

Plus :

867 - 8493

Plans

867

we use probs of the corpus in the old corpus, since when we want to incorporate the new into the old corpus, this will tend to give min. total cost.

01: 1160.40: successful objects (with a freq. & their use freq times and combining them. [concept. need be the only comb. of instructions is "degenerate"]

Now the big trouble (I think) with this, was that it is a seq. of problems, and I got ^{a bunch of} some new problem types, my trial mutations of the old, successful ppm, I would use the whole top. seq. thru!

This can be avoided in 2 ways (this "devry" is source (at least temporary) joy) ① At first, use a "non-stationary" ^{i.e.} ~~operator~~; supposal ^{that} after ~~the~~ the r^{th} problem, our last operator begins to fail. Then we devise a new operator that comes into play after $t=r$ only, that is to be used only on problems only. The probs of coding can, how, ~~be~~ ^{be} ~~obtained~~ ^{obtained} by averages over the ~~past~~ ^{past} code of ~~previously~~ ^{previously} successful op.

② After one or more ^{codes for} ~~new~~ ^{out of them} "new op" have been made, one tries to incorporate ~~them~~ ^{them} into the ~~old~~ ^{old} of the op., w.o. using the "absolute time" marker. is successfully done, of course, only when the total ~~of~~ ^{of} the entire op. has been reduced.

③ V.G. Select a set of "typical problems" from the past corpus use ~~these~~ ^{these} to test new trial ops, instead of running the ~~entire~~ ^{entire} top. seq. These "Typical problems" can ^{in which order they are to be tried} be ordered, so that the expected real time to failure for new trial ops. is minimized. ^{Note: these "typical probs" will not} We can use ~~these~~ ^{these} set of ~~ops~~ ^{successful} problems to either devise a modification of the old op. non directly, or to use this for trial on the "incorporation phase" i.e. ②

I think both of these ideas are very much used in humans.

→ A standard way of doing ②: Devise a method for recognizing when ^{problems} objects of

plans 868

01:67.40 : "th. new type" occurs. i.e. in what way way T most previous probs. Th. "recognizer", hrs. most cost less "time marker", or else one gains nothing. Hvr., a "time marker" may

Cost of Th. "time marker": This marker can just symbol that indicates that th. T.S. after that point will have treated difrently. If one uses r such markers in code length k , they will each cost $\frac{r}{k}$. This corresp.

every r truncation of a T.S. — and I suspect th. might often be better to use r cost $\frac{r(\pm + \pi)}{\pi}$ truncation cost.

→ results that are 100% accurate anyway — say only 80% th. new problems are amenable to th. "new method," it's such a case that 20% are amenable to th. old method. In a case, a time marker, dividing th. T.S. into 2 sub-t. of difrent. statistical properties ~~with the same goal~~ can be good.

Later, if one could devr. a method to recognize th. 20% with ^{hy. ent} ~~numerical~~ degree of accuracy, one could dispense with th. time marker — tho even with th. time marker, such a recognizer would be v. g. In fact, if I had a recog. scheme it would have to be very good percentage-wise, for me to be able to trust it on th. very large previous corpus, because it would make a numerically large no. of errors (tho. percentage-wise, small). Th. time marker would certainly aid considerably — unless th. recog. scheme was very close to 100% accurate — (as it might be

2 **MTM**) [One can view th. time marker as one type of "recognizer"! Th. Problem of how to combine partitions of a ~~set~~, to obtain optimum discrimination, is discussed at great length in IR \approx 8180ff

Have TM able to work probs. many difrent ways, and store all of th. hy. cost ways. Th. exact mechanics of th. \leq cost evaln. is immed. clear to me, but certainly we would want to store alternative methods, so we could use them, or ~~just~~ combs. of them

10-8-60

Plans 869

.01: 6840: for new trial methods.

Th. "Char., Methods" ideas of ~~DATA~~ EPM 1110.01
be a rather reasonable formalism to put this machine in

.05 Another idea is to try to use direct machine
~~because of their~~ speed.

.06 For these problems, and try ~~sets~~. - it may be possible

GENERAL PLANS: (10-8-60): data - learning

1) Do the simple Pgm. Th. simple nb PSG of Dart 46

2) Do some IR Pgm (around Dart 67). If any results at all are obtained, this will be a reason
"pot boiler". Th. "Medical diag." prob. may be easier

to get data for; since it has been done on a com but w.o. "Learning". Try to get names of people who did it, so I can get their data.

A/O Try to get SM data from U of Chi people to use for project! Kill 2 birds with 1 stone!

→ 3) → work on "Human: aided" TM of Plans 1160.20 - 869.05. Th. will be getting us into Th. home stretch!

4) Try to work Th. out on a very "hi level" before getting to details.

(4.26.61): Th. Pgm descr'd in ZTB 141 (and Dart 46) isn't var. Its main values are outlined in code 262.30.ff. My present idea is that Th. IBM 709 Pgm is almost ready for debugging, and I have had it almost key-punched, I shouldn't finish it now, but go on to (3) or some preparation for (3). Th. main poss. justifications for finishing Th. present 709 Pgm (a) Make AF feel I'm doing something material. (b) see

4.26.61

T.M. 8

Plans 8 264

01: 69.40 : if ~~any~~ the costs of R. upon data are about right
 in ppm. (d) I have gotten much of it key-punched
 ≈ \$50 (e) It could be of some use - e.g. feeding a set of
 R. ppm of ^{Dart} (B) 300, also some of R. applies of Dart β
 might be useful.

Rebutal: (a) ZTB 138 ^{ZTB'S} may make a big splash and keep
 happy. Also, I've written ~~139~~ 139, 140, 141. If they're not happy
 I can't waste too much time just trying to satisfy them, and
 needed time on serious research. Also, I don't really need
 money. (b) Apparently (code 262.30) it would ^{not} do this. (c) S
 true, - but I suspect that I'll not use Fortran anyway, ~~but~~
 my R. serious work, but something close (if not identical) to
 (d) Do not throw good money after bad. Occasionally, hvr, it is
 for public relations (e) Hard to say. - I may actually want
 use it to find sets of ngrams to be used to make better "un-
 trys" in a Hill climbing T.M. - but until I have the
 need - don't ppm it!

So, if I don't finish R. Dart to ppm: what next? It would
 be a bad idea to write some reviews, to familiarize myself
 what I've done, to prevent needless repetition. Also, ^{written} a review
 is always good to have.

4:27.61

Proby 8265

Plans

.01: : Some impt. things that I want to consider ahead into a big final project.

1) Th. ideas of Plans ~~...~~ . . . β 1160.20 - 869.06

2) Th. Observer, operator machine of EPM β 1110.01 - 11
This machine looks like I could get it to do things Th. way human seem to, by getting a reasonable set of ops and obs, combination rules.

3) BSTM : I'm not sure as to Th. relationship of Th. Co. MoTh. of ind. ~~inf.~~ inf. to this machine.

4) Th. ~~own~~ problem of getting optimization of a T.M. w resp. to a goal, in a rather direct way. ~~This~~ This comes up in Th. SM β "betting rule" problem.

5) ~~I~~ I would like a T.M. like TM704, that uses ^{basic} machine in a very direct way.

4.28.61

T.M.

Phaus

01:265.40:

SN On the BSTM and "integrated T.M.'s" ...
 is some discn. in BSTM section on the possy. of not
 (looped)
 a hierarchy of T.M.'s, but having a single T.M. with
 recast. schedule, that gave F.B. ^{as a funct. of responses} a long seq. of
 values. The exact nature of the F.B. funct. was ...
 In such a T.M., the division into hierarchys is not nec.
 "E" " " " ^{also,} we can, presumably "put R. To
 hill manually".

SN: Perhaps ~~was~~ before working on this BSTM, work
 an ordinary recast. T.M. See if a genl. solu. to
 problem is, almost automatically a solu. to BSTM
 providing one uses the rite "sequencial" recast.
 schedule. There may be much in BSTM of

SN For an ordinary recast. T.M. Perhaps draw
 a model (of a recast. net T.M. (I may have
 done this) — ~~then~~ see if I can modify the genl.
 principals, so as to make a T.M. built around
 properties of a stored prog. computer (e.g. T.M.)
 very I think working on ① and ② simlty would
 be very helpful

See various discns. of Hill-climbing T.M.'s

4.29.61

TM

flu

.01: 266.40

Some of BSTM (JS in particular), was idea that if one had a $TM_1 \approx TM_2$ situation, say, TM_1 wouldn't, perhaps, have to be very "brittle" to "get

ANN A kind of TM_2 that might be an adequate substitute for TM_1 , say, would be represented by a large, binary matrix. (This could be true if TM_1 is a ^{neural} net, or a ppm.)

Reinforcement is either +1 or -1. (Good or Bad). Various patterns are tried (\approx Binary nos.) with various inputs. Say

I_j is the j th binary no. that represents the j th trial TM_1 .

Say I_j is the j th ~~trial~~ input.

$G(I_j, T_i)$ is the i th ~~trial~~ ^{reinforc.} (+1 or -1) of the output I_j is fed in to the TM_2 assoc. with T_i .

TM_2 tries to extrapolate to get a $T_k \rightarrow \sum_j G(I_j, T_k)$ will be max.

A simple way to do this: TM_2 uses "correlational coding" (as in IR section) to obtain input grams within T_k .

so that TM_2 can predict $G(I_j, T_k)$ ~~without~~ as well as poss. without knowing I_j . TM_2 then tries T_k 's for which

this prediction is ^{$G=+1$ with prob. 25} as close as poss. to 1. ~~any~~

The method of discovery of good grams, that is discussed in IR would be used.

If the T_k are grams, then we have much a priori info on what grams are "good" — e.g. ^{Pr. vector} a gram will tend to remain if it is translated along T_k . Also, ideas about ~~what~~ what grams constitute subroutines, etc., are impl. This kind of info should be built into TM_2 's ~~matrix~~ extrapol. methods. There is a lot of disc. of this sort of thing in early "Dart" section (~~to~~ Dart B(130))

2.29.61

Plan

01: 267.40: So, using these T_2 's as TM_1 's, I could do TM_2 . This $TM_1 - TM_2$ comb., will then be a hill

TM_1 will have all possl. orders available and will \therefore be "universal"

\therefore I think, with a suitable Trig. seq. and recnt. method \therefore be able to get TM_1 to be a $TM_1 \equiv TM_2$ combination. This method would ^{give} recnt. ~~for~~ for working long sets of probs. correctly and fast.

At the present time, I have some idea as to what TM_1 would look like and also what TM_2 (correlational coding) would look like. I like to consider the pair ~~as~~ operating together, as a TM_1 and have TM_2 be universal, with some sort of simple ~~h~~ hyper TM (perhaps just myself), to improve TM_2 . I don't know how to ~~make~~ make TM_2 universal (— unless the universality of TM_1 automatically makes TM_2 universal) — this may have been discussed in (JS \approx Jan 5, 1959). I.e. TM_1 would ^{be automatically} modified ~~into~~ so as

take optimum advantage of the fixed form of TM_2 . Perhaps has to have a minimum amt. of Goodness for this to take place.

Then O.K. — say I have this fixed TM_2 , and a universal TM_1 that is optimized with resp. to a ^{mean-} speed-of-correct-response (last n -trials) ~~recnt.~~ recnt. criterion. If I do some "manual" placing on the hill, I will also have to put TM_1 on a hill that includes some hill-climbing techniques!

269.15

35
136
SN

On Network TMs:

The big & conceptual problem in TM 's was how to make the analog of (defining a useful sub-network in a sequentially prog. computer. One way, is to have each tree net be describd. by Polish notation, so that defns of sub-nets

269.01

2.29.61
phang

268.40: can be made and used.

How, another way to do it, that may be a way its
If a particular "subnet" or "sub-operator" is very useful,
reinforced by having its functions (randomly) duplicated in 11
So a highly reconf. ~~of~~ sub-op., will be duplicated in 11
(these 11 duplications may not be exact duplicates, but will be functionally
usually the same). These duplicate ~~of~~ sub-nets will then
to be used in other sub-nets ^{which} which will be activated w
net-as-a-whole, gets other input types (other problems)
How, my ideas about the last # aren't very clear as yet.

15: 268.35: By going into a bit more detail on this 2 stage TM,
be able to see if its at all reasonable to expect it to get
clever.

In a slightly more general form than 267.01 ff (More in the
of TM704): We have a universal machine, with a ^{many-computer} vector \vec{V}
determines its operational function. [\vec{V} may be the ^{"instructions"} program of a seq. comp.
or the xtra. instructions for a Turmac; or the ~~xxxx~~ ^{micro} connection str.
of a ~~the~~ "neural" net, or...etc-]. ~~xxxxxxx~~ This Universal M
is called TM₁. ~~TM₁~~ TM₁ is fed various inputs and
various outputs, and is given a G value, for each (I, O) pair
TM₂ looked at the set of known $(\vec{V}, G(I, O))$ quads., and
this, using a rather long-run, reconf. schedule, makes the next
TM₂ usually doesn't ^{use} a very clever ~~or~~ inductive methods. \exists the
method of correlational coding, or some even simpler methods
may be used. G may often have only the values +1
- or TM₂ may depend on G info to that \exists range.

It may be poss. to show that a simple random net TM, with 1
connections, that is reinforced with a proper temporal ~~net~~ weighting, is a
special case of the above!

An refinement that may be very useful: ~~the~~ In certain
270.01

2.29.61

Plans

01: 269.40 : Cases, there is only 1 vlt answer — this for all other replies get $G = -1$. Or there may be ^{some other} simpler by TM_2 , to figure out $G(I, 0)$. In such a case we may want TM_2 to be able to "exercise" TM_1 by putting the same inputs to TM_1 , again and again, and suitably his replies. The point here, is that all this "exercise" does not sample size — and we want to get optimum reply performance on a (small) sample size.

SN: In many (perhaps most) extrapolations we are given Br. parsed c's of a lang, and do make inductions on new s's. Some examples are computer pgms, LISP, ^{operators} Mathematical expressions in Polish notation, Spoken English as a lang. seq. — Usually the lang. seq. is arranged for the child so that he learns phrases that are meaningful, and then he can assume these phrases are usually pos's whenever they occur.

The parsing may not be optimum, hvr, and if we had $(\alpha(\beta(\gamma(\delta))))$ [Greek letter are ops], we might do better with $(\alpha\beta)(\gamma\delta)$ — i.e. if $(\alpha\beta)$ and $(\gamma\delta)$ were ops by frag.

Some impt. ideas relevant to a good 704 TM:

- 1) The possy. ~~free~~ of a BS TM, using only 2 ^{universal} TM_1 , with suitable reconf. ~~same~~ routine, based on history of (I, O, G) sets.
- 2) Use of th. IR ^{upm} discovery routine for hill climbing, ~~to~~ to w/ optimum pgms for this TM_1 .
- 3) In addition to ^{upm} discy, use all other regularities in seqs. that are to be used for computer pgms. Express pgm as a set of ops in Polish notation. → see 7)
- 4) A "good lang. seq." is one that ~~unwilling~~ encourages th. "vlt" ^{data} ~~data~~ ^(pkts) ~~pkts~~ with ~~rather~~ ^{rather} ~~short~~ ^{short} runs of guesses.



Plan

01: 270.40: 5) The idea of making $G = +1$ or -1 only, as a correlational code (like IR code) for predicting whether will have $G = +1$ or -1 . \rightarrow I think it may be then possible to construct programs that have a priori a hy exp in any particular case. (see 10)

6) There seem to be some good suggs. in ^{TM 701} 730.01 ff. Also see ^{long} Pr. review of BSTM. \approx PL 450 529.01

7) (From 3): \otimes If one has a pre-parsed corpus, it is easy to get a suitable / ^{stochastic} WPSG — i.e. Pr. xtn. probys are easy to get. to get a WPSG that fits much better, human norms should be defined. This will not ^{change} the predefined set of s's at by the gramm., but it will make Pr. obsca gn. corpus have — which is very impt. \otimes If one doesn't define these I think Pr. resulting "fit" of Pr. pramm. is not very good — e.g. w.o. these "defs", one wouldn't recognize figures of speech, etc. [this idea is relevant to

8) Actually, I should first do an analysis of Pr. applicn. of Pr. (Correlat.) lang. to Hill climbing. Then try to extend Pr. m to Pr. s's with pre-parsed corpus.

9) **A GREAT** Goulou of Th. IR lang: Instead using norms of phrases as Pr. prediction element with Boolean combs. of them, — use "properties" of all kinds, Boolean combs. of them. By suitably choosing Pr. properties it will be automatically easy to construct objects of hy exp

All sorts of properties can be considered — e.g. Pr. presence of certain norms, or ops, or ops used in various ways, or Pr. presence of certain kinds, etc.

At a "hyer level" we will want to have a "property generator" This device will combine properties and other abss., to produce

Play

01:27:40: new properties of hy expected "U". We have
 a rather natural way to get the ψ U of a property
 the actual U of a property will also be assoc. with
 which it can help create strings (\equiv pgms) with the
 Perhaps certain properties will be assoc. with certain
 betw. probys involved in the MC generation of the trial

- 10) ^{from (5)} The idea is to first link the "properties" to the generation process used in the MC gen. of the pgm. strings. This involves the of dividing the "G" into 2 levels, then getting the proby a gen. string having a G of +1, for arby. input. We may be able to make an approx. generalization for ⁿarb. G space by giving non-equal wts. to all G's. (Ordinarily, for G = only, all G = +1's are gen. = wt.)
- 11) A refinement of the 2 level G idea! Have n ^{the set of known pgms} A good ~~many~~ property is one that gives a very skew distribn. on n levels and \therefore \downarrow coding costs.
- 12) Since many successful (G = +1) pgms will have many long ones in common, we should use a search method for them, that is signif. diffrnt from the methods dec'd in "IR". Hrr: note that we start using the "correlational coding" used in IR, use the method of Dart 46 to find the signif. ones — we may catch the long ones at that stage, or we might a special pgm that looks for long ones that are repeated > 1 time in the corpus. The putting of the corpus in numerical notation (say radix 27) and comparing "size" of subsequences starting at various pts, might be useful. The trouble with Dart 46, is that it may mal-parse a corpus, so that repeats of some fairly long ones are not noticed.

5.4.6)

T.M.

Plan

.01: 272.40!

Note that I can arrange to have the program
 if I write them suitably. Then the problem of
 and property deriv (for props = Boolean comps of ngrams) will
 be simplified. For this pre-parsed form of pgrams the
 — operator T_M form of TM of EPM 31110 would

July 25, 61: Recent work (273.1.01 ~~to~~ to 318.40) has been on
^{probly "code"} ^{latter part in BS.}

empt. probs. involved in what I had considered to be a "final"
 — Considerations of a T_{M1}, T_{M2} comb. — /conds. under which
 (open loop) would deriv that something like T_{M1} = T_{M2} would be
 good idea. Much work had been devoted to deciding what
 Goro for T_{M2} to use on T_{M1}, so that $T_{M1} = T_{M2}$ become

Also, much work done on devising some new types of
 for use with operators that are to be improved
 for various kinds of hill-climbing probs. in h.r. and S.M.
 these codes being devised to prevent ad-hocness.

~~Next~~ Next, (BS318.25), it seems clear that any essential
 "deterministic" TM would probably not last long because of comp

30 failure — ~~and~~ so the random-net type would have to be
 worked on more. (unless there is a way to get my "Z" to
 do work — which is, in ^{a certain} spirit, a "deterministic" computer
 in the sense that I can use its principals to construct what
 essentially a very large "determ." computer, that has an
 large reliability.

One of the basic motivations for the careful analysis of Goro
 for T_{M1} and T_{M2} — was that I wanted T_{M2} to be able to
 supervise T_{M1}, ~~is~~ with soft. competence, so I could leave the
 system alone with a set of problems (tag-seq), and not 319.01

July 25, 61

Plans.

.01: 273.40 : have to supervise the system myself. Also, I can
for hill-climbing
myself, w.o. having to worry about whether my trials were a.h.

— And also. I wanted to be sure a $TM_1 = TM_2$ comb.
find some solu. But "satisfied itself" — but did not satisfy



~~Image~~ It appears that the final TM will be

net T.M. If I made a TM_1, TM_2 like 313

I could make R_j the den. of the connections (or the "con
input lines") of a "random net", and have the "data

$TM_1 - TM_2$ comb. optimize the net (as an operator).

.15 However, even so, I will, eventually, have to understand
nets, and above all, how they can be made to corresp. to T
for the uses I am putting Turmacs to — i.e. how to
"definitions" in a ~~net~~ β net, ~~net~~ and how to imple

.22 any other coding methods I can think of for Turmacs

.23 So, a seq. of problems that should lead to a final, work
random net TM:

① Study and ~~some~~ summarize my work on codes of various ty,
Try to gaulz. the work as much as possl.

② Write Gores for TM_1, TM_2 and for various kind
Randd. op. T.M's. In the case of the latter, devise ^{some} type
seqs. and see if the rand. criteria are reasonable

③ Perhaps make a TM_0 to solve certain types of pro
(e.g. arith. learning) and let me be TM_1 . Make so
that TM_1 uses the rite Gore on TM_0 .

④ Make a TM_0 to solve probs. with a continuous exten

→ as ~~313.04~~ 313.04 - .10.

July 25, 61

7 Mo.

Plans

01:31940 : ⑤ study ^{Boolean} network machines and random β net
 preparatory so that I can make a corresp. bet
 and Turmass (as 319.15--22)

⑥ Some work on HR and SM — and possibly
 the coding of continuous quantities.

→ ⑦ Design final "random net" TM, using R.
 Th. EPM section has some pretty good ideas on random net TM.

Also, ^(met recent) $\frac{1}{2}$ in Dart ^(I think) is a long section on the
 electronic realization of a recent. for random nets using
 an H drum analog-wise. Also see recent work in section on "Rand
 for good ideas."

SN

What I want to do is make a temporary
 ordered list of things to work on, culminating in the
 final TM. 319.22 is an attempt at such a list.

An impt. Q is to what extent work on any of
 intermediate projects should be completed. — i.e. into

review actual, formal Σ TB, ^{computer} programming, etc.

The point of completing some of R's intermediate
 projects past R's point of apparently max utility,
 that a large project (e.g. final TM) may often

more rapidly completed on R's whole, by ~~the~~ actually
 completing some simple progressively more ~~the~~

projects first. ~~MM~~ Ordinarily, an important

sub-goal of these completions, would be to get
 continuance of funds w/o extra interest in R.

July 30, 61.

T.M.J.

Plans

321.90 work. At the present time, such goals are
Small, ^{positive} (if not negative) weight.

A sequence of projects that are rather specific
culminate in a reasonable, malfunction-independent T.M.
The following are all operator T.M.'s, with recent goals
~~for each (1,0) pair.~~
for each (1,0) pair.

① Operators are ~~described~~ described by strings of symbols
and each such op. is given a probability by some rule
based on my a priori ideas about which ops. are
"reasonable". Various ops are tried, for the strings of
and they are each given a G value by the Gorc. This G
Somehow includes an emphasis on "more probable" ops.
Also, somehow, the G's of past ops tend to influence
probability values in new trial ops. We try, of course,
to get an op. of max Gorc. This Gorc is
think something like the ^{expected G} operation that the given op.
will have for the next (as yet unknown) input
or the exp. G. averaged over a seq. of inputs
in the far future.

② Same as ①, except that instead of "the most
op" we simply select ops that are described by strings
of symbols of minimal length. The set of symbols is
fixed and finite.

Aug 1, 61

T.M.

Plan

01:323.40: — So, if we build complex computers in
get plenty of computation before failure. — equiv. to
of human computation — or ~ 300 ~~yr.~~ yr. brains!

How many Zog's will to simulate hum. brain power?

Brain: 10^{12} comps/sec. 5-Zog's: 10^6 comps/sec —

So we would need $\sim 5 \times 10^6$ Zog's!

We can reduce the requirement by ① Saying Zog
complex as 100 neurons. ② speed up Zog by

— So this gives $\frac{5 \times 10^6}{10^2 \times 10^3} = 50$. So Hum. brain equiv. to ~ 50
these rather fast computers. (we would have to triplicate, so say

But anyway, the moral is: that from these very rough con-
we can guess that we can get useful T.M.'s w.o. using random
Perhaps by the time I get a ~~non-random~~ ^{random} non-(random net) computer
going, I will have ~~some~~ gotten some good ideas on how to do the true
net T.M. Or perhaps I can make a T.M. big enough to work on the
problem.

30 The type of T.M. that I am closest to being able to build
is a M.T.M.: We feed in math problems as inputs, and expect
answers as outputs. We have a routine of selecting the
"R. more probable" ops. that give R. ^{known} answers for all the
past cases. \rightarrow A ops of type machine might be

From such a T.M., could we ~~never~~ develop a tower
a hill-climbing T.M. ? — i.e. an h.c. T.M. is, perhaps
in general, is capable of dealing with more realistic situations
and temp. says. than ~~is~~ an M.T.M.

Aug 2, 61

T.M.G.

Plan

01:326.40: ① A simple kind of h.c. TM: Have 2 levels of G (the median G) and use 2 correl. lags. to ~~the~~ ^{a priori} divide trial levels. Devise various "qualities" to use in the correl. lags.

could also be of the sort that will help create new trials

An example of such a TM: TM_0 's are p.g.m.s. R represent operators. Inputs are g.u. and each R gives for each input. G is evaluated externally (usually partly on of speed of response). TM_1 tries to create R's of maximum mean G (or some other Gorc). ~~to~~

Note, here, that this ~~is~~ is an example of a noisy h.c. I think many (if not most) are of this type. It may be always poss. to get rid of all of the noise by comp (in some way), (the "Expected" mean G in view of the known in

② A more complex TM that tries to find the relation between the empirical G and its trials, and thereby ~~maximize~~ climb the hill with fewer samples. The cleverness of the trick depends on the "cost per sample" to TM_1 is opposed to "cost" of computing time.

③ An additional refinement to ②, would try to ~~predict~~ get the probabys of all poss. $\langle \text{next input} \rangle$'s.

④ A further refinement would hill-climb with a G that was based on \hat{G} (only probabilistically known) inputs of the distant future, or upon a mean over a set of such inputs.

⑤ A slight refinement of ④ would try to define more exactly just what I want in the way of optimum TM behavior in the long run. It might be something close to

W, Aug. 2, 61

J M J

Plan

.01: 327.40 : a large man G in th. far future. Perhaps goal would be that of developing TM as rapidly as possible with as short a top. seq. as possl. into a dev. that could solve real problems - i.e. problems that I actually want to solve and can't ~~do~~ w.o. a TM. A sign of this type, would be a S.M. w/ H.R. T.M.

.10 Much work on just what goals I want for a TM that is good

Now, for further "Plans" work: Write out ~~in~~ more detail on th. MTM of 326.30 and th. h.c. TM's that follow it. if they do, indeed, form a reasonable sequence, so that a model is worth while for itself. - keeping in mind, that a ~~step~~ ^{step} left very often a very difficult project is ^{total man hrs} accomplished by leading up to it by a seq. of simpler tasks. In particular, th. coding methods of code 273.1.01 - 3 & should be examined, since they are very relevant.

.25 (this is Rē MTM) : Rather than th. MTM of Dart Ref use an MTM utilizing ^{whatever fast} th. functions are in R. arith unit th. computer that is to be used. e.g. have it learn w

$((18 + 341) - 8) \div 3$ is, by numerical examples.

The problem will be one of constructing operators "similar" to those that have been successful in the past - when have to modify an op. (i.e. when th. op. has come to a point that it doesn't respond to properly.)

Th. problem of top. seq. construction is paramount in MTM. Any skill I gain here, will probably be directly applicable to h.c. TM's.

→ Also, I want th. abstractions ^(= categorizations) used by th. TM to

Aug 2, 61

Play

01: 328.40 be very close to those that I use.

Th. try. say. — get a ruff try say.

abss. I use, then add or modify exa

that TM can learn these abss. properly

.07 (1) Th. simplest h.c.TM (of 327.01ff). Here, various "qual" be used to categorize objects to predict whether their G be above or below Th. median. MTM (of 330) could be case of TM (1), if we ~~get~~^{give} successful R's \rightarrow G's and unsuccessful ones G's of 0.

Th. trouble is, this isn't quite what I had in mind. MTM — i.e. I expected to ~~have~~^{not more than} one comp successful R available at any time. — Even so, Th. may be formally quite close.

How, I do feel that the coding methods used in R.M are very likely to be useful for the h.c.TM problem. Even if it would seem that this most be true, since MTM recog certain "regularities" in the successful R — and these regular most be useful for h.c.TM to distinguish betw. R's of high G (i.e. $G=1$) and R's of low G (i.e. $G=0$; unsuccessful R).

At any rate, I do feel MTM and h.c.TM to be related, and it would seem worth while to work on both Th. some time — perhaps modifying both until they be more similar. It would seem that Th. problem. of design a good try. say for MTM, should also be impt. in h.c.TM.

Another imp. Q about h.c.TM is whether it will be able to climb noisy hills.

App 3,61

Plan

01:329.40: (C) On MTM: Say I were successful in

fairly good MTM - (a) just what would such a machine do
how would this be a step toward h.c. TM?

(a) I would have a Tng. seq. consisting of a seq. of inf.
known proper responses to these inputs. It is R's job of
to devise an operator (\equiv seq. of machine instructions) that

(I) Expresses th. relationship of input to output for all cases

(II) This operator is of ~~small~~ small cost (not necessarily minimal) \sum actually R's

must be $>$ a certain amt., depending on R's size of R's
seq. and R's "info" contained in R's relation betw. its
and outputs. This is Q is discussed much in Code 23

The operators, R, will incorporate in itself, ^{at least} all of R's heur.
tricks and abstractions that I think are useful and not
for solving such problems as I'm interested in.

I may or may not want to ~~see~~ compute to see if R's
of R is by end with resp. to R's known tng. seq. See
ideas on how to do this approximately - or get upper bound
p cost threshold.

(b) To ans. this Q, I should ~~perhaps~~ perhaps write a desc. of
what I'd want R's optimum h.c. TM to do, (there is much discn.
this problem in \dots) and then list some approx solns
that might be close to ~~the~~ R's above "idealized soln. to
MTM. Th. "approxns" will be both in goal (a) actual attain

Some kinds of approxn. that might be considered:

(1) 2 or more levels of G and correl. coding, using "properties"

(2) I think there are several approxn. ideas in
(3) Th. pure h.c. prob. (w.o. trying to predict inputs)
and R's noisy hill, and R's noiseless hill.

Aug 3, 61

Plan

01:33:40: Perhaps what would be useful, would be to
 several ways in which MTM is a special case of
 Also, it would be useful to note in just what ways
 are used by hc TM — since this would make clearer its
 to MTM.

Note also, that there is another kind of hc TM — the
 probably not interested in just now. This hc TM is ^{very} related
 SM and H.R. In this hc TM, the TM proposes ~~all~~
 input objects x_i (~~the~~ vectors) and is given G values on
 the ~~obj~~ goal is to find an x_i with max G as soon
 (this goal has to be more clearly defined, but that's the
 anyway.

How, I'm not so sure this is diffrnt from what I want ^{now} in
 the R 's corresp. to the x_i 's — and I want an R that will
~~continue to have~~ have ^{expected} $h(\bar{G})$ for the "next input."

There are various kinds of hc TMs — one's which consist
 the nature and structure of the "inputs" to varying degrees.

030 (1/2) Woops! Another / intermediate step betw. MTM and
 hc TM: A MTM that ^{can work even if there are} ~~occasional~~ occasional mistakes in the "try"
 Such a TM will have to have a much longer try seq. how. The
 is ~~aim~~ to minimize the total cost of Φ the operator $\textcircled{2}$ a is
 deviations from the predictions of the operator. This is ess

a digital version of the generalized continuous curve-fitting
 This problem is not at all far from the "simple" MTM, but it has the advantages
^{numerical}
 all R 's can be given a "goodness of fit" ~~measure~~ with an "almost fit" being
 useful. In the pure MTM, we throw away info about ~~the~~ "near misses"

Aug 4, 61

Plan

01. 331.40 : In general, what I can do now, for any of
is to write out a partial ~~to~~ solution or type of
from entirely intuitive considerations — ~~then~~ ^{then} ~~the~~ ^{the} ~~cod~~ ^{criticism and re-}
of ind. inf. will suggest a quantitative formulation.

05 ⊙ : While 330.01 expresses R_i goals of MTM, (and to some
of "noisy" MTM ($\equiv \text{⊕}$)), R_i methods by which this goal
achieved are not made clear. These are R_i "Search methods"
"heuristics". The method for MTM that I had in m.

20 Another big problem that I'm not too sure of is R_i Q.
how good are non-optimum sets of codes in making pred.
Can R_i use of such ^{any} code be made equiv. to an optimum
code, using a corpus with somewhat different regularity
in R_i "distant past"?

I did much work on this problem recently — I think
when I was asking just how good predictions I
expect from my 7090 pgms ~~for~~ for "coding with binary
definitions." I think this pgm. was desc'd. in ZTC
See Cod 311.15 for newer discn.

30 → 25 follows: Say one has an R_i that satisfies R_i type
up to T_1 . At T_1+1 we get a counter case. We then
to add to the operator R_i , Methods for recognizing
 R_i new probs. differ from R_i Odd. (\equiv an obs. operator) \oplus New
sub-ops. for dealing with R_i new kinds of probs.. Both \oplus
are constructed using ~~the~~ costs based on R_i statistics of
the R_i to which these new ops. are to be adjoined. Remember

Aug 4, 61

J. M. J.

Plan

01:332.40 : to include th. costs of punctuation!

02 If th. new codes cannot be taken care of at cost, then ~~there is a way~~ a TM will try to recover of th. more recent parts of R_i (or use some of lower cost codes that were found) in attempts to th. statistics and \therefore costs for th. new set of R_i . This is called "back tracking." ~~Back track~~ Back track facilitated if non codes are retained that are too close to optimum, — as well as retaining opt. codes. \rightarrow see 334.20-30

The above seems to be close to th. way humans th. requisite search for codes.

01 hc TM : What I had in mind, intuitively, is like this:

02 A hc TM that doesn't seem too far from ^{my} intuitive con of th. way humans do it, and seems close to ~~the~~ M (or, more exactly, to $\frac{1}{2}$): We take th. ~~known~~ N (F pairs of known best G , and try to "fit a curve". This "curve" is an operator, and ~~is~~ (I_i, O_i) of th. G be "close" to an empirical G (I_j, O_j) by having I_i close or identical to I_j and ~~we~~ have O_i close or identical. We make N "as large as possl." — ~~for~~ so as to \rightarrow

36 size, yet if N is too large, mean G will \downarrow . \rightarrow also see

37 Another, more general way to do this: Use all known (I, O) pairs, and choose 2 op. \rightarrow its (I, O) 's are close to th. empirically G pairs, and distant from th. low

Aug. 4, 61

T.M.

Plan

01:333.40. G pairs.

This letter seems fairly good, and ~~it~~ ~~is~~ ~~more~~ ~~available~~ ~~info~~ ~~than~~ ~~Pr. P~~

consideration ~~for~~ ~~more~~ ~~available~~ ~~info~~ ~~than~~ ~~Pr. P~~

.03

→ ①

A top. seq. for h.c.TM that would facilitate learning (be fine for early top., if not all top.), would be to th. some input to TM repeatedly, until he gets a of by ~~G~~ ~~for~~ ~~that~~ ~~input~~. Other, more sophisticated to be more like th. parent-child relation. e.g. giving recast. for th. learning of phrases in

Another advantage of Pr. h.c.TM of 333.37, is that th. idea "closeness" is used by humans and it would be well to do in a TM.

I have done some work on th. h.c.TM of 333.37 previously (these notes, but I don't know just where. I think ~~perhaps~~ I to use negative wts. for statistics on low G (I,0) pairs

.20: 333.18

Note that a "part" of an op. (i.e. its response to a certain sub-set of inputs) can be, slightly modified by other ~~either~~ either ① a recogn. op. ② changing an "action" op. ③ Changing th. ^{nature} action that is called into play by a recogn. op. ④ 2 or more of Pr.

Also ~~if~~, we can leave ~~part~~ part of an op invariant by doing 2 or more of Pr. above — so we may want to do one to fix op part of an op, then do another to cancel ~~Pr.~~ ~~evil~~ ~~wrought~~ ~~by~~ ~~this~~ ~~change~~.

.30

①

A simple approxn. to 333.30 is 333.22-.36. ~~These~~

.31

If we don't use th. "closeness criterion" and have our op. actually pass thru ~~th.~~ as many of th. best N (I,0) pairs as possl., we will simplify things much, and get a lot closer to TM $\frac{1}{2}$ Another useful modification see Part 340.05 for objection — also Part 342.05 for

335.0

Aug 4, 61

Plan.

.01: 334.40 is to use "closecross", but only the simplest, m

(e.g. physical resolution ϵ). This will \uparrow our samp.

.02 Cost in time and memory.

Note that this is ~~TM~~ - even the simplest type, will be a lot more than MTM. This is because the nature

"the N (I, O)'s of best G" changes with time. I one discards old members of the set as new ones

This means that if the genl. method of 332.30-.40 is be used, that a lot ^{more} revision (and perhaps "backtracking" (333.02) like 334.20-30 ap.) will

to be done than in ~~MTM~~.



= Epist MTM

A big ref. on an operator, observer T.M. is EPM 1107

and 1110.01 ff in particular

It seems to end on Epm 1143.40.

1110.
1143.

.25 But while this idea is fine, and I may, ultimately

I will at first start out by deriving how I, intuitively solve certain learning problems. I will first write how I solve these inductions - for the entire type seq.

Then I will devise a "log" to express these ideas formally and use the coding method to evaluate things and see how it works quantitatively.

So now, write a review of just what I expect to do in the way of MTM and the TM, giving as much detail both, as I can.

Aug 7, 61
Plays

.01: 335.40:

O.K. - So first:

TM₀ : (Refs: 328.25; ^{Tells what TM₀ does.} 330.01; ^{tells proposed method for TM₀} 332.05)

Anyway, TM₀ is a MTM: Tug seq. of inputs, and assoc. co

No errors allowed in tug-seq.

of TM_{1/2} : (Ref: 331.30) Same as TM₀, but ~~it~~ ^{it} ~~has~~ ^{has} a few occasional errors

Seq. are permissible. Actually TM_{1/2} involves a more exact of th. efficacy of th. kern. of the operator as a kind

"short code" for ^{an} ~~an~~ ^{input-output-set} corpus. Work o

should be done in such a way that TM_{1/2} will be and easy. I think that this will be possible.

TM₁ : (Refs: 327.01^{to 328.19}; ^{this also derbs future poss. refinements} 329.07; 333.22^{to 334.02}; 337.31) This is a sim of h.c. TM. ^{[This] ref. gives further th. direction of further developm}

more powerful h.c. TM's. Code 273.1.01 - 311.90 also has much on th. Q. of just what is th. Genc. of a h.c. TM if we want it BS. Also BS section has much on this Q. Essentially it tells just what ultimate goals (of excellency) an h.c. will have.

no → A very basic Idea: That an h.c. TM - in addition being able to BS. itself, can be trained to do anything one would like. I think there is a ^{few pp's} section in TM704 that ~~do~~ outline some tug. seq.s. for an h.c. TM, to get it to do various diff't, interesting things.

→ Another very Imp't. idea ^{of} Q. that needs more work. How ^{much} prediction accuracy can be expected from a pu. non-optimum coding method? ^{332.20 discusses this a bit.}

→ Perhaps a main goal of th. presently ^{contemplated} seq. of TM's, is that I can make a h.c. TM, that can BS itself. I will want to 337.01

Aug 7, 61

TM₀

Plan

01:336.40 develop a TM₀ and TM₁, in 11, so as
 TM₁ will be developable toward a CHTM of Th. 14. This
 can work problems ^{like} ~~of the type that~~ "prog. improvement"
 ∴ th. problem of improving itself.

→ Methodologically, the most imp. idea, is to first
 th. operations lang. seqs and operations of these
 on a very intuitive level, in ord. lang ~~etc~~ (≅ Eng)
 then slowly put in more quantitative form. (i.e. 335.)

Work on TM₀, and TM₁, will be continued in Th. 14
 Section i.e. Dart 338.01.

→ On ^{Dart} 341.35 ^{to 342.04} it is decided to work on RTM (reconfd)
 because this TM is capable of anything a human can do
 ∴ can be ~~see~~ ~~#~~ BS TM ← also th. hours a
 close to human hours, since humans are close to, if
 not exactly, a ~~#~~ RTM. Also RTM can be ~~#~~ taught
 Q's in English (≅ QATM), which is, I believe, a good form
 very powerful.

Actually, th. TM, that I've been contemplating is a ~~#~~
 - but I wasn't sure that I knew how to make a BSTM of
it. Hvr, since we can ask a RTM (→ QATM) to "improve
 himself", RTM can be a BSTM.

→ Dart 344.15 discusses a very imp. idea in a RTM, i.e.
 what extent such a RTM can ~~be~~ ~~learn~~ eventually learn very fast
 even tho its routine hours for search are not v.p. Dart 345.10
 extends this concept to "reliability" (≅ redundancy), cost, and wt.

Th. Sep 2, 61

~~Plan~~ Plan (396)

01:337.40: A lot of recent work in Dart, has been in attempt

Th. possy. of Dart 344.15 - ie. a simple / TM with ^{operator} _{automatic} simple search procedure, that somehow is able to develop equiv. to a very clever search procedure.

At Th. present time, I'm not too sure as to just how work, tho. Around before Dart 386 I was fairly optimistic it was possib, and had Th. idea as to how it might be done.

.15 The discussn. sort of culminated in Dart "Th. system"

Dart 381.03 - ^{387.35} Th. desirable features of Th. system listed on Dart ^{387.06 ff} - followed by a few of Th. impt. probs.

At Th. present time, I'm not really sure as to how much an L3. \Rightarrow L3 must be able to deriv. all possib. ^{generally} reg in Th. set $[M_i]$. Hrr., even if this part of Th. sys doesn't work, I can still be L3 for a while and still have all advantages of 387.06 ff, except #3. ^(sort of) ie. speed in search, as typ. seq. continues.

But Th. remaining advantages are ~~not~~ considerable

Also after I write up a few L3's, I will have ideas on how to improve Th. automatically, and perhaps get Th. TM, to do this improvement of L3. - or just make a hier order search TM to improve L3. With such hierarchy (even w/o $TM_i = TM_{i+1}$), Th. power of Th. may be enuf to work signift. probs. - such as externally asked and / self improvement.

Dart 387.35 ~~that~~ Th. system doesn't try to anticipate Th. kind of probs. Seems to have Th. most serious criticism of Th. and goal of this system. This will certainly slow Th. system down in Th. typ. seq. Plan 396.35 may have way out of this.

F. Sep 8.61

Plan

.01:395.40:

Th. "Probs" are fed to \mathcal{M} as inputs, to find ~~an~~ ^{simple-as-poss.} operator that will work all probs. correctly time. ~~These~~ Trial ops. (called M_i) are at first constructed by \mathcal{M} of Th. basic set of computer orders (or operations, or \mathcal{M}). This results in a seq. of ops. \mathcal{M} - each of which has been up to a certain pt. in Th. seq.

(second) order TM?

After a few such ops have been obtained, L_3 (order TM) looks at, say, the last 10 successful ops. tries to find regularities in them that would be useful in reducing Th. no. of trials nacy. for Th. next problem. At these regularities will be nps defs and perhaps nps de \mathcal{M} . As more these operators get bigger (more complex) we begin to look for more complex regularities (since \mathcal{M} is larger "corpus"). Ideally, any regularity that I think of intuitively should be "accessible" to L_3 . The exact meaning of this isn't entirely clear to me, but

.22

Th. regularities L_3 observes should be expressible in Th. of ~~Th.~~ parameters (or other features) of a creative \mathcal{M} ^{Th.} \mathcal{M} . That is to create Th. M_i trials. One big prob will be to design a universe of Grammars G_3 , \mathcal{M} \mathcal{M} is possl. Also, we want to be able to construct regularities in G_3 , to see if they hold in Th. set. Essentially, we want Th. new trial regularities in G_3 to be formed by combining Th. old successful ~~Th.~~ regularities.

.35

It is possl., at this point, for L_3 to notice Th. sequencial regularities of Th. sequence M_1, M_2, M_3, \dots This would get around the objection #5) of Doc 387.35. Hvr., at first, we wouldn't bother with that.

(see for ...)

Plan

$(M_i \text{ is } t. \text{ op. that has been successful on all I/O pairs up to } t_i)$

$M_i \text{ is } t. \text{ op. that has been successful on all I/O pairs up to } t_i$

01: 396.40:

TM₁ (\equiv HCM \equiv RTM)

If we get

properly (as sh. foregoing pp.), then we can modify

2 RTM (\equiv random TM) directly.

In MTM we have these ops, M_i Also

G_3 (a 'creative' gram) for making new trial M_i 's

These can be carried over directly ~~to~~ from MTM

In MTM, L_3 can test each M_i created, since

~~each~~ single correct responses are known for all inputs. On

basis, a particular trial G_3 can be evaluated - i.e. how

a particular G_3 is very good if it takes only a few

trial M_i 's to find one that fits completely all the

past inputs.



In RTM, an externally evaluated G is given to

for each response to an input. At first, TM can

estimate sh. \bar{G} and σ_G of a trial M_i , by feeding

it inputs of 'known output', until some outputs of known

G are obtained. In more advanced machines, TM

will try to get an approx. functional relation between

G and say (I, O) pair, by seeing how "close" a

(I, O) in question is, to other (I, O) 's of known G

(Later, more general forms for "fitting" G as

a funct. of (I, O) will be used). Using P_i 's for

form of $G(I, O)$, TM can then evaluate M_i 's

approx. very quickly, by using a few ^{previous} ~~the~~ inputs as

signif. and evaluating P_i expected G 's of sh. result that (I, O)

vary by/

and evaluating P_i expected G 's of sh. result that (I, O)

In RTM, we can then get G_3 to try to create

F. Sep 8, 61

Plan

01.397.40: M_i 's of \bar{G} . As ~~is~~ in MTM, L_3 will
in \bar{G} . M_i 's of known by \bar{G} .

Th. \bar{G} M_i of \bar{G} ~~will~~ ^{known up to that time} usually be able
make a response to any input.

.05 There is some confusion in \bar{G} . above discuss. of L_3 -
for MTM and RTM. At first ⁽¹⁾ spoke of finding "reg"
in \bar{G} . "good" M_i 's, then somehow incorporating the
regularities into \bar{G} . creative stock. G_3 .

Later ⁽²⁾ spoke of trial modifications of G_3 - as
of hill-climbing problem - with ~~the~~ hill ht. determined by
with which G_3 could find a "consi" M_i (for MTM)
a better than before "hyper \bar{G} than before" M_i (for RTM)

The ~~this~~ ^{this} is probly a less el. soln. I think that probly
this will be used at first. A possy. is to use ⁽¹⁾
find good trial modifications of G_3 (ie. by looking for "reg"
in \bar{G} . "good" M_i 's), then using ⁽²⁾ as a less el. soln
of \bar{G} . efficacy of these modifications of G_3 .

The plan for MTM, ^{first} is to construct a typ. seq. leading
~~the response~~ some interesting capability and ^{leading to the requirements of} some interesting
kind of induction. Next, I will write a seq. of M_i 's
(~~or~~ or several such seqs) that "solve" \bar{G} . typ. seq., using
 \bar{G} . basic computer orders or ops. in a "minimal" fashion. I will
try to make \bar{G} . solns. ^{at each primitive typ. seq.} use abss. that were useful in solving
earlier probs in \bar{G} . typ. seq. I will try to devise defns.
so that \bar{G} . solns. I have gn. are compact and perhaps close
to "Minimal".

I will then try to devise an L_3 that can search for

For Sep 8, 61

Plan

01: 398.40: "regularities" in the "latest" of the M_i 's, and to create defns. to make the cost of coding the defns. ~~is~~ ^{always} this is ^{by defn.} ~~passl.~~, follows from the fact that found is a "regularity".) (2) Modify the Gram, so that it produces the desired "good" M_i set ^{by} ~~the~~ cost as passl.

All of this manipulation of G_3 , the search for M_i in the M_i , will be done at first "by hand" - I find a good way to mechanize it.

Def The finding of regularities in the M_i : This will be done by starting with simple regularities (\equiv reps), then use the successful reps. to combine with each other to produce new total reps., ~~which~~ which are then tried and then etc.

The thing that I'm vaguest about is G_3 - its form and mode of modification.

In the top. sep. - try to keep things on a pretty inner level before deciding what kind of order a/o ops to use.

Do the same thing in discussing "reps." found by i.e. keep it intuitive (i.e. in English) as long as possible.

30 An example of what "intuitive" form of the type might mean: "First MM will learn the meaning of ~~the~~ +, then - x and \div . Next, that - is the inverse of + and vice versa
" " x " " " " \div " "

Next, combined operations - evaln. of expressions containing +, -, x, \div and parentheses.

Or use Polish notation throughout. Unfortly, this makes the associativity and commutativity of certain operations have low opprip.

F sep 8, 61
Plan

o: 399.40: Next, ~~simple~~ solve simple linear equs. - 1 on
" " " " " " 2 "
" " " " " " 3

The above typ. seq. may actually be used to give inter
induction probs. e.g. How big a jump would it be from

10 3 on knowns to, say 5 unknowns? → 401.18

If the above typ. seq. is inadequate try 395
plans

prabs. that might be made into an adequate typ. seq.
An Alternate typ. seq. would go from linear equs in 1 unknown to quadra

Things to discuss further:

1) It is poss. to order the trial M_i 's rather simply, in looking for a
Def is "consi" (i.e. consistent with all previous (I, O)'s) in MTM. ^{Make}
Defining h 's and g 's that \uparrow search space \approx sort of me
apripd of M_i 's. Discuss interaction betw. h 's, g 's
"into in R. (corpus up to now)".

2) The details of the action of L_3 or G_3 are not clear. I
to have 2 diffrnt ideas in mind. (a) "Randomish" Mutation of an "almost so
 G_3 to produce new trial G_3 's (b) Noticing regularities in R
set of "Good" M_i 's and incorporating these regularities into

I should be able to merge these 2 approaches. (398.0
with)

Also, I should define G_3 , L_3 , \approx and "R. hyper
TM" more exactly.

3) 396.22: "That all regularities that are intuitively conceivable (by me
in R. set of good M_i 's) should be 'accessable' to G_3 ".

The meaning of this must be clarified. ~~After~~ One way to
this would be to get examples of $[M_i]$'s and express
the regularities in them intuitively. \Rightarrow Also these regularities
must be creatively expressible by G_3 .

For more general ideas on this consider diffrnt. forms in
which the M_i might be expressed - e.g. in LISP, in 401.0

200 Sep 10, 61.

~~400~~

Plan

.01: 400.40 conventional 1 or 2 or 3 or 4 address comp. orders, system, etc.

.03 [SN] An example of a "sequential regularity" that should be realized by G_3 — or should be "wired into" G_3 : say M_i are in th. form of recognition ops \approx (\equiv obs) x fun ops (\equiv ops) as Dart 360.11 ff. Then G_3 should realize that th. form of M_{i+1} is related to that of M_i 's having either (a) one new Ob. or (b) a simplification of th. set of previous obj's. And a new Op $_{i+1}$. And it should be ~~wired~~ built into G_3 , that (a) is more likely.

.18: 400.10 : I suspect that th. top seq. of 399.30-400.10 would be better constructed, if I worked backwards — i.e. Take th. final induction problem I want VM to solve and list th. "concepts" that would be easy to solve it in a reasonable ~~small~~ ^{make} no. of trials (Note that it might be expedient to ~~list~~ ^{make} several lists of concepts — each of which is adequate.)

Again, I would like to do this by starting out as intuitive & literal as possl. — e.g. th. sort of notes used in th. discn of Dart 390.35-391.35 — Note also ^{there,} how to work backwards

Note that th. writing of th. top seq. in "intuitive" makes it possl. to xlt. it ~~to~~ directly into any comp. lang. type. So — perhaps for studying L_3 , I should keep th. seq. $\{M_i\}$ in intuitive form and write the regs. of $\{M_i\}$ in intuitive form first — then, consider various formalisms in which ~~$\{M_i\}$~~ ^{th. M_i} and in which G_3 may be expressed. Continue in Dart: Dart 402.01

W Dec 6, 61

Plans

.01: 701.70: Th. A.F. seems to be breathing down my neck.
 ← Big Wheel at AFOSR.
 Milsaps facts that publishing in reputable journals is the

thing. Th. Q is, how much ~~time~~ am I expected to spend a year?
 Say 2 serious papers for Info and Control -

1 for I and C, and another in some conference proceeding.

This latter is good, if it is an "invited paper" - because I can spend less time preparing it. This latter seems

2 papers a year would not be so trivial as to be "journal

Th. 2 papers can be, for myself, good "review articles"
 actual papers presented will be a cut-down version of my
 i.e. no refs. to TM notes, ^{little} discn. of reasons for various
 weak pts. of discn.

More specific present plans: first get 2TB 191 out. This is already written and has only to be proofread.

.20 Next write I.C. paper ... This will, for me, be a rather
 demn. of th. coding method and various variations, also much discn.
 weak pts. in th. theory, and work that must be done. Also to

specific applics. to specific problems, that have been worked
 to some extent.
 I.e. ① Bern. seq. ② Coding with binary data. ③ Coding with non data (I)
 to work this out quickly ④ PSG coding ⑤ Non-Dim. PSG Coding. ⑥
 for the IR ② Presence or absence of strings, and β combs. of strings ⑤ some
 rel. of these
 with frequz within th. doc. ~~with frequz~~ ⑦ Th. soln. to th. B.G. prob.

.25 ⑧ Coding of continuous data ② Correlative ^{How to use randomly chosen params.}
 for printed or written letters. ③ Th. genl. finite history Markov chain.

.35 Next, write Chicago SOS conf. paper (Typ. seqs. for mechanized
 For my own use, I should write a very complete review of my
 methodology, and th. reasons for th. choices involved. In particular,
 a fairly detailed outline of much of th. expected future work.

- For th. SOS conf. paper, just give part of a typ. seq., - some
 solns., and some ~~state~~ solns. to th. prob. of generating solns. There

Plans

G

01:484.40: is no particular virtue in writing the paper
 - with the "latest developments" - since, as it is, 89.
quite long and adequately meaty.

Dac27,61 ^{most recent} On the work in "Dart": (see say Dart 42.01 - 480.01)

idea was to follow very closely my intuitive method of working problems
 quantifying the probs involved by using epsilon (= coding Method)
 Unfortunately, I fear that in this recent work I have strayed from
 intuitive notions, and have gotten into an excessively rigid
 notation. The problem of the 2nd order TM ~~was~~ that was

- that of PSG'd WOQ - is fairly distant from intuitive prob.

The formalism used at the first level ~~is~~ was that of an operator calc
 + bunch of (unordered, I think) substitu. rules. Rules of this sort

close to the intuitive ones, for probs. like arith. eval; solving prob
 solving ^{linear} n eqs. in n unk.; solving quadratic eqs. I'm not

sure there are many other prob. types that can be worked this way
 (i.e. in which this way is close to intuitive method). Instead of attacking

the higher order problem directly, try to find other kinds of
 problems, in which other formalisms are more close to the intuitive

Also, consider the 2nd order problems as being 1st order for another T.
 try to obtain intuitive-like formalisms for them. The idea is to

2) set of ^{unified} formalisms that appear to be adequate for all probs.
 think of. Note that for some probs, it may be necessary to have

long temp. seq before the solus. within the formalisms used have reason

Also note that the 2nd order problems are somewhat distant from
 1st order probs. The 1st ord. probs. are ops. - The 2nd

probs. may not be ops.

Also, I'm not too sure about how this idea of nth order TM is related
 to my ordinary concepts of intuitive regularities.

Plans

01: 485.40: A rather old intuitive concept of a TM: (this for recent ~~work~~ section in "Dart"). At a pn. time "total method of coding" - that includes parsing and coding in terms of a set of abs. that have been found to be ~~found~~ new data comes in, TM. can ~~work~~ ~~code~~ code this data in these abs. in a routine way.

In addition, TM examines small samples of recent ~~and~~ and recent partially and totally coded input for regularities. Finding any of them amounts to a ~~new~~ new ~~one~~ but must be confirmed on th. basis of a larger sample. ~~man~~ man and animals, this "larger sample" is usually obtained by waiting for more data input - but it can, in principle (and probably in practice) be obtained from th. record near past, as well.

A Q. is: Does there exist a set of code symbols (with and a set of ~~things~~ ~~to~~ look for)'s \rightarrow using th. technique of ^{finding} defining new abs., we can by iteration of method, discover "all" regularities in a corpus? If cannot do it for "all" can we do it for a fairly large set of regularity types?

Essentially th. method is that of ZTB141, except that instead of looking for new zgm's (or even ygm's), we ~~the~~ look for ~~various~~ various other regularities in th. corpus. Also we try various parsings of th. ~~coding~~ ^{corpus} looking for higher order abs.

It may be possible to pattern this TM very close upon my own apparent mental processes, since this is apparently th. way human work.

As for th. coding itself: each abs. is gn. a probability which
492.01 492.01

Plans

01. 491.40 . is \equiv its post. Th. by post abs.
which scan ^{in ll} and code th. corpus as it comes
human CNS, certain ops. have such a by post, to
data runs thru them like transducers and actually a
(that to be irrelevant) to be lost). Other ~~WPSG~~
abs. are tried in turn (sequentially) in order of P

Also certain abs. of very low post are tried at random
small ^{recent input} samples. test of an abs

Actually, this allocation of trials is not entirely on
of post, but partly on th. basis of how much time
equipment is needed to implement th. trial. —
a sort of "op. res." prob.

Th. Monte Carlo aspect of th. coding rule is only evident
th. ~~low post~~ ^{higher} coding methods. For th. by post abs., th.
and parsing is almost certain.

An apparently good source of problems for such a machine,
text. Here, th. old abs. are ~~the~~ whatever parsing rules are known
along with their parser posts. Say we use a WPSG to start.
could more gramm. rules of this sort be derived? Other rules
are Chomsky's "xfuns" — that are ^(perhaps) common ways in which many s
~~are~~ parsed. Actually, I don't know it, at present, I have enough
of things to recognize" to be able to deriv. many PSG rules
— there is hvr., some possy. that I do — i.e. see recent w
in "Dart", in which I was ~~using~~ ^{to test} th. palindrome lang. and
 $a^n b^n$ as "test langs" ~~th.~~ ^{th.} power of th. methods prop
th. trouble is — using this approach, I get into th. PSG deriv
prob. — which is a diff. prob. I would like to take a try. s
in which I am ~~not~~ more or less could solve all of th. prob
w.o. much diffy. Possibly if I used English as input, I could find
other, non WPSG-type rules that would be adequate.

Plus

8493

.01: 792.40!

In particular - / context dependence

Things to look for in trying to find more param.

Substitu. probabys upon in a WPSL - this mits look

Using previous history of substitu. at a substitu. pt. mits

Some of Chom's "x fun. roles".

Another early idea, was to have TM learn one lang. L1, then learn another L2 ("learn" in Pi. sense of being able to predict

well as possl.), then learn ~~x~~ from L1 to L2. This

MT learning, is ~~like~~ not too distant from input \rightarrow output

or Stim \rightarrow response learning.

How, at first, perhaps, try to apply this idea to regularities other than ~~the~~ human lang. sequs. (which are prob. rather by state of devlpmt) - try noticing regularities in series - like S.M., or H.R. If human lang. is

stick to more directly ~~notice~~ intuitive regularities ^{types} - then

lots have been found, see if they are strong enuf to do

2 test PSG's (a^n b^n and Pi. Palindrom lang.) - for human

Such a seq. of tests mite be: 2 gm defts, ... n gm defts, optimizn of

~~units~~ upstts (i.e. define an upm as one that usually comes before a certain word

of by post), ~~perhaps~~ notice then define new upstts by combin

upstts ~~and~~ notice a certain set of new upms usually precede

previously define upst. - Notice structures - i.e. certain general forms

~~first~~ as's of basic phrases - Find ways in which various param

roles ~~to~~ could have been decrd (but use Pi. grammar rule: types of cl

grammars, not Chomsky. - This is because th. classical gram. ty

probably easier to decr ~~for~~ humans ~~using~~ "conventions" (for h

techniques.)