

Perhaps Tikoff section: ~~Probability~~ Conditional Probability Distributions.

spec. (285.22) (292.00-23)

Start out by dirbug d funcs, s-funcs:

Prob theory is only one way to express D-funcs. ~~Equation~~

(well no: say $x=f(t), y=f(t)$ implies $y(x)$ but can be many valued, not a function.

For s-funcs there are ~~many~~ several ways to express them;

1)

AZ (at approx) ~~Does~~ can assign prob to any subset of k variables. How dist from Coast pd? $P(x,y)$ v.s. $P(y|x)$? $P(\vec{x} \rightarrow y)$

SN AZ can assign prob to any subset of k vars, k scalars

$(y=f(x))$ assigns prob to $P(x)$. This is a ~~stochastic~~ pd prob is fixed by

choice of parameters is formalizing of P_c assignment. I want to be able to try various poss. pd's on all $y=f(x)$'s. By adding just i more input, (essentially "R") we get a (perhaps) universal s-func. Consider \vec{x} -set of k scalars. $y=f_i(\vec{x})$. We know in AZ,

\vec{x} is fixed on all such f_i . Here, in $y=f_i(x,R)$ t function depends on R and

t . P_c of y depends on t . (AZ assigned) P_c of f_i^2 mult by $2^{-|R|}$.

In AZ, the local R (input, any input) must be a prefix set, so. Then can tell where

that input ends. AZ stop symbol is a common way to end — but ~~AZ~~ does ~~it~~ really give to pd that want?

SN Looking at §2.1 "immune problems" (in FDSA report)

It starts out looking for a L such coin. — Also a section §3 does same.

Try P is: On p 14, after eq (7) $\sum_i P(h) P(1-h)$

"When we find a suitable set size n ."

This O^i becomes part of the updated GCPD (General Conditional Probability Distribution) and can be used to guide L search.

Which is, indeed, possible to run an L search in this way, we will describe a search technique that ~~seems~~ seems to be much faster

~~rather~~ than L search

Continuation: Given a problem, (G, n, a) ...

Changes needed in §3.1 (02 prob) [change Name of §2.1]

We want to find O^i 's such that

$$\sum_{j,l} P(O^i | G^{j,l} | \tilde{G}_j, t_j, F_l)$$

is as large as possible:

(\tilde{G}_j, t_j) describes the j^{th} optimization problem, to find, in time t_j ,

an x such that $\tilde{G}_j(x)$ is as large as possible.

$O^i(G^{j,l} | \tilde{G}_j, t_j, F_l)$ is the probability density ~~then~~ $P(O^i | F_l)$ ~~of~~ F_l ~~of~~ O^i ~~of~~ F_l

SN See how an "updatable" now and! Spending $\frac{1}{2}$ time on updating is $\frac{1}{2}$ time on searching is perhaps not so slow in WON! — looks like P^i 's!

14.20

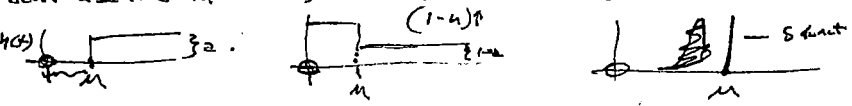
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I want to discuss how s-functions are to be realized in QATM!
 Have new section on "stochastic functions"
 Do give examples of 2-3 param $h'(t)$ curves as a way to get a s-function
 from a 3-output d-function. In section A very much ~~very much~~ improved update technology,
~~refer~~ refer to this section for examples.

SP:

Since I really don't use α or β in $h'(t)$ of t : update function (S 2.1 output)
 Why not use $\alpha = 0$ 

would I get ϕ pc for some cases?

I would! For h' I would almost always get ϕ except for when $h' = h$ at
 which point doesn't matter.

So it looks like I will have to model α , even tho I don't care about its value - it does
 contribute much to "Goodness of fit".

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So while Eq(1) is correct, Appendix A tells how to get types
 of d-functions, not to ~~functions of~~ $q(t)$.

We can start our TSD w. d-functions (MTHS TSD), but eventually, we will want
 to work on s-functions. At one time I thought it was necessary to get h' to do
 prod'n. of s-functions - ~~but now I know~~ so it could do ordering. But recently,
 its become clear that update can be done w. pure d-function. I still don't have a clear
 idea how to construct vector outputs of functions, using economical codes, so all vector computation
 share same costs. → 28800

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N.B. GHTI assumes (not $\frac{P}{C}$ ordering is optimum) assumes β pc's are uncorrelated
 i.e. just knowing one trial fails, tells nothing about other pc's in variant. ~~The update~~

re optimization of P_{PC} \approx $(TD)^i(QA)$ takes advantage of correlations between trials.

In WOV/GHTI, t -ordering of h' 's is by max values of

$t_{PC}(\theta) = \frac{h'(t)}{t}$. Each h' has a t value so P_{PC} is Max.

to largest θ for a trial, θ is ordering index



If we then determine trial to ~~then~~ extend out to that max θ pc's and stop, then
 then $\frac{P}{C}$ obtained is exactly right for t : **GHTI!** The only part that causes it
 to fail, is P_{PC} correlates of t pc's set to rands.

We say "suppose" because the Gambling house Proverum assumes P_{PC} ~~success~~ success probabilities
 of P_{PC} trials are uncorrelated - i.e. when one trial fails, ~~the~~ P_{PC} ~~success~~ success probabilities
 of other ~~trials~~ trials do not change. In our update scheme, we will take advantage of P_{PC} so
 correlations ~~to speed up your search~~ to speed up your search.