

Th Ap 2, 64

TMJ

Plan

01:776.90

II Th. Transp. Seq. :

There are several poss. types of Transp. seq. that I may want to use. A few of them are:

1) TM starts with ~~very limited~~ ^{children's} text of ^{wide} spectrum of ^{limited} vocab. and very limited / ^{grammatical} str. ^{are fr.} very freq. to be sure TM is looking for Relevant tags. Text develops as rapidly as poss. to Re. pt. at which TM can use a (Eng \rightarrow Eng) dictionary. Next ~~sub-goal~~ ^{sub-goal} in Transp. seq. is TM's ability to use an encyclopedia. Another sub-goal of much import, is TM's ability to get some info from figs. of speech and other strange expressions like 765.10. Another sub-goal is TM's ability to deal with various typo errors, and grammatical errors, and literally "meaningless" ss.

For each QA in Th. Transp. Seq., TM's "successful" completion of that QA means that it has found a way to obtain Re. info answ. (or a set of s's equiv. to that A), by assigning a hy. proc. to that (set of) answ(s). Also, it must have a search routine for finding such an op. in a reasonable time. If TM is unsuccessful in either of these pts., he will present an output query stating what Th. situation is. Th. operator (\equiv client) then must present accepted solns. to these probs. on Th. "special input channel". With Th's extra info in Th. corpus, TM will try again to Th. prob. in Q, and present Th. procs. of solns. and search time resulting from Th. new suggestions. Th. then decides whether his input suggs. have equival.

Input suggs. will take Th. form of A by proc.

Th. A could have been

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01: 777.40 : Corpus. Suggs. must also show how such a rule have been found by a not-excessively long search. (Sugg would be a sugg. for TM₂).

At any time TM_1 consists of an operator that on any input Q , and processes it with R . Known text produce an output A or a set of output A 's of R with P -costs being listed. If unable to do so, TM_1 will state that it cannot, and tell how far it has gone toward goal.

TM_2 's job is to improve TM_1 's work in R . sense of and more use as well as "accuracy" of products. This is with resp. to R . goals of answering R : Q 's.

20 2) TM_1 starts with R . simple text of 1) (777.09), but is gn. at first, not QA 's but R . problem of xltg. of Eng. into ~~SL~~ SL. Also tng. for $SL \rightarrow Eng$. When TM_1 has learned R 's, R . problem of xltng Q 's into SL is taught. ~~Next, an integration of R into~~

~~processes in which TM must go from Q to A~~

Next, TM is gn. a text in SL, also Q 's in SL, is taught to ans. Q 's in SL. At first simple Q 's in which R . A 's are just a subset of s 's in R . text.

Next, more complex Q 's are considered involving xfmns ~~among~~ among several s 's in R . text.

Next, R . processes are integrated by Eng. text Q 's in Eng. and A 's in Eng. At first TM does this

by: (Eng text \rightarrow SL Text; Eng $Q \rightarrow$ SL Q) \rightarrow SL A 's

At last, TM_2 finds more rapid ways to find acc.

all off these intermed. steps. R . SL lang.

for TM is a necessity

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TM



Phn

01: 778 to: → Th. tng. seq. of 2) (778.20) could be used as expt to that of 1) (777.04) — and probably will be.

In any case, Th. "sugg. channel" should be used to make TM "understands" everything he learns.

3) A third kind of Tng. seq., in which we start with fairly complex text, and very elaborate explanations on Th.

"sugg. channel." — with Th

Hr., to do this, we must have a large corpus of such for TM to start on — so he can legitly "find" various success rules for relating text and Q to A. This is very diff to do, because for just one QA, TM will have to devr. an enormous no. of rules. These would not only be rules directly used in Th. sentence involved in Th. QA, but would be rules used in "understanding" other ss, so that rules directly relevant to Th. present QA will have adequate samples.

→ so this kind of Tng. seq. doesn't look too hot.

III TM's Internal operation:

As was mentioned, TM, consists, at any time, of an operation, Rel. xfm any Q, along with Th. existing text, into an actual A & set of actual with their corresp p costs. So TM, is an/pgm. with 2 arg.

② Th. text (including all known QA's) ③ Th. present Q.

Also, assoc. with TM, is a symbolic pgm. or set of pgms.

of TM, in Th. sense of CMI, or pgms are induction

of "regularities", or devrs of induction methods.

equiv: to devrs of part or all of Th. corpus.

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T.M.J

Plan

.01:779.40!

Consider, e.g. R. problem of learning $E_{ng} \rightarrow SL$
 TM is given R. corpus of Eng, SL pairs, to start out
 Then, ^{TM₂} starts with R. first pair in R. corpus and tries
 find an op. that xfms it into R. corresp. SL (or equiv.)
 TM is given R. rules for certain equivs. betw. SL exp.
 Actually, TM₂ wants an op that will do R. for all R. of
 R. corpus, but he is starting with a "part op." This part
 will complete R. $E_{ng} \rightarrow SL$ xfm. wherever it is applicable. The
 is, it ^{should} produce no errors in MT.

Ops. will usually (if not always) be constructed of "obs" and
 An "ob" is an observer. It looks at a string and has an
 output that is 1 or 2, ... or n. This is a decision op.
 and tells what ob 2/o op. to apply next to R. string in
 "ops" actually xfm. R. string into a new string.
 All obs. are a subset of ops. All obs and ops
 be constructed by combining other obs and ops in standard
 ways, so obs and ops are easily ^{defined} in terms of
 obs and ops that have been defined. This makes/proves
 easy to compute. Also, it may be useful, for proof or in. to
^{define} obs and ops recursively (like LISP).

My tentative idea is to how this would work: For this small
 corpus, I give TM₂ an op. that works. This is TM₁
 Also, I try to find a minimal decn. of R. op. I
 I do R. by try to find sub-obs and sub-ops. build
 them hierarchically 2/o recursively. - So I also
 "decn" of TM₁ (779.35). From this decn, there is
 to compute its post.
 This form of TM₂ must conform as much as
 to how I do R. in MT.

Th. A p2, 64

TMX

Plsn
• proof 780.20

01: 781.40 : Next, I add a few more (Exp, SL) pairs, in which old rule will ~~not~~ not work. TM_2 must then modify R. a rule to obtain one that will work for all cases of

~~779.35~~
I then give TM_2 R., modified form of TM_1 (symbolic form of 779.35) that will work R. new as well as old pairs. I also tell TM_2 how he could have looked for ops. so he would have found R. "correct" one in a reasonable length of time. This search routine is also gn. in "symbolic form" for TM_2 has a fixed routine for going from R. symbol of TM_1 to R. actual pgm. (i.e.: this amounts to a compile

I continue to give TM_1 new Φ pairs, that involve new rules, and I try out TM_2 's rule to see if it is adequate to find R. new TM_1 rule. If it is not, TM_2 's desc. must be modified. This is R. job of TM_3 , but at first

I do R. modification of TM_2 and express R.'s modifications in "symp. form".

At some by value of n , TM_n will be max, but will be a fairly fixed routine.

In the final PMTM, TM_2 or TM_3 will be one of R. "modes" of TM_1 - so PMTM will be able to attempt to improve itself.

At first, it will not be poss. to use R. "special supp." (776.22), because TM will not understand Exp. or SL.

For this reason, it might be best for TM to learn SL text first, so that he will understand "SL

poss. and \therefore be able to properly interpret supp. phases

Plan

01: 78.40: The IR aspect of QATM operation: As data comes in text, TM codes this info partially — i.e. it assigned index terms various sections and subsections. These index terms are for quick retrieval (i.e. isolation of ~~the~~ parts of R. ~~text~~ that very probably contain R. info) with resp. to expected and, at first with resp. to past Q's (since these are an appropriate Q's).

As new kinds of Q's are asked, new index terms are added and old ones may be discarded (depending on R. format) and old text may be reindexed, if it seems to p. R. depth of indexing (i.e. R. amt. of info contained in the index terms), will depend much upon R. material, upon R. speed/needed when a Q is asked. If low response speed is O.K., then shallow coding will be used, and much material will be retrieved and "read" to answer a Q.

SN The importance of an all over analysis of TM operation with resp. to an "ultimate goal" is imp., in that much of R. operation of R. hum. brain, and one's intuitive observation on it, are involved in problems that are not pure index. E.g. R. IR problem involves ~~the~~ use of index terms, a varying ~~and~~ preliminary (first pass) depths of coding would be relevant to a "pure" ind. inf. machine whose was simply rapid ans. of a Q, with work starting after R. Q was asked.

Info is stored in hum. brain (and probably in TM), in v. ways — depending on R. speed (and accuracy) of access. For inteq. used info, knowl. of where R. info is obtainable. Say R. name of R. book. Even this will not be simple and is a simple role to go from R. Q to R. place (its in ~~some~~ encyclopedia). The role

Sat Apr 4, 64

TMY

Plan

01:782.40: If R. info could be deduced from a simple genl. plus a few params., then R. rule and params. are stored. Very rapid access is not needed.

In all cases, if rapid access is needed, R. info is in a form in which it may be most quickly used. Also, it is in a way that makes it very likely that R. situation or Q. requires it, will indeed, call it up.

Probably a direct, // access mem. on many index (like ~~Σ~~ coding) would be very fine for much of mem. The so-called "assoc. mems" could also be used.

For very freq. used facts, R. facts themselves are stored of their source. Then, if R. text repeats a stored R. fact is not rerecorded.

A human stores info and parts of info with resp. to R. kind of Q's he has been asked, and with resp. to R. kind of Universe he has seen (or ~~has~~ ^{thinks} he has seen) in. R. past. These ~~mems~~ facts make it poss. for him to store a very small part of his input, w.o. danger of error in reconstructing R. supposedly observed scene.

If, however, one a person is shown a scene marked different from ^{any thing} what he's seen before, & /o asked Q's entirely different " " " " been asked before, his abs. way lead him to entirely wrong results. His reconstruction R. scene will be wrong.

E.g. If he's only seen info with B.W. then he can use 2 f pts./sec. to characterize. He sees info with B.W. > f his char. pts./sec. will give a false result.

TU AP 7, 67

TM's

Plan:

01:783.46 : IV General Idea of this TM :

Ideally, we could start out with a TM that had with it a priori info that a human has at birth - including the hours and genl. principals of induction. We could give this TM a Tug seq. n to that which a human child ordinarily recvs., (with slight modifications in accord with FB info we get by asking Q's of TM). From this we could eventually get TM to read Eng., and to improve (even to "improve" ~~some~~ general computer pgs., and finally, to improve itself).

Avr, since we're rather uncertain as to what info human has at birth (- also this birth info and methods may be closely linked to the technology used in human ^{possibly} and \therefore inappropriate to an "electronic" (say) machine), we take the follg. tack:

We start out with a simple tug. seq. and do more complex probs. At first, we tell TM how to solve the probs - using as "minimal" a vocab. as possible ("min." in the sense of hypcost). We also give TM the principals of induction. We consider our "explained" a soln. to a prob. "adequate", if, after we've given the soln., he can then solve n probs. ~~is~~ using a relatively loose search.

We hope that after we have given TM enough probs. along with their "explanations", he will learn within him, just about the same info that we have after that child was able

To Ap 7, 64

TM

Phu

01:784.40 same probs. as TM. R. child and TM would, hr., be arrived at this pt. via diffrnt fgs. seqs.

After we have gotten to that pt., hr., the seq. of child and TM will be about the same except that we can get more detailed F B for TM - TM being more "open" to us. (2) we can tell TM how to solve a prob. and give him an "explain" and know whether that explain "good end" for him.

Tu Ap 7, 64

T.M.Y

Plan

.01:785.40

V Expected Diffies

n.c. prob
→

1) Local Maxima : This is always a big diffy in any or schemes in which one takes a previously satisfactory pt. and tries to improve it, or ~~to~~ make it acceptable to a slightly changed envt. — Some standard tricks:

a) Retain several solus. to each prob. (this is normal anyway). This will make it possl. to have very many essential diffn "near by" trials — i.e. trials with hy. chances of success

b) Retain history of 'hill climb, so that one can back a way, and try for ^{new} soln. to old as well as new probs. is sort of automatically done in suggestion a), but it may be necy. to "go back" in R. sense of finding more non-satisfactory solus. for old probs., before solving R. present prob. satisfactorily.

.20

c) Devise a method for detecting how R. new prob. from R. old, so one can recognize it and find a special soln. to R. new prob. Later, try to find ways to fit this "ob" into the entire soln., by simplifying R. ob — so that it need not recognize only R. new type of prob. — but recognize any prob. in which this new kind of soln. is needed. Also, one may be able to simplify R. to R. present prob.

This, at first, a.h. soln., can perhaps be related with other a.h. solus. that have occurred in R. ^{prev}

W Ap 8, 64

Plan

Don

(787)

TM



01: 786.40: . Expected Diffys (cont).

2) Th. Eng \rightarrow SL phase:

a) I don't have a good SL layo. to use. See work of Dan Bobrow, in Monk's Group; Darlington in Yng. See "Loglan". J. Perry's "Tulag. Abstracts".

b) I don't know just what a genl. form. such xtra. rules sh

~~3)~~

3) I have R. Eng \rightarrow SL ~~Phase~~ ^{Phase} for both text and A's. — Then R. SL ^R for relevant and R. xfun of R. relevant data. R. A of R.

~~Wanted to know:~~ I will have to do this in the ^{el'is} ~~at~~ first, but would like time to do non-el. operation

with R. option of forming his own SL layo if he want

I don't know just how to get TM to go from el. non-el. operation in R's. ~~operat~~

4) There are several "tips. of speech" and other "diffit" types of expressions as 765. ~~to~~ ff. In theory, TM could learn understand what these mean by ^{finding} ~~understanding~~ Rags. in R's and in R. Text + QAs. The exact way that R's sh be done isn't too clear to me.