



## Structure, Probability, Entropy

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September 22, 2014

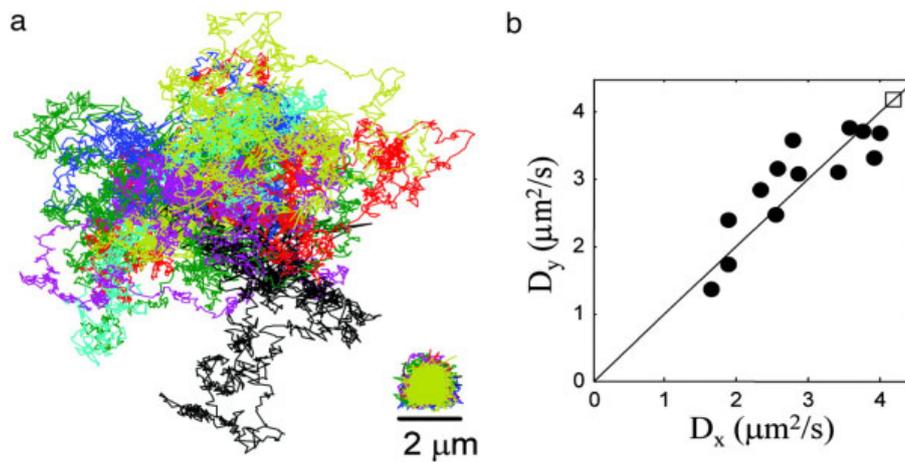
Rituparna, a king of Ayodhya said  
 $\approx 5\,000$  years ago:

*I of dice possess the science and  
in numbers thus am skilled.*

More recently,  $\approx 150$  years ago,  
James Clerk Maxwell said:

*The true logic of this world is the  
calculus of probabilities.*

*All the mathematical sciences are  
founded on relations between phys-  
ical laws and laws of numbers.*



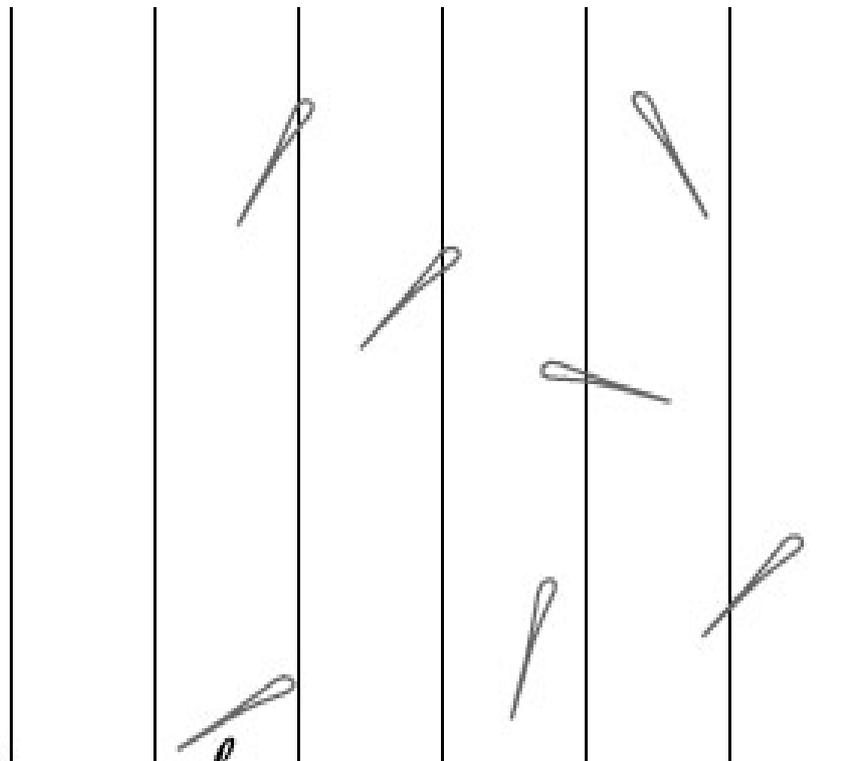
*... small compound bodies...  
are set in perpetual motion  
by the impact of invisible blows...*

*The movement mounts up  
from the atoms  
and gradually emerges  
to the level of our senses.*

ARTICULATED BY...

TITUS LUCRETIUS IN 50 BCE  
AND EXPRESSED IN NUMBERS BY

THIELE (1880), BACHELIER (1900),  
EINSTEIN (1905), SMOLUCHOWSKI  
(1906), WIENER (1923).



## SYMMETRY IN RANDOMNESS.

Most (all?) of the classical mathematical probability theory is grounded on *(quasi)invariant Haar(-like) measures*.

(The year 2000 was landmarked by the discovery of *conformally in-*

*variant* probability measures in spaces of curves in Riemann surfaces parametrized by increments of *Brownian's processes* via the *Schram-Loewner evolution equation*.)

The canonized formalisation of probability, inspired by Buffon's needle (1733) and implemented by Kolmogorov (1933) reads:

*Any kind of randomness in the world can be represented (modeled) geometrically by a subdomain  $Y$  in the unit square  $\blacksquare$  in the plane. You drop a points to  $\blacksquare$ , you count hitting  $Y$  for an **event** and define the probability of this event as  $area(Y)$ .*

(This *set theoretic* frame conceptually is similar to André Weil's *uni-*

*versal domains* from his 1946 book *Foundations of Algebraic Geometry*.)

If there is not enough symmetry and one can not *postulate* equiprobability (and/or something of this kind such as *independence*) of certain "events", then the advance of the classical calculus stalls, be it mathematics, physics, biology, linguistic or gambling.

#### ON RANDOMNESS IN LANGUGES.

*The notion of a probability of a sentence is an entirely useless one, under any interpretation of this term [that you find in 20th century textbooks"].*

Naum Chomsky.

An essential problem with prob-

ability is a mathematical definition of "events" the probabilities of which are being measured.

A particular path to follow is suggested by Boltzmann's way of thinking about statistical mechanics – his ideas invite a use of

*non-standard analysis*

and of Grothendieck's style

*category theoretic* language.

Also, the idea of probability in *languages* and in *mathematics of learning* deviates from Kolmogorov-Buffon ■.

FIVE ALTERNATIVE AVENUES  
FOR IDEAS OF PROBABILITY AND  
ENTROPY.

1. *Entropy via Grothendieck Semi-group.*
2. *Probability spaces as covariant functors*
3. *Large deviations and Non-Standard analysis for classical and quantum entropies.*
4. *Linearized Measures, Probabilities and Entropies.*
5. *Combinatorial Probability with Limited Symmetries.*

### ”NAIVE PHYSICIST’S” ENTROPY

*... pure thought can grasp reality... .*

Albert Einstein.

*...exceedingly difficult task of our time is to work on the construction of a new idea of reality.... .*

Wolfgang Pauli.

A *system*  $\mathcal{S}$  is an infinite ensemble of infinitely small *mutually equal* "states". The logarithm of the properly normalised number of these states is (*mean statistical Boltzmann*) *entropy* of  $\mathcal{S}$ .

The "space of states" of  $\mathcal{S}$  is NOT a mathematician's "set", it is "something" that depends on a class of *mutually equivalent* imaginary experimental protocols.

*Detectors of Physical States: Finite Measure Spaces.* A finite measure space  $P = \{p\}$  is a finite set of "atoms" with a positive function denoted  $p \mapsto |p| > 0$ , thought of as  $|p| = \text{mass}(p)$ .

$|P| = \sum_p |p|$ : the (total) *mass* of  $P$ .

If  $|P| = 1$ , then  $P$  is called a *prob-*

*ability space.*

*Reductions and  $\mathcal{P}$ .* A map  $P \xrightarrow{f} Q$  is a *reduction* if the  $q$ -fibers  $P_q = f^{-1}(q) \subset P$  satisfy  $|P_q| = |q|$  for all  $q \in Q$ .

(Think of  $Q$  as a "plate with windows" through which you "observe"  $P$ . What you see of the states of  $P$  is what "filters" through the windows of  $Q$ .)

Finite measure spaces  $P$  and reductions make a nice category  $\mathcal{P}$ . All morphisms in this category are epimorphisms,  $\mathcal{P}$  looks very much as a partially ordered set (with  $P \succ Q$  corresponding to reductions  $f : P \rightarrow Q$  and few, if any, reductions between given  $P$  and  $Q$ ); but it is advantages to treat  $\mathcal{P}$  as a general

category.

*Why Category?* There is a subtle but significant conceptual difference between writing  $P > Q$  and  $P \xrightarrow{f} Q$ . Physically speaking, there is no a priori given "attachment" of  $Q$  to  $P$ , an abstract ">" is meaningless, it must be implemented by a particular operation  $f$ . (If one keeps track of "protocol of attaching  $Q$  to  $P$ ", one arrives at the concept of *2-category*.)

The  $f$ -notation, besides being more precise, is also more flexible. For example one may write  $ent(f)$  but not  $ent(>)$  with no  $P$  and  $Q$  in the notation.

*Grothendieck Semigroup*  $Gr(\mathcal{P})$ ,  
*Bernoulli isomorphism*  $Gr(\mathcal{P}) =$

$[1, \infty)^\times$  and Entropy.

*Superadditivity of Entropy.*

Functorial representation of infinite probability spaces  $X$  by sets of finite partitions of  $X$ , that are sets  $\text{mor}(X \rightarrow P)$ , for all  $P \in \mathcal{P}$  and defining Kolmogorov's dynamical entropy in these terms.

*Fisher metric and von Neumann's Unitarization of Entropy.*

Hessian  $h = \text{Hess}(e)$ ,  $e = e(p) = \sum_{i \in I} p_i \log p_i$ , on the simplex  $\Delta(I)$  is a Riemannian metric on  $\Delta(I)$  where the *real moment* map  $M_{\mathbb{R}} : \{x_i\} \rightarrow \{p_i = x_i^2\}$  is, up to  $1/4$ -factor, an *isometry* from the positive "quadrant" of the unit Euclidean sphere onto  $(\Delta(I), h)$ .

$P$ : positive quadratic forms on the Euclidean space  $\mathbb{R}^n$ ,

$\Sigma$ : orthonormal frames  $\Sigma = (s_1, \dots, s_n)$ ,

$\underline{P}(\Sigma) = (\underline{p}_1, \dots, \underline{p}_n)$ ,  $\underline{p}_i = P(s_i)$ ,

$$\text{ent}_{VN}(P) = \text{ent}(P) = \inf_{\Sigma} \text{ent}(\underline{P}(\Sigma)).$$

LANFORD-ROBINSON, 1968. *The function  $P \mapsto \text{ent}(P)$  is concave on the space of density states:*

$$\text{ent}\left(\frac{P_1 + P_2}{2}\right) \geq \frac{\text{ent}(P_1) + \text{ent}(P_2)}{2}.$$

Indeed, the classical entropy is a concave function on the simplex of probability measures on the set  $I$ , that is  $\{p_i\} \subset \mathbb{R}_+^I$ ,  $\sum_i p_i = 1$ , and infima of families of concave functions are concave.

Spectral definition/theorem:

$$\text{ent}_{VN}(P) = \text{ent}_{Shan}(\text{spec}((P))).$$

*Symmetrization as Reduction and Quantum Superadditivity.*

LIEB-RUSKAI, 1973.

*H and G: compact groups of unitary transformations of a finite dimensional Hilbert space S*

*P a state (positive semidefinite Hermitian form) on S.*

*If the actions of H and G commute,*

*then the von Neumann entropies of the G- and H-averages of P satisfy*

$$\text{ent}(G * (H * P)) - \text{ent}(G * P) \leq \text{ent}(H * P) - \text{ent}(P).$$

*On Algebraic Inequalities.* Besides "unitarization" some Shannon inequalities admit linearization, where the first non-trivial instance of this is the following

*linearized Loomis-Whitney 3D-isoperimetric inequality* for ranks of bilinear forms associated with a 4-linear form  $\Phi = \Phi(s_1, s_2, s_3, s_4)$  where we denote  $|\dots| = \text{rank}(\dots)$ :

$$|\Phi(s_1, s_2 \otimes s_3 \otimes s_4)|^2 \leq$$

$$|\Phi(s_1 \otimes s_2, s_3 \otimes s_4)| \cdot |\Phi(s_1 \otimes s_3, s_2 \otimes s_4)| \cdot$$

$$|\Phi(s_1 \otimes s_4, s_2 \otimes s_3)|$$

*Measures defined via Cohomology and Parametric Packing Prob-*

*lem.*

Entropy serves for the study of "ensembles"  $\mathcal{A} = \mathcal{A}(X)$  of (finitely or infinitely many) particles in a space  $X$ , e.g. in the Euclidean 3-space by

$$U \mapsto \text{ent}_U(\mathcal{A}) = \text{ent}(\mathcal{A}|_U), \quad U \subset X,$$

that assigns the *entropies* of the *U-reductions*  $\mathcal{A}|_U$  of  $\mathcal{A}$ , to all bounded open subsets  $U \subset X$ . In the physicists' parlance, this entropy is

*"the logarithm of the number of the states of  $\mathcal{E}$*

*that are effectively observable from  $U$ "*,

We want to replace *"effectively observable number of states"* by

*"the number of effective degrees of freedom of ensembles of moving*



balls”.

- *Classical (Non-parametric) Sphere Packings.*

- *Homotopy and Cohomotopy Energy Spectra.*

- *Homotopy Dimension, Cell Numbers and Cohomology Valued Measures.*

- *Infinite Packings and Equivariant Topology of Infinite Dimensional*

- *Spaces Acted upon by Non-compact Groups.*

- *Bi-Parametric Pairing between Spaces of Packings and Spaces of Cycles.*
- *Non-spherical Packings, Spaces of Partitions and Bounds on Waists.*
- *Symplecting Packings.*

*Graded Ranks, Poincaré Polynomials and Ideal Valued Measures.*

The images as well as kernels of (co)homology homomorphisms that are induced by continuous maps are *graded* Abelian groups and their ranks are properly represented not by individual numbers but by *Poincaré polynomials*.

The set function  $U \mapsto \text{Poincaré}_U$  that assigns Poincaré polynomials

to subsets  $U \subset A$ , (e.g.  $U = A_r$ ) has some measure-like properties that become more pronounced for the set function

$$A \supset U \mapsto \mu(U) \subset H^*(A; \Pi),$$

$$\mu(U) = \text{Ker} (H^*(A; \Pi) \rightarrow H^*(A \setminus U; \Pi)),$$

where  $\Pi$  is an Abelian (homology coefficient) group, e.g. a field  $\mathbb{F}$ .

*$\mu(U)$  is additive for the sum-of-subsets in  $H^*(A; \Pi)$  and super-multiplicative for the  $\sim$ -product of ideals in the case  $\Pi$  is a commutative ring:*

$$\mu(U_1 \cup U_2) = \mu(U_1) \oplus \mu(U_2)$$

for *disjoint* open subsets  $U_1$  and  $U_2$  in  $A$ , and

$$\mu(U_1 \cap U_2) \supset \mu(U_1) \sim \mu(U_2)$$

for all open  $U_1, U_2 \subset A$