

T.M.

array they are. For this reason, if they are in the "wrong" part of the array (distant from where □'s will be) they will not "fit" or "apply to" the □'s that are asked.

.05 (from α 81.11) The method of 69.10 does apply to all cases and is automatically ^{but nec. automatically} used in the method of α 81.16.

According to 69.10 we collect data on all points and make up ngrams. We then ~~we~~ keep freq. counts on only those ngrams which ~~we~~ have ^{at least one} ngrams in them that has been applicable to an example with a □ in it.

* I'm still not entirely satisfied that α 81.16 is not ad-hock, or that it is in any sense optimum.

15

Consider the full prob: One is given a very ~~big~~ big set of examples with □'s in them. Then one is given a Q with a □ in it - what is the optimum prediction?

It would appear that 81.16 is inapplicable - the q. is - is this a "legit" problem to ask of T.M.?

Perhaps: It is a q. of whether we want to categorize the □ sqs. with only other □ sqs. or with all sqs.

In this case, we have no choice; we must cat. it with all sqs., since □ occurs only once. In general, it is best to cat. it with □, if the sample size is large and.

If not, we must cat. it with all sqs. We may look upon ~~each~~ each of the 2 types of cat'ization as having a certain ~~specificity~~ mean reliability (for ∞ sample size) along with a lack of reliability due to finite sample size.

A good way to use both kinds of data is to use a weighted mean of the 2 kinds of data. The ^{rel.} wts. will depend on both sample size and expected reliability. I think the □ type should be expected to give zero mean error, whatever that

.01 Before we start on this review, let us clear up this

□ problem. Th. discussion on α 73.13 and α 73.31 is not correct. Th. method of α 69.10 is not as good as was that on α 69.10 — on th. other hand, it may, in th. long run, be best. Th. reason why th. discn. of α 73.31 is wrong! In lita of th. new way ~~of defining~~ of defining ngmsts:

1 is of hy freq. ∴ th. digm $\boxed{1}$ will not be very useful in prediction.

.11 On th. other hand use of th. method of α 69.10, will cause T.M. to use trigrams to solve "=" problems. It isn't clear that th. method of α 69.10 is meaningful, if applied to ngmsts, other than ngms. (e.g. digms, trigms, tetragms, etc)

.16 A possible way: Use ngmst. freqs. averaged over all possl. points in th. array, but use th. U's obtained with only tentative finality. For th. final judgement of th. U of an ~~ngmst~~ ngmst, use its average over only interrogation positions i.e. its "score" in answering questions actually asked. This sounds not too bad. Th. "all-over averaging" gives a kind of apriori U estimate.

It must be noted that th. questions ~~are~~ raised here, are essential Q's. This is because ultimately, T.M. will be shown many examples with no □'s in them, and then, occasionally ~~is~~ a □ will be given. I think that if T.M. uses sufficiently good ~~ngmsts~~ ngmsts, of suff. large n, it will be possl. for T.M. to get good predictions by averaging all data over all squares₂ in th. array. I think, however, that th. ultimate U assignments must give much weight to th. efficacy of th. ngms in ~~the~~ answering actual questions.

Th. reason why I think "good" and "large n" ngms. ~~can~~ can be used in this way: "good" ngms, will have in them, info about just where in th. α 82

Also, this enables one to evaluate str. and utpsts, to some extent, independently of th. presence of ~~other~~ good ngms that they did ~~not~~ have a hand in th. creation of.

If this solution to th. problem proves unsatisfactory, try to find th. equivalent problem in real language and science, and solve it. - Then translate back to Math T.M.

from $\alpha 77.28$

We still have 2 kinds of words:
 first kind.. Ngms that ~~consist~~ consist of ngms that are divided into 2 parts.
 (See $\alpha 79.25$ for description)

second kind Ngms that may be combined to form Ngms of th. above type. - e.g. one may concatenate 2 of them, and have R. first predict R. ~~and~~ second.

Words of R. first kind (after R. division is taken out) may be used as trial words of R. second kind. - And, of course words of R. sec. kind are normally combined ~~to~~ by str. xfmns. to give words of R. first kind.

Well, it looks like I am now ready to do a detailed description of ~~the~~ Math T.M. up to ~~it~~ ~~with~~ ~~carry~~ line).

When going past this point, remember, if there is a certain abs. you want T.M. to make, then if it is a really good abs., there will be many problems that are either unsolvable without it, or solvable with ~~the~~ much difficulty without it.

If these problems are gn. to T.M., and there exists a ~~sequence~~ sequence of abstractions that will lead T.M. to it, then T.M. will eventually get there. - So take heart, oh intrepid voyager into lands unknown!

All you have to do is really understand th. probs. you want T.M. to solve - know the abs. you want T.M. to use ~~as~~ as you would know thyself!

T.M.

reliable this implication is.

This seems to solve the problem. There is some Q.

about book keeping - and whether one should count certain

α 80.09

cases twice or not.

Actually, this implication "both ways" is unnec. - we do, however, have the exist. rule that if $a \rightarrow b$, then it is not unlikely that $b \rightarrow a$. In terms of

ngms: If $\frac{a \ b}{c \ d}$ is a hy U ngm. i.e. if

$\frac{a \ b}{c}$ usually implies $\frac{a \ b}{c \ d}$, then $\frac{a \ b}{c \ d}$, $\frac{a \ b}{c \ d}$

and $\frac{a \ b}{c \ d}$ are good trials for / ngms. ^{useful}

[SN] Incidentally, this ~~new~~ way of assigning U ^{of an ngm} on the basis of the ~~the~~ predicting power of that ngm alone, seems to be in the spirit of the D.G. soln. that simply chooses the "Best" ~~the~~ ngm.

25 An ngmst, then, is a set of objects, ~~that~~ ^(Engm) Each object may be divided into 2 parts. The ~~presence~~ predictor, and the predicted. In each case, the presence of 1) implies the presence of 2). The ~~strength~~ reliability of this prediction determines the "goodness" of an ngmst.

34

~~from~~ from α 73.12

Also, this tends to solve the "swamping out" problem:

i.e. how can one recognize ~~the presence~~ of a good ngm. in the presence of another ngm. that is as good or better?

The answer is, that often and perhaps always,

the ~~presence~~ U of an ngm. will be indep. of what other ngms one uses along with it. α 80

of them, is that one of these words may be related, in an implicative way, to another. ~~The~~ Several of these words may form a space-time config., so that if one observes one of them, then the rest are likely to follow.

It may, indeed, be, that ~~the~~ ^{a set of} words, of R. second type always form a config. and that this config. of words is a word of R. first type. Also these words of R. second type may be ~~seen~~ also of R. first type.

This suggests that a, say digm / is "good" if

a) 1 usually implies 10 and b) if 0 usually implies 10
 It would seem that there would be "one way" implications for some types of ngms - e.g. if a) were true, but not b).

What might happen: That one finds some ngms that are reasonable "good" - that R. implications work both ways or one way for a natural division of that ngm into parts. That ~~one~~ eventually, if one is to make much use of this ngm, one will make a statistical study of R. implication relations betw. its parts.

A "good" ngm, (a type 2 word), then, is one (with clearly defined parts?) in which R. implication relations betw. its parts, (as found by a statistical study) prove to be useful in prediction.

What this implies for Math T.M.:

One has all of these trial digrams. One tries them for implication both ways. The one's that are good at least one way, are kept, and indexed as to R. way in which R. implication ~~is~~ works, and how α 79

T.M.

e.g. say 3 completions are possl. :

1 α , 1 β , 1 γ . If 1 α and 1 β

have zero freq. , then Ph. freq. of 1 γ isn't very critical.

A possy: Just retain ~~ngms~~ ngms. of hyst freq. — discard those that are of low U.

Another possy: Use sh. ~~≠~~ zero freq. ngms and give them hy U. This will be opposite to ~~various~~ language, but ~~may~~ may turn out o.k. in sh. end.

An ngmst. can be characterized in Ph. foll. way :

say A α , B β and G γ are members of R. same ngmst. Then this means that if A, then A α is probable, that if B, then B β is probable, and if G, then G γ is probable. This conforms, I think, to Ph. idea of a ~~word~~ word corresponding to an ngmst.

28 This discn. seems to clear up a lot.

Perhaps I am confusing 2 or more kinds of "words". One e.g. is Ph. use of Ph. term ~~Automobil~~ car (\equiv automobile). One sees part of Ph. car, and one can extrapolate to Ph. rest.

Another kind of "word" is "cloudy day". A similar word is "rain". We have learned to correlated "cloudy day" and "rain".

The word "car" is a ^{space-time} config. of elements } ~~if~~ if most of them are present, Ph. rest will probably be present.

The "cloudy day", "rain" type words, are also space time configs. — they may be words of this type (and usually are), but sh. imp. characteristic 278

T.M.

~~The foregoing is in order for a 1 dimensional stochastic process.~~

~~For a 2 dim. case:~~

is partitioned into ~~n~~ n sub sets, ~~each~~

S_i ($i = 1/n$) Each subset S_i is divided into

S_i^1 and S_i^2 . If one starts ~~the~~ st. system at

any digit in S_i ($S_i \equiv S_i^1 \overset{\text{Booleant}}{\oplus} S_i^2$) one will, at $t = \infty$,

~~end up~~ end up in S_i^1 .

There is a good discn. of Markov chains in fuller.

p. 307 ff. on ~~sum~~ 318 starts discn. of accessibility of states to each other.

The foregoing is for 1 dim. case: For a 2 dim.

stoch. process:

The ~~this~~ is an interesting problem, I don't think it is relevant to what I need just now. What I want is, indeed, st. relation of digram freqs. to implication probabilities (correspond to transition ~~prob~~ probabilities), but this can be done without any of st. theorems on Markov chains!

Suppose I want to predict $\begin{matrix} 0 & 1 & 1 \\ 1 & 0 & 0 \end{matrix}$.

I have th. freqs. of

~~11, 10, 01, 00, 11, 10~~ 11, 10, 01, 00, 11, 10 etc.

[This problem was treated on 252, 53. - no really good ans. was obtained. - th. difficulty, was that th. digms of zero freq. proved to have hyst U. This is not in accord with common language, where word names are given to objects of hy freq. and hy U.]

If there are > 2 possl. completion digits (which is usually st. case) then th. ugn. of zero freq. is U, but th. other ugnms are also imp. - Not necessarily!

compute transition probabilities. ~~rel.~~

say D_{ij} is \mathbb{R} . (freq. of \mathbb{R} . i^{th} and j^{th} digits)

say $P_i = \sum_j D_{ij} = \sum_j D_{ji}$ is \mathbb{R} . long time/freq. of \mathbb{R} . i^{th} digit.

$$\sum_i P_i = \sum_i \sum_j D_{ij} = 1.$$

say T_{ij} is \mathbb{R} . probab. that if i at time t , then j will follow at $t+1$. $\sum_j T_{ij} = 1$.

Then $P_i T_{ij} = D_{ij}$

and $\sum_k D_{ki} T_{kj} = D_{ij}$

so $T_{ij} = \frac{D_{ij}}{\sum_k D_{ki}}$

but to go from T_{ij} to D_{ij} one must solve n^2 equs in n^2 unknowns. (There are n different digits)

say $\vec{T}^* = \left(\frac{\vec{D}}{T} \right)^\infty$ then

$$\vec{P} = \vec{T}^* \vec{P} \quad \text{also} \quad \vec{P} = \vec{T}^* \cdot (\text{any vector}),$$

T^* has \mathbb{R} . same eigenvectors as T . Providing any state can go to any other state

T^* has only eivalue of absolute value 1. They come in conjugate pairs, since T^* and T have only real components. If T has complex eivals. of abs. value 1, then $\sqrt[n]{T} \rightarrow T^*$ does not converge. It oscillates periodically, or "almost periodically".

If T^∞ does converge then T has only eivals of value 1, if their eiv is of abs. value 1.

If T has 2 or more eivals. of value 1, then T is degenerate, and any linear comb. of ~~these~~ \mathbb{R} . corresp. eivacs. is an eivac. of $\lambda = 1$.

There is some Q. about just how one gets th. ~~comp~~ digits partitioned into sets that cannot go to one another. What must happen, is that the set of digits $d_i (i=1/n)$