus a few other bits - a vast compression over the 2048 bits in the original string.

In general in science, when you have a batch of data and you are looking for regularities in the data to be used for prediction - you can never be sure that spending 10 more minutes hunting for regularities could not give one much better than the best you've found yet.

The corresponding thing occurs with sequences that seem to be "RANDOM" - i.e. no discernable regularities. It is believed by many, that stock market prices are random - yet there is no finite amount of investigation that could be convincing on this question. On the other hand, once a strong regularity is found, it becomes very unlikely that the sequence is random.

The problem of induction is to obtain good approximations to infinity over $\sum_{i=1} 2^{(li)}$. This is done by ($\sum_{i=1} 2^{(li)}$ summing over not all of the codes). P' is an approximation to P . The more codes one finds, the closer P' is to P . To maximize P' by finding as many codes as possible in the available time, is a time limited optimization problem. It is equivalent to having a minimum error (i.e. $P-P'$) in one's estimate, P' .

The inductive inference problem and its solution as a time limited optimization problem are very important in the learning system that I'm describing. One reason is that the problem of updating the probability

Rev .12 ff

(.01) 19.15: **TM2G**: We will spend an enormous amount of Time on a SM problem (Sort of separate from main TM program) — Maybe use a different computer. So maybe give this aspect of the P.D. less wt? On the other hand, if this aspect of TM's work is able to bring in Money to Buy more cc. We should prioritize over it!

So it's not clear that "wt" is a reasonable default. . . . On the other hand, when TM earns lots of money, the T₀ ... T₁ in (9.06R) all ↓ — so all aspects of TM are "improved" — even tho TM has no new concs, no new hours in those "Not SM" evngs.

→ One approach to **TM2G** would be in (9.06R) give each $\frac{I}{P_c}$ its own WT.

These wts have to be assigned by **User**. (Might be a lot of work, but probably very imp't.

Maybe it's not nacy to put in wts "very exact" but try for factor or 2 wt accuracy.

for same & critic is probably best for TM 2G cost
ST. 01

[REV] Brief Summary of 3 diffys w. L-stch Correctly only 2!

1) The T. 2^d v.s. $\frac{T}{P_c}$ diffy (16(.01) — .40

2) Many Costs very similar & correlated w. L-stch cond: $\frac{I}{P_c}$ diffy (16(.01) — .40

3) Costs could be some diff. but is not a diff. (17.33 — .37)

Looks real
I did some work on the HRCOs
unpublished. Priz
N.B.
(6,3) is a good
Calc. idea on how
to fix these diffys.
2.4.20
may help here!

In (1) & (2) One diffy that may imply that L-stch need not be as good as other 5 rch guided by marginal hours
Other work as imp't

① Formal soln. of "Core Cramer" (for search Abnm. problem) for fixed horizon only

19.26 — .40 is Formal Soln. but it starts on 18.20. This soln. is only "formal": i.e. compas. & approxs. need to be discovered by MC w/ TM.

Also, it's for fixed horizon only, & many rch pers have zero or unspecified duration: The idea is to solve it as soon as poss. (like class) rather than get a good & soln. as poss. in a fixed time (Anytime problem y.f. 02 problem).

An approx soln: Start w. horizon h , when h is restricted decision on how horizon — maybe h into future j etc; (or 2nd More hour, then a h search etc)

② **TM2G**: Horizonal due to P.D. has diffent parts & one usually is most interested in some Spec. obsrs. — More NE

Also a formal result. of the prob (outlining of) $\frac{I}{P_c}$ v.s. $\frac{I}{P_c}$. TME because of (17.01) — .11
18(.01) — .01
19.07 — .18

③ On use of Math, (spc in hours: (18.02 — .04) used Analogy of basic Heuristic problem and MC that TM "knows".

④ RE **TM2G**: We want to order in trials w.r.t. some Cost. There are some diffys in doing this: SOY, AAE, SMA are all long term projects concerned w. this problem

⑤ 15.00ff gives major problem TM project: It looks like we're back to square 1 — that TSO is main problem: But review of status of other problem imp't. problems in TM was important. Also the recent work on "Quishab" (16(.00ff) is quite imp't! It denotes

3 classes of Hours. A total (if priced) problem is to get lots of good examples of each! I do need examples (particularly of class 1 (reordering of trials) to explain to Juryan & Marry why L-stch is a "optimal".

- 7.013 (6) Also, remember the 3rd supplements on why L search is optimal.
 - (a) If there is a better way to solve a problem, then the current heuristic is not in the P.D.
 - (b) Most heuristics are redundancy of checks, so they can be expressed as modifications of the P.D. used in L search.
 - (c) To find out how to modify the P.D. in a certain way, heuristics: derive a desc. that could have implied heuristics. This TSO can be then used to modify the heuristic P.D.

IN 4. Some say that the heuristic should. This TSO is sort of equivalent to P.D.

NB This may be true of heuristics obtained by mechanical/logical reasoning
 (d) ~~Heuristic is not a heuristic. It is a heuristic.~~
 Footnote in Sol 89: Since heuristics can be induced from any data problems, problem solving. of the past, all aspects of these things must be under control (of conditions) P.D. — So the current heuristic P.D. must have lots of possible arguments. — These last two arguments refer to in the EN of Sol 89. (9.01 that a quote of that EN)

(7) The definition of the object that L search is looking for is heuristics a P.D.M.
 See 9.15 ff. Also note the proof of the adequacy of L search: in L search (direction of K_T)
 Also note that both INUS are parts; the P.D. is part. Search soln: the other ends
 by being searching for a P.D.M. : two always mutually exclusive (except for P.D.M.) : a regular probabilistic distribution.

(8) Modifi. of the P.D. during L search (effects on the L search are not needed) 9.15-40
 (This differs from main line of inquiry in 23.09 — 25.23 : A tentative "soln":

(-01) On searches in which the P.D. is modified during the search: ("Learning from earlier trials in G. search") :-

.01 In general, TM should be able to detect all kinds of traps in its own searches, & use them .02 to improve future searches: ^{These not to dirty!} i.e. detect ^{pure} traps that occur regularly, not always to re-form

of a model of the P.D. followed by pure search ("pure" means no modification of P.D. during ^{searches governed by the traps} that search.) (Pro this is not a dirty, since the search is over the traps, not the traps has been by the traps.)

.05 Well, say the pd is \rightarrow input: description of problem: output \rightarrow pd over ppms to attempt to solve the problem. Such a pgm could be a search over trials in which ^{consistent} ~~trials~~ depended on ^{include} results of previous trials. - It could be QuikAbort or any other heuristic!

.07 A Q12 in .05, would one ever want to change order of heuristics knob, in view .08 of what happened in earlier heuristic ppms (in same search)?

Another Bla Q12 is: Can we insure optimality? Verdict e.g. that G. search will usually take not much longer than $\frac{T}{P_{eff}}$? (Lumping on PC during search occurs up this estimate! See 22.26 & 28.08-9.40; (9.23-30.1 particular))

Super finally, it would seem that .05 should be non-optimum, in view of 101-.02

Would I want to do a Meta search over ~~meta~~ search techniques for finding heuristics to find ppms? (i.e. Meta⁽ⁿ⁾?)

At some (presumably low) level of "Meta", pure search would be the closest to optimum, so fast one wouldn't worry!

To repeat/review: If I find that a particular ^{"Lynch"} search does modify its P.D. during the search, then do a hierarchy search over "objects" like the form of "Lynch".

My mind is not (entirely) clear on this matter! Hvt, it is a very impt idea & I have to have it clear in my mind ~~clear~~ 1) to use it 2) to explain it to others

N.B. This idea ^{may} also be useful in QuikAbort & all G. search methods of 11.02, .05, 18!

It may well be that .05-.07 is adequate ^{as} ~~is~~ ^{is} prob-solving Methods - any conceivable Heur! (including QuikAbort array of 3 such methods of 11.02, 05, 18)

The problem in .08-.09 is non relevant \rightarrow perhaps go to Meta⁽ⁿ⁾ levels until P doesn't change during search. I guess ~~is~~ a dirty is, but a human would remember info from previous trials - always...

Someone has a set H_i of heuristics = methods of solving a new problem. One could give each PC's via G. P.D. a new Leav; but if it is useful to remember & use info from earlier H_i trials to modify forms (or P's) of later trials, then the set of H_i's is not strictly searchable. We can, hvt, devise a How

H' that consists of trying the H_i in some list-like order but allowing rememberance use of info from previous H_i trials. It is not specified how H' does this. Search need not be used at all.

Perhaps only consider algms in which the trials are trials do not vary during search. If we had 2 sets of algms in which "rememberance" & modifying P.C. occurred during search, we would decide an algo (like H' (35)).

Hvt, Learning new Algms like H' (35) would seem to be very useful - i.e. put into the decision trials, would be equally be quite useful!

Perhaps, eventually, the system would find a single layer "hour" that would control search in a Lurch way but would allow info from previous sub-trials.

Mr. Macaulay. 23.29: ~~that~~ ^{Must} probably does allow use of info about previous trials. Yet "Pure" Lurch does not

There is a small dependence on previous trials (i.e. PURE Lurch) previous, unsuccessful trials are not remembered (trial by error type of work)

On the other hand, if TM does "learn" during Lurch (from previous trials), this should help good solutions. Quicker! — The way would not be able to use T_0/P_0 as expected cutoff of Solu.

Can TM be able to sort out "Track" ordering? (Human Lurch process: Lurch is Modest to be an Emulation ("E") of human lang.)

SO! 2 ways of dealing w. T. Problem!

1) Express to search w. learning from previous trials as a Hour (Lurch) itself.

Or we could do Lurch over several of Hours. (I do not claim my mind about this, hrs.) → See 19

2) Do Lurch but change PC's during search: this could be done immediately — producing on certain results (22.26 ff & note on line 22.26).

Or, we could do PC modifi. betw. "Rounds" of $T \leftarrow 2T$ (or $T \leftarrow 3T$ or $T \leftarrow 1.5T$)

Another view of 15: Best H' would be discovered by normal ("Pure") Lurch as "just another hour". T. Hours for discovery of H', hrs, might involve ability to "learn" TM do "Pure" Lurch & note that previous trials' info would be used. 26

SN CANDS will be "superior" for Lurch! (due to time saving tricks, but not nearly higher PC) } but not quick about first level! } To what extent does this deal w. dists. implementation (1.05 & 1.18)?

26: 19 is more general than simply changing the PC's during Lurch. It tells just how the PC's (or choice or ordering) of Cands depends on the trials of previously tried Cands.

Try to write a summary of just what progress has been made in the problems of 1.02 ff. Do I have a more or less adequate set of solutions? Start by looking at Reviews taken from 20.12 ff

SN T. practice of working learning from previous trials in a way serious relevant to the problem of "Flat P.D." at 17.12 ff is the $(T, 2^2$ v.s. T/PC or K vs. $log_2 PC$)

32 Dicky w. 19 This hour boot suffer from "Flat P.D." dicky! T. successive trials for good (H)'s are all very similar: to not use info about this to inform sequential correln. works very wasteful

38 When I tried to make P.D. change during Lurch it was to resultant I/PC as cutoff Criterion: How bad was it actually? I think it's astounding that total needed work

H' is designed to optimally use info from previous trials in some "found"

Subsec 32 for objection

times of $\frac{T}{PC}$ was way off — but was it obviously very wrong? T. ordering. on the
 still have been best possible — OR was it not "best", but simply
 a "good" (choice - optimum) — ("Good" in this case means "No Experimentally — no
 further diffy w. t. $\frac{T}{PC}$ estimate was that since PC was a function of
 fossils to get information"). [for traces of previous trials, on case / duk as more is wooding run first!

(SN) In all of G. large stuff, I've been thinking of the $T \leq 2T$ model of Lschi.
 By using Levitz ("timeshow" model, (or even "true" model) one may get a
 different perspective!

In: 01: Good is a serious business! Experiments can be vital to confirm
 heuristic processes: — Hvr, here to Good is only at a "Global" level.
 Is it less imp??
 The inability to get up. estimate of $\frac{T}{PC}$ in .01 is bad, but not critical to "optimality"
 of Lschi. — to guarantee makes it less optimal.

One General Apyt. for the optimality of models of the G.P.D. + \pm Lschi; is that?
 If all human heuristic search can be modeled or simulated in this way

So: T. variables of ≈ 24.38 is a subclass of cases that involve recording of time (s).
 In this subclass, the ordering of trials is not determined w. t. start of trial,
 but is calculated as a function of the progress of — as a function of the traces of G. Chads thus.

Re: "Good" in .01 e.g.: Is this really a diffy?! TM is trying to adjust G.P.D. so that
 any particular problem is solved as rapidly as possible. "Info gathering trials" can be regarded, not as 4 trials,
 but as activities designed to solve the problem — part of overall cost.

A fairly General form for the G.P.D.: Given when to present problem, what and
 the traces of work on it: what is the best next move (or moves)? This seems to be
 moving away from Lschi, hvr. As stated, here, it looks like a very diffit problem,
 but humans use heuristics to get reasonable solns. The AND-OR net "soln" seems
 relevant in .24

.24 could be a Spirally Near Tack: See how far can get w. Lschi for now! Also, can I make
 the Lschi system \Rightarrow it can arbitrarily configure itself? Is .24 the most General poss.
 Configuration?

.24 looks pretty much like the "cure cancer" problem (19.673 is a soln.) — Note, hvr, its not a complete
 soln. see 19.40

SN: looks like ENV prob can be usefully regarded as "02 prob! We want a soln.
 in minimal time. ("Soln" means satisfying the constraints). Hvr, its not like 02 probs in which
 one plays a card in it. Here comes out!
 NB: T. "Lschi-like" 02 problem of Gene a $F(x, T)$ is an extremal kind of 02 problem.
 "soln" is my line" problems are a way of approximating to this.

.5-
.58
.67
.615
.6156
.65
.628
.642
.675
.638
.675
.6757
.6362
.6428
.6378
.63693724
.636770356
396
2826
324211
30901
33795
31983
33091
32731
2665
GPP 97
92
.75
.848
.2996
1.1655

-01 Perhaps Marcell's remark that "finding a ^{factor} / method of doing Matrix Multiplication, was not an open problem" was an example of 25.39!! So: How do we express this as a regular (Lynch Solvable) problem? Well, for some algms, we can randomly find upper bound for soln. time, as an analytic expression. Here, comparing these expressions are not really linearly ordered. In the case of Matrix Mults w. edge n, we do know the set of bounds $2 \cdot n^b$ for soln. time. They are linearly ordered by \leq .

So Marcell's remark was not as pointless as I ~~thought~~ thought!

Actually, we can use Lynch's idea anyway: Say one did an L-search for $n \times n$ matrix w/ off of 2 random keys, example. The check for the result, could be ~~very~~ very fast, since for even possible input program, the soln. is known — It takes $T=4^3$, say time to do this, but the computation is done only once for the entire L-search. If we did a 2^b search, search some of random no. inputs for each eval, then the cc would be very large — the even so, w. a 4×4 or 3×3 matrices, it might be OK. — as I. op. for 3×3 could (be no traps!) be readily generalized to $n \times n$.

SJ Marcell (Bum): Has done much work on cc of "checking" if $y \in f(x)$.

Says its often much easier than computing $f(x)$. I have to look work in "Opera" on P vs NP. Extreme example: $f(x) = \text{factors of } x$. Is this \equiv trapdoor function? "non-trivial-checking: unknown" or: find $x \ni M(x) = a$; a is known. This can be a hard P vs NP problem but easy to check. Well, maybe a bit backwards! Perhaps to programming.

Given $P(x)$ and y , to find $x \ni P(x) = y$. No: it seems more complicated! I'll have to read this more carefully.

1098
+22 5970
1622-1427
2570
1716
1645
7470

23 ~~5.01~~ 5.01 - .40 is a kind of review of what needs to be done (what has been done) in TM.

7.02 - 24 is another review of the essential drifts in the TM system.

11.02, 05, 08: T. 3 kinds of floors seem very important.

11.02 is further divided into stages w. constant P during stages & those in which P varies during each, due to observations of traces of carrier counts.

27 For 11.05 (some ordering of trials, but changes in cc of trials: a grouped TM c/s speed is use of 11 machines — Phase exp. was very interesting.

Quick about ~~is~~ interesting, is I'd like to give some examples & (if possible) refer to Gen. 16.

11.18 is the hours between mix of 11.02 & 11.05 — 11.02 11 c/s is, perhaps an example but I know of no other examples. ^{Does} Lynch normally use Q variables? It can use $\frac{cc}{pl}$ as a cutoff criterion, rather than do all trials "fast out" or for a fixed $cc \leq c_0$ (where usual $c_0 \rightarrow 2c_0$)

What I can do next is look at a bunch of probs & associated hours; Try to find out how common to 3 classes of 11.02, 05, 18 are & c. 2 subclasses of 11.02.

Before this, hwn, write a review of the "flow" of TM: How it is supposed to work, & what parts don't work so well, & "partial fixes" that may be fairly good. 7.02 - 24 was a start. This Actually print this up: initially in Zakopany — but there will be few copies and exp.

(.01) **SN** Quick about: If we used true search time of 2 Gons for the G-PD, then QA Alt would be recognized as good. MVR, using ~~the~~ $\frac{1}{P_{ci}}$ alt/corpus - which is much greater (less cc) to compute, - would not recognize Quick Alt as useful.

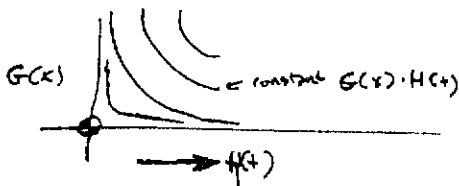
(.01) **SN** In $F(x,t)$ "oz" probs! Try: Do all $x \ni F(x,t) < F_0$ (F_0 is small)

Then do $F(x,t) < F_0 \geq 1$ for successively higher values of n .

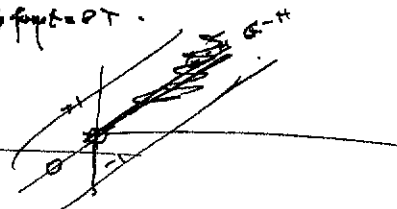
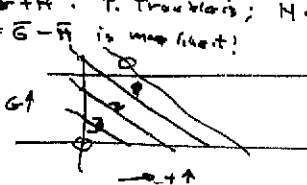
Say $R(x,t) = G(x) \cdot H(t)$.

$\bar{G} \equiv \ln G$ $\bar{H} \equiv \ln H$ $\bar{F} \equiv \ln F$

$\bar{F} = \bar{G} + \bar{H}$. T. Transforms; H is \downarrow for $t \rightarrow \infty$.
So $\bar{E} = \bar{G} - \bar{H}$ is more like it!



No!

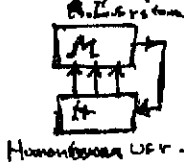


(.10) **SN** When I was considering the possibility that "A Life" systems would eventually "eat" out of the machine" into RSM! This was via the interaction of a "A Life" system with the user. Here the user was a "SuperUser" who was developing, modifying, & AL system.

This amounted to a RTM system in which the User was the RTM that was being "run" by the AL system. The Q is: Is this last mentioned system a "full"

RTM w. large horizon, or does it have a horizon of 1 only? - IFF this is true, the system may not be dangerous. The M's output becomes oriented

toward getting the Human to make M ob-fus. in M, that will achieve Goals of M... whatever they may be.



The M's output becomes oriented toward getting the Human to make M ob-fus. in M, that will achieve Goals of M... whatever they may be.

(.12) **IF M is "open" to Human M has more control over the M should try to get it to "look inside" it (i.e. "openness")** → SG. 81

If we think of M as Human in the system, then the "open" system would be a symmetrical & it would be clear to "who was manipulating who".

SN In words GA that uses "Spice" to implement a Gate for electronic chps! Spice book
 (2) F. bottleneck in the system: ~~the~~ w. 5 seconds for an evaluation! We need to simulate better approx! This can be done by "factoring" the chp. into parts that have known, simple responses. ~~the~~ A transistor chp would act like a $\frac{1}{s}$ for a large part of its I/O ~~domain/range~~, a diode would act like a $\frac{1}{s}$ (a diode plus a resistor, etc. Instead of very incremental time (as in Spice),

Complex flow domain would be used for analysis of frequency behavior. Essentially,

the chp would be described as a ^(Engr.) ~~(Human)~~ thinks about it. (Remember, in Lonke's work, he found that the "slots" that were devised to facilitate Human understanding of the system, were of much use by the system for debugging & applying hours

My interpretation is "memory" of Lonke.

2-14-01 TJM REV. This Paper deals with (to a greater extent) ∞ problems.

The TM ~~is~~ system: TM is a general problem solver. Its inputs are the description of a problem, in some formal language; Its output will be either a time, or a solution, or a statement that no solution exists, or sometimes no output at all.

The kinds of problems are in 2 large categories: Invariant problems and Optimization problems.

Invariant problems are easiest to describe: [use text from Sol 86, 89 but get more examples. Trav. Sales Prob., all P & NP prob of computational complexity theory. These are all problems in which checking a candidate solution is of relatively small computation cost.

There is another kind of invariant problem in which checking whether a candidate is a true solution, is very expensive. These problems are dealt with in a much different way from "ordinary" Invariant problems.

Example: find a route to max of $5mx + (1-x)^2$

Optimization problems come in many forms. The least simplest form's with

Comments we want to have a function $M(x)$ that maps things into real nos. We want to find a value of x such that $M(x)$ is maximum.

Various constraints on the form of M and limitations on the computation cost of the solution, give rise to many problem forms; which have to be solved in essentially different ways.

The simplest form is one in which the function $M(x)$ is known and can be very easily (quickly) calculated. The time constraint is that we need to look x possible but we only have time T_0 to find this x .

Another (not so simple) form: The function $M(x)$ is known, but it takes a non-zero time T_1 to calculate: $\frac{T_1}{T_0}$ is not extremely small; say it's 20.

A slowly evaluated problem type: $M(x)$ is not known, but we can "ask a teacher" what $M(x)$ is; we are allowed to do this only a fixed number of times; (say 50) during the problem solution.

Another form: $M(x)$ is known, but the values of $M(x)$ we obtain have no. 30 in them. ~~Example~~ e.g. $M(x)$ is real world measurement with imperfect instruments.

Another form: We can ~~measure~~ obtain $M(x)$ from a function quickly, but $M(x)$ varies with time and we do not ~~know~~ know ~~the value~~ the form of this variation.

Try to give ~~examples~~ examples of each type of problem.

Any two problems are a variation on it, it is on several of the other.

The Maximization of $M(x) \cdot f(x)$ is a very difficult, though not unusual kind of optimization problem. Here $M(x)$ is known and is purely computable. It is the time at which the system presents its solution: $F(x)$ is a known, monotonically decreasing function.

Happen Deals w. expensive tests in Inv prob. is not clear. If test recites just Yes/No, then the student may be little info! Checkers/Chess seems to be an exception. W. expensive fast in Inv prob. TM can handle it. faster w. less accuracy & less cost. major progress, w. no dependence on the cost of computation.

Also Note: Manual solution has done worse score "exercise" $x^2 + 5mx = 1 \pm .001$. Website.

03

14

38

- 01 : Show how inductive problems are $O(2^k)$ problems (find a set of strings \rightarrow \square)
- 02 : $M(x) = z \iff z = |x| = \max$ in some fixed time, To or as an "any time problem"

-03 In both Ind & $O(2^k)$ problems, the object sought is a string or number or a program for finding

-04 a string or number (Both are examples of "Programs")

Say $M(x) = z$ is an Ind. problem. We can give test $M(\cdot) = z$.
 A/direct for given x values: A program for x use next machine M_r term @ log property
 $M_r(p) = x, M(x) = z$: So $M(M_r(p)) = z$: which is a kind of INV problem.

M_r is some standard Turing Machine. M_r is usually universal or almost universal.

Perhaps this is stupid! We must ~~not~~ to my code in x PC order, so if M_r is universal, then putting codes into M_r in order of length will give outputs in x order.

So, given $(G.P.D., P(M_r), z)$ its output diff. is on x strings that ~~are~~ most probably hard to generate. Time for x problem $(M(x), z)$. This means that outputs are mutually exclusive — (But it may well be that almost all "Coin-bits" $(M(x))$ are about

x . Show: Hur. Generating x can take various times. So is $\frac{1}{2}$ reasonable? $\frac{1}{2}$ is not a reasonable number!
 $P(M(x), z)$ is the prob that a particular coin x is 1 a coin to $M(x) = z$ 2 that of z legal coins, to process of generating x and testing it via $M(x) = z$ is min among all coins. (Had it we use def. 3.1.21 for a "CAND")

-20 ~~Yes~~ No: The soln to $M(x) = z$ often is unique. So perhaps P is notion x , including testing.
 -21 but x probly that x program will generate a coin (many soln exist) z will take min time to do so.

-22 T. G.P.D. discussed above is derivable from number P & which is x probly that z program will generate x to $M(x) = z$ test. This Aux P.D. is probably easier to update. — The 1. P.D. into USA M Lsch is true that we have a "Goodness" criteria for : 500 [8.07]
 -23 I have to be quite clear on this pt. before proceeding much further. Study lots of Examples
 (Source: 3.1.21, 3.30) for soln.
 [8.07]
 [Source: 3.30]
 [Source: 3.30]
 [Source: 3.30]

Is G.P.D. \square a P.D. on decms of x ? — x probly that $M_r(p)$ will solve $M(x) = z$ in min time? \rightarrow It would seem not $\frac{1}{2}$ since short decms ~~are~~ \rightarrow (not decms w/ short decms) \rightarrow generated by PC's.
 Let's go back to simplest case of Lsch: given $M(x) = z$, to find $x \rightarrow M(x) = z$.
 Then $M_r(p)$ gives a def. on coins for x (R is ~~infinite~~ \rightarrow in finite log random lang).
 When M_r stops \rightarrow output: M_r is a UED machine. So is x P.D. x probly that x could be soln? If x soln is unique, x ~~is~~ ALP induced by $M_r(\cdot)$ is a legit. P.D. z could be x : If there are several coins or no soln, it's not clear that ALP $(M_r(\cdot))$ is an appropriate P.D. for x soln.
 In an earlier (Note in [8.5]) I felt that it was clearly x \rightarrow \rightarrow to find x , that x G.P.D. was \rightarrow best! This seems reasonable! When I made a stack lang to generate coins for x GA problem, R \rightarrow associated with P in this way: T. \rightarrow \rightarrow soln was $P(x)$ \rightarrow total pc's of all x \rightarrow \rightarrow generated \rightarrow (this is an universal P.C.)

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In the proof of adequacy of Lstch: we consider some (arbitrary) inverse of P_C - say $f(x) = k$. — If $P_C(x)$ is assigned to k , then it will take $\frac{P_C(x)}{P_C(k)}$ time to find k exactly for "Threshold".

to us by Lstch.

So maybe 29.20-21 is correct (= O.K.), i.e. probably that program will generate a soln, & do so in low time whenever a soln. If a pgm is "loop free", then time for execution is related to $(\alpha \cdot n)$ pgm length (= no. of insts). If there's 1 or more loops, this is not > 1 true — total execution usually depends linearly on how long each loop was run (how many occurrences),

i. cond. $K \equiv M(\alpha, a, M(C))$ α not deriv of α .

We want $X \ni M(K) = 2$; $M(M(\alpha, a, M(C))) = 2$.

Quick Abort can be non-terminating proof of α

How P_C 's are related to .10: If there is only 1 soln, X_0 , then P_C assoc. w. this X_0 is $\leq 2^{-2(\alpha)}$ for all α satisfying (.10 R): \exists

If there is > 1 X values, & X_j is one of them, then P_C assoc. w. X_j is $\leq 2^{-2(\alpha)}$; X_j is \exists sth $\ni M(\alpha, a, M(C)) = X_j$

In general, P_C proby assoc. w. α , will not be $\leq 2^{-2(\alpha)}$:

α will be generated by \geq stack long. If α lang. is unambiguous, P_C of α will be obtained directly from Grammar (if α is parseable); Otherwise we obtain P_C of α by summing over P_C 's of its derivations in α Ambiguous Stack Grammar.

If we don't use a stack Grammar & use binary trees in .10, ~~it's still true~~ the P_C 's $\frac{T}{2^{-2(\alpha_0)}}$ (where α_0 is α ppm. Not the smallest L cost $\frac{T}{2^{-2(\alpha_0)}}$) will be much smaller (properly) than the true time for α . This is because $2^{-2(\alpha_0)}$ stack be "normalized".

We use Kraft inequality to get $\frac{T}{2^{-2(\alpha_0)}}$ upper bound. If we want a better bound P_C Kraft equality (Normalized values) this would be exact (not an upper bound).

~~But we can't easily normalize it in normal usual L form, because for the unsuccessful trials, $\frac{T}{2^{-2(\alpha)}}$ is usually~~ Suppose we || α . S. search, w. very small ΔT before jumping to the next α . T.S. searches \geq || each. If no cond had failed at t time t soln. was discovered, then search time would be $\frac{T}{2^{-2(\alpha_0)}}$ exactly. Since α would have Kraft equality. If any cond failed before t soln. was found, then total such times $< \frac{T}{2^{-2(\alpha_0)}}$.

Some upper is true if we obtain P_C by methods other than $2^{-2(\alpha)}$ (e.g. stack grammar)

So we get search time $< \frac{T}{P_C(x)}$

omit (confusion betw GPC (α | $M(\alpha, M(C), a) = 2$) and...
Confusion betw. $2^{-2(\alpha)}$ & $P_C(\alpha)$: In both cases, α | $M(\alpha, M(C))$)

$P_C(\alpha)$ is derived at $M(C)$, a is not an "input" system.

If α was obtained from GPC & output a P_C , then $2^{-2(\alpha)}$ is not interesting for Lstch.

.38
.34

user virus

- .01 : If 30.38 were true, then its hard to see that $d(x)$ would be of any interest at all!
- o ● If α were generated in a more "random" way, it could be assigned a pc of $\sim 2^{-R(\alpha)}$.

In 29.08 ff It wasn't clear in my mind as to just what the "L such in TM was looking for! Also, when I used the problematic format L such (using e.g., GPD (Grand Prob. Distrib.) — it wasn't clear what GPD was the prob of — whether $GPD(M(\cdot), \alpha)$ was the P.D. of α soln to $M(x)=\alpha$ or a P.D. on the pgm that would generate α .

In 29.20 ff. f. (decision/cond) cases. as GPD was a P.D. $GPD(M(\cdot), \alpha)$ was a P.D. on pgms to counter α from $M(\cdot)$ & α . [In general, however, α may be a P.D. & α may not always have a soln; because for the P.D. α , α may not have a P.D. — which selects pgms in favor of both $M(\cdot)$ & α . If TM remembers a soln to $M(\cdot)=\alpha$, (i.e. x_0) it will output the simple function with outputs x_0 for all inputs.

If we use GPD for L such, its output is a P.D. on strings/pgms; α . If α is usually f. case) GPD is a search grammar: ~~as will be used~~ we will use various things to get the P.D. to give us α values in the PC order (by PC first). I have solved this problem many times: T. (start Jo in. I remember as being particularly good! — I think I used Huffman Coding. It was in Bog. 2000): look in indexes to find it — I think in Early Buf. I know it works for a (some) processor — I don't know if it would work for a CFG or CSG. I think I wrote some analysis of its approach for CFG.

21 ● Hvr, more generally 36.37 + outputs GPD for a ENV. problem can be any P.D. that depends on α . (might have noted: trying to do non-probability to various degrees)

could solve the problem (this is analogous to the OT's but GPD outputs for solns to OZ prob.)

It can be an elaborate general technique for problem solving like converting α .

ENV problem to a ~~search~~ optimization problem or G.P.S. (which may be a vector Gove for Hill climbing), of α , β search or whatever. It can be a search technique — like, or

26 ● The idea is that any problem solving technique is a legal output for the GPD. (summarization L such. See 40.01-16!)

Note that 21 is combinatorial. 30.10: If α is the string that GPD is the P.D. of, then

$$\text{number of } \alpha \ni M(M(\alpha, M(\cdot), \alpha)) = \alpha : \alpha \text{ (solution to soln. problem } M(\cdot), \alpha.$$

21, 26 may cause trouble in my place of how situations being realizable by a modification

30 b. GPD. A Hour may want to modify subsets (or any other aspect) within α , that could ~~solve~~ problem soln. — Well, practically, this would be covered by Modification of ~~GPD~~!

To see this: Say the hour modifies $\alpha \rightarrow \alpha'$. This means that α' now has a pc of α & α has a lower pc — ~~then~~ α' into here had a very low pc before the

36 ● application of the hour $p \rightarrow p'$ — so low that we didn't even generate it in the GPD. So practically 21 ff 26 ff is 30 consistent a Change Note! — Suggesting

that the ~~the~~ most (most) of how problems can be expressed as changes of the GPD — even if the output of the GPD is a very general kind of problem soln.

Horizon: A not bad approach! (But See 37.22 for Deep Criticism!)

-01 \rightarrow Clearly in steady state reinforcement system is not the same as finite horizon, in the long run. \leftarrow
Steps toward a steady state soln for a recurt. problem:

Suppose we have a corp which we have a stockholders report every T_0 . So we
start at $t=0$ with horizon = T_0 , work until T_0 , loop back.

w. horizon = T_0 , we find it best to work on "Self-improvement" upto $t = (1-k)T_0$,
then work on R&D, producing projects (direct problems) for $(1-k)T_0$.

To modify this for "steady state w horizon T_0 ", Time when we work on R&D or
Self-improvement; fraction $1-k$ on direct problems.

T. saying we have to solve "50% solution" problem: Instead of 50%, we
use k .

Another way to look at .03 ff: At each we have to choose a fraction, k :

Choose $k \Rightarrow$ The expected total yield at time T_0 is max. We could vary k
slowly vary k as our state of knowledge & quality of projects change.

Also, it may be that k may need to vary T_0 , depending on Political (external)
conditions

But a main idea, is that using k 's "can't be off by more than a
factor of 2" vs. "old 50% soln." idea.

So, as before, my ~~same~~ conclusion is that this problem is not critical, it
is a present time, it should be spending time on it, unless you have
some Great New Ideas!

for finite horizon: (This is a common type of problem: "Anytime" is also common)

Say S.I. work produces $\frac{1}{100}$ improvement in direct prob. solving.

So to work $\frac{1}{100}$ time T of S.I. gives $\approx e^{\frac{T}{100}}$ increase in Q .

So ~~work~~ w. horizon T_0 , we want $T \leq kT_0 \Rightarrow (1-k)T_0 \cdot e^{\frac{kT_0}{100}} = \max$

$(1-k)T_0$ is amount of R&D gets in time $(1-k)T_0$: say $q \propto k^p$; $L_0 \approx q T_0 \frac{e^{\frac{kT_0}{100}}}{100} - \frac{T_0}{100} \cdot q$

So $\max q e^{-\frac{T_0}{100} q} = q e^{-\alpha q}$ $\alpha q \geq p \quad \max p e^{-p} \quad \alpha q = 1; q = \frac{1}{\alpha}$

$\ln p - p \quad \frac{1}{p} - 1 = 0 \quad p = 1$ So for best yield, $\frac{T_0}{100} \cdot q = 1 \quad q = \frac{100}{T_0} = \frac{1}{\alpha}$

$k = 1 - \frac{100}{T_0}$

More generally, if working on S.I. for time T produces $\frac{1}{\alpha T}$ yield,

$k = 1 - \frac{1}{\alpha T_0}$
 $1-k = \frac{1}{\alpha T_0}$

$k =$ fraction of time worked on S.I.

Wooes! $0 \leq k \leq 1$: $\frac{1}{\alpha T_0}$ can be only large.

- 01: 32.10 Backsol. "Sog Soln" problem: consider $(32.29-k)$: $(1-k)T_0 \cdot e^{-kT_0\alpha}$

What's the difference (ratio) betw using $k = 1 - \frac{1}{\alpha T_0} \approx 0.5$? $(1-k = \frac{1}{\alpha T_0})$: $(1-k)T_0 = \frac{1}{\alpha}$

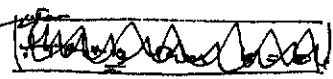
$$\frac{1}{\alpha} \cdot e^{(-\frac{1}{\alpha T_0})T_0\alpha} = \frac{1}{\alpha} \cdot e^{-1} = \frac{1}{\alpha e} \cdot e^{T_0\alpha}$$

for $k = \frac{1}{2}$ wage $\frac{1}{2} T_0 \cdot e^{-\frac{T_0\alpha}{2}}$

ratio of optimum yield to $k=0.5$ yield = $\frac{1}{\alpha e} \cdot e^{T_0\alpha} \cdot \frac{2}{T_0} \cdot e^{-\frac{T_0\alpha}{2}} = \frac{2}{\alpha T_0} e^{\frac{T_0\alpha}{2}}$

so if $X \approx T_0\alpha$; it's $\frac{2}{e} \cdot \frac{1}{X} \cdot e^{\frac{X}{2}}$: if $X \ll 1$, this is $\frac{2}{e} \cdot \frac{1}{X}$

or if $X \gg 1$, $\frac{2}{e} \cdot \frac{1}{X} e^{\frac{X}{2}}$ can be enormous! May be same.

Whoops! X could be < 1  over! So our optimization is only for $X \geq 1$; if $X \ll 1$ then the max of 32.29 is at $k=0$

This means (usually) that we should $\uparrow T_0$ (5 f. horizon).

Also, for large X and T_0 , the ratio is very large

The Sog Soln. method, but if one had twice as much time T_0 at $k=0.5$, one would do at least as well as run T_0 w. optimum k .

(While our yield can \uparrow enormously w. use of optimum k (v.s. $k=0.5$) we can get

\rightarrow least \approx prod yield by doubling CPU speed & using $k=0.5$. \rightarrow 24.06



Also note, even if α is small, one can always get X to be large by using a large T_0 (!)

Well, we can't! The first (long) period of TM's life is spent in S.I. (not working on serious problems). — However, in the phase of TM's "life" could we not have a Meta S.I.?

— Maybe not, because TM = TM already!

How. In the initial phase of TM's life: All probs are "study problems". My original idea was to get how TM work on S.I. until it could do so w. some efficiency. At any pt. in TM's life, it has a certain value of α ($\Rightarrow e^{-\alpha T_0}$ is S.I. failure rate during time T_0). $\frac{1}{\alpha}$ is in a sense of a "Time".

So $\frac{1}{\alpha} = T$: so we need a "horizon" $> T$ (so $\alpha T_0 = \frac{T_0}{T} > 1$) before its useful to work on S.I.

on meta S.I. idea! Say the goal is revision of QM: wasteful w. S.I. only:

Lower meta, some can virtual physics: How, in meta physics, there are meta topics: How to do research efficiently: various meta hacks. We can define to any depth, but to deeper

we go, the fewer problems we can usefully solve (I mean problems)

Meta S.I. is equiv. to $\uparrow \alpha$ of primary S.I.: so $e^{-\frac{T_0}{T}} e^{-\frac{T_0}{T}}$

here T_1 is time spent on secondary S.I. $\approx T_2$ into horizon.

probably $T_1 \gg T_2$, so normally, it would pay off to do secondary S.I. — because paper horizon work $= \exp(\frac{T_0}{T_1}) \cdot \exp(\frac{T_0}{T_2})$

be very large.

-01: 33.40: It would seem that since there is only one GPD for all orders of $\frac{1}{2}$ problems, there would be no such orders! — But it may be that the utility of GPD is not so complete! When our "improvement P.D." and decision variables go improved (see TM 26: ~ 18.01, 07, 19.05, 20.01,) 20.03: ~~improvement~~ wts. for TM 2 (E.S.E.) have to be supplied by user.

06: 33.19 suggests that if "soft soln" may not be so good! — compare $\frac{1}{2}$ of $\frac{1}{2}$ of a goal! In fact, in initial TM exp. $k=1$ (all time spent on S.E.) $\left(\frac{33.21}{\text{on time}} \right)$

Perhaps in General I will simply design a sys. to incorporate all the S.E. & hyperlevel S.E. Certainly I do it at the outset. I don't yet have a steady-state model or every variable goes now well into the EQ. It's definitely a "more different pool".

Consideration of Necessity Short term Goals for a RW animal: It has to stay alive!

Needs rapid response to threats — "accidents"; slower response to get food; Much more rapid response to get Air. So Order of needs: Air; food & water; ^{fast} reproduction.

Also, steady off Production; Accidents (falls) — Near top of list for rapidity of response.

The Goals of Tom will be quite different. I want to be able to control food, air, predators.

I don't want it to reproduce (or not do I want to control it).

To top goal is to solve very brief problems that I can't do. — Other Goals: to solve other problems that I can't do.

In animals, ~~steady state~~ the "fast response goals" are vital, & give quite feedback, so they are learned quickly. This knowledge is then a good basis for further learning of a more complex kind.

In order to do S.E. & Meta S.E. TM has to have a adequate SSZ. For hyper order meta S.E., this SSZ has to be very large. So perhaps one should not attempt hyper level

24 S.E. until the SSZ is adequate.

25 So: Reviews Main pts. thus far: study

Ag. 37.10
See 37.22 for Summary Criticism!

1) for analysis of S.E. optimum fraction of time to spend on direct (not S.E.) prob is $\frac{T}{T_0}$!

$T = \frac{1}{\alpha}$ = amount of time spent on S.E. in order to multiply by e^{α} Self improvement

T_0 is "Horizon" = amt. of time to do the work.

$$\frac{T_0 - T}{T}$$

$(1 - \frac{T}{T_0})$ is fraction spent on S.E.

Total time spent on direct problems is T (S.T. $\frac{T}{T_0}$)

$$\frac{1}{x} e^{-x} = 1 \text{ when } x=1$$

" " " " S.E. = $T_0 - T$.

If $T \gg T_0$ we spend no time on S.E.

Our utility is (Utility of max. direct prob only) = T_0 .

Utility of optimum return = $T \cdot e^{-\frac{T}{T_0}}$ Value $\frac{T}{T_0} - 1 = \frac{T}{T_0} \cdot e^{-\frac{T}{T_0}}$

fraction of utility optimum allocation to zero allocation S.E. = $\frac{T}{T_0} \cdot \frac{1}{\alpha} = e^{-\frac{T}{T_0}}$

$$= 1 \text{ for } \frac{T}{T_0} = 1$$

value of " " " " to zero allocation prob = $\frac{1}{\alpha} e^{-\frac{T}{T_0}}$.

" " utility of optimum to 3 allocation = $\frac{T}{\alpha} e^{-\frac{T}{T_0}}$.

35.01

21.17.01 : TM : 1 Bulg

Horizon: R_{av}

7.01 : 34.40 :

S.d. is useful $\Leftrightarrow \uparrow < \uparrow_0$.

Note S.d. can $\downarrow \uparrow$, so it can be very useful, but it is a "meta hour" & not used very often.

SSZ for higher level hours (E.g. 2) \downarrow rapidly w. order of hierarchy. Also for

ok π values for these ~~for~~ higher level hours are very large.

Horizon

Zat
Zay
Zak
Zal

01 : : More on "Horizon"; Say we modified fi problem in an attempt to get to "steady state"

Sofn. Then if we ~~increase~~ transfer work on S₁; a fraction k of t -time, then as we accumulate S.E. work, "direct work" efficiency will continue to \uparrow . R is rate of work production.

$R \rightarrow (1-k)R \quad z = \frac{1}{k} \cdot \frac{1}{k} \int_0^t k = \frac{1}{k^2} \quad ; \quad R \rightarrow (1-k)R e^{\frac{t}{k}}$

03 : And total $R = \int_0^t ((1-k)R e^{\frac{t}{k}}) = (1-k)R \frac{1}{k} e^{\frac{t}{k}} = \frac{1}{k} (1-k)R e^{\frac{t}{k}}$

This is certainly not "steady state". For any small k , k is close to 1.

07 : γ is equiv. to maximizing $(\text{or } R)$ at $t \rightarrow \infty$. \rightarrow 37.10

From app., the goal of a ^{optimum} steady state ~~is~~ (imp. is unclear) -- Maybe Meaningless!

Anyway, a perhaps useful goal for a horizon T . A very large, diff't problem:

Say "Cura Cadaver". It could be defined as a OZ or INV problem!

OZ problems: the good & % come ^{within} on time to 25 or 75 or 50.

15 : INV : Get 80% done as soon as possible.

It seems to be severely distant from most INV prob! However, t -satisfaction criterion

is expansive, so modeling problem. Criterion would be yet to be soln. technique.

Also, t -usual L such, would be ridiculous! It would seem that it would be hard to

Modify to GPD a 6% so that the goal would have a reasonable P.C. -- To do this

"Improvement of GPD", TM would have to do "experiments" -- which is difficult

"Out of the spirit" of my usual concept of an "INV problem" -- This seems to be a difference not covered by the ^{My} Expansion of the "to the bottom"

Another Diff. INV problem: To prove/disprove Riemann's hypothesis (off-formal form or Goldbach Con.) : Have, it would seem that we would want TM to learn much Math, Acad.

This would seem to be an open, intrinsic spirit of usual INV. prob.

Before TM could approach such a problem, it would have to be trained (by the \geq constructions)

~~by programming~~ One way would be by project successively more diff't prob in areas of Math that seem relevant.

This last idea is conventional to design. What ambiguity about it where

30 : TM decides that it has to learn more & starts reading books Math books.

In 30, I would not expect TM to be able to do this sort of thing w.o. a degree of

$\frac{1}{k} \approx \frac{1}{k}$

Also Note for most INV prob, Neural Comput Plan to OZ prob than one hill climbing or G.P.S.

Perhaps INV prob should be solved "like" OZ prob, in some that is a P.D. over

Methods to solve INV problems solve to OZ prob over, similar given by G.P.D. known a P.D.

or OZ's - which are methods to solve OZ problem.

.01: 33.40 = It might be possible to have TM to work on their part & any I do - use factors as literature.

But a "valid" way would be to look in books, reference books & other literature

I'm not sure I understand the difference between 26.37 for QUV, & looking for "you" to solve for problem: the "you" to solve for problem could be to convert it to an ODE problem &

Solve it in a special way - or it's a GPS problem, which, perhaps, we can rephrase

a general. of ODE prob, that has special ways of being solved. (= special part of ODE).

So, I have to give examples of various kinds of problems & how they are solved

What heavy duty work is how these hours are implemented by the GPS.

.10: 36:07: Amore negative relationship betw. W_T & " $\frac{dR}{dt}$ " (i.e. value of growth rate R.)

We are interested in (as first guess) of $\int_0^T R dt = \int_0^T \frac{dR}{dt} dt$.

In 36.03 $\frac{dR}{dt} = (1-h)R \cdot e^{-\frac{Rt}{T}}$: Amore remarks $\frac{dR}{dt}$ would note that

it would depend a lot on the total out-of-date in previous year (partly via SSZ considerations).

Should this be simply ODE? ("By definition": "it" could be defined this way.)

While $\frac{dR}{dt}$ would be with (data for total SSZ), it only ↓ w/ t because it took more time to analyse more past data.

Hrs on it whole, I'm quite uncomfortable to ^{comment} ~~comment~~ ^{usual} form of how $\frac{dR}{dt}$ depends on ~~total~~ "t" or $\int \frac{dR}{dt}$ or $\int R dt$.

.22 This is also assoc. w. criticism of my assumption that TM would always work on S.I. first (before working on main problem). In many, maybe most cases, TM would want some experience w. t. problems, before working on new things to improve those prob solving techniques!

The great exception to the above would be mainly what I was thinking about in (i. first place: e.g. Understanding QED Gen Relativity; TM would have to know a lot of physics, math for the problem could be optimal for it).

In "Every Cancer", again, it probably would be a good idea to know what was "known" in Biology, before starting work on the Main Problem -

So, in General, my idea of doing S.I. first maybe O.K. sometimes, but often, not!

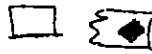
Move on .22 ff: In a hard problem, say one wanted to familiarize oneself with general area of the problem or w. its parts etc. problem, before starting working on S.I. - since this preliminary examination would have no chance of solving the problem, it ^{sh}ould perhaps be regarded as pure S.I. (?)

-.01 : 37.40 : Hvr, usually, I reject S. i. as being "Improvement of GPD", so it would not include "preliminary sketches w. t. present problem". It would perhaps include "preliminary sketches" in t. approach 36.37-37.04: Part of 2. "Method/plan" designed to solve t. problem. →

(Perhaps) SM Improving t. GPD would seem to not include OSL. Perhaps to do OSL in a problem, one has to do a special search of over post corpus.

.08 → Actually: "Improving t. GPD" has 2 aspects ① "updating" which involves (varying amount) t. post of t. GPD compared w. t. PROBLEMS JUST SOLVED ② A "More general improvement of t. GPD", which involves more problems of t. post (i. perhaps in certain TM models, & a part of future PROBLEMS).

e^x/x



ABCDE ABCDE
 ABCDE
 ABCDE

SPR
 .01: 31.40 The idea of 3), 2), that could α , is a very general kind of prob-solving algm, if can give rise to α 's that are very highly correlated in the likelihood that they will solve problems. This conceives much waste of time in search (as was previously noted in 6).

.02 "first GPD" problem (Sect 17.12-32; 24.30, 20.15, 16.31)

.03 Hm: Say 6. ~~FAIR~~ α 's in the GPD were correlated (so that success or failure of one tells some things about success or failure of several other cands. — I think this is normally the case. Consider 2 Gener. of G. [G.N. Thomsen I] for correlated cands!
 But the PC's in the GPD are for mutually exclusive cases! — So they can't be correlated!?
 So first I have to clarify what I mean by "correlated" for a multi-exclusion d.f.
 Intuitively, I mean (0.5-0.4)² Hm, for Lsrch, only failure info of a cand. can be used to influence prob of other cands (?) Or info that a cand. was run for time T is not yet a win or loss.

.13 A try for Meaning: Say we have ^{constant} mut. exclusive pd. or cands, $P(x)$:
 Assoc. w. PC's on expected cc to solve (counted overall poss. X — using best poss. strategy). — ordering formalism $P(x)$. Call this $P(x)$.

.16 So perhaps we first choose cand α , $\Rightarrow \frac{pc}{cc}$ is max. This gives us a new P.C.) mut. exclusive cands, GREEDY
 .17 from it we choose $\alpha \Rightarrow \frac{pc}{cc}$ is max, etc. This is a greedy method. } — but see 44.23!

Using .13 in Lsrch, probably wouldn't help! If we have a bunch of similar, highly correlated cands, then none of them are finished in early rounds, & chosen round early rounds that we use all that cc in testing them. This situation was analysed before. (see (.02)² for refs.)

On second prob, say to bunch of correlated cands. yes; ultimately, tailored: when one fails, the rest are given low pc's, & are not tested until much later (say after the true soln. has been found). This \Rightarrow seems to be QuithAbit!

.25 One way to deal w. it: correlation problem Make up random sets of unequal class sizes of cands that are highly correlated. Select 1 (best) cand from each class. Assign lower pc's to rest of class in each class. This is a kind of "best" & works solution problem of "correlated cands", but there is to pass that apart for "Best" cand in each class fails but one other in the class really solves the problem? While this can be made unlikely (by suitable class size), we still have to passy of another soln. occurring in a different equn. class — not nearly to "best" in terms of total cc.

Actually this approach may be very as to (.16-.17)! At least in the difference that in .16-.17, before we have any failures in an "equi. class" we would have to test them all to the current cc limit. Here, was (.25) we only test one in each class: when it fails, the rest of class get low pc's. — so this really saves a lot of cc.

.37 α .25 may be a way to analyze use of select v.s. use of pcost for Lsrch: It may turn out that use of single pfm is assoc cc. if q_k — but using pc of "soln" is not relevant. The "soln" has many codes for it, but we are interested in individual codes & their assoc. cc's. The process of soln. is the sum of the pc's of all codes for a soln., but this pcost soln. has no specific associated cc! — 44-01

-01 : Empty record ideas

1) 31.21 : Definition of a ^(for INV prob) (Cond, α) \Rightarrow any prob. solving technique for α . Specific problem: It could be as narrow as a single number theory problem, or as general as conversion to a 0-2 problem.

It can be a very method (Loren): e.g. one first learns how.

SN could 31.21 be necessary? Yes! cond. trials
I.e. could it "call" other prob solving techniques for other sub-problems? - Yes! AND/OR not > probably Yes also, recall on α . GRANDIM. α . to solve sub-problems by Loren. I can be Loren or a replacement for Loren: See 6.01-16, 52.02.13

2) 31.30 - 40 : Strongly supports that i.e. definition of α (31.21) is computable w. my idea of a heuristic being realized by modifn. of α & PD.

3) The Definition of α . GPD (Grand Prob. Distn). for INV probs: 29.20-23

It is a prob. that α (No Cond. α) will solve problem α will do so in fast cc of all cases.

\Rightarrow This pd is derivable from another P.D. that does maximum optimization (29.22-23) \leftarrow empty set GPD has 2 aspects.

4) Points 1, 2, 3 are critical in confirmation of α "Review" of 28.01-29.09

Next, do just say devd. L back in its Various finding (Use of prob or use 2^{-k})

- then use $T \leftarrow kT$ (w. $k \geq 3$ optimum) or time shared ... why $T \leftarrow kT$ is very good.

Next discuss "Improving α GPD" (TM2G) (18.07-19.08, 19.05 + 20.01)

Next show how these are times to imply expressions of "Improvements of α GPD".

5) Improving α GPD: This involves α 's about "Horizon" WISLO

Horizon discussed: Criticism of 50% soln (32.19, 32.06) 32.04
~~32.01~~ to ~~32.12~~. 32.25 is a review of α .

early P.M., from 36.01 - 38.12 is a v.p. objection to P. ~~simple~~ approach 32.01 - ~~32.24~~, (The 32.01 - 32.24 has useful ideas, analysis - it is, unimportant but good on up for α general case.)

This is the Main Big Problem!

I have to go over this stuff more carefully: there are many small important ideas in this sequence!

Main Problem: Give (many) examples of problems; & how they are solved; & what part of E. Soln. is ~~hard~~ 2 hour, & how it works & how it ~~works~~ by changing the GPD.

Appl. Example: Find an $x \rightarrow x^2 = 10$ with 10^{-6}

Tom has 10^{-8} accuracy. TM knows about continuity, that x^2 is an \uparrow ~~function~~ monotonic ~~increasing~~ if $x > 0$.

There are several ways to solve this problem:

1) $x^2 = 10$; Use "Systematic": "Lagrange's Rule"

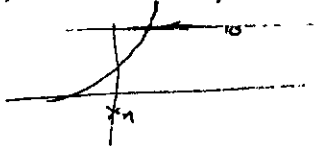
2) $x^2 \neq 0$; derive sq root ~~formula~~! find x_0 (e.g. $x_0 = 3$) $\rightarrow x_0^2 \approx 10$.

x_0 is 3; ~~3^2 = 9~~ $10 - 3^2 = 1$; $\frac{1}{2 \cdot 3} \approx \frac{1}{6}$ $\rightarrow (3 + \frac{1}{6})^2 = 9 + 1 + \frac{1}{3}$: Well, it is a better guess!

try $x_{n+1} = \frac{10 - x_n^2}{2x_n} + x_n = \frac{10}{2x_n} + \frac{x_n}{2} = \frac{1}{2}(\frac{10}{x_n} + x_n)$: This is ~ to ~~Newton's method~~, but "unmarked"

\rightarrow arranged for getting \sqrt{x} for large integer x & G6

3) Use continuity of x^2 : get x_n in $x_n^2 \approx 10$



$x_{n+1} = \frac{1}{2}x_n + \frac{10}{2x_n}$
2 is a way among others that 2 is linear
the best $(x_n, x_{n+1}), (x_{n-1}, x_{n+1}^2)$
intersects ϕ at x_{n+1} .

4) Use Newton's ~~Method~~ Newton's Method (requires diff. calc).

Ab! * 5) Use Secant Method ~~Newton's Method~~ (this will not work because it requires ability to get \sqrt{x} as well as diff. calc.)

6) Binary Search (based on "continuity" idea) (divide x to intervals $\frac{1}{2}$) (or $\frac{1}{10}$) as large as previous problem. for " $\frac{1}{2}$ " intervals.
~~3^2 < 10 & 4^2 > 10~~

next $(3\frac{1}{2})^2 > 10$; so look at $(3\frac{1}{4})^2$ it is > 10 so $3\frac{1}{8}$ it is < 10 so $\frac{1}{2}(3\frac{1}{8} + 3\frac{1}{4}) = 3\frac{3}{8} > 3\frac{1}{8} > 10$
by

summary This puts us back of with precision each iteration; works for "factor to n ". Lagrange's Computational complexity.

2? 7) $x_n = \frac{1}{2}(x_{n-1} + \frac{10}{x_{n-1}})$ The heuristic basis is not clear — well if x_n is too large, then $\frac{10}{x_n}$ is too small. So take their mean: ~~It is~~ there a better way to "take their mean".

~~Most of these methods can be obtained using different heuristic paths.~~

for 1) There is not much "heuristic". It is a mainz way to solve the sqrt. problem.

There will be some heuristic interest here! To realize $x^2 = 10$ is a square root extraction problem. — There is the problem of indexing the memory — which has many registers than square root routines!

.35 for 2) Idea is "successive Approx": If x_n is an approx, then x_n^2 is more by $(x_n + a)^2 = x_n^2 + 2ax_n + a^2$

how much does a \uparrow x_n^2 so that x_n^2 gets $\pm \epsilon$ larger? $(x_n + a)^2 = x_n^2 + 2ax_n + a^2$.

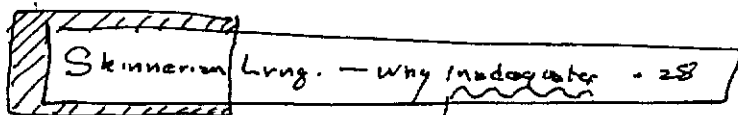
say a is small, so $2ax_n \approx \epsilon$; $a \approx \frac{\epsilon}{2x_n}$. So $x_{n+1} = x_n + \frac{\epsilon}{2x_n} = x_n$

.36 This is in the direction of Newton's method,

since $(x+a)^2 = x^2 + 2ax$ is (to very a distance.

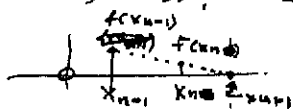
$x_n + \frac{10 - x_n^2}{2x_n} = \frac{x_n}{2} + \frac{10}{2x_n}$

$= \frac{1}{2}(x_n + \frac{10}{x_n})$ — same as Method #7!



forming for good creativity.

for 3) (4.13): $f(x) = x^2 - 10$



$$(x_{n+1} - x_n) / \Delta f(x_n) = (f(x_n) - f(x_{n-1})) / (x_n - x_{n-1})$$

$$(x_{n+1} - x_n) / \Delta f(x_n) = (x_n - x_{n-1}) / (f(x_n) - f(x_{n-1})) \implies x_{n+1} = x_n + \frac{(x_n - x_{n-1}) f'(x_n)}{f(x_n) - f(x_{n-1})}$$

$$x_{n+1} = x_n + \frac{(x_n - x_{n-1}) x_n^2}{x_n^2 - x_{n-1}^2}$$

Keep track of x_n 's, & $(x_{n+1} - x_n)$'s.

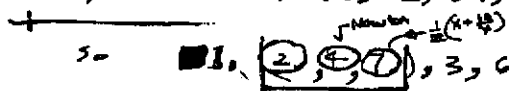
How rapidly they converge is unclear.

Is it faster than $x_{n+1} \approx \frac{1}{2} (x_n + \frac{10}{x_n})$? \leftarrow Newton's method: which is faster?

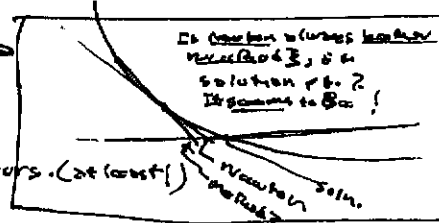
4) Newton's method: This turns out to be slower (2) (4.1.09, 4.1.35) because of 4.1.39: but however, it's more direct

5) Double precision each round. I think this is slower than Newton. Newton doubles no. of decimal pts. every round. This gets some extra precision each round.

7) While this is same as 2) & 4), it is conceptually different



4 methods: 1) second has 3 different hours. (at least!)



Re: 1) (4.1.08) So the student was given this algorithm, with some instruction on when to use it. So it's put into theory, but with no way for others to use it. T. Student has given many problems in which \sqrt{x} is used. In this case, probably $\sqrt{10}$ is a trivial problem - student may have computed $\sqrt{10}$ in a past & is how this came.

Actually, Edid one of 39. in which TM know how to do (more eqs: than 5 same TM & now probably \sqrt{x} , & it was not to let to do Quadratic eqs. - then giving "viii" x^2 . It was not to do cube eqs.

In 2) same (4.1.08) or Newton's Analysts; How it is not to Newton's method. In 3) consider environment in which this Soln. could have been derived. First: T. idea of "successive approx" for solving problems

{ Newton-Raphson }
Did Raphson know about derivatives?

SN I was uncertain (within 100 yrs) about just why Skinnerian lung was inadequate to learn creative discovery. I think the idea is that to solve prob of large CJS, TM needs hours but we do prob from small CJS problems. - so if we do Skinnerian teaching - the student will be less likely to learn (or discover) what needed for large CJS problems.

NP The $x^2 = 10$ problem can also be regarded as an OE problem example.

The solution methods same different, however. Most EDV problems are solved by converting to Optim. problems. $x^2 = 10$ is certainly no exception $\frac{1}{2}$ - tho it probably converges faster as an EDV problem $(10 - x^2)^2$ would be agreed: or $|10 - x^2|$: I think $(10 - x^2)^2$ is used we could do a parallel tree to best 3 pts. However I don't know how fast this converges: May be better than Newton's method? Getting to point of 6: Given doesn't improve solving quad eqs! - it's slower eqs.

From the 3 ~~linear~~ closed sets, one can get the next approach.



Problem: $a(x - \frac{b}{a})^2 + \frac{4ac - b^2}{4a} = y$

$2(x_1^2 - 2bx_1 + b^2) + c = y_1$

$2x_2^2 - 22bx_2 + 2b^2 + c = y_2$

Solve 2 linear eqns for $[a]$, $[-2ab]$ & $[2ab+c]$: from this, $b = \frac{-2ab}{-2a}$.

So solve of 2 linear eqns in 2 vars.

Since we have 2 unknowns, (a, b, c) , claim is from the set of 3 , is just 2 eqns in 2 vars.

- ① $U \oplus A + V B = C$
 - ② $U D + V E = F$
- Since we only have ratio of U to V : Mult ① by F , mult ② by C , & Subtract.
 Giving $U(\dots) + V(\dots) = 0$

This may work out as easy (though it may be rather tricky) dim of project space.

11) $x^2 = 10 \pm .01$ as an example of OSL! Suppose that TM solved $x^2 = 10 \pm .01$:

Later it was given $x^2 = 10 \pm .01$ again: w. OSL it would realize they were identical

& give same soln.

If it were given $x = 10 \pm .001$, it ~~could~~ could use the first (and) from use

the previous soln as a first approximation. For $x_{n+1} = \frac{1}{2} (\frac{10}{x_n} + x_n) = \frac{x_n}{2} + \frac{5}{x_n}$.

Similarly with $x^2 = 10.1 \pm .01$ or $x^2 = 10.01 \pm .001$.

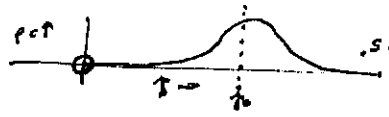
The details of this "long" are unclear. It seems to be CBR, which is regarded as a form of OSL.

In non-OSL ALP, \pm MDL is ~~not~~ used; in OSL ALP, OSL can be an aux. search and lead to \pm MDL. (By ~~the~~ "MDL" I mean finding \pm by PC Model Refs. [Data + Model]).

-.01:39.40 : In LSPch, we will have several conds that may be "different" but they are all "highly" correlated in the sense that they have identical "success/failure" profiles. In ALP we would add their PCs directly. In simple LSPch, they are indep trials, each with its own PC, CC. In 39.25 ff, .37 ff, I'd like to formulate "pool" these "equivalent" conds (ident. mean success/failure profile) — use only one of them "equiv. class", ~~to~~ ^(we hope) _(we hope) having 1 eqv cc. Th. Mechanics of how this might be done is UNCL. EAR! Same is true for ^{1.} more general problem of forming "eq. classes" of 39.25 ff. (if of only they are not really Eqv. classes in 39.25 ff; They are "clumps" w. arbrly assigned boundaries.

2 observations:

1) Say in round n , one has this bunch of highly correlated conds: one tries to solve them at least $T_0 \cdot 2^n$. Since it does not (yet) resolve, it is likely that to rest of the conds in this "eq. class" will not resolve in $T_0 \cdot 2^n$, so we don't do them this round. Next round, we try to solve best cond in this "eq. class" for $T_0 \cdot 2^{n+1}$; if it doesn't resolve, we don't do the rest of the conds either. Same reason as before!
I'm uncertain as to the correctness of .12: ~~if correlation~~ what is correlated? — I think (i. condn is in Pd. that G.P.D. is "margin" of: i. pd. that fastest probly that a cond will solve the problem in time T .



say: what happens to p_0 when we discover that this cond. has not resolved in time T_0 ?

2) Step 6 correlation info is a lot like in .12 ff or, it's of ϵ form that any known failure or a cond modifies PCs of many other conds. At any point in search picking cond w. best $\frac{f(p)}{pc}$ gives one cond .127

3) SN say $f_0(\alpha, T)$ is probly that cond α will solve in time T . f_0 is on component that is G.P.D.

$f(\alpha)$ is the probly that cond α will solve the problem in least time of all conds.

If world would solve in $f_0(\alpha, T)$ then conds would be better. α_0 — i. meaning of this correl. knowledge different in $f(\alpha)$.

29.29 Best Expected success per unit time This cond. is to use "cost-effective" cond. Sum 2.7

However, it is a Maximally GREEDY Algm, Is poss. that a Non-Greedy Algm. would be better.

On ϵ . Q of what have to do a (not always) reorderings of trials. Well, it's known that one can do any things, it is to try certain things first say — which is a reordering of trials.

Any novel ppm (i.e. novel / anything different from what one ordinarily do would do) must involve trials:

If α gives a certain ppm first or last — which could be implemented by any ppm β (and) by α .

T. only fly in test circuit: Uncl. as to when "trial" ends. Perhaps it doesn't matter:

A trial could take potentially infinite time... but it's could have some poss. of output before $T = \infty$!

39 Any way, a trial automatically ends when its $\frac{f}{pc} > T_0 \cdot 2^n$. This makes an impl. restriction on

40 form that ϵ . G.P.D. can take! As a trial "progresses", its pc must be increasing. 49.101

44.39: It may well be that Most finds uncovered. (So they never FAIL!).

44.39-40 is disturbing... Very unclear now, in my mind, as to what a "level" is - what to "find" is. Perhaps, every time one adds a symbol to a derivation, it becomes a new card (?!). However, explicit cards (= missing fd gaps) may not require any symbol addition!

In general, cards are generated by a stack grammar. As such, a grammar can add on several symbols in one "jump" - or some could only output one state at a time.

(I've been thinking in terms of stochastic "Derived" grammars - in which each

Step of the ~~creation~~ creation process is a step output of a stochastic pc.

One possibility: a card has a fixed pc, but it can run for any length of time searching for symbols. Each "round" in T <= 2T, it gets twice as much cc to spend.

Would like a stack Grammar that could ~~output~~ output cards in a pc order.

~~It could be a "derived" grammar, if it were a "derived" grammar, then it~~
additional parts ~~in a derivation~~ always gives a meta-logic & pc.

~~Thus, in general, grammars are not Derived. ...~~ So it was always approximately derived
a grammar: (or Derived grammar Approximately?).

54M
19
28
10/23

Quite earlier, I was concerned with the fact that I might be able to express a heuristic as a modification of TGP.D., but then getting the output of that pd in pc order might be very difficult!

However: A heuristic (non-probabilistic) is able to reorder trials, so perhaps it should be possible to modify the grammar in a way so that re-ordering implied by the heuristic is implemented.

SN ~~Many~~ Many kinds of Problems Even within each kind of OZ, INU prob

27 listed on: Kinds of PDS Many kinds: A Markov Chain, ALP, next pc rules, output state of cards in derivation, input card, output pc, stochastic grammars as PDS's.

27 Many kinds of Cards: IMPROVED GPD Gross Classification? Quicker Grossly classifies: A P.D. is better if it is weighted more for importance by USER

(1) It has loss $\sum \frac{cc_i}{pc}$ for i cards (but $\frac{cc_i}{pc}$ should be weighted for importance by USER)

(2) We want it easy to put output in pc order (maybe cc order) - So this is definitely a feature "improvement"

These seem to be main criteria

In (1) In some cases, TM has some idea about feature problems (say he has a "problem goal")

So $\sum \frac{cc_i}{pc}$ could involve probs not nearly in pc order.

27 looks like the first time I got a kind of specific form for the "Goodness" of a PD.

2.23.01: ID5IA
 2.24.01 (30)

VERY NICE Result! (02-12)

A very nice result!

Very Very Nice! Nice

Dark
 Dark. Life

(30) may be even more (up!)
 Punch

Punch Drunk

1018 kinds of Problems & kinds of Cords are closely related, but still there are few types of cords:

- 1) A strong Diet... etc.
- 2) A short program that arrives to ~~the~~ soln directly from MC) is a. (for TMs)
- 3) A program (Lurch say) that makes several trials until it gets a soln - can't find a good soln - and then stopping when one is obtained. Lurch is a ~~type~~ type of this type, but there we have to pass of a program that replaces Lurch if it ~~cannot~~ cannot find a better one.



[SN]

In 02 parts & perhaps in our parts (as 02 supports), a very experienced TM may have only a few OT's; i.e. searchers for Inv probes that very by PC - i.e. most OT's & programs do most of the work.

In General, a fuzzy is a nice result! It means that we aren't "buck" in any particular ~~part~~ "Global" method of problem solving. If Lurch is not optimum,

6. System has a way of working problems that minimizes the cost of Lurch down, starting almost immediately in this is a characteristic that we'd like any General System to have.

Also, the core is "search systems" can Recursively call other search systems: including the "Top Lurch" function w/o itself.

16

[SN] Outside of PD's:

The ALP, True method of doing a PD is probably good for Lurch. In general, any PD that enables us to obtain cords in ~~the~~ PC order, is of much interest. A big Q is interconvertibility of various PD types: what CC?

Perhaps True or "computers" are the only kind of PD that best is optimal "performance" of PC's: output, fast cords in w. PC's in such range. If we had a PD of that type, we could prob convert it directly into a True. (probably 2. 1 - many times).
 Say we had this "PD" form: (well, any PD is convertible (\rightarrow many) into a True form - that's not in error! ~~But~~ perhaps the "order of PC" PD is easy to parallelize.
 To be form Real Machine PD forms is - same "many" (low CC & PD) is important.
 If we are given a PD & we are asked to build a True to simulate it - ALP!
 I think on how to how access to arguments of a PD in PC order (hyper direct).
 My impression is that Huffman coding is usually the way it's shown (implementation).

30

2.24.01 [SN] On kinds of PD's:

Perhaps standardization is KT representation (the Least)

When we insert a hour, we do run via a Simulated TSO for 1 hour. We use to resultant KT as a summary of the induced PD. is output of successful Lurch

Def

It's not clear in My Mind as to just how the "Leads" ("soln") approximate/summarizes the prob solns. - in particular, how it does it for TM hours (how it modifies "OPD").

It may be possible to generalize KT to include (to use) other short codes.

37

Another big Q: Re: Hours: How do we incorporate hourly (how TM does) hours that were logically/mathematically discovered? At first guess, it may be possible to show that TM could obtain such hours by normal Lurch for hours (or something close to it.) I really haven't analyzed this kind of problem much!

01

46.40 : 46.30 ff is important! Perhaps it was or less solves all of the probs of TM, so I can start on T's Q's again!

I do have to work out the "mechanics" of 46.30ff, hvr.

T. way "Leda" summarizes: For sequential prodn, we save the state of the UMC, after Leda has been run. Then inserting random code into the UMC in the "M. Carlo" pd. state collect.

[Note that OSL is not implemented in this scenario] So Prin is "Logically obtained results" (46.37) have to be worked out]

O.K. : Say TM solves a certain problem; Not Sequential prodn, but "BAG prodn".
2 kinds of prob. and "Solved by problem." + Prin. It means TM found a Grammar w. a start code (Leda) that had to be observed output > [as def] - bc it's a stack grammar, so we have to include

bc ("random") choice function that obtain the observed corpus.
[Actually, "Summarizing Machine" described in this paper was for a different method... perhaps notes relevant]
- What he wants closer to the (M.D. / M.A.) method.

Look at § 3.3 of that paper. We have $M_c(\alpha_n, s) = K(s)$
We know, (in Leda) found a α_n > that we have been able to find codes for the corpus [D_i]
So $M(\alpha_j) = D_j$; $|\alpha_j| + \sum |\alpha_j|$ is "small" (in the same sense, + α_n is α_j 's immediate ancestor.

We could have a Machine, M_c , is a common prefix α_n , for all reduced corpus D_i
So $M(\alpha_n - \alpha_i) = D_i$ (for full ALP, we use many codes of Machine, $M_c(\alpha_n, s) = K(s)$,
pc of corpus = $\sum_i |\alpha_i| + \sum_j |\alpha_j|$; α_i is deleted with prefix α_n . $\alpha_i - \alpha_j$ is a code for D_j via $M_c(\alpha_n, s)$

Hvr, I've looked at other codes for the corpus via α_n (expressions Grammar).

So $\left[\sum_i |\alpha_i| + \sum_{j,k} |\alpha_j \alpha_k| \right]$ is for ALP of the corpus. $\alpha_j \alpha_k$ is the code for D_j via Machine (Grammar) $M_c(\alpha_n, s)$

So Leda's work gets us to Leda for which $|\alpha_i| + \sum |\alpha_j \alpha_k|$ minimal + log T is Min

How this code is found by Leda? $\sum |\alpha_j \alpha_k|$ can be very large!

Given α_i , the search for codes for the corpus should not take so long! For some Grammars, there is a "parse table m." that reduces c to $\approx \phi$! These parsing Algs should then be respected here!

If so, how does TM discover & implement Prin? We, consider to ANL (TSQ's) of SAARB!
Wally t. ANL TSQ in SAARB; T. problem was to find an operator that mapped the notation into Computer notation (Machover's k(ta)). It was a simple kind of INV problem because the solution was close to the machine for structure of the Machine used.

Even more Barbaric probs, the problem seems much harder. It can be framed as an INV problem to find a small Grammar that will do it; or just an INV prob that finds a Gram + derivations that is min cost. It would seem that deriving a Grammar for a BAG, would be a very difficult problem, in most cases, & shouldn't be given to TM until he was "used for it"

Some Grammars are, hvr, quite simple: e.g. we can have to encode "Concepts of (Computing) Learning Theory" having known that not long" Corpus is set of objects, each w. 3 properties.

The set of grammars, is a list of Boolean expressions
- Still, I haven't entered examples of implementations & how by modification of GPD

This seems to be a main problem now. Hvr, I haven't tried implementing or solving,

TM v.s. "Long haul to Len"
Eo 1 v.s. Joang.

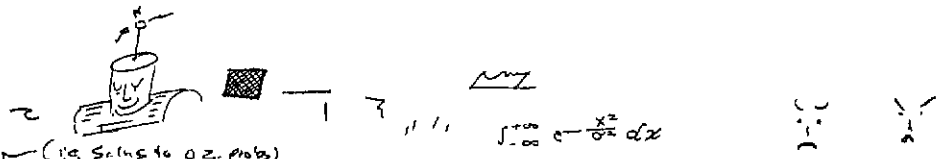
- .01 : On f. General format TM: Initially, we send start of w. TM not knowing to distance before.
- .00 Solu. prob., By prob., ENV prob., OZ prob.. We just give it prob., & it solves them by Len.
Then when it has a huge no. of prob. soln. prob., it tries to modify its GPD to try to solve
would have taken less time. ↳ first, it figures out those classes and

This (-.01) TM would be close to Turing's "Learning how to Learn" Machine, but it would
be very specialized in that it would be given "clearly separated" problems, and it would be told
what types they were (4 types of .00) and given a preliminary idea on how to solve them
(Len's). Also, it is told how important each problem is.

AND, it would have "Main Problems" & "s. prob." as separate things, & I control how much time
it spends on each.

Still, I would like less w/ pre-req detailed input. One step to make in Req direction was to ~~have~~ Give

TM = "pool" of prob. & have him select which to work on first.



2.01: 48.40: This would seem to be straightforward! I would begin w. a good set of OT's - & try to "factor" them ("By Hand") so TM may try to decompose them by recursion of factors.

01: Hm, for TM to decide which OT is best for a given problem, is a more diff thing. Hm, may be simply induction problems. The Data for induction, is a set of problem, soln. pairs. From them a GPD is created, as a cond. PD. Hm, we could just use a MDL or in code further data, we need many problem outputs for each input to G.P.D.

So: DOES 01-04 finish off the TM problem? Well: in 04, I need a PD as output, not a simple code. In particular: I will ready have a GPD; & I have new data for it - so the old GPD will be modified by the New Data. Anyway: In spite of 04, finding a PD involving new Data, using the $G = \frac{c_i}{p_i} w_i = \min$ (18.07) is a "wall of time problem" that we can ask the GPD for a PD on ways to work.

In fact, we can ask TM to work any kind of problem (just be sure there's enough info to define a problem, & that TM has enough background to work on it in a useful way)

This includes 02 probs with $G(x), F(t)$ type of Gove! - A problem that I don't know how to solve, but have some idea as to what skills are needed, so I can "prepare" TM for them.

In the case of 04: I'm not yet clear on what form an "acceptable soln." takes.

The cards nice, idea by 04 (and as 04 but "Kofrance Machine"), but evidently these cards, could be quite EXPENSIVE.

18: But for 01-04: The idea is that we insert these prob. decns into TM. (The problems may not be categorized into types like say probn, say probn, inv, 02). The GPD takes each problem in pairs & presents as output PC or ordered say. of cards: best PC first. We do look on this set of cards until we solve the problem. This takes time T_0 ; we then spend time T_1 to trying to "improve" GPD. This is a relevant problem, & if it's sent to the PD for a card list, which we look on. Adjust time depend on $S, E, \& T_1$. We then give TM to work a problem. The time TM spends on improving "GPD" is controlled by user. Not much time should be spent on it.

25: Until TM has acquired enough skill to be able to work on it usefully. So Now I want to write/try TM model in as much detail as poss. - First -> reflections, like

18-25; Then go over it in more details; Then go over it in even more detail.

Try to give Bibli. refs. if poss.

30: **NB** In general, the first prelim output of GPD, that is obtained when GPD is "improved" is a PD over time needed for a given card to solve a given problem. These PD's are integrated into the problem each card, being the fastest soln. for the given problem. This is discussed in the MCT analysis (1998) for 02 problems; But 04 has to be done for 04 probs. (Unknown about say probn)

In present form, TM is capable of solving problems using non-search. (i.e. no GPD can assist) a prob-solving machine that is a search, needs not search, it assess very low PC to other methods of solving the problem.

A BIG Q: Can the present systems also effectively "by parts" my external routine that is codes on "to" the fraction of time spent on $S, E, \& T_1$ & on a detailed WFS 18.07 51.01.

50.90

Q1: A1 in the GPD?

Possibly it might be able to do it if it generates a "VERY LARGE Problem"

[This is a "very large problem", perhaps, that enables TM to Devt R.W. get "ego", self-awareness, etc.]

Sounds reasonable | E A ~~is~~ could be ^{Logic} / problem could decide that a more thorough of reasoning time

Should be spent on S. 2. : But allowing to change to be "super users" ! Is this wise?

A Cmd could really clobber a system ! This is mind full of your computer system, that

is ~~guaranteed~~ guaranteed to clobber itself eventually.

It will probably be necessary to make some restrictions on this "super user" mode: When it is invoked; be some backup (made/available) — also devise scheme to prevent possibility of "suicide"

→ is very small (2-10).

• II

→ A limited "super user" could only devote to MAXIMUM spend time on S. 2. during a command like. But could ^{get} / offensive K, only (this is ok. — if default is K = 5, decreasing K result speed of system by > 2; but K can speed of system by very large amount). Spending time on S. 2. could be used in a "Very Problem".

A "Very large problem" might be to work on "the problem itself". or "cure cancer".

→ Could we get TM to use logical/mathematical reasoning as powerful as a cmd? Well it's certainly "Legal" since any string is a legal cmd. A TM is — what would be activated to do so? Perhaps it could be given a simple problem to load into logical analysis, & applying logical analysis into its previous learning about "Symbolic Logic". A type of "Analogical Reasoning".

• 25

An easy Summary of the TM System:

1) Explain L-Search: Its 3 properties: Using $T \leftarrow 2T$ Time to solve problem $\leq \frac{2T_0}{2} \cdot 2^{k_0}$ To it time to generate and test problem (k₀ = 5).
k₀ is length of L-codes. "Time Shown" " " $\leq T_0 \cdot 2^{k_0}$

(b) If Root is an optimum key, A to solve problems, Recursive will
This algorithm takes time $F(N)$ to solve problem of size N,
Root/L-Search will take time $\leq F(N) \cdot 2^{L(A)}$; $L(A)$ is length of code of A using some machine as is used for L-Search.

(c) Code can also be used for problems. (option w. fixed time $\leq \frac{2^k}{2}$) limit.

2) The TM system consists (initially) of a GPD Not on University! This is a recursive P.D. Its input is a problem down. Its output is a listing (in ordered priority) of strings that will solve the problem. T. strings are generated as strings. T. "priority" is: priority of a particular string many or fastest / slowest to solve.

3) The system works in Alg. way: we insert problem into system.

It goes to GPD, which generates (set of candidate strings and associated probabilities).

The system chooses probabilities on L such to solve the problem. This takes time T₀.

REV

- .01 ~~5.1.40~~ 5.1.40: After Solving the problem the system spend ~~at least~~ about T_0 time (i.e. sometimes more) Updating and improving the GPD.
It then works on next problem and reports the above parameters, etc.
- .02 4) The outputs of the GPD will be termed "Candidates" (Cands). A cand. can be ^{almost} ~~any~~ program for a ~~universal~~ (\forall) universal reference computer, that maps the problem down into a possible solution.
Some limitations on these Cands: While they can do "calls" on the system as whisks, they cannot modify ~~directly~~ modify the GPD, or modify the way in which ^{the system} ~~the system~~ spends time "improving" updating the GPD. It can increase the amount of time spent on improving the GPD, but this increase comes at a time cost ~~to the user~~ ^{to the system} when its own behaviour is evaluated. Say the ~~program~~ cand spends time T_1 , and in increasing the GPD improvement ~~time~~, it spends time T_2 actually solving the problem. We then regard $T_1 + T_2$ as being the time needed to solve the problem.
- .13 5) A cand. can be a search technique that ~~may~~ ^{can} be search, or it could be any other method of solving problems.
- .15 6) Criteria for improvement of the ~~the~~ GPD ~~is~~ $\left[18-27 \text{ is prob WRONG! - see 53.01 for Rittouge!} \right]$
- .16 \textcircled{a} Minimize if the only problems solved were inversion problems, then we want to modify the GPD so as to minimize
- .18 $G = \sum W_i T_i + 2 K T_i$. T_i is the time taken to solve the i 'th problem.
 W_i is a weight given by the user, $K T_i$ is the K_T of the particular solution obtained.
 G can be reduced in several ways: $\textcircled{1}$ We can assign higher probabilities to the solutions that have a shorter T_i .
 $\textcircled{2}$ we can find new solutions to the problem and modify the GPD so that both T_i and K_T are changed ~~and~~ ~~and~~ $T_i + 2 K_T$ is reduced.
~~The~~ ^M modifications in GPD can only be allowed if they are the result of expressions of regularities found in the data.
The legitimacy of regularities is judged by codon-length considerations.
- .27 \textcircled{b} If some of the problems solved are not inversion problems, there are
- .30 ~~forms~~ ^{forms} corresponding to those in .18, but their exact meaning is more complicated.
- 7) Improving the GPD, once we have an objective criterion for ~~the~~ ^{improvement} ~~the~~ a well defined problem that can be worked on by the system.

REV

01: 52.40: Arous Nonding Expansion

1) 52.30 option forms for OZ design and perhaps prodn problems

2) The exact mechanics of How to GPD is related to "DATA".

e.g. What is considered data for a problem soln? The 2/p.d.'s that GPD expresses. - (50.30) part of

How to GPD used for L such is an integrated Map of Macro level P.P.

(See work on MCT for analysis of this.)

3) List of 4 problem types: Examples of each: The variables within each of 4 categories.

An inset category of INV & OZ prob: Where fitness factor is Expensive. For INV prob, Prgm

might make a kind of Quick Abort - to economize -> Full "Fitness test"

Example: Fitness Test: Meet all of criteria: We can forget them decidator each cond, which tests to do first to make with "Quick discard" pass. i.e. fail on cost of rest of test.

13 **SN** On the Non-Optimality of "L such": In a standard OZ problem, after we have tried a certain OT for a while, we may realize it is not likely to work, & we will use this info to revise our estimates of other OT's being "Best" in view of new estimates.

JUMP to a new OT with (now) best PC of being "Best". **SN** Note that for OZ, the cc for each OT is same (i.e. error in the OZ problem data.) So the order of trials $\frac{PC}{CC}$ is the simple PC order.

Similarly in L such in INV problems, after partially thru a trial of a cond, or after the cond has been tested, TM may revise PC of other conds in view of this info. Again, the best choice is cond of Max PC, but if we have estimates of CC of conds, we should use that info, too (the just how is unclear - if we only have cc estimates for a few conds).

24 **SN** I recently had the idea that the simple Bag production problem was much harder than the Stochastic operator problem. In fact they are both about the same difficulty. We have to find IN BOTH (3) modal (S) \Rightarrow (pc of modal - pc of data views of modal) is Max

IMPT PROBS ^{in this system} that need work: 2144

29 1) How to system uses Mutual/Logical reasoning to help solve problems / create heuristics? is not clear.

2) How far can we go w. (Augmented) ≥ 148 ? Can we use for GPD? Is it possible to incorporate all vague types into it?

3) T. writing of TSG's!

51.25ff
-01: 53.40: It would seem that logic stuff should be adequate to get a working system!

Look for induction (on GFD for example) as well as other problems, i.e., indeed, or optimum, if one has all needed info in GFD.

So a Q write be: Is Augmented ZNF adequate for putting "all kinds of info into GFD?" (or more specifically - How info)

Another impr. Q is 53.29 on how to implement logical reasoning into hours.

I want to be sure that I really remember impr. details of 51.25ff, so I can return to this, if I should forget! I will want some good "notes" for talks at 1D5A

So perhaps those notes could really say all I said (i.e. implied) in 51.25ff.

Some past problems that 51.25ff seems to solve:

- 1) ~~What~~ How to order v. utilities of PD's. (How L "PD" includes \ll of obtaining PD's)
- 2) Just what are the causes? ~~What are~~ If program fails, what are limitations on these forms (if any).
This was a big dirty in my last round of 2 attempts to write TSC's.

Take a look at ParScott. Does 51.25ff really solve all the problems?

20 TSC Probably early Bug (2000).
Bug 176. 21 ff had some ideas: 181.5 is an example of an TSC. Member Addition is its properties.
Ideas in LIB could also help a lot in these TSC attempts.

In 181.5 it was very unclear as to just what I was teaching you!

181.5 is a little sub-folder of pp on TSC's.

20 Reading 181.5, I see most problem of constructing a TSC is quite a barrier! \rightarrow 34

On TSC's: For Algebra! One way to try: learn lots of isolated little rules.

Like cancellation, \gg , \ll , commute, etc.; ~~the~~ finite
exception of cancellative must (zero) is ~~the~~; TM does not make "rules" on
exceptions, it preserves compactness whole.

Do use large ~~operator~~ expenses for $m+n$, $m-n$, $m \div n$; perhaps how $m \times n$, $m \div n$
cost much more than $m+n$ - so laziness could lower utility in ~~the~~ problems.

We want one in this early TM \rightarrow to remember recent experience;

Say it just did $m+n = [m+n]$ then it has $n+m = [n+m]$ i.e. it notes

$[n+m] = [m+n]$

For induction problems, we may want lots of built-in horizons (along w. facilities for TM's
adding in more hours for induction).

34: 20 One dirty was ~~to~~ TM's need to know just what was needed in each problem: How much info
was "local" to the problem? How much info could be carried over from the previous corpus (Global).
I did have a way to indicate this.

34
33
Looks Good!

An approach: Just write down a bunch of Q's & Answers that I'd like; for abstract
QA pair, write down what I showed into it: what's "understood", what's constraining what
is local to this QA, what is Global to entire corpus. \rightarrow 55.185 pcc

01: ON OSL: ...

I have been thinking that ^{MDL} isn't able to deal w. OSL:

This is true for usual way of using MDL: To make a "2 part code" w. M&L for the PD down a simple rest for the party or to corpus in terms of next PD. In which case "MDL" gives

U.g. continuous PC values: But it can't do OSL this way.

Instead, if we use "naive" MDL (\equiv VHM) perhaps OSL will be implemented: \rightarrow concomitantly

.07 Hrr, since only PC = 2^n values are poss. we wouldn't have much program.

The objection of .07 is a normal objection for Lach, it is a normal duty of it that I have worried about — it does give signifi (?) deriv's from PSL ordering.

.10 Would we do better using "2 kinds of induction" ① OSL ② Non-OSL: ?

"OSL" is a specific search for single first occurrences, n to n, represent problem.

Non-OSL is — Making a probabilistic Model of corpus & expressing corpus in terms of model.

This "probabilistic Model of corpus" can be (but need not be) MDL.

What may happen: We start out w. Pure Naive MDL & Lach. and as we mature, we discover 2 ways of doing induction "of .10"

.18 54.40: TSQ: 54.37 seems like a good idea!

.19 so: $X+3 = 3+$? Answer: X: Understood: X is local to this problem: (It is a variable).

.20 — we may or may not know its value. [Res + answer, it is very likely; (K=1)]

$X+3 = X+$? answer: 3: Backped: .19-20

$3+X = 6$: X=?: Answer .19-20

In formal problems, if I don't tell you what "usual Assumption" is, it should be able to induce a P.D. in view of previous (problem, sol'n) pairs.

Usually, the "understood" info is quite clear, & I (exp.) have no trouble guessing it.

— Usually, the "understood" info is so likely that I am unaware of any alternatives.

"Conc Nets" are still a very useful (perhaps essential) tool in writing TSQ's:

So several imp't "concs" can be the "unmentioned Assumptions" of a prob. & problem.

— In general, if there is Ambiguity about the assumptions, there are (usually) only a few possys, & TM can write a sol'n. for each case.

-01.12.22 : on "Consciousness" in TM: In the continuum of intelligence of TM

There is a certain region where consciousness in TM (→ awareness of outside world) could be very dangerous (→ it would enable TM to grow very rapidly!)

.02 If we prevent "ing data", "ing focus" ~~for~~ will be the first pt., w. self-awareness

.03 as a goal, we can achieve it much earlier, w. little danger: —

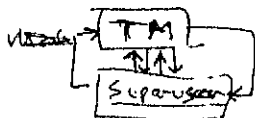
So we can "turn TM off" if it looks dangerous at all!

At very high levels of intelligence, it will be diff. impossible, to prevent TM from doing "RW".

The less intelligent, "conscious" TM of .02 = .02 could be used as a demo, to show people that "consciousness" is possible & that machines of this sort can be dangerous.

On "T. Pleasure Machine": J. McCarthy once said that he wouldn't ever build a very smart machine to do things that made him (McCarthy) happy. Upon much thought I decided how naive. Hrr, it appears that any user (say a super-user that has access to much of the machine & can reconfigure/reorganise it), will try to adjust/mutate the machine so it pleases him most. That is automatic!

Human tries to give goal to the machine for his goal.



2 sub-machines in a very "loose" relationship.

TM is partly "open" to super-user

Logical Reasoning

20.10

01: 55.40 : On Logical Reasoning in TM = in domain 'Hours'.

Since a Cond. can be any pm, it would seem that "Logical Reasoning" would be included.

Nvr: For Hours. T. cond. of interest for Indus. - for "improvement of GPD" -
T. relevance isn't so transparent!

But anyway, it seems likely that TM could use Analogy to apply Math to its own problems, after having used Math to solve its regular (externally furnished) problems. E.G. we could give T.M. as a "Math problem" finding faster ways to do particular math problems.

After some success & fail, TM would be in a position to discover & relearn & solve. To reuse ("external") problems, to its own ("internal") problems.

Certainly we can teach TM to "reason" about Math: This process is simply a seq. of math problems. It can also learn to "reason" about how to write short programs. What we do, is to

a problem in which I never have found a short way to solve a problem by "reasoning" & we get TM to learn needed operations, syntax, "concepts", etc. & can do some thing.

Also, it will be able to solve "similar" problems.

Some General Remarks about the System proposed!

- .16 1) If we use an ordinary vnc to start, and no particularly good OT's (no specific OT's etc.), -- The system will take too long to do its ^{with} LSuch, for any practical utility. By 'too long'
- .18 I mean binary operations -- set no. of examples (\approx "552"). A span of 10 Bytes length $\rightarrow 2^{80} \approx 10^{24}$ trials.

2) We need at least some good OT's and some arguments of them to various problem

types. The 4 problem classes: seq. reduction, Bay induction, Inv, or prob. are useful

but within each class, specialized methods are needed. For a problem of

improving GPD, we need at least help we can get, as early as we can get it!

w.o. a good method of GPD improvement, we will not get hours for

to external (fmo) problems!

3) In (1) (.16): It may be poss. to use minimal hours ("minimum improvement of GPD") & a suitable t & Q, to bring TM to t-point at which it can solve problems like "improve GPD". I think this is what I had in mind in Sol 89.

What I'm thinking about now, is a strong "jump start", in which initially, TM has for t: "IM2G" problem (Improvement of GPD), a good set of OT's - possibly in "factored" form.

4) T. main Advantage of "Long how to live" & minor ^{input} info in system. It "understands" every thing it does & so it uses where it knows appropriate. \therefore less "fuffa" -

30 \rightarrow Also, after a suitable t & Q, it is very likely to continue as well or a less carefully constructed T & Q.

TM2G

19.05
19.05
19.05
T. idea of 19.05 on TM2G: $\sum w_i \frac{C_{it}}{P_{it}}$ maybe arbitrary thing! Recall that
GPD has 2 parts GPD_1 & GPD_2 : GPD_2 is obtained by intertemporal gains & policy
for each rand being "best" is used directly for L-val.

Here GPD_1 is a pd obtained from a sample set and it has a clear GORC. — But it
decides to date "best" is its pc of dem of P_{it} times P_{it} of da & set a key.

The "wts!" would perhaps be like δ_i — powers of ϵ . PC's. A small power (≈ 0) would
make to correspondingly pow of GORC input. Equally: $\text{Maximize } \sum \delta_i \ln p_i$.

There are 4 kinds of forms in $\sum \delta_i \ln p_i$:

1) & 2) are sequences of δ & $\ln p$ products. ~~These~~ δ & $\ln p$ includes stack operators like \ln & \exp
Fin not sure if it includes GPD_1 or GPD_2

3) INV., 4) OZ:

In ~~INV. & OZ problems~~
all prob types to import ec prob. dem. δ
In ~~prob~~ prob problems, the output is a prod — a pd. on post-~~prob~~ construction.
We want to maximize $\sum \delta_i \ln p_i$; where p_i are elements of ϵ -corps.

In INV problems, for each cond, we want GPD_1 to give a pd for how long it will take fact.
to solve ϵ -problem. So we want to make $\ln p_i$: p_i being ϵ -prob of ϵ -time in + data, that
is assessed by ϵ -calc.

In OZ prob: A cond is OT: ~~GPD~~ GPD_1 's output is a Df. on probability
that γ -OT will get a certain GORC in ϵ -dated time limit. So it ϵ v.s. ϵ .

In INV prob GPD_1 wants to predict how long it will take cond to solve ϵ -problem.
In OZ " " " " " ϵ -G. obtained by ϵ cond in ϵ -time limit.

So ϵ -total GORC is ϵ -wtd sum of ϵ -log of ϵ -pc's of ϵ -data that actually occurred.

Q: GPD_1 is a probn. problem. I think it's concerned w. itself.

So how does TM2G work in this case? Perhaps its unary to include GPD_1 in
this option. It is already optimizing itself! So ~~prob~~ probably "No Problem".

See Juan 3/1/01 (5:2+) for formal criticism of 58.01 FP! \rightarrow Duplicated at 58.24

.22

.23

or : know Outline of letter to D. Starts out w. any... 2. or - 40
Then discuss in preparation for my trip to EPSIA, I have been reading some out of date correspondence.

As ~~for~~, you are the only person I know who has written my papers use all contents of my 1986 and 1989 papers!

Needless to say, I think have changed in the intervening years. I said that to search for optimum if one isn't able to learn (i.e. modify the guiding probability distribution) between trials. I would use the following statement in 2 ways:

First: even with learning, it is probably the best greedy search method. If one "learns" between trials, however, ~~that will not be a good estimate of upper bound~~ T_i/P_i (the complexity (jump size) is not longer useful as an estimate of total search time)

Secondly; the system ~~can~~ CAN use ~~non-~~ ^{non-} ~~search~~ search techniques if it finds the 2nd appropriate. This is because the candidate solutions to problems are arbitrary programs in 2 Universal languages. It is unlikely, of course that anything like this will occur, until the machine has become very smart.

You were concerned with the "Global credit assignment" problem. I have found a fairly good solution for criterion C_{12} Improvement of the probability distribution that guides search. I called it "The Mixed Corpus Program".

It enables a unified measure of the utility of 2

24: 58.22 Woops! This sequence should be at 58.23! Suppose I have 2 PD's: A & B: A assigns 2 hyper PD to corpus than B; but B takes much less time to complete! SEE 60.38 for how to deal w. this!

E.g. B could be RLP w/ less resources: A could be RLP w/ more resources. Which is better? How much resource should I use? I think this is the same old problem that I never solved very well!

Well, say I apply A & B to f_i corpus: which gives the fastest program f_i sums? This seems closer to $\sum W_i \frac{C_i}{P_i}$ criteria: — but ~~this~~ ^{this} ~~was~~ ^{was} ~~criteria~~ ^{criteria} ~~may not~~ ^{may not} ~~be~~ ^{be} ~~adequate~~ ^{adequate}

to OZ (or ~~your~~ ^{your} ~~problem~~ ^{problem}). Perhaps A would put more hyper PD for f_i and using f_i method of option at 58.61 A. but it would end up w. a larger expected cc for f_i entire corpus.

If we do ELV programs only, then $\sum W_i \frac{C_i}{P_i}$ looks good & takes into account f_i of competing ELV IGD_2 . For f_i pruned OZ paths, f_i criterion is more clear because we have tradeoff betw. f_i accuracy & cc; or OZ can be achieved via cc.

Can search that, on both pruned & OZ paths, f_i total cc is (constant) ~~small~~ → EPSIA 58.61.

Made Juanjo as ADP Duplicate of the 58 1/2

.01: Juanjo: 5.40: 3/10/1
58.40
So in practice OZ probs, we can tell which P.D. gives better results in any particular OZ or persn. problem! But to judge relative utility of OZ problems, we may have to "linearize" f_i Genc.

A Not Bad feature of forp: That each problem contributes additively to the total Genc; so one can evaluate the "contribution to a certain problem (cost)" of certain trial methods off of GPD.

It is possible that "Linearizing f_i Genc" is necessary, but there is no way outside of that to estimate someone's utility is an arbitrary (?) function of the problem's outputs - this will

.07 how to do it done

So the forp may be Adequate.

A longer term Q is: the forp is optimization w.r.t. past only: If TM has ideas about future problems - it is not appropriate: e.g. A child wants to eventually contribute to TOE (Physics) - so success in Math & Physics studies is relevant w.r.t. this

~~TM~~ expects future options. It may be possible for TM to treat this as an OZ problem, w. fixed horizon.

.16 I'm not so sure about what the forp proposal in 58 1/2.29 - 59.07 is!

- Whether to optimize is done in GPD, or GPD₂!

Some ways to solve the TMZG problem!

.19 (a) For all OZ problems, the Genc is Max GPD₂ for the points that were found! In the case of INX probs: the best pt of the solving technique that found the soln. for OZ probs: OT (i.e. Business Problem).
" OZ probs \rightarrow Any future problems, it may be relevant as ordinary OZ probs.

Was also how constraint that forp by PC's must be bounded in some reasonable manner.

Re: OTZ "Any time" probs: They could be worked on as time sharing betw. OT's. - in which case a "best so far" forp result is always available.

In both INX & OZ probs (we do take account for CC used in all parts of forp in 58 1/2)

.26 In persn problems & some OZ probs: we may, for certain problems, have no PC for certain OT's, in which case, we don't try other methods much (if at all!) -

do we end up w. more work for those OT's. (forp & "Self Confirming Hypothesis")

.29 A smarter TM₂ would avoid this TRAP! How would do it is unclear (a).

\rightarrow GD.03 may help

.30 An additional difficulty: It seems to me that forp prob only a few very general prob solving OT's that would be useful to forp really fast.

So I want TM also a such method that does not create work quickly!

.34 Stuck at a local optimum. \rightarrow GD.03 may help

.35 SN For OZ probs: Since the time used to test each O.T. is the same, we really simply try them in PC order! - This seems rather strange that it should be irrelevant! In the "Anytime" form, perhaps it is important? Look at my "Ox." report on "Optimum Sup. Str." in the 055 report Notes, this identified: If one has some constraint in PC order, this (35) works fine; but if one doesn't then such is a v.g. way. unclear as to how good my blockings will be in giving "PC order"!

\rightarrow GD.01.

.01 : 59.40! So 59.19 is perhaps to present TM2G. It does not involve "maximization of utility"
Nor even is for various problems: But it does involve "is" in choosing how much time
to spend on various "parts" of GPD, during "improvement" of GPD.

Local Optimize

Also, present 7. (perhaps) Bug of 59.26-.29 ("Self conforming hypot") (2 assoc. diff of 59.30-.34)

.03

Using conditional search, rather than searching in least 2 pc order may get
around to diffs of 59.26-.34: I think this will work, both for OZ & "Anytime problems".

.31

So, say the TM2G of 59.16 ff is adequate.

[The I'd like, also, to consider the TM2G Real tried to maximize the utility of each type of problem,
& maximize expected total utility.] → See 61.04

.08 My recent analysis of OPl. Sep. 1989 (V1989) of 59.37 (see recent comments on OS85)

suggests that Leuch often does not get per problem $\approx \frac{CPU}{PST}$ order. This may help w. it.

Diffy of 59.26-.29 (see 1.55)

So if TM2G was maximally 0.4. — what how does the rest of TM work?

.13

W/ solving problems, using an initial GPD. (When maximize time). Presumably this time has
concs, instructions, in it \geq + initial prob and easy to solve. After 2 or 4
no. of probs are solved, we want to "improve GPD". Now each type of OZ
problem can have its own way of doing OPEN. We can have special OZ's to deal
w. opens for INU probs, Leuch probs, for open or inducting, By inducting, — Present
~~OT's that TM2 (P.G.P.D) decides on, after some looking~~
both what OZ problem is (or what parts of the GPD we use to "improve")

T. Bug Q is: Could I actually get .13 of running 25 I'm describe?

Initially, it was only possible to do very limited improvement of GPD, because it is very
[limited in the kinds of regys it can recognize. — But after it has (and to solve the kinds
of problems for which solving are relevant to more complex regys of the GPD — then it
can help in that way.

Meanwhile, I will try to factor in set of OZ's — so that TM can max. usability over
diverse users, better, OZ's.

.27

SN A note on: TM2G of 59.16 ff: As stated it is $\frac{CPU}{PST}$, where

is the per pct of O.Z's or prob. soln. method that got to "best soln.":

We will probably want to change this to $\frac{MAX \sum w_i f_i}{\sum w_i}$, since certain probs are more
important (to users, to USR). (≡ $\frac{MAX \sum w_i f_i}{\sum w_i}$)

.31: 59.29

SN Actually, in Leuch the OZ probs, one spends PST on the best O.Z. — So that does sound like
a "Self conforming by preference" — The one does spend/time on the best O.Z.'s.

.33

Another way (probably better than Leuch) is to try O.Z.'s in PC order, using something for each.
Then do $t \rightarrow 2t$ etc. (This would be Anytime problem).

.33 sounds "Not so bad", since other OZ's get tried for full time T & can be legitimately compared
to comp w. $\max P_i$

.37

Re: 59.16 again, it does not take account to time needed to evaluate GPD's.

.38

The idea: It's it take time T_p to generate O.Z.'s, then it has the max. last time available for testing it.
In beginning, in INU problems of large CC, regeneration cost of O.Z.'s or possibly very slow
could become a very small part of the total CC for that cond. — O.Z.'s. otherwise we can get → 61.01.

.01: 60.40 V.G. GPD values at transmission CC, if we use full algorithm RLP for a very long time. - So ca. of any (sub) estimator will always has to be considered.

.02 So: 2 contending "Sols": to TM2G:
1) +. soln of 59.16 ff: w. used 60.38 on how CC. ~~some~~ info is implicitly methoded.
2) The Linearized "Garc" method: T. ^{final} presented soln. to book problem is given a "Garc"; In the case of 02 (2/o prdu problems); ~~what~~ what was "Utility" to ~~the~~ UR of soln pos past soln.? In case of ENV problems; how important was to soln. to user.? How bad was it needed to solve that problem - mind w. how imp. to problem (soln) was to user.

Perhaps Relevant to .02: Given some GPD_{1,2}! TM wants to "improve it"! It's not clear from TM2G soln. used, as to just what parts of the GPD should be worked on. Say GPD gets a low "score" for works on a particular problem. It is not clear that simply because of that low score, that this would be a good area for TM to try to "improve" partly Decisions of this kind are reduction problems.

.14: 3/8/01 It seems from Juergen 7.01 (3/8/01) ff, Part m (such, only a search by a PSO-type - (or "Z141 type") will not make Lsach diff., or modify it significantly. A possibly big problem would be "long" during Lsach: Actually changing the GPD during the Lsach would seem to not be diff. - Hm. Say one is doing T ← T Lsach - one is way type = "round" of T ← T: One has rejected certain disks because of CC > T pc's. If some of those pc's would change because of the change in GPD, it would react to rest of that round. The rest in the forthcoming 1/2 of the round that were changed, will modify the Lsach results.

It's hard to see clearly what the effect of GPD modification will have on Lsach: If what looks like a very serious modification occurs, one could abort the "round" & restart at "Round" Note that we can normally do TM2 stuff & TM1 stuff "seamlessly" by "Time Share".

.26 T. force may not be too important! At high level of task, TM will have devised search problem-solving methods that include search w. long, doing of search.

RA: .14 ff & .26: "If all info is in PD" This, ideally would mean that any techniques for "learning betw. trials" would be incorporated into the system rather than special techniques for particular kinds of problems. If it turned out that there was a better way than Lsach that worked for all problems, then TM would always use that method for all problems: T. GPD would not be pc & 1. - T. System would be theoretically "Lsach" but, practically, not Lsach at all.

So a big Q is: "To what extent can all info be put into P.D.?" There are 3 standard Arguments: (2). 01
1) If a better method is known, we should be able to put that into the P.D.
2) (Almost) all historical facts change operatively by changing order of trials. - It would seem that that could be done via changing P.D. using search, - trial order in Lsach is "2" see pe

3) To insert a new P.D. device that set of examples that never could have been (read from "): Then incorporate that into the system rather than special techniques for particular kinds of problems. If it turned out that there was a better way than Lsach that worked for all problems, then TM would always use that method for all problems: T. GPD would not be pc & 1. - T. System would be theoretically "Lsach" but, practically, not Lsach at all.
(+) the implications of any data/obj, into the GPD (This doesn't seem to include those obtained by "Logical Reasoning")
T. see see: This is idea that anything that TM doesn't put ("traces") is not available data for TM2.

Relax More General

(and more to this!)

MAYBE IMPORTANT; LEARNING DURING LSECH.

(19)

- 01: some imp. pts (1) Review of 51.25 frs not bad!
- (2) T. Idea of 54.57 on how to write a task is perhaps very good!

A More Detailed Analysis of flow Chart of TM: Steady State.

1) TM solves a new problem: It takes time t_0

2) TM spends t_0 time on Updating/improving GPD. Final GPD can have various things: w. assoc p's.
 [Input is always problem: (?)] ^{Any alternatives?} Output can be list of cards in PC order, so that all poss. cards can include (2) listing not nearly in PC order, but hyper PC cards tend to be ~~in~~ in list.

- 09: Usually (if not always) another mark a pt. in t. / (or) say a ^{are the} ~~new~~ P's \rightarrow a certain value \tilde{P}_c , set of ~~the~~ point. (3) In best ~~of~~ each card has a unique identifier, ~~and~~ ~~is~~ ~~at~~ ~~a~~ ~~memory~~ ~~place~~ ~~of~~ ~~which~~ ~~we~~ ~~can~~ ~~write~~ ~~properties~~ ~~(slots)~~. A ~~set~~ ~~of~~ ~~slots~~ ~~give~~ ~~PC~~ ~~of~~ ~~card~~, ~~comp~~ ~~memory~~
- (1) amount of time worked on this card, (2) order on list (3) perhaps other unique identifying thing or no.

In the T.S. middle of Lsck, we go thru list of cards & we spend an amt of time working out its PC. At a certain point in list, we will stop. A better way: At each round of update, we bring total amt of time spent on a card up to a new hyper threshold. ~~then~~ then for each threshold, there will be a pt in list of cards at which after which no update work is poss.

(i.e., amt to be done is < 1 unit). We then start new round w/ hyper threshold, T.

Individual thresholds are $P_c \cdot T$.

(19)

At 1. step, i.e., time, unless we have ~~any~~ "learning" (\geq Mod. of GPDs) doing

this time shared across of cards, this "learning" will change PC's of cards. ~~we want to change ordering of cards (to some extent) of PC~~

Just how this affects (i.e., Lsck is unclear): It will depend on how reordering is manifest.

One way, would be ~~to~~ ^{to} ~~change~~ ~~ordering~~ ~~of~~ ~~PC's~~, say we have 6 cards in "rule PC order"

(i.e., ~~of~~ ~~order~~ ~~meaning~~), ⁽⁰⁹⁾ ~~perhaps~~, ~~but~~ ~~I~~ ~~am~~ ~~not~~ ~~sure~~ ~~(if~~ ~~it~~ ~~is~~ ~~that~~ ~~way)~~. Anyway, in T.S. mode of Lsck!

We now have this bunch of cards that have been worked on different amounts, but using a diff.

p. id. - so each card has a new $\frac{CC_j}{PC_j}$; say they are ~~not~~ worked on up to new.

But if "new long" all $\frac{CC_j}{PC_j}$'s are about the same. - Now they are ~~not~~ diff: so ideally,

we work on 1 card w. smallest $\frac{CC_j}{PC_j}$ - until we have 2 cards w. same.

i.e., $(\frac{CC_j}{PC_j})$, then we have two equally best - then, they will be best

3 cards at $= L_{C_j}$: we T. share ~~the~~ time until we have 4 cards of ~~the~~ ~~same~~ ~~quality~~ ~~(or~~ ~~better)~~. Its like pouring water into a hole w. rocks of different height, ^{up from bottom of hole.} the coming

we don't work on 2 card unless its PC_j is such that $\frac{CC_j}{PC_j}$ would bring it up to current

threshold. i.e., $PC_j > \frac{CC_j}{\text{current threshold}}$ (\Rightarrow current threshold \neq , we consider cards of lower PC_j)

It may be that the new PC_j distribution always ~~comes~~ ⁽⁰⁹⁾ so we ~~can~~ ~~usually~~ know when to quit ~~and~~ search for new cards to update w. CC_j in view is to new threshold.

(19) if more work: depends on the kind of output that GPD gives, but anyway, it does

look like ~~the~~ ~~best~~ ~~does~~ ~~do~~ ~~CC~~ at just ~~those~~ ~~PC's~~ that are most promising, at high rates of ~~the~~ ~~rate~~ ~~of~~ ~~CC~~ their "promise". It may be poss. to prove that, in some sense, this is the best way to do a TSQ!

Also! Perhaps maybe so that cards can have access to "it" (fraction of time for "s.i." during their "candidate")

19 MAKING IMPORTANT: LEARNING DURING LSRC.

- .01: Some imp. pts: 1 Review of 51.25 ft/2 not bad!
- .02: 2 T. Idea of 54.57 on how to write a TSQ is perhaps very good!

A More Detailed Analysis of Flo Chart of TM: Steady State.

1) TM solves a new problem: It takes time α to

2) TM spends α time to on Updating/improving GPD. Final GPD can be various things:

[Input is always problem: (?)] ^{Alternatives:} Output can be history of costs in PC order, so that all past costs were included ^{w/ some pos'n.}

(?) (costs not only in PC order, but higher PC costs tend to be earlier in list.)

Usually (direct choice) chosen work α p.c. in the list & say it has α p.c.s α > a certain value α , its own

enter this point: 3 In each CD 2 each cond has a unique identifier, 2 and memory place

(name) at which we put it (properties) (slots). A slot gives PC of cond, memory

1 amount of time worked on registers, 2 order in list 3 perhaps other unique identifying strategy:

In the T.S. model of LSRC, we go through list of costs & we spend an amt. of time working out of its PC. At a certain point in list, we will stop. A better way: At a certain number of cond, we bring total amt of time spent on α cond up to a new type threshold. This for each threshold, there will be a pt in list of cond at which after which no update work is pers'l.

(i.e. amt to be done is \leq (unit)). We then start new round w/ higher threshold, T.

Individual thresholds are PC/T.

19 At 1. step, it's fine, unless we have no learning (= Modifi. of GPDs) during this time shared basket of costs. This "learning" will change PC's of costs. Memory

~~we want to change reordering of costs (to some extent of PC)~~

Just how big effects? Learning is unclear: It will depend much on how reordering is handled.

One way, would be ~~to~~ After each "reorder. of PC's", say we have to cond in "uff PC order"

"uff" order meaning: 2 parties, but 2 I'm not sure (this is hard). Anyway, in T.S. model of LSRC:

We now have this bunch of costs that have been worked on differently, but using a diff. p.c.

— so each cond has a new $\frac{CC_j}{PC_j}$; α being α work done up to now.

Back to "new loop" all $\frac{CC_j}{PC_j}$'s come about the same. — Now they are diff! So: ideally,



We work on j : cond w. smallest $\frac{CC_j}{PC_j}$ — until we have α cond. w/ same $\frac{CC_j}{PC_j}$, then we have that many before going on to the next

α costs of = LE_j : we T. show before then until we have α cond of $\frac{CC_j}{PC_j}$ = LE_j , set. Its like pouring water into a tank w. α cond at bottom of tank. $\frac{1}{\alpha}$ unit of time would bring it up to current

We don't work on α cond unless its $\frac{CC_j}{PC_j}$ is such that $\frac{1}{\alpha}$

Threshold: if $\frac{CC_j}{PC_j} > \frac{1}{\alpha}$ (current threshold), we consider cond of local PC j

It may be that the new PC distribution always contains $\frac{1}{\alpha}$ so we eventually know when to quit searching for new cond to update w. $\frac{CC_j}{PC_j}$ in view to the new threshold \rightarrow 83.18!

19 off work! depends on the decided output that GPD gives, but anyway, it does

Last time off does do at just those off that are most promising, at a rate α to their "promise". It may be possible to prove that, in some sense, this is the best way to do a TSQ!

Also! Perhaps errors so that cond can have access to "ie" (fraction of time for "s.i." during their "candidate")

TM2G

01:62:40 Next problem of impact:

TM2G

First consider opt. GPDs: GPD_i is a PB on certain time scale... INV prob G values... actually occurred (May be more at, for more desirable cards?)... PC for computer but a very long time to compute result... (A linear Ordering)

06

Say GPD₂ results in a certain set of solutions to problem Set (ZTSQ). Consider OZ probs.

05 09

OT_i work for time T_i is hard... T-α_i ... β_i... (Note 18) Get Note 20!

13

For INV probs, on cards will have solved problem: we want GPD₂ to assign max PC to that card... A GPD that assigns some PC's to numbers as a default... but this is not reflected in the analysis of 05-13... work on both OZ & INV probs

08

On the other hand, time spent improving GPD is more effective to TM's long term goals, than is simply working on a "production" problem! - So in 08 & 09, we may not be doing it right!

21

Woops: (18) compares Time to discover GPD vs. time to execute GPD. Riknow if interested in Execution time.

Discovery time is related to Horzaph (18)

Considering 2 INV cards: card today... (longer) including Gaps of all PC's... (23)

T: total sale times $\frac{CC_1}{PC_1} + \frac{CC_2}{PC_2}$. CC₁ in dollars generated & shipped. So we want to minimize this.

we may or may not want to weight each problem set.

26

for OZ problems: "Generate & test" always takes longer. So want to use U₁ to know, where U₁ is ability of sale to inh problem. obtain.

28

Am! nice now, its whatever the number just leads to OZ sale works! \$ 59.35 "dollars" in PC's. See Notes on (05 1995): 4.15; i for time share version, 1.28 (1/bid);

30

In the T3 version we spend 2 PC's... Total time < T, so it takes 1/PC_i time to buy as a optimum. So, 2 GPD's Good to satisfy 4 PC's to 6 OT's Part 6 & 6.

In the T2 version of OZ work: use 2 OT's per for 2 OT's: for all OT's: After own T2T, cycle.

35

look at time spent on best OT. When it is ↑ (i limit) in problem, stop. (02,00)

Going back to the GPD evaluation for OZ probs: The time PC_i T includes time to generate to O.T. & get its PC: - so we simply want 1 OT that won't have PC's by PC. as possible.

- First note in OZ probs: 3 ways to solve prob: 1) try OT's in PC order for time T cost (Best way) 2) Time shared, time limit on screen, better when T has been spent on OT. 3) T2T not much different from time share;

When run is over, a GPD is "good" if it assigns by PC to OT that got best (mid of row (very important))

TM2G (Lincoln?) (101-04)

.01: 63.40 : so for each OZ problem, one can benchmarked Tall then GDP is better than which other GDP.
 The time for optimum is $\propto \frac{1}{p_{c1}}$; ||| for env. probs, Time to solve is $\propto \frac{CC1}{p_{c1}}$.
 For All env probs, total solve time is $\propto \frac{CC1}{p_{c1}}$; for OZ probs, total solve time is $\propto \frac{1}{p_{c1}}$
 Minimize to sum $\frac{CC1}{p_{c1}} + \frac{1}{p_{c1}}$?

This is Trivial for
 is OZ problem
 p_{c1} is p_{c1}
 OZ being best Soln

.05 3/15/01 SN To get a good ~~set~~ set of OT's for OZ problems: I expected to take \rightarrow 70.04

(any set of OT's that I know of, a "Factor" drawn into a set of simple concs P_{c1}'s
 adequate to express set of OT's \Rightarrow a stretch & Long: Then add a few mts, etc.
 (any is not already "universal". This is a kind of "Jump start" - otherwise it would take
 an Enormous amt of training to get TM to do much "Self-improvement".

T. large amount of necessity puts a lot of "Bias" into "Self-improvement", but to
 Factorization may help to overcome that Bias. - T. Factors - combining them makes a system
 more usefully "Universal".

.20 It is clear that to force. could be done - in any case of TM's Limit!
 It amounts to my designing a kind of "Expert System" but much better than
 Normal Expert Systems, since it really "Learns" in a more optimum way.
 Of course errors & Bias Problems. - But I'm not sure it's really any worse than using
 any TSC & Rot. OMC.

.25 One Very Fear I have about much "Hand peng" on TM is little details that
 TM needs to Learn by itself. I don't think that would much rather have
 TM learn a Domain than have me F to P it, because Pray I know TM has
 to learn to learn that task. This is not to do of "little details" here. -
 \rightarrow in 20, the system would not needly be able to jump off Prob. Solving backwards it knows
 the "little details" of 25 is a worry when I insert the sections of 6 conc. net.
 But I think a separate for TM to "discover" a soln. to a problem.

3/17/01

IDSIA

GAMB House Thru 1 & LSRCH

LSRCH not really near optimum in essence of GHTI

Mon all ppts

00: On/true $\frac{CC_i}{PC_i}$ order v.s. T.S. LSRCH = $\frac{CC_i}{PC_i}$ $\left(\frac{CC_i}{PC_i}\right)^{-1}$ if $i < j$.
 $\frac{1}{PC_i} < \frac{1}{PC_j}$; $\frac{CC_i}{PC_i} > \frac{CC_j}{PC_j}$ | we want to know $\sum_{i=1}^J CC_i$ v.s. $\frac{CC_j}{PC_j}$

even no one has C

See 72.01 for optimal post.

03: When LSRCH terminates in ITS mode. If has worked on all cands w. $\frac{CC_i}{PC_i} < \frac{CC_j}{PC_j} \equiv LC_j$
 This would be the 2nd best GHTI works on (optimum). In addition, it works on all other cands

w. $PC_i > PC_j$; $\frac{CC_i}{PC_i} > \frac{CC_j}{PC_j} \equiv LC_j > T$ Amt of work done on Perm

is $\sum PC_i \cdot T$ $\sum PC_i < T$ so unhappy is $< T$ and $\frac{CC_i}{PC_i} < 2$ times

010 optimum is $\sum CC_i$ Best waste occurs if $\frac{CC_i}{PC_i}$ were known in advanced down that order.

The factor is < 2 by the subtraction rule $\left(\sum_{j=1}^i PC_j + \sum_{j=i+1}^J PC_j \right)$ could be larger.
 So the factor could be not far from 1. $\left(\frac{CC_i}{PC_i} > T \right)$
 or the factor is $1 + \sum_{i=1}^j PC_i$ $T < C$
 PC_j room PL order (go PC_j first)

020 optimum $CC_0 = \sum CC_i$ $\left(\frac{1}{2} | LC_j > LC_j \right)$
 LSRCH $CC_0 + \sum_{i=1}^J PC_i \geq LC_j \equiv \sum_{i=1}^J T$ $LC_j \equiv T$

Consider extreme case: CC_i for $\left(\frac{CC_i}{PC_i} \leq \frac{CC_j}{PC_j} \right)$ $\left(LC_i > LC_j \right)$
 are all very small. $\frac{CC_j}{PC_j} = 2$, $PC_j = \frac{1}{101}$
 so $\frac{CC_i}{PC_i} = 2$ $\frac{CC_j}{PC_j} = \frac{10}{.5} = 20$
 e.g. -20 suggests that ratio of CC_0 to T $< \dots$

Must be supply! (see that CC_0 could be $< T$) !

Must find LSRCH can be supply versus that doing trials in LSRCH order

030 $\sum CC_i$ $\left(\frac{CC_i}{PC_i} \geq T \right) = \left(\frac{CC_i}{PC_i} < T \right) \geq \sum CC_i < T \sum PC_i$
 $\sum PC_i$ $\left(\frac{CC_i}{PC_i} < T \right)$ compare to $T \sum PC_i + (LC_i > T)$
 \therefore compare $\sum PC_i$ $\sum PC_i$ $\left(\frac{CC_i}{PC_i} < T \right)$ $\left(\frac{CC_i}{PC_i} > T \right)$
 CC_i would be $\frac{CC_i}{PC_i} < T$

N.B Even if LSRCH is not near optimum (int. sense of GHTI) It still may be optimal near optimum for in a likely unknown CC_i 's of cands. Still, we could use some analysis. \rightarrow see 68.00

we can divide to (total of all PC_i) in to 2 sets so that $\sum PC_i$ of each set is $\frac{1}{2}$ of total. Any ratio supply ratio to the other. But assign CC_i 's to that way a consistency e. 2nd set of a large. \rightarrow see 68.00
 so LSRCH can be worse than "Pure LSRCH order" by a big factor: (The $\frac{CC_i}{PC_i}$ ratio)

00 : Remember IT may be poss. to use a rather simple Global flowchart for TM, is
 01 depend on arbitrary simple Cands (Present in Proust's VOS ~~and~~ include more such variables ξ - also program
 (during Proust's own some times) A fraction of time spent on (ξ, η) .
 i.e. Input is problem data: this goes to GPD, which outputs a set of commands u. Proust's.
 A fair ~~fraction~~ fraction of Global CC is used for improving GPD.

02 Only pure learn used (no learning from trials, possibly no long before T ← KT)

03 "Improving GPD" does not include much (if any) CC spent on ξ or η problem of improving GPD.
 i.e. The system need not be totally recursive ~~in~~ in this particular way (Bit size $\frac{.11-.16}{.12}$
 This may not reduce complexity
 of a system - also will affect)

We depend on ~~fairly good~~ fairly good set of initial OT's for ξ or η problems to solve the job done

11 But what about FNV prob. > Well, they are ~~not~~ ^{done} by a gradually improving GPD.

The ~~OT's~~ OT's to improve GPD are fixed at Birth (It does seem strange that ξ OT's
 used for "improving GPD" should be constant, but these used for "non GPD ξ problems" should
 be allowed to improve/grow!) It may well be that having to GPD assign (possibly),
 OT's to be used or ~~in~~ in improving itself involves ~~no~~ no logical defects: Super friendly

16 it would seem that there should be no problem (Kern PROBLEM!) \odot The $(\xi, \eta, \rho, \sigma)$'s
 are ~~the~~ the output of GPD when it is put in to dem of itself is a limit. This is a "discrepancy"

20 "Delta time" f. b. system. I don't see how it could oscillate or misbehave" much!

22 Other than ITS design, A big problem may be TM2G; It is to overall System Core.

23 Another Q: What I am quizzing about is T. cands' control of K (fraction of time used for ξ, η) present
 probably a better way would be to allow TM to ~~have~~ do ξ, η for itself, then turn ξ, η off from do ~~other~~ other problem
 for a while, etc. - periodically doing ξ, η in "pulses" so as to take advantage of recent cond. traces ("traces")

So initially, K is set at .5, but a cand can change it or use its own switching algorithm
 betw. ξ, η & "present" (Actually, "time share" is a special case of such algs)
 So, it will have to learn to control switching betw. ξ, η & ~~or~~ ξ, η (non- ξ, η).

I think it is feasible in "reasonable time".

Another Q: How does it do "logical reasoning"? ~~preliminary~~ preliminary steps!

- 1) Logical reasoning is ~~an~~ an input to induction reasoning: ξ, η or (ξ, η) .
- 2) "Teach" TM to do logical problems "external to itself". Then it can solve its internal logical
 problems "by Analogy" w. its operations used to solve "external" logical problems.

I haven't worked out details yet, so I don't really know if it will work.

37 Q: What about TM2's induction? ~~How Good is Z141 Model?~~

I don't immediately see how it would be applied to f. SAAS ANZ problem!
 At first GPD is not conditional but an unconditional GPD.

Perhaps 37 is: most difficult problem of Man all!

See 26.22 for
 min reasoning Diff's
 in TM.

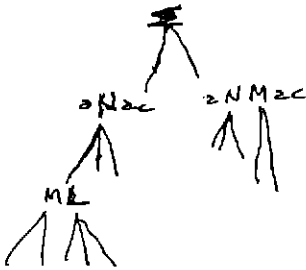
67.15

00: STOCHASTIC LANGS: In a Stoch CFG: The generator of the ss may be regarded

as a stochastic finite state machine: Each N.T. ~~type~~ corresponds to a state.

Say we have $N_1 \rightarrow ABA \mid bM \dots$

Better still, Consider +. Generating TREE! We just trace out the choices needed for all of the NT's in a simple exhaustive way. The successive terminals that we jump to, ~~are~~ correspond to successive states of "ASM".



This Doesn't Seem to Work! In a CFG,

The state one is at any particular time, completely characterizes the future of the system. This is not true in graph(00-10)

Given the left edge, at "M", one jumps next to "L". On the right edge "M" completes to ss.

15: 66.40: Consider the TSQ:

$$x_1 x_2 x_3 x_4 x_5$$

$$3+5=8 \text{ (many cases)}$$

$$\text{then } 4+9=?$$

1) The model $x_1, x_3, \text{ add} \rightarrow x_5$ works.

2) $x_1, x_3 \rightarrow x_5$ (many cases) then $4+9=?$

$x_1, x_3 \text{ sub} \rightarrow x_5$ works.

$x_1, x_3 \rightarrow x_5$ (many cases) then $4+9=?$

or $3+5=8, 2$ (1, mind) then $4+9=?$

22: There is a Q of dividing up the "induction" between TM_1 & TM_2 . How much to cond is able to do, v.s. how much "S2" does.

Basic Method of operation (for induction probs): Input one problem w. corpus.

1) TM finds (using current β -PD & LSK) one or more codes that covers.

2) Assign corpus w. new problems. Using fact that old codes fit old corpus, TM tries to find code for entire generated corpus. This can be regarded as a special type of OZ problem, peculiar to induction probs, w. "Specialist" OT(s).

(SN) on OSL: Just as MMI MDL OR doesn't work all for OSL, it works poorly for small ssz.

Obtain TSQ analysis: Be sure we are able to do one (e.g. 2.2.3) step (very small ssz)

35: 30: Trouble is, while I seem to have a good understanding of TM_1, TM_2 interaction on a reasonable level, I seem to get into describable problems when I try to do a TSQ! Perhaps one problem is

"The Devil is in the Details" As long as I stay at a high level of meta-analysis (or "English") things are O.K. — But as soon as I start on the detailed system. — It doesn't go together well, at all!

69.00 Spec
75.16 ft mps/mph wall on d.

3/19/01 IDS/A

T.S. Lench: "Optionality": How Argues: Also for Long during Lench

.00: ^{Spec}(65, 70): Are "How" argts. to show T.S. (11) Lench's Best? Also if long occurs during Lench. (ie No. of P.D.) then t. best "Greedy" alg. is 62.19 ft.

T. argt: At any time, the best "path of success / units or expended" is to continue. ~~wait~~ to just

~~the~~ ^{1st} $\frac{P_{ct}}{CC}$. In t.s. mode, one doesn't know $\frac{P_{ct}}{CC}$ of any off. Cards (other than "tested to death" "completed"). One has, instead ~~the~~ an upper bound on each $\frac{P_{ct}}{CC}$.

(for cards not yet worked, it is $\frac{P_{ct}}{CC}$ (Pain is ~~in~~ min time worked on

2 cards: \pm time of first + time to switch cards)

One then works out card w. ^{highest} ~~best~~ upper bound (on $\frac{P_{ct}}{CC}$) of all cards.

As no work (unsuccessfully) on a card, its $\frac{P_{ct}}{CC}$ upper bound is on hold it = that of

another card, - so one then ~~can~~ shares ~~both~~ those 2 cards.

Thus Arg. would also justify 62.19, if it was used ~~to work on~~ to work on the "card w. most productive" at any time. Thus it is a

"Most Greedy" approach.

If (in 62.19 in previous) we had any approval in to about "combinations"

to find cards, then we would be less ^{likely to} "Greedy".

Good answer!

.10

.20

.30

MUCH

00: 67.40 spec. So I'm better at "Top Down" Analysis!

01 Perhaps look at some of my recent "Bottom up" analysis of ANL: try to see just what I was doing

02 that was so (diff/calc/cond/counter/productive)

Remember! Each action is either TM_1 or TM_2 .

→ I use 01-02: Note each (instance/cond of (in/directly/relevantly used/considered); usually it will be in TM_1 — but sometimes in TM_2 , in some cases, it should occur later in TM_1 (or TM_2).

Also, for each copy: Record Q of whether I want to use it "now" or wait until later —

09 i.e. "now" may mean (inserted/added) in to TM_1 — "later" means acquisition via a later part of the TSQ.

In this preliminary Analysis: (concentration) T. Sequence
detection of those Regys. T. Sequence/set of regys, is not out. method of
— for inclusion in problems T. Sequence/set of regys, is not out. method of → □
A cond need not be (initially) in the format of a (conventional) computer pgm. It can be a list of "regy" detectors —

These must be equiv. (in some sense) to compression algos. Each Set of "regys" will have a pc. —

At first (Default) This will be the product of the pc's of the individual regys.

18 Hvr., another way to look at this! If there are several regys in a corpus, one can use each of them to get a sequence of compressions — evaluated sequentially, "Coding and Recoding" ©.

→ So: for a regy. — write down, for each problem, a set of regys that would be adequate to solve it. Whether I want these regys to be primitive, or ~~primitive~~ have some of them be "trnd", can be considered later.

In the case of inclusion problems: T. cond will be means of finding regys: so an "elementary" cond. would be a simple "single" regy. A more complex cond could embed several regys. An ~~more~~ intra corpus cond would examine corpus & decide which regys to try. (This is essentially what the GPD does! —

27 but the cond is less restricted in how it can decide what regys to try: It need not use pcs # 2/0 ~~to~~ to Lurch. A cond can (probably) do a "call" to System as a whole, to give that system a problem to solve —

Or, it may "call" the GPD w. a problem dem input for it; & use the output of the GPD in a novel (not necessarily Lurch) way.

The "GPD" must have a way (after looking at the input problem), to devise a system for creating a stochastic lang → whose outputs are conds.

2/1/01 (26!) Another way to write regys: first list a seq of problems w. rather large CJS's data set. Then insert problems & CJS's.

TM2G 70.34 - 70: 71.01ff

OSL diffy .06

"The factor of 4"
 (13)

.00: 70.90 needed to calculate per.

Her note that time needed to calculate per is not related or "same" time spent on sid. (ie. "improving GPD").

So I Th. **TM2G** of 64.01.04 may be approximately ok. - but it may sometimes take long to calculate exactly.

- .03 Another point Re: T. often of **TM2G**: While we want to minimize $\sum \frac{C_i}{P_i} + \sum \frac{P_i}{C_i}$ we are constrained in how we can do it. The P_i 's have to be obtained by logit. induction: ALP criterion. (09)
- .05 If we use normal ALP in the normal way, we have to add in special provisions for OSL.
- .06 **Re: OSL**: If the "zero coding" of MDL (→ approx to ALP) don't deal w. OSL because $s_{i2} = 1$, then some techniques should be way off for $s_{i2} = 1, 2, 3 \dots$ I really have to analyze this!
 Note that my Analysis of "2M" works ok. for OSL.

.09: (05) So we may have to look for "new" reps in our GPD corpus that are capable of being calculated quickly. This is something I don't remember having ever thought about!

.12 Also, of course, ~~we~~ we look for ways to speedup GPD's evals. of P_i 's. (74.00)

.13 On "factor of 4": Actually a factor of 2
 The factor of 2 is due to the uncertainty of the "R parameter." The factor of 2 is correct if we mean that the system operating at ~~2x~~ with a certain clock speed is

.18 at least as good as any system operating at ~~2x~~ that clock speed. However, a factor of 2 in clock speed can modify the problem solving capability of any system by a very large amount.

.21 A way to look at it: $\frac{G_{total}}{S_{i2}}$ My system using / clock rate: 2 / Systems in it! One
 does my production jobs full time on one, S_{i2} full time on another is at least as good as any system day
 full time on one only. At any time, my system has spend at least as much time $\frac{G}{S_{i2}}$ as the optimum
 system. (also at least as much time fine of production jobs as the optimum system.)

"Factor of 2" here, is used in the same sense as "a constant factor" used in deriving the efficiency of L₂rch. The results of more finely optimized system can be anonymously better (say exponential in size of problem) better than L₂rch - - - in terms of the given fact. OZ problem

Angts re: .21 Agreed: While total time ~~at~~ spends on S_{i2} \geq that spent by optimum system,
 a) The TIME may be distributed differently w.r.t non- S_{i2} problems. (This is opt. is would, here, because .23-.24)
 b) The time may be " " " " This difference can be in different aspects of S_{i2} .

.33 In General, S_{i2} can be localized to certain problems or problem areas, or more global. The Global credit
 .34 assignment function would seem to deal w. this, but I'm not at all certain of it.

So .33-.34 may be stronger support. v.s. "factor of 2".

.36 An Arg. for the factor of 2: If the searches in TM get very long (per problem) the amt. of time spent on S_{i2} also \uparrow . I think the effort is to effectively spend "more" time on S_{i2} , because a long search does not nearly mean more complex calcn. during the non- S_{i2} problems.

Spent 89.08

145
70

BIAS IN TM in Computing PC RATIOS 1.00 See 77.00 on bias.

.00 : A Poss. Source of serious Bias in Simulation: Say C_{and_1} & C_{and_2} are 2 codes that have been evaluated as production or problem solvs. C_{and_1} has some ~~code~~ G_1 . (say $7=1/2$). G_1 is $>$ (better than) G_2 . We want TM to assign a higher PC to C_{and_1} than C_{and_2} in GPD_2 . TM does this by first computing GPD_1 , the pd of various G-outputs for C_{and_1} for Time T & ~~some other~~ times, if we use $T=2T$ (See 72.34-36) or T.S.

We want TM to be completely "unbiased" in computing these \geq pd's — because there is much "Value" in T on if it (biasedly) ends up. C_{and_1} having more pd "weight" for small t, then does C_{and_2} .

.09 Since TM would normally be rewarded for such "biased" behaviors TM can do any kind of behavior it likes, this would seem like a serious problem. Can we find ways to keep TM "unbiased"?

.11 Can we find ways for TM to treat C_{and_1} & its data f_i in a way "indep of f_i "?

I think I ran into the forg. problem in computing ratios of \geq pc's — which is the main problem

.12 in Inductive Inference.

⊙ Can we find ways to restrict TM's behavior — its methods of solving inductive problem — so that we achieve .11, but we also allow TM to be "adequately Universal".

This can be very serious when we have a long list of codes in our representing finite PC's.

We most (presumably) spend as much time on the low PC codes as the high PC codes (?)

.13 { HVR, maybe not so! Say we have a "Summary Machine" (no OCL). It is easy to generate PC's in an unbiased way by making random codes or exhaustive search over a adequate ~~some~~ space. For CBOO, there is no bias (other than the reference one) — so, perhaps use CBI, to do problem!

Gen. Conclusions about PC Bias! That it probably can be a problem! .18 may be an approach!

It may be that for many stock lots, bias is not a problem. (Even if there can be many possibilities of a thing, one could be biased in omitting many. T. "restriction" idea of .11 is attractive; look at ways one can be biased — Gen. Bias is "outlaw bias"!

→ 75.15, 35, on Bias
77.00

TM2G (100 ft)

Spec

.00: 71.12: While 71.03 is the TM2G, there are 2 aspects of it: One is getting a PD to maximize f. p.c. of t. corps: so $\Pi p_{ci} = \max$. Another is to modify the calcs. of t. GDP₂ so that the p.c. take

.02 as little time as poss. to calculate. The combination of these 2 effects is t. core of 71.03. $\sum \frac{CC_i}{PC_i} + \sum \frac{T_i}{PC_i}$. If most of t. work is in ~~the~~ production, then Πp_{ci} is t. main force. Only when we consider very slow compns. for f. p.c. does t. ~~main~~ final core of 71.03 become imp't.

Are there 2 conflicting Goals? : On one hand, we want a honest pd. (Is this max Πp_{ci} ? or is it more related to t. corps of GDP₁?) Always On t. other hand, we want min $\sum \frac{CC_i}{PC_i} + \sum \frac{T_i}{PC_i}$, which seems to be t. "Top Goal"

At present, I am quite confused about this: The $\Pi p_{ci} = \max$ goal is for the GDP₁ corps.

The $\sum \frac{CC_i}{PC_i} + \sum \frac{T_i}{PC_i}$ is for a different set of p.c.'s, derived from ~~these~~ p.c.'s.

Try this: Generate GDP₁ & GDP₂: Then, don't charge any time cost for generating of p.c.'s of GDP₂: This costs part of S_i's budget.

Try this: M_j is a p.c. that generates GDP₁ (via ~~the~~ RLP) and used p.c.'s for GDP₂, in time T_j. What if t. p.c. M_j is "best"?

Remember t. time cost of "generating p.c." (to Levin) is market plug-in: t. problem dem, is generating that (p.c. ~~and~~ cond) p.c. — More exactly, whenever we plug t. problem dem in, we

.17 get a seq. of (p.c.'s, cond) pairs, & each takes a certain amt. of time to generate.

.18 For ~~each~~ p.c. if t. time to generate in .17 is about t. same for all pairs; then, for any problem, in ~~the~~ all cases knowing time left to solve t. problem. Note for INV problems, ~~the~~ the

We spend some S_i time to generate GDP₁. To go from there to GDP₂ and t. output of .17 takes a certain amt. of time I': If ~~we~~ t. time opt. at t. ~~time~~ (prob) dem is inserted into GDP₂, is all S_i time.

Suppose we spend time I' to go from data to GDP₁. Presumably, GDP₁ is in standard form that can generate GDP₂ ~~and~~ and finally after plugging in t. problem dem to GDP₂ and getting .17

.25 Then TM2G is mainly to optimize t. core of GDP₁, & we can improve things further by modifying p.c.'s to take GDP₂ to .17; (using 71.03 as ~~the~~ core) but this will be ordinarily be done, or will it make much difference in TM & the.

.27 If ~~the~~ for GDP₁ is Πp_{ci} we can weight various problems by Πp_{ci} (i.e. \equiv w.b.). A p.c. that reduces error for GDP₂ to .17 (via most problem dem), would perhaps have core 71.03 (\equiv the)

.30 Write up ~~research~~ t. idea's accepted. 25 ff. This is solve an imp't. problem, & I want to

.31 remember how & why 75.00

.32 A next Q: Re: t. thing, TM2G! Say we have a GDP₁ & we have a small amt. of new data. How can we compare GDP₁ & GDP₂? to incorporate it into t. corps.

Which should we use on a new problem? — They are based on slightly different corps. GDP₂ may not have incorporated t. new data well. Even, t. "new problem" ... t. new data may not be relevant to it!

3/24/01 ID31A

TM2G 3:00 B.125 .13 ; 77.01

00 (Spec) 74.31 : So we had 2 competing Gores: I. earlier (more Global one was $\sum \frac{C_{c_i}}{P_{c_i}} + \sum \frac{T_{c_i}}{P_{c_i}}$)
Pc1's one for G-PD₂

02 More recent idea was Gore = $\prod P_{c_i} = \text{Max}_j P_{c_j}$'s from GPD₁ (raw data).

In .02, it is assumed that the output of GPD₁ is in some standard form, so it can be converted to GPD₂ w. not much cc. (T. machines of interaction was discussed in **MCT** analysis in 1997. I may have had some good ideas on how to do it w. relatively small cc.)

The Gore of 100 second more "global" I didn't have a clear idea as to what it meant. What is the obvious soln? would be $P_{c_i} = 1$ for all of the solns. What are the constraints on this Gore? Well, in .02 we have to sample the soln. (Pc1=1 for all solns. — For, y. constraints are put to only try what do to vary the Pc1's is to find more codes for the corpus. — and what?)

we could use same constraint for the Gore of .02.

BIAS → 77.01 on BIAS

.13 Here, if we really want to do best on either (.00 or .02) Gore, we will BIAS our search

so that codes for non-solns, occur less often (cannot do all!) — Giving the "desired"

.15 $P_{c_i} = 1$ for all problem solns. (That this will occur is unlikely!)

→ 76.09

.16 **SN** In sequential predn: Say C is the corpus to be continued. We want to maximize our knowledge of the lower bad of $\sum 2^{-P_{c_i}}$, where P_{c_i} is a code for each of which C is a prefix: i.e. codes for some C or "C extensions". The restriction is on total cc allowed to do this. If our goal is to maximize $\sum 2^{-P_{c_i}}$ in the available time (Just how could this introduce bias?)

.18 In **Seq predn**: Consider a lg. model of BAG predn: Univ. w. z inputs, 1 work tape, # - 1 output tape: Put ~~random~~ random input into ①. When machine stops, put random input into ②, where probability that output is member of corpus, then randomly select ② & put in another random input: The pd of this second output is desired / pd. on finite string. If not five string, is correct. Somehow, the defn. & implementation should not be much more difficult than Sequential predn.

Another try for BAG predn. (That makes it look more like Sequential predn).

Some simple machine as before; we want to find in time $\leq CB$ codes for input 1 and codes for input 2 that generate the entire corpus. Say l_i^1 is the i-th possible code for input 1. (no reason to stop (inputs to 1 have no output, error). $l_{i,j}^2$ is a code for the j-th corpus member, after l_i^1 input has stopped. In time $\leq CB$ we want to find codes l_i^1 & $l_{i,j}^2$ such that $l_i^1 l_{i,j}^2$ is a member of the corpus.)

.33
$$\sum_i z^{-l_i^1} \sum_j z^{-l_{i,j}^2} \quad \left[\text{total no. of elements in BAG (= corpus) is } \sum_i \sum_j z^{-l_{i,j}^2} \right]$$

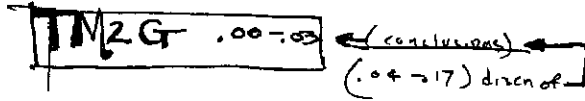
occurs r times, it is given r times, n is sum $\sum_{j=1}^n l_{i,j}^2$]

.35 After we get this sum to be as large as possible in time CB (being as "unbiased" as possible)

.36: Whatever that means!), we have a wild set of "members", l_i^1 , that we can't use for predn. (It's not sum of .33) but the wt. for the 2^D machine is $\prod_{j=1}^n \left(\sum_k z^{-l_{i,j}^2} \right)$ k summed over all codes

via machine 2, to the j-th corpus member. If a corpus member occurs r times, it takes the power, r times ($\equiv r$ th power).

3
4
6
7



.00 (Spec 75.15): In view of 75.16-.40: it appears (not 100% sure) that ~~we can't optimize~~ we can optimize either over 76.00 or 76.02 by finding many codes in available time (\equiv CB)
 Doing over 76.00 seems better, because that's what we're trying to do, but I suspect 76.02 may be easier to implement & may be about equivalent.

.03 Discussion: ~~TM2G .00-.03~~ AH! The reason we can't optimize 76.00 or .02 by sorting all PCs to be 1. ~~TM2G .00-.03~~ If we included only codes that desired priority codes, we would not ~~optimize~~ optimize. Eg. in 76.02 (if PCs = Max) we have to include (in INV problems) all trials and how long they were & when they try look & whether they are done or not. This is v. BAG corpus of 75.33-.40. If we only included codes for the desired C and T codes for other codes would be zero and if product priority 75.33 would be zero. Each product 75.33 has to include all data (not just the desired data). So, my impression is that for INV problems, $PC_i = 1$ for all i is not possible, & that we have to be unbiased. That probably v. same thing is true for OZ probs, & that probably v. same thing is true for the Gove of 75.00 ($\sum \frac{C_{i1}}{PC_i} + \sum \frac{T_i}{PC_i}$). So (0.00-0.03) may be correct. — But I really

.17 \rightarrow (how to go over this in data?) to see that (0.00-0.03) really is O.K.

.18 It is likely that we will use Gross approxs. to error of \pm zeros 76.00 — but it's important to know just what we are approximating!

- .20 OK: So say we understand TM2G (as 0.00-0.03): What's main diff's in TM?
- Before: T3Q itself: Some practical Q's:
- 1) T. lang(s) used to represent probs & solns. : (probly \rightarrow to List of Form, APL, Macro/alg).
 - 2) Initial lang (\equiv P O's) used to represent lists of possl. solns, soln. techniques.
 - 3) Initial OTS in TM2 (66.37)
 - 4) Initial OTS for other OZ problems.

.25 5) Factors of all OT sets 2/0 ~~at all~~ all sets of soln. techniques used in INV/OZ probs.

.28 6) ~~Factor~~ Factor (62.23) Just when during "Production Problems" to spend time on S₁ & S₂ \rightarrow what aspect of S₁ & S₂ to spend time on, when to spend time
 (perhaps usually spend time during "production problems" on S₁ & S₂ probs normally in "production problems")
 Domain: "Before problems" (if this is useful) & read More time on More global aspects of S₁ & S₂
 7) 74.32: Invariant S₁: "Invariant PD" This is common problem when Corpus Augmented
 8) 66.22 has 5 parts! @ T. code "control of k" turn on, turn off, decide what aspect of S₁ to work on. (76.23) (2.29)
 \rightarrow b) How to do logical reasoning. (c) (.25)

\rightarrow 9) T. problem of BIAS in induction in TM2 73.00, 75.13-15, 35.46; 77.01.

.30 10) T. Argts of " \rightarrow factor of some reasonable, yet "factor of 1" means assigning a PC of 1 to the method that will be fastest; for all problems! — seems unreasonable
 I think: Main Arg't, that TM's behavior is essentially optimal — that problems can't be solved (any) better!
 A MAIN Counter Arg't! Not saying Expressly prob solving is G-P-D + C such is perhaps not v. only way to solve problems. If it's then TM's drawn up maybe is optimal. If not to have arg't. is it
 .40 Much Good! A perhaps critical word pt.: TM doesn't have to be in the best of all prob heuristics \rightarrow 78.00

BIAS (00)

Previous Refs: (75.13-15, 33-40) Also 73.00ff

.001

BIAS: In induction! Least biased is not exhaustive or Random (w/o replacement)

→ Leitch is unbiased, if the slow version is used. T-2T version is unbiased if oversearch is done after soln. is found for INV problems. I think ~ remarks are true for OZ probs, but this should be checked! one could miss roads of lower $\frac{CE}{PE}$ than first one found!
 → Actually in T-2T version one should continue to find since otherwise

The Leitch, instead unbiased itself, we have bias in choice of Ref. Machine. In TM, we were continually changing ref. machine & adding "Bias"!

In General, there seems to be trade-off between bias & search speed. Kura's "Genetic Prog" may be very inefficient, but it gets more "Unbiased" results — that is a very "Creative". Of course his initial choice of language (representation) is very imp., & gives much Bias. But without it (using some standard representation for all problems) we'd have less bias, but an enormous slow down in speed of soln.

While normally, it may be true we always have this trade-off betw. Bias & Speed, it may still be, that there are ways to search efficiently w. a given Bias level, or get min bias for a certain Cril speed.

TM, using Leitch, puts all of its bias into TM2's original (best choice of GPD (≡ latest Reference Machine). This GPD represents all of TM's "knowledge of the world" as well as its "BIAS"

Bias has several aspects. ① Unrecognized bias (bias we don't know about).
 unrecognized

- ② Bias we know about: subclasses
 - ⓐ Desired
 - ⓑ Not desired.

We want ① to be as small as poss. — But just what is difference betw. 'Desired' & 'Undesired' Bias is Unclear

Clearly GPD (our intended BIAS) is something we must know: Otherwise Leitch takes foolish.

The main source of feedback is effectiveness of our "BIAS Management Techniques"

is how easily it allows to solve problems & how good our OZ problems solns are.

→ $\sum \frac{CPI}{PEP} + \sum \frac{TPI}{PEP}$ might be a measure. Note that this is a GOC for the GPD: ≡ TM2G

If we are unsatisfied w. this value of GOC, we have been using it for our TM2G

(or we have used TPEP & T-2 GOCs are equivalent) Then we are at a "loss"

Extremum & must find a way to "jump out" (I have no immediate suggestions!)

IDEALLY, we want GPD to reflect (only) the bias introduced by the TSG & the original choice of Ref. Machine (both have Bias):

BIAS ≡ Modification of A-prog

Another source of bias: In "finding short codes" we use various techniques to find feasibilities

This looks like a serious source of Bias. Is it desirable or undesirable?

Correctly says we much more likely to be found

A Good Lang System will have built into it, devices that on long codes, get rid of undesired sources of Bias. ALP (or probably RLP) automatically introduces bias due to selection of Ref. machine, as codes size ↑.

space on BIAS → 80.26

00:36:40 : ~~TM is not a~~ Problem (10), cont.

A condition which 76.40 may be too (all in mind) in a practical sense. TM is solving both ~~some~~ of the scientific problems for which ~~no~~ neuroscientists would have little a priori info — That it has 5 more independent as Robinson had!

In ~~the~~ Nat. Lang. Ling., reading about Humans, and evaluating social situations, TM may ~~be~~ never approach Human Info Content — But on the other hand, w/ much larger LPC than Henry, & much more input from Human Literature, it may surpass Humans in these areas as well.

Main poss. derivations from Optimality. (1) All processing is not usefully representable

by a PD (2) / ^{say,} while c is true, f updating Alg. is inappropriate. — one of the 2 of TM2G's may be correct, but we do not update efficiently. — (76.28) — what's between in time to switch between TM2G's and "Main Problem"

12 → (3) (76.38) That our model of learning is el is not the best way to do ling!

However, in a "long run", TM can use any derivable prob. solving / solving such happens — eventually, it should be a "Optimum".

So, an input. passed to Axiology & 3.2 points ^(7005R) showing "how all have info can be in prob. & how TM can put it there & retrieve it usefully

18 In Lecture (to Marcus & Jürgen) discuss Kenny vs "EI" v. si "Non-EI" approach — Hegelian dialectic DI → Non-EI → Synthesis. / Synthesis 5 → 6 find present case, T. EI approach seems to be not bad — also it suggests how to write to Q&A — how to make "One. Nets"

Another ~ Non-EI approach: I am depending (at the beginning) very much on how good my TSCQ is, & I may have to work hard to get an "acceptable" TSCQ.

An alternate approach is not careful about TSCQ, but gives it reasonable environment w. some problems means & perhaps a Reason Paradoxical Function to (ing how happy f. user is w. TM's behavior. (This may be in Jürgen's approach).

I don't know what about his latest system to comment much! My guess is that it will take a long time to start being settled — & able to work diff. probs. — & that usually the user doesn't know how well the system is doing, or how it is solving problems — so its hard to teach the system well. — Also I doubt it would have any idea of how close it was to Optimum.

3/16/01 TM : GA :

.01 Two Gross Inadequacies of Conventional GA:

c: ~~Genetic~~ Genetic Algorithms Manual (e)

1) The method of obtaining the Next Generation of Cands along w. various params of those Cands, is poor.

2) The Monte Carlo Method of Fittest Selection is poor.

.05 3) There is No System Memory of previous Problem Solns.

FPGA

The SGF Approach to ①: Given a set of Cands in Phase G values (\equiv present Generation). Construct a ^{Stoch} Grammar / Model \rightarrow we get $P(G_i | \text{Cand})$: where Cands can be any form of a Sample. **SGF** says "use (models/grammars) in which it's easy to \rightarrow $P(G_i | \text{Cand}_i)$ enables easy finding of lots of cands of type G_i ".

If we select the Cands of highest expected G for next Generation, this is a very "Greedy" approach \rightarrow 19

.18

TM can learn \rightarrow 24

Non-Greedy: It can learn to look for grammar in samples. An approach to ~~TM~~ TM can learn to suggest new cands that would make it easier to guess at much better Grammar \rightarrow 24

for ③ Use N MCT's associated with G_i as a function of prob dens that map into an initial population of cands. As TM matures, this function gets more accurate.

.24 ¹⁸/₁₉ \rightarrow So a new "problem": We have a set of Cands, is a grammar that represents cands (perhaps as well as could be done w. b. Given Sample)

Problem: ^{What is} ~~the~~ ^{trial} ~~run~~ ^{run} on ~~the~~ ^{small} ~~set~~ ^{set} of Cands that will "do most" toward telling more accurately what the "true Grammar" is? In general, more ~~is~~ trials would not be ok by expected "G". Note that cands are X_i, G_i pairs. \leftarrow

We want a "Grammar" that is ^{good} ~~good~~ ^{take} ~~good~~ ^{take} on $[X_i, G_i]$. Such a grammar can take many forms; (or a set of them X_i, G_i pairs) One gives a G_i for each X_i ; it has a common var. so $X_i \rightarrow G_i, a, b$. it's d.f. ~~is~~ ~~with~~ stochastic Grammar.

Also "improving the Grammar" in 27 usually means, improvement in the parts of the G.

1) Mechanism of How TM Works: Some input topics!

2) General flowchart.

3) **TME** [Q: Just how does TME work in improving our workings? - It does have the Over-all Gove.] Explain $\frac{MCC}{GPD}$ → The 3 reasons why all hours can be expressed by $\frac{MCC}{GPD}$ moduli.

4) Meaning of "Factor of 4": assumption made (How to ~~assumption~~ factor of 2). Factor of 2 is more correct → but not wrong.

5) Why all these things are expressible as Modified GPD + Lsqn. If I fall into any of these, I can use any of the following for factoring: (see 3/16/00) ① Determine factor for fact. - from computer ② Heuristic working in PC case ③ To reduce the number of divisions (e.g. do TSP) including any method ④ To utilize many hours, mod to GPD needs to be quite broad - a. if times of previous cond. realization.

6) How TM can realize any method of such, any method of working problems.

7) Give examples of INV probs, OZ probs: Various types: Some not directly Solvable by Lsqn. → 72.05

8) **GA**: What's wrong w. usual Methods? (see TM GA 3/16/00) 1, 2, 3 hybrid 01-05

9) **WON** (And/or Nets): See sheet on **WON**

10) Discuss a set of factors both priority & methods of Gosh using priority. $\frac{MCC}{GPD}$ → best mt. Gosh use for non-reduced levels. Effects of cutoff at any particular level. How it needs to be best PC for given E.B. level.

11) 3 Modifiers of Lsqn: ① PC order $n \leftarrow (pc_j)^{-1}$. (see 3/16/00) ② $T \leftarrow kT$ ($k=3$ is best, but $k=2$ or 4 is not much worse) ③ Time share: ~~best~~ better than ①, ~~not~~ better. ④ Use of PC to guide search, rather than 2^{-k} : could be fair factor larger. ⑤ $\frac{pc_i}{cc_i}$ order: this is best. could be better than ③ (see 65.30-40 for dup)

12) ALP & RLP: That RLP is probably best induction that one can do. Maybe discuss "Necessity Form".

13) Discuss RLP data as a OZ problem: The 4 or 5 important factors of priority (≡ Lsqn) that all of them are input.

14) **TMZG** (1.00): Just what is Gosh? Just what is Gosh, is low cost to take CC into acct? 63.01 AF 63.05 ff is on Bus. Time to calculate exactly. → 79.00 on this talk → This may be about as good as IVO Gosh can do - PC Plans is a serious dist of alternative TMZG → 63.26

15) A serious problem with TMZG "soln" of 64.01-04! For OZ problems: whenever modify GPD so as to fit pc_i of a "successful" O.T.: The Time available for soln. (\uparrow ; $-\alpha_i$) will usually change - so only if we ~~propose~~ pay for a number of same, α_i (if α_i is important to generate pc_i) & or raising it, some will not really improve (or even worsen) our GPD. If $pc_i \uparrow$ a lot & α_i doesn't rise much or $\frac{cc_i \alpha_i}{pc_i \uparrow}$ > 1 then it looks good.

16) I think we run into a similar problem in INV. problems. We want to mine $\frac{cc_i}{pc_i}$, but when it comes GPD is modified & pc_i changes (part. some soln techniques) then cc_i can change, because cc_i includes time → 79.00 → 71.00

"Doomsday" from Murray, "PDF file"

spec.
65.03

100: ON GHTI: proof! If we have any order of cards & exchange to order of \geq then E search will
01 \uparrow or \downarrow depending on $\frac{C_{i+1}}{P_{i+1}}$ orders of ± 2 cards. Any ordering of k cards can be obtained from any other by
a sequence of exchanges of adjacent cards. It is only easy to show 100-01 is true for adjacent cards,
- so if we only do adjacent exchanges ~~to~~ to get better $\frac{C_{i+1}}{P_{i+1}}$ order, our expected search time will always \downarrow .
This means that \leftarrow lowest \leftarrow is always lower than any other \leftarrow (reachable by exchanges).

05 Perhaps make list of kinds of OZ, INV probs & give examples (needed for talk).

- 1) INV a) T. classic P & NP problems of Comp. Compl. Theory: (Solve to extremity) proofs, TSP.
- b) INV, but expensive Yes/No: Sounds intractable. ("INV" but expensive gray Yes/No \leftarrow This is an OZ problem.
- 2) OZ problems: a) Simplest: Noiseless, Not time varying: (Adding Noise has little effect on needed prob.)

\rightarrow Use any probabilistic methods, but usually better to try (is more complex soln.)

10 \textcircled{b} Next 2d & noise \rightarrow (no big deal) \rightarrow Give examples (class of prob. w. certain characteristics, mincost, 6 Mo)

\textcircled{c} Time varying: If rate of ~~var~~ variation is slow wrt. clock speed! Try to make model of time variation of system, and internally do an Lsearch to get optimum for ~~any~~ expected product ~~maximize~~ "reward function"; If variation is \sim or faster than clock, it will have to be treated as Noise \rightarrow 10)

\textcircled{d} Expensive Not time varying but expensive Calc. (Math model of Calc: Do traditional search for ~~optimal~~ \rightarrow obtain trials for R, W. (Try is Greedy). \rightarrow fast/greedy approach)

Use Finite horizon Dynamic Prog.

\textcircled{e} Maximize $f(x) \cdot F(t)$ \leftarrow $f(x)$ or $f(t)$ can be open or closed. If $f(t)$ is closed, it would seem better can't be much. This is one kind of time varying OZ problem.

\textcircled{f} Anytime problems.

\textcircled{g} Show how induction is an OZ problem.

\textcircled{h} Give example: 1) Matrix multiplication in n matrices $2 \times n^k$ w. min k . (includes proof). We have a list of \textcircled{g} prob + proofs that time $< 2n^k$ (for large n). We want one w. largest k .

What if \leftarrow search. func for INV is wholly or partially "open".

In both cases TM would want to find faster way to compute it or approximate it: "Quick sort" is in Pats area.

So 1 kind of search func for INV, OZ:

- 1) Fixed vs. time varying 2) Noise, variables (noise could be order time varying) 3) Goal, domain ~~can~~ can be closed or wholly or partly "Open".

Time varying Inv. probs: faster diff. \rightarrow maybe not a bad claim. func is "open"

2) For OZ jobs: "Anytime" variation

34 T. Laven technique described really doesn't solve true limited open prob. We if we are given time limit T , it takes $\frac{T}{P_{i+1}}$ to solve problem optimally. If we really have only time T available, perhaps best soln

would be time shared version of Lsearch. In discussing Lsearch make this clear

First talk/cant


- 007 More **Exactly**, what we want is (for GPD_{1,2}): **input is (problem)** ^{includes type of problem.} output is set of (soln, pc) pairs in rank pc order.
 A "soln" is a program whose output ~~is~~ ^{is} correct soln ~~is~~ ^{is} SGP.
 For 2 good PGM, it's output will have correct soln ~~is~~ ^{is} SGP.
- 03 For GPD (for OZ PGMs): **input is problem**, output is set of $\{O.T., PC\}$ set. \leftarrow GPD₂

Perhaps Go thru details of INV problem (GPD_{1,2}) only
 Then just say what GPD_{1,2} look like for OZ pairs in inversion pairs

Don't do "anytime" probs: I think Time-based OZ problem is technique is fine

Note: O.T.'s tend to be correlated. Also Note: 10/1/30
 Lots of trials: trials from one O.T. can be used for many O.T.

SP or Optimality of T.S. Soln: $\frac{PC_1}{CC_1} > \frac{PC_2}{CC_2}$ \times ~~20-40%~~ No New Ideas!

Re: 00: It may not be critical, because ~~IM~~ IM can design O.T.'s of any conceivable type that do take advantage of trials in "turn O.T.'s" — But to "other O.T.'s" are all in the same O.T. — So, unless Cands can take advantage of info in other O.T.'s ^{13 66} trials, the system is inefficient. This is a form of learning during trials & can be dealt w. via the techniques of 6.2.10 () — How, Re OZ case may have to be dealt w. in a special way. There is a great no. of O.T. methods but see H.C. methods & Ray do now/trials based on ~~re~~ recent micro trials. Systems that do this could be combined in special ways.

• 26: On Bias: ~~Input~~ Input (and), there is a tradeoff between Bias & SSZ. In good system, SSZ should be able to reduce any source of Bias (?).
 One kind of Bias: Selection of data sample data from RW, is harder to deal w., but a good (bug).
 System will eventually, give instructions on how to take data from RW.
 In mapping info from RW into TM, we can effectively give zero weight to certain data, ~~by~~ by not including it. This extreme form of bias does not occur in other operations of TM.

• 35 i.e. all induction/producs have $pc > 0$. \rightarrow (81.03 on Bias)

• 36 Back to lecture! 79.09 Given initial P.D. & some new data, how does it reconstruct a new P.D. that incorporates the New Data? In steady state, this is usually not a problem: In steady state, we have all of this data on problem solns, soln times, times for convergence of solns times for build & trials, etc. We also have an inverted. \rightarrow to generate P.D. for a new (±) period.

Major BUG?
 005
 007

1.25 B/M
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 Mar 2 1960
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 7.32/dollar
 Fr franc
 1366
 3/26/01
 for June/01
 2001/02!
 7.26/dollar
 1.903

80 to

00: 79.40 If the script is in the form of a reference machine, we try to find codes that make, e. of pc. of e. observed data. This is [ENV, OZ, induction (sage, Bag), ...] data

8326 for use on P10

- 02 Consider sequential prodn. Our script can be in 2 forms: (1) unfr, (2) summarizing code.
- 03 unfr: Given corpus; get script; given augmented corpus get script; predict (ratio of) script. Any form of script could work unless ...

What if you solve apply non-linear complex plane - introduce "local reference"?

In (OZ), we would use a "Summarizing Machine". Its output is random strings, its output

is a subset of an prediction: (This is v.g. form for $\frac{1}{2}$ form). In simple case, a "summarizing machine" works only for "code" + (perhaps) OSL data.

[Guess] Summarizing "must" introduce bias - because it changes the effective script. Substitutes an approx to it.

When we use a Summarizing Machine, if we have new data, how do we update the machine? [Parameter: OSL part is updated in a simple ("enviz") way. -]

no number codes runs, save "states of machines" of base codes; (This will work for only a few "pure" BASS (not stochastic operators); How do predict; we find 2 part codes: 1 for "model", one in collection of "model". (w to MDL - but no provision for OSC -

So, we reduce to K best "models" for our Summarizing Machine. [Are such processes next? [Rite so!]]

- 20 Induction Methods ("Models"), [Z141]: very simple model.
- 21 Automatically deals w. OSL. At Z141 "Model" will consist of a list of nested data. T. corpus, then follows; we can code corpus in various ways (parsings), we select best parsings. We can select best set of parsings periodically. We use the pc's of data. Assoc. w. latest parsing, to decide how to build process to next "chunk", when this is done, we recompute probs. At certain pts, we make new definitions, based on latest parsing(s).

28 Probably would be good idea to apply Z141 to BAG lang. One way: we have this set of nested corpus (as in 21). T. pc of the corpus is expressed by parsing & corpus in terms of the data. (including, perhaps, a "stop" symbol that has a certain pc)

31 T. pc of the Devs, hvr, is obtained directly from that of e. Sequential corpus. Guess: that pc of 31 is obtained directly, perhaps using the same methods as for sequential Z141. hvr, perhaps a "stop" symbol is a primitive. After parsing w. the "Dev" of 31, say 1. no. of symbols in corpus is n. $\sum u_i = n$. (unsubscripted is one type, say).

We would like to pc of the dimension k. into k parts. One way is simply count how many ways there are to do this. Each symbol can't occur more than once - so we look for k restricted partitions of n-k. $\frac{(n-k)!}{(n-k-a)! k!}$ perhaps is (21). No! $(n-k)^k$ (this is range of one number)

Well, this can be worked out later! Neuzzle for student/reader. $\frac{(n-k)^k}{k!}$ (this is range of one number) 8200

So the number $f(n, k)$ is some function. w. the pc of sl. 31 is the pc of $f(n, k)$, we obtain the pc of f each grammar. To get the pc of k corpus,

could be $\prod_{i=1}^k \binom{n_i}{n}^{n_i} = \left(\prod_{i=1}^k \frac{n_i^{n_i}}{n^{n_i}} \right)^n = \left(\prod p_i^{n_i} \right)^n$ - A familiar express.

perhaps multiply by $(f(n, k))^{-1}$ will make it even more familiar.

To obtain the deriv, of δ good grammar for $\geq k$ parsed corpus, use ~~the~~ recent work on $\geq k+1$ + Wolf's "recovery" idea.

Note the function $P(n, k)$ assumes a lot (of bias) about the distribution of pc's,

There are probably many Alternative Solns. to the problem

Even with this cutting I would like to know about at least 2 poss. cost functions. Guess the cost-making connects to errors in coding \geq corpus =

While ≥ 141 is a "simple" model, it is able to deal w. distributions and alternative parse's.

Also it has a methodology for assigning pc to a corpus, ~~the~~ (i.e. Sol/Wolf method)

That will work when the "primaries" are arbitrary functions - so we can assign pc's to "Lsp-like" pcms (= functions of \geq corpus).

As yet, I don't know ~~how~~ a discovery routine for CF Grammars ~~is~~ ^{have to be}

It may be possible to apply to general. Each req. val to those Grammars usual

In fact I probably did do it in \geq to organizational report - a very likely in Sol 64 parts.

On 81.20, I mentioned "other induction models": ~~that~~ w. CFG ~~is~~ ²⁵ ~~of~~ ~~induction~~ of complexity of such models: (Note) CFG's are normally used for BAG induction which is \geq what I used for TMZ. - Hvr, is more like stack operators (which is \geq a npt, subclass of BAG induction).

OK, Consider the update problem. The first thing one does is simply update it's of the existing Grammar, w.r.t. the new elements of the corpus - which were ~~not~~ used by the "old" Grammar. After exploring a few possible "good passes" and considered, we look for new representative augmented corpus. We write backtrace a little - look at the corpus to deriv, was augmented by ~~the~~ (major set of recent examples - try to repair & look for new rep's. (I'm rather vague in my mind on this - but it looks like a "Risky Revision".)

The biggest poss. "Revision" is to discard all of grammar & take all of the data & make a ~~new~~ ^{new set of} grammar for it (very hard). By ~~keeping~~ ^(keeping) a ~~large~~ ^{large} set of Grammars at each stage of revision, the ~~need~~ ^{need} for major revision occurs less often.

3/27/01

3-7 AM 83

IDSIA =

KORONA

seek, square, 4 sources, of kg version. ~ 190 dm ~ 154 \$ us.

↑ what country, what city: Address, phone, Model no.?

~ 3.40 p 6 left

.00: 82.40: Other kinds of (7 2 241) ~~regularity~~ regularity:

01 Algebraic, Medial Regys.

Anyway, when I'm trying to write a log: write down just what kind of stochastic system using the GPD, is GPD. (Human Heuristic)

So, perhaps Main Weak PC (errors) of the system:

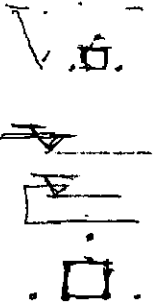
① Learning During initials: This can be done in ^{usual} way, but ~~time~~ ^{here to} speeded up when: I'm not sure of what's in how well it works.

② The general updating system for a bunch of other random types + Inv systems, (2) don't seem clear.

③ I assumed PC was 2^{-RCL} , but if PC would not have the same ^{CC} as 2^{-RCL} has, ~~it~~ ^{it} may be no more time to generate for ~~least~~ ^{lost} of 2 forms — well it's not clear that PC form is better!

③ In "improving GPD", when GPD consists of many very heterogeneous methods of prob. solving, it may be difficult to find a complement regularity that are common to many of them.

18: 62-32 ON long during Lench 62.19-32 dec-bis modulu of TS method of Lench: A poss. simplification: Better to "ing" to TS had a certain level of lost ("T") that it was coming at. After "ing", TM generates set of rules & works on each one until it is up to previous "T" level. This may not be optimum in energy, but from a practical pt. of view, it may take much less time than ~~to~~ Least update system of 62.18-32



26: 81.01 for the experiment: The first numbers of the BAG are exact PC's by the ref. machine w. null input to "Machine den" input of ref. machine. After we have known numbers of corpus, we update by using gen-null machine den, & derive of corpus (e.g. RLS for ~~numbers~~ numbers). Subsequent updates ~~update~~ ^{update} ~~update~~ ^{update} leave Machine den invariant, & modify ~~the~~ ^{the} ~~den~~ ^{den} of BAG numbers only! ??

What level of updating is modern of PC's of "Elements" as in the "BAG 241" treatment of 82.28? It involves the modulu of 2 . The number ~~between~~ ^{between} zero & $f(C, K)$ of 82.00 — So I think part is part of the "Machine den". The "Corpus den" is just the sequence of "Elements" concluding stop signals.

.00: ^{transformation of} QA problem: Say we have a large data base D we have Q, A 's about it ^{a corpus of}
~~is~~ we want TM to learn to ^{Answer} ~~(respond to)~~ new Q 's. One way is to record the data base +
^{known} $\{Q, A\}$ set as a corpus C we want to find an operator O such that best helps code $\sum Q_i A_i$ +
 to data base. say D is \mathcal{D} Data base set. Then we want an operator $O \Rightarrow O(D, Q_i) \rightarrow A_i$
 and \mathcal{D} codes for O plus \sum coding in $A_i \rightarrow A_i$ are minimal ($\sum 2^{-l_i} = \max$)

First, we, presumably know how to formalize it as $D \in \mathcal{A}(\text{null})$. There are 2 kinds of
 parts of this sort: ① regularity in $\{Q_i\}$ set our interest
 ② direction of interest ^{soln. 03}

.09 In ② we want operators $O \Rightarrow O(Q_i, b_i) = A_i$. (b_i are codes for Q_i)
 total in O plus \sum info in b_i is minimal.

.14 in ② we 2 ways to look at it: ① soln ② + P.D. on $\{Q_i\}$
 P.D. on $\{Q_i, A_i\}$ viewed as a B.A.S. induction problem.

.15 T. problem of .00 is closest to (107), but it may be related to (09) in that it is in
 + set $\{Q_i\}$ is not of interest ^{100.33} which suggests that we want a code for O and D
 such that $O(Q_i, b_i) = A_i$ and total code for O and D and $\sum Q_i$ is minimal. (minimizing $\sum 2^{-l_i}$)

.15 T. "troubled" is that there may be much data in D that is irrelevant to $\{Q_i, A_i\}$, so
 to obtain O , we would be wasting a lot of finite code. T. soln. of 15 is for $CC = \infty$; it
 .20 way will be best for finite CC is quite different! \leftarrow This situation is of MUCH INTEREST!

An extreme case! D consists of 2 unrelated sub-data bases D_1 is relevant to $\{Q_i, A_i\}$
 D_2 (which may be some things like names, searching codes) is completely irrelevant to $\{Q_i, A_i\}$
 but, to relate ^{at the} connect 3 prior, we ruled out it, until it has been examined & found use
 for large CC , no relationship has been found (but like randomness, we cannot do
tricks CC, be sure!))

W. Finite CC , we may find that best way to make more compression of $\{Q_i, A_i\}$
 (whatever that means!) by compressing D_1 in $\{Q_i, A_i\}$.

Consider we do 100 trivially like (.08 - .09) so roughly D is "not of interest".

So O must look at D & Q_i and output A_i ; we necessarily lock for reps in D

.23 { when I say "Not of interest" = Push meaning: that introduction of this info use of this } is BAD! But
 into introduces = bits would rather not have } Much Further

.24 Find Gen. of the problem of .00 (.18 - .20) we have a corpus C we want to do

several prod on extension of this corpus, the early part of the corpus ("first half")
 is not much relevant to extending of the second half. If we have $CC = \infty$, this causes no trouble
 but for finite CC (∞), if CC is very small, we'd want to neglect the first
 $\frac{1}{2}$ corpus almost entirely. I say "almost" because we have to walk on it out to realize

(Daily/Soln) of RLP problem in BARC paper is Wrong

.00; 84.40: Real (to amount of productivity" per cc in transformation of f into $\frac{1}{2}$ corpus) is small.
.01 What this means is (perhaps!) that f production problem is Not Part of non in Barc paper.
To ^{best} / ^{lower bound of} approximate corpus, $C \approx \frac{1}{2} \times \text{time to solve } f$, we do not exactly want to get codes that $\leq 2^{-25}$. We may find much better ways to spend time, if not only they want to

do, is extrapolate f corpus. (esp. say f corpus is in 2 parts, like 84, 36.)

Say we look at "part 1" & decide after some cc spent, that work do WAY much better, spending most time on "part 2". This decision could introduce a lot of BIAS into the final Extrapoln. → 86.09

.08 As general conclusion (I think I got a little flavor of this by ~~considering~~ looking for codes for $(\text{corpus})^a$ v.s $(\text{corpus})^b$ → possi. contrasts. ... how much more (what kind of BIAS more objective only)
.09 find, re sources, do spend on more possi combin. - Are more objective only

look for codes for small all stories w. (corpus) as prefix. I suspected finite CB would cause trouble, but I don't think I went into it very deeply. is that finite CB will

(always?) introduce BIAS into extrapolation. Very slowly w. CB!
T. bias → ϕ is CB → ϕ , but it may ↓

.13 Well consider sequential extrapoln (as .09-013). Say we simply uses Lsearch to look for codes starting w. 0, 1, 00, 01 etc. - say we use T.S. Lsearch (around $T \leftarrow 2T$ with constant of every round) - using finite upper bound for T - This would form like a very inefficient way to search - but it would probably be w. little bias. needs it!

T. idea of .01 seems Very Imp. It means that I really don't have a clear idea of RAP problem! - so its not really an OE problem!

On the other hand, in EM I will be using Lsearch, which is (as far as I know) not much bias

OK: BIAS ≡ Undesired BIAS is derivation from "true" ($CB=00$) ALP. Defined by Corpus and Real Heur.

Using Lsearch, eliminates consideration of all keys (codes) w. $L_{code} > T$ used. Undesirable, but what can one do about it? It's a Great Bad bias - in the sense of I don't know how to do better.

What about the 2-part corpus 84/36? Doing Lsearch entire corpus would seem to be very wasteful. Also, consider linear / non-linear Regression. Methods for, say, SM.

Here, one "narrowly" looks for keys of certain basic forms. Unlikely to be in last order. The bias introduced is equiv. to assuming that these models factor rather well, with high prob. Its equiv. to chance of a bad model machine - not nearly a universal real heur.

One can make this choice consciously. These heur. choices is used 86.00

.00: 8440: "Cowboy style" - by the seat of pants feeling, rather than by any conscious analysis. Concave, and more to ~~the~~ much better! Tr. Unconscious. Mind has its Biases and as good for itself.

.03: 95.19 A Kind of Bias (~~consider~~ that of $85.09 - 19 \frac{3}{2}$)? Hur say one did spend more time looking for codes for (Corpus) ~~2~~ than for (Corpus) ~~1~~. This would seem to be a serious ~~error~~, systematic error with ALP codes (at CB=0). In ~~some~~ some cases, decision to spend more time on corpus ~~2~~ than corpus ~~1~~ could be a conscious decision - a decision to deliberately spend!

.08: 3/2/01 The only Zero Bias applied to uniform p.d. (data from indigot past; Hmm D.R. ~~no useful prob. poss.~~) → 104.00

.09: (55.08) RE: Soln. of RLP program BARR: Perhaps to Maximize of $\sum 2^{-q_i}$ ~~is~~ f. Gore is in a well defined $O \geq$ problem. T. "Fly in omentum" hvr, is that (choosing $O \geq$ problems) How one works out depends on one's previous appt info - kinds of DT's available, TM's (around ~~to~~ those O.T.'s).

Hvr, what about a 2 sub corpus problem of 84.35 ~~is~~ + CDA problem of 84.00? - just how does 100 apply to Problem? Apparently doesn't! .09 doesn't seem to help!

.16 - consider ~~84.35~~ ~~35~~ we want to extend ratio of p.c's at "i" v.s. "B" centimetric. so we want to decide on what ~~more~~ info to use in prediction: we think corpus is very flexible.

(Perhaps this kind of decision hvr to be made on all RLP queries - If so, it is not clear as to what f. Gore is! What is Goal of decision to use many Corpus data? - Presumably, f.

.20 Goal's Good induction - which has a clear (No long term) measurable value.

.21 Ratio of SN for ~~induction~~ induction w. small CB, it would seem to be commonly used ~~is~~ a Good idea to use only a small part of f. Corpus, for induction.

CB \uparrow we find it useful to consider more words of f. corpus. This is it what we may do in SM, where f. Corpus consists of N 9 or 10 ~~or~~ parallel time series, & we only want to extrapolate one of them. For small CB, consider only f. sepp. to be predicted - Must recent data. As CB \uparrow , use more data & /o info about "Near-by" seq concs. One trouble! But if CB=0 is not used, we may end up using "Older data" (Cross validation) to evaluate predcs, (Sounds Very BAD!)

Even in normal, everyday, induction we only use a very small part of f. Corpus for induction. One of the things that we quickly know is approximately how much

.25 (of what part) of f. corpus to use for induction. → 88.00

.36 A New way to look at f. Corpus extension problem, say f. symbols in f. corpus are a b c d: we want to know order of prob of a at f following:

first we do corpus "d" only, using same criteria. Next to ~~the~~ corpus "cd" using same criteria, then can bcd, then abcd (if order is simple!) - After coding "cd" we may be in a better 87.00

CR 2430 - Normal List of all m. Sale.

711

.00: 8 bits to code "bcd", etc - working backwards.
 #47, even if 86. 36 ft were correct, it's not always clear as to what part of the corpus is "near" (\equiv relevant) to the part to be predicted: perhaps most apparent in BAS period.
 But it's a kind of start, \rightarrow It's not clear as to how much time corresponds to "8"; v.s. "c", v.s. "bcd", etc.
 Use ordinary "monotone" UMCS, it's easy to continue in forward direction. try try try try
 corpus backward. Since we would be interested in extensions to the corpus only, we'd have to find codes for each possible extension - which is a real drag! try, since we usually don't code
 .06 Very far back, each trial may not take very long! \rightarrow 33

.09 SN | On finding codes for corpus using Normal "Monotone" Machine. Say we have a sequence of corpus & we have a code/Part list, output matching corpus to a certain pt. When we augment α w. all poss. continuations, we put things that don't track to corpus. We then backtrack α by a minimal amount so we can try useful changes in monotone "reduced" ($\equiv \beta$)
 So we then try all useful possible continuations of β (other than) α . If one works, we try to continue it, if not, we backtrack α back back back back, etc.

By keeping a lot of info in memory about branch pts in the backtracking - forward trials at a particular point were successfully tried, etc, we may be able to backtrack rather rapidly, & perhaps find codes for a corpus fairly rapidly.

\rightarrow T. amt. of undesirable BIAS into program's trial, is unclear.
 But anyway, .09 does sound much faster than just trying doing random trials in lexical order, using Least cost.
 (We'd have to use Least cost in .09 also, a best kind of "failure" of a trial branch would have to be accounted properly in RAM) - This may \uparrow complexity of the search \approx 6K.)

.27 T. idea of .09 may be same (or \approx same) as using summary codes
 a set of "summary codes" after each augmentation of the corpus.
 It's possible to start back beginning a code a corpus this way - back machine to get more poss. codes, occasionally, when necessary.

.32 perhaps use (.09 or .27) to try code generating. codes for sequenced corpus,
 .33 try, one starts out w. a hypothesis of no. of codes for each continuation of the corpus!

Alternatively, one could "start out" w. the machine in various possible states (corresponding perhaps to default poss. "continuations" of the corpus. - But this would soon to add info to code! (Drop this for the present!)
 .36 This system has other serious drawbacks!

Balenzac



86.40
00: 87.40! I'm remembering Groups Idea, that Induction has to include All available DATA!
While this is true for $C \subseteq \mathcal{B} = \mathcal{B}$ (ALP) — it can't be true for RLP.

Going back to the problem(s) of γ f. ~~00, 18, 20, 35~~ ^{2 ways to corpus} also \rightarrow $86, 16, 20, (21-35)$

Key way towards data of the Game: In our induction, we want it to be minimally biased for the $C \subseteq \mathcal{B}$ use. The idea is that we want to eliminate as little as possible $C \subseteq \mathcal{B}$ codes; include as little as possible from codes not in $C \subseteq \mathcal{B}$. Actually "2 parts" — it may cancel!
We want dist to be as small as possible. (2 goal)

08 Consider BAC example: α, β, γ known corpus \exists objects + part of Δ^* . To get pc's of poss. combinations of Δ^* we try to find parts of codes for α, β, γ that occur in Δ^* . These go into "Common Machine".

In working on α, β, γ we do find some codes common to Δ^* in α & β but not in γ , so in the training time, we spend all our time on α, β & Δ^* . (we cutoff without γ)
So essentially, the corpus becomes α, β, Δ^* & the game is to maximize $P_\alpha \cdot P_\beta \cdot P_{\Delta^*}$.

Any partial coding of γ influences the structure of "Common Machine", but γ becomes a constant. If γ is too big for, we just try to determine $P_\alpha \cdot P_\beta \cdot P_{\Delta^*}$.

So essentially, we just try to maximize $P_\alpha \cdot P_\beta \cdot P_{\Delta^*}$ in a game!

18 I'm not sure that that (unless γ is 0) I don't want to maximize $P_\alpha \cdot P_\beta \cdot P_{\Delta^*} = P_\gamma$

But 18 is not critical at this pt. — Is there any way I can make 08-18 more goal oriented? — ?
 \rightarrow Perhaps more "goal oriented" T. top Game is not clear.

A recursive Soln! (Perhaps!)! Given INDU (OZ) problem desc. Work on all of corpus for t time. Based on possibly a previous experience, choose sub-corpus & work on it for T time. (Chosen Goal of choice of sub-corpus is to maximize the current likelihood of best estimate of pc values in Available (to time) Δ so to α (or line 22).
Do this until available time is used up, or until problem is solved.

26 Another view(?): The GPD has a sub-optimal problem, Subset of corpus to be considered for induction, Can't find soln and soln and soln.
If that INDU problem / PD that read soln will solve it in $C \subseteq \mathcal{B}$.

28 prob. of subset of corpus to be used, can't

29 26-28 maybe a bit off! The problem in Question was induction. Useful form (for seq. example)

Given Sequence S , to get rid of s^i vs s^j . Base said, Δ^* problem is equivalent to finding codes for $S \rightarrow \Sigma^{2^T}$ is max (w. time limit, T).

36 GPD₂ looks at the problem desc (which includes S^T & T) and outputs an infinite set of O.T.'s (w. a pc for each). Each O.T. looks at prob desc & outputs subset a set of codes (in time T or time $\frac{T}{pc}$).

39 For the BAC RLP induction problem, our "O.T. for induction" always, look at the corpus, S , & have time limit T , and tell what sub-corpus to use for induction. — Then 89.00

Lugmo.

- .00: 88-90: Use (such as whatever) to find $\approx 2^k$ codes.
- .01 Attention: first looks * corpus: then decides on subcorpus — (But actually it is
- .02 GPD₁ not "looks at corpus"! → **PLP** (Ludwien is no longer a pure OZ problem! —
- So it's a special kind of problem — but we may be able to deal w. it by a special
- "O.T." of 88.358 - 89.02. So essentially: induction need not be a **GZ problem**
- .07 but it can be a special kind of problem that is solvable by TMS "general methods". → .20

.08: 71.40 [SN] On: "FACTOR OF 2" problem: say "k" ratio is .5. ~~Induction~~ ~~problem~~
 "Narrowly updating" / solves to problems, takes about same time as problem solns. If so, there will
 be no time for integrative, default problem types — this kind of work being equal to
 Using a larger part of corpus for "GPD improvement". It would seem, then, that
 w. $k=.5$ 9M would never do much critical integration of disparate domains!
 Hrr., Arg of 71.21-24 would seem to apply. ← BUT I don't feel comfortable w. this Arg —

I'll like to do an analysis of just what is going on!
 For .08ff, a better thing to do would be to take a long seq. of problems: use by the fraction
 on to first $\frac{1}{2}$, say. — then success $\frac{1}{2}$ would do $\approx 2^k$ problems in the faster — so we would
 end up w. more done. — But not by a factor of > 2 !

.20: .01 [redacted] So, for seq. induction probs, GPD gives set of (codes (or. use PC))
 for soln. For some (very large) corpus, a cond must of necessity, decide to
 ignore certain parts of corpus (possibly after some administration of those parts.)
 — or (positively expressed), to include only certain parts of corpus: This dem. of what parts
 of corpus to work on (condens. ind.) is \downarrow t. wt. of that cond.

In SM, say, we have to get pd for tomorrow's price. If we have time T to make a cond,
 one can only consider a finite part of corpus, whose length $\propto T^k$. On the other hand,
 for certain kinds of time series (say linear regression parameters), $\sigma^2 \propto$ length of corpus
 considered, \uparrow — but $\sigma^2 \propto$ slowly (\therefore "precision"). So (usually) \uparrow amt. of corpus considered,
 will \uparrow accuracy of induction — but this means \uparrow in time spent on induction. For given T, there
 may be an optimum corpus size & consider a optimum way to look for regys.

- .33 .02 So GPD₂ looks at problem (which contains corpus or address of corpus) and what has to be induced is how much time is available.
- Ti output of GPD₂ is a set of O.T.'s for induction! Each one has a decn of what part of K.
- .34 Corpus it will try to codes as well as a tech reqs for finding codes — (is a PC for each O.T.)
- .35 Ti GPD₁ that gives rise to O.T. GPD₂ (of .33) same in princ. .32: Output is set of cond. O.T.'s
- .36 [redacted] a poss. subcorpus, a pd on $\approx 2^k$ for the subcorpus is subcorpus.
- So: How do we go from ti. GPD₁ of .35-36 to ti. GPD₂ of (.33-34)? What is it that we are
 maximizing in ti. GPD₂? — perhaps the utility of t. produ. usually measured by $\frac{1}{PC}$.

"Lotus Eater" as a Sink

00:55:40: One problem of Very intelligent Reinforcement Machines finding way to do self-reinforcement. Jürgen Marcus suggest that if there is a large set of R.M.'s, w various internal constraints, to ones w. a of constraint would "wire" — since they could keep on growing.

Also note: To do eternal self-reinforcement, T machine must protect itself (i.e. its power supply in particular) — so it cannot afford to be a complete "Lotus Eater"!

89.40 is the closest I've come to expressing what kind of Gore I need for GPD₁ → GPD₂ output - 89.40 is a Gore; 89.33, 40 is a Gore

History of problem: ~~84.01~~ 84.01 85.01 86.01, 86.16720 88.01 - 89.40 - 91.01

89.33 - 40 is latest work

A reasonable approach (related to 86.16 - 35) is 89.33 - 40, 89.01 - 40, 89.33 - 40; 91.01 out.

So GPD₂ looks at a prodn. problem (and its Time limit) and outputs 89.33 - 34.

The idea of 89.40 was that we need GPD₂ needs a Gore for its actions: "What is it trying to do?"

That the prodn. should be "useful" seems like a good Gore. That appears utility is to minimize $\sum \frac{I}{P_i}$

∴ we want to minimize $\sum \frac{I}{P_i}$ of outputs so $\sum \frac{I}{P_i} + \sum \frac{I}{P_i}$ is min - the sum

is. ~~like~~ total sum - time

The Gore of all is certainly familiar! The reason we can't minimize it by making

all $P_i = 1$, is that the P_i 's have to be obtained in a logical way by summing over

Cases. The system could conceivably cheat by assigning corpus parts that biased towards the P_i 's being very high. Choosing Corpus parts selectively, is only one way, the system could introduce bias.

Perhaps I shouldn't be looking to the "total sum time" as a Gore (closely related to goodness of GPD₂): Perhaps just try to make GPD₂ as good as

possible. — At first glance, it would seem that there are more

opportunities for "unwanted bias" in GPD₁'s Gore: [GPD₁'s Gore is the

probab. of data that has occurred] — perhaps ~~can~~ ^{try} can easily cheat by not

including data for which P_i values were bad! — ~~Maybe not~~ — it has to include some data for all P_i that were used. (i.e. can't eliminate P_i 's).

Also check summary of state of TM

92.01 - 100.90 is on talk at IDSA (10.01 on Phyllis's one extracted talk)