

0: (Spec 297.50): If the ratio of root space to coil space densities is constant (w. change of no. of coils) then the root space ratio is $\frac{\pi}{2} r_2 \approx 3.57$, (from the ratio of the constant coil space as well — which seems unlikely, but may be true).
 I think I ran into difficulty trying to compute the vol. of root space. I took random pts in root space & took mean of the Jacobian. Unlucky, & while this would eventually converge, it converges very slowly. This is because there are many (I don't remember how many) very small regions of very by Jacobian — 7. empirical result was that the variance of ("estimate up to now" seemed to $\propto t^{-0.45}$).
 So I'd never (or for long time) ever find true mean w. any precision.

$$\frac{\pi}{2} r_2 = 3.57079$$

$$\sqrt{1.8996} \approx 1.378$$

$$\frac{\pi}{2} = 1.5708$$

0: Anyway, it may well be that for large n (no. coils) the results for coil space are quite different from results for root space.
 Also, it's not clear ^{as to why} Jacobians should not be particularly large in any regions — even on edges of root space!

15: Anyway, this "Gauss problem" is interesting & important. I should study it in a problem (Group; SF/IV Battery averaged 297.38, etc).

0: I think the whole pitch of Non-parametric Statistics was to take params on variables ^{in which} people agreed on a prior. I think there may be many params ~~for~~ for which people agree on a prior. These may be "built-in" to humans or into the physical laws of our universe.
 → I think that people normally are unaware of the arbitrariness of the assumed prior.

26: I am pretty much unable to work on these (15) problems I can ever before. I'd idea of the Scientist continually updating his prior is important.

27: **ON THE OTHER HAND!** In 50178 I had this idea
 $b = \text{no. of coils}$
 $n = \text{no. of data pts}$
 $b \text{ cost} \approx n \log n$
 A log is a no. characteristic of the model. It may be that log for must linear & non-linear cases w. large n . So cost $\propto n \log n$. But the interest in the problem was just that the cost per new coil was just log — same for every coil!

30: For me the problem, coil space is quite small dimension.
 It would seem that the subset of models is quite small and simple! But the only thing one needs to derive the model is the set of coils — w. costs that depend on precision needed — ∴ dependent on "n" is simple way!

Hum! The Apriori of Ω^2 need not be simple. — it is indep of no. of coils, so we compare the cost of the models for different no. of coils. The smaller Ω^2 is, the less into we need in each predn. Say we have a "Gauss area" of Ω ,

NIPS

HMC

"Because the constraints on f roots, I'd guess at f d.f. being uniform in root space —

But f : Root \mathcal{Q} (for correct answer) is How to use "previous Experience"?

However, because of n. val \rightarrow complex roots being treated differently (2.96.20-28), it's not entirely clear as to what a "uniform spread in root space" means!

Anyway, the solution to the Gompertz problem is the "same" as Soln. to "Griv":

In the case of Griv, we have data of past known universe, sometimes using the "GRV" (data) sometimes not. R is given via a pc of f : symbol "drive" as used in coding/proc.

R is pc extrapolates into future & is updated whenever each time we use other P data.

In a similar way, we can "parameterize" the selection of "f. root of cost".

For small n , use only 1 param. — say pc is same for "1 more param" under of n . I guess there are many other parameterizations poss. — uniformity in

root or cost space is an aspect: Also distance of set of roots from "stability" manifold boundary in the loc of interest.

Also, to add a new coil! look at d.f. of previous coils, R 's could give a reasonable input for f . way one! Say, use Gaussian d.f. so M, σ would characterize it.

If coils tend to have other d.f. (not Gaussian) use P d.f. w. 2 params — maybe more! (instead of e^{-x^2} use $e^{-|x|^k}$ to get tails of varying "fatness")

What Corpus should be used to parameterize our "Ecom probly"? In the case of SM, use any security that seem "n" to those being predicted.

A not bad corpus would seem to be simply the major dividend part of the present stock.

Too far back to use f . have coils over entire corpus, but ok to

use to parameterize choice of next coil! (Say use X window — on

any other window of sequentially finite width). — (for a given window width, R , we could only have a certain maximum no. of coils!) **WOOPS!**

Actually, there is interaction betw. W (window width) and no. of usable coils, so how would the way \geq grip \downarrow in no. of coils is a way to tell what W is (to some extent, but not probly not exactly).

.30 — not at all clear! This might be possible. — for larger W , only 3 coils possible, \rightarrow not obviously good for such a large corpus. For smaller W , we can use more coils, but \rightarrow not clear periodicity of relevant data. **not obviously correct!**

It might be interesting to calculate descriptors for a few cases

contrast root space w. coil space results,

38 $\sim SN$ In that STERN, Baseball battery averages, case, I had results critically dependent on behavior of \geq prop for ≥ 10000 or something. "behavior when" characteristic. How should they be treated?

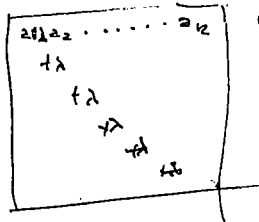
Spec 303.00

N.Y.S

HMC (How Many Coils?)

One could express each set of k successive z 's as a fixed matrix times previous set of k successive z 's. So we have a $k \times k$ periodic matrix - which converges if all of its

$$|\rho(V_i)| < 1.$$



what's Def of this Matrix? looks like $(z_i + \lambda) \lambda^{k-1}$.

Getting this $k \times k$ matrix in. too hard. fast way to decide convergence because our root is

successively squaring the matrix! This amounts to doubling to jump each time -

Even tho matrix mult takes k^3 steps, the squaring effect will always win! So many matrices don't get > 1000 , say, necessary divergence.

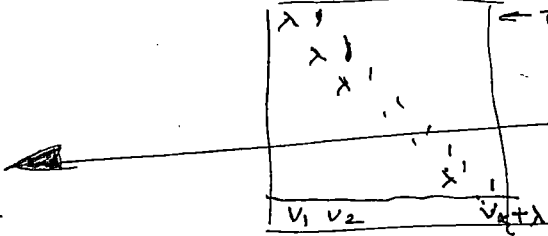
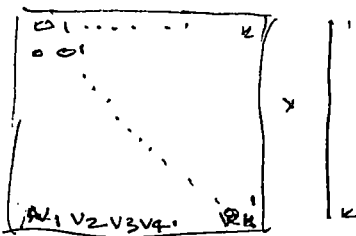
Another trick I remember! For a $k \times k$ matrix, the first k powers have all info our needs for hyper powers (- Is this true?)

To square a $k \times k$ matrix k^3 mults on k^2 dot products, so doing it n times is k^2 dot products,

equivalent to $k^2 n$ jumps. which could normally take $k^2 n$ dot products. So to find λ $\frac{k^2 n^2}{k^2 n} = \frac{n^2}{n}$

So for large k , it's when immediately efficient say $k=10, n=10 \frac{2^{10}}{10 \cdot 10} = 10$. for $k=5$ $n=10$ it's 20.

To correct matrix for $\vec{z}(t) \rightarrow \vec{z}(t+n)$ is



← This is the matrix.

insignificant! ~~...~~

Remove the λ 's

to get a finite sum
can be repeatedly squared

To write the power, I have to know Max/Min of each coil E is data, $V_n = 1$.

$$n=1 \quad (x-\lambda_1) \quad \blacksquare \quad V_1=1 \quad V_0=\pm 1$$

$$(x-\lambda_1)(x-\lambda_2) = x^2 - \frac{V_2}{V_1}(x+\lambda_2)x + \frac{V_0}{V_0} \lambda_1 \lambda_2.$$

$$V_0 = \pm 1 \quad V_1 = \pm 2 \quad V_2 = 1$$

but these λ units are restricted.

$$\left. \begin{matrix} V_0 = +1 & V_1 = +2 \\ V_0 = +1 & V_1 = -2 \end{matrix} \right\} \text{two best poss.}$$

$$V_0 = -1 \quad V_1 = 0$$

I wanted all of this out in previous work on HMC. perhaps try to find it.

For $k=2, 3$ maybe I can do it by hand!

For about $\frac{1}{2}$ off. coils, e. fall \pm range will often be poss. for $\frac{1}{2}$ off. coils, may be only $\frac{1}{2}$ is possible so ϵ . less volume of stable pts will be $\sim \frac{1}{2^k}$ for $k=10$ $n=10$ $2^{-5} = \frac{1}{32}$. - why

Also note that many of the divergent pts will diverge rapidly, so little time spent there.

BUG??! ~~...~~ If I used $k=3$, coils = 1, 2, 1 The roots are $\frac{-1}{2} \pm \frac{\sqrt{3}}{2}i$ and + seq. should not

diverge! But since $k=3$ is ≥ 1 , it would seem that iteration would always \uparrow values for it would diverge!

Start w. 1 $\rightarrow 1, 3, 6, 12,$

$$1 + 2A^2 + A^4 = \dots$$

$$A(n+1) = A(n) + 2A(n-1) + A(n-2)$$

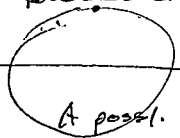
$$A^{n+1} = A^n + 2A^{n-1} + A^{n-2}$$

$$A^3 = A^2 + 2A + 1$$

\Rightarrow I may have been wrong on char. func wrong! check this out \rightarrow

Nies

00 : On my work on HMC How many coeffs problem: I assumed that the roots were real, & roots had to be in conj. pairs / Is this true?
 I guess that $x^3 + 1 = 0$ has 3 roots: 2 conj pairs, 1 real.



$$(x-i)(x+i) = x^2 - 2ix - 1$$

A possibl. way to get density of ~~solutions~~ "states" forms in conf space:

pick random pts in finite space & think each conf is uniform betw its limits.
 for each pick, we can either solve a poly. eqn or do an integration of coeffs (zero noise)
 for a long distance we see it flows up (like Poincaré brat sets). The edges may be fractal, but the structure is volume of a legal region may not be hard to estimate.
 for k coeffs, do we start w a random vector?

10 Given an analytic expression with y^k iteration: perhaps there is an easier way to compute it.
 Perhaps RMX does vector dot products

13 $a_1 \dots a_k : z_{k+1} = \sum_{i=1}^k z_i$; $z_1 v_1 + z_2 v_2 \dots z_k v_k$
 $z_{k+2} = z_1 v_1 + z_2 v_2 \dots z_{k+1} v_k = z_2 v$

15 Its a circular prob: we only keep the k values of z_i & and overwrite our values of some of v with dot product.

There may be a fractal attractor w/ the starting vector z_0, \dots, z_k . I can try this empirically;

Say I have a fixed \vec{v} : I try various initial \vec{z} values & see if converges or diverges.

It might be possible to figure out what will happen, from simple analysis — i.e. Why are roots outside unit circle cause divergence.

22 From an analysis of .13 - .15 I think the value at any pts. is a linear function of values of $z_1 \dots z_k$. So we always have k coeffs & correspondingly, k roots, so we get k roots at each successive iteration, and a simple function of the previous set of roots!

[if .22 is true, I can do an analysis of $\vec{z} = 1, 0, 0, \dots$; $\vec{z} = 0, 1, 0, \dots$; $\vec{z} = 0, 0, 1, 0, \dots$]

I do k runs. If any of them diverge, then I assume divergence.

try $\vec{z} = 1, 0, 0, \dots$; $z_{k+1} = v_1$; $z_{k+2} = v_1 \cdot v_k$; $z_{k+3} = v_1 \cdot v_{k-1} + v_1 \cdot v_k \cdot v_k$; $z_{k+4} = v_{k-2} \cdot v_1 + v_k \cdot v_{k-1} \cdot v_1 + v_1 \cdot v_{k-1} \cdot v_k + v_1 \cdot v_k^2$

4 ; v_k ; $v_{k-1} + v_k^2$; $v_{k-2} + v_k v_{k-1} + v_{k-1} v_k + v_k^3$; $v_{k-3} + v_k v_{k-2} + v_k v_{k-1} v_k + v_k^3$; $v_{k-2} + v_k v_{k-1} + v_k^3$; $v_{k-1} + v_k^3$; v_k^3

(looks like Pascal's Δ) ; $1, 1, 1, 2, 1, \dots$
 May not!

$v_k v_{k-2} + v_{k-1}^2 + v_k v_{k-1} v_k + v_{k-2} v_k^2 + v_{k-1} v_k^2 + v_k^4$
 $v_{k-1}^2 + 4 v_{k-1} v_k$

So our v 's is 1 instead of $k-1$ with "1". No Pascal's on line .40

.40 With suitable complex notation, I may be able to write a closed form in some special form of form.

NIPS

ALP General. "IF & ONLY IF" for ALP of Conv. Pnm.

50

- 1) On to "Inverse Conv. Pnm." (1) IS ALP to any soln? (2) If μ, t seq, generator is known w/ $P = P(\mu)$; is KL distance of soln. w/ $P_{opt} - P(\mu)$ to best that can be done? **BEST**
- (3) If we allow larger errors, say error is greater than $-P(\mu)$ by factor, k or by additive amt. α .
- (4) I'm not talking about "asymptotic" results for infinite seq length, but short finite seqs.
- (5) How large is the deriv. space of pnm's that give seqs a distance d , from P ? Pnm's generated by μ ? This could be very useful!
- (6) Perhaps Get Marcus to work on it!
- (7) A Guess; Prob for large SSZ , (i.e. Big Coins)

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Some Quick notes on Univ. d.f. of Continuous functions of a real (or complex) variable:

We can descr. Continuous functions w/ a finite no. of bits.

One big uncertainty is in the prep of continuous params.

Alan suggested using floating pt. notation! This does give a finite sized space (unlike low ends). Perhaps use Very large word. Since one integrates over all derivs of functions that are const w/ P_t data, using ability to size of word may become irrelevant for word sizes \rightarrow a certain amt.

HR: data, using ability to size of word may become irrelevant for word sizes \rightarrow a certain amt.

HR, my guess is that if the space of interest gets very large we have to pay for it — by the prep of smaller values decreasing.

Another tack: Somehow by the context of the problem, one has some idea as to

Size of the space.

\rightarrow Re floating pt. notation! My intuition is that one should use ordinary floating pt.

That one initially tells what the upper bound of the space is like, and from there wants precision, you get it by integration over many points. !! Values.

Data: \pm (upper bnd) ones as a prior Uniform dist.

26

One apparent funny diffy: Linear regression: when one introduces 1 more coin, the rcost depends much on one's upper bound for the coin.

A poss "Soln" to 26: rcost of one more coin is always "d". "d" is found by

trying various values on one's coins & selecting a value that "works best".

HR: The value of α may depend critically on the f. coins; on how many coins are needed

to predict!

For many variables, small values are always of high exp.

For linear Regression, one can assume "stability" so roots of char eq. are within $(unit + \epsilon)$ circle ("+" to allow for some "instability"). For each coin, one can set max and min values. A root guide would be uniform spaced best. Must Max for all coins.