Interaction between Truth, Belief and Knowledge as the key to entropy measures of statistical physics

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What is entropy?

Entropy operates in the interface $God \leftrightarrow man \text{ or } Nature \leftrightarrow man \text{ or}$ system \leftrightarrow physicist studying or actually observing.

Entropy assigns to the true state x of a system the complexity of the system in that state as seen by man. More precisely, the entropy of x is given by

$$S(x) = \min_{y} \Phi(x, y)$$
 where

 $\Phi(x, y) =$ description cost (or effort) needed to describe the system when the truth is x and your belief is y.

So: Entropy is minimal description effort

Note: Focus on Φ with S as a derived concept. Another important derived concept is redundancy (divergence) D defined by $D(x, y) = \Phi(x, y) - S(x)$.

Problem: $\Phi = ?$

Answer requires knowledge about nature, the world we operate in, and input from man who should design efficient description strategies via descriptors (see later).



truth x

belief y knowledge z

Regarding z: outcome from extended observations, synthesis of extended experience, knowledge.

x and y will be distributions over some alphabet, A: $x = (x_i)_{i \in \mathbb{A}}, y = (y_i)_{i \in \mathbb{A}}$ and similarly for y.

Possible worlds

Key postulate: There is an interaction between x, y, zof the form $z = \Pi(x, y)$. Π characterizes the world. Some possibilities:

 $\Pi(x,y) = x: \text{ classical world},$ $\Pi(x,y) = y: \text{ black hole},$ $\Pi(x,y) = qx + (1-q)y: \text{ mixtures}.$

Assume that interaction acts locally via interactor π : $z = \Pi(x, y)$ with $z_i = \pi(x_i, y_i)$. Always assume that $\pi(s, t) = s$ if t = s (soundness).

Examples: $\pi_q(s,t) = qs + (1-q)t$ corresponding to classical world (q = 1), black hole (q = 0) or mixtures.

There are many other possible interactors, but:

under consistency the π_q 's are the only ones.

Consistency:

 $\sum z_i = 1$, for x, y probability distributions, $z = \Pi(x, y)$.

What can the physicist do?

We assume that the world is known to the physicist through the interactor π .

Every observation entails a cost (or effort). The cost depends on the event being observed. An event with high probability has little cost. Thus define a descriptor κ to be a decreasing function on [0, 1] with $\kappa(1) = 0$. Also insist that κ satisfies the normalization condition $\kappa'(1) = -1$, corresponding to a choice of unit.

Interpretation: $\kappa(t)$ is the effort associated with observations from an event which you believe occurs with probability t (equal to some y_i , say).

Total description cost, $\Phi(x, y)$, when truth is x and belief is y is given by

$$\Phi(x,y) = \sum_{i \in \mathbb{A}} z_i \kappa(y_i) = \sum_{i \in \mathbb{A}} \pi(x_i, y_i) \kappa(y_i).$$

The perfect match principle (PMP)

To design κ , hence Φ , apply perfect match principle:

PMP: Φ is the smallest when *belief matches truth*: $\Phi(x, y) \ge \Phi(x, x)$, i.e. $\Phi(x, y) \ge S(x)$.

Theorem Given π , the only *possible* descriptor satisfying PMP is the solution to $\frac{\partial \pi}{\partial t}(t,t)\kappa(t) + t\kappa'(t) = -1 ; \kappa(1) = 0.$ If $\pi = \pi_q$, this becomes $(1-q)\kappa(t) + t\kappa'(t) = -1 ; \kappa(1) = 0$ with solution $\kappa = \kappa_q$ given by $\kappa_q(t) = \ln_q \frac{1}{t}$ with *q*-logarithm given by $\ln_q x = \begin{cases} \ln x \text{ if } q = 1, \\ \frac{1}{1-q} (x^{1-q} - 1) \text{ if } q \neq 1. \end{cases}$ However, this only satisfies PMP if $q \ge 0.$

Comments on the proof

The differential equation comes up via standard variational principles (introduce Lagrange multipliers!).

Re π_q , κ_q : Failure of PMP for q < 0: simple direct counter examples with a 3-element alphabet.

Validity of PMP for $q \ge 0$: via PFI, the pointwise fundamental inequality, which states that $d(s,t) \ge 0$ where

$$d(s,t) = \left(\pi(s,t)\kappa(t) + t\right) - \left(s\kappa(s) + s\right).$$

Indeed,

$$D(x,y) = \sum d(x_i, y_i)$$

and PMP really says that $D(x, y) \ge 0$ (with "=" iff y = x). Writing up the formulas with π_q and κ_q in place of π and κ one finds that a simple application of the geometric/arithmetic mean inequality leads to the desired inequality $d(s, t) \ge 0$.

Comments on result:

• Strong argument in favour of the view that only possible entropy measures of statistical physics are those in the Tsallis family: $S_q(x) = \sum x_i \kappa_q(x_i)$

 $=\left(\sum x_i^q - 1\right)/(1-q)$ for $q \ge 0$. Classical value for q = 1, and for black hole, S(x) = n - 1, number of degrees of freedom (n = size of alphabet).

• Φ more important than S, gives more, e.g. shows which are the feasable preparations, viz. those of the form $\mathcal{P} = \{x | \Phi(x, y_0) = c\}$, and Φ also assists greatly in finding equilibrium distributions (via Max-Ent, and even without introducing Lagrange multipliers – instead a natural, therefore better, intrinsic approach is used). (more on next slides)

• The roles of God and man (nature and man, system and man) clearly separated!

• Formulas are mathematically attractive, e.g. lines up with popular Bregman divergencies ... there "must be some truth in them"!

 main outstanding issues are: interaction, how? description, how – via coding as in classical case (à la Shannon) ?

Equilibrium calculations

Background theorem \mathcal{P} any preparation (set of *x*'s). Given x^* , y^* such that: $x^* \in \mathcal{P}$, y^* robust, i.e. $\exists h \forall x \in \mathcal{P} : \Phi(x, y^*) = h$, and $y^* = x^*$. Then x^* is the MaxEnt distribution of \mathcal{P} (and y^* the MinRisk strategy for the physicist, i.e. $\operatorname{argmin} \left(\max_{x \in \mathcal{P}} \Phi(x, y) \right) = y^*$).

Proof: Assume $\Phi(x, y^*) = h$ for all $x \in \mathcal{P}$. Then $S(x^*) = \Phi(x^*, x^*) = \Phi(x^*, y^*) = h$, and, for $x \in \mathcal{P}$ and $x \neq x^*$ we have $S(x) < S(x) + D(x, y^*) = h$. The min-risk part is proved just as easily. \Box

Define: $y^* \in \mathcal{E}(y)$, the exponential family of y, if $\Phi(x, y^*)$ only depends on $\Phi(x, y)$. In short:

$$\mathcal{E}(y) = \{y^* | \exists \xi : \Phi(x, y^*) = \xi(\Phi(x, y))\}.$$

Corollary Put $L^{y}(h) = \{x | \Phi(x, y) = h\}$. If $\mathcal{P} = L^{y}(h)$ and x^{*} , y^{*} are given such that: $x^{*} \in \mathcal{P}, y^{*} \in \mathcal{E}(y)$ and $y^{*} = x^{*}$, then x^{*} is the MaxEnt distribution of \mathcal{P} .

... continued

Let \mathcal{L}^y be the class of non-empty models of the form $L^y(h)$ with $h \in \mathbb{R}$. The associated exponential family or equilibrium generating family, is the family

$$\mathcal{E}(y) = \{ x | \forall L \in \mathcal{L}^y \exists c \in \mathbb{R} : L \subseteq L^x(c) \}.$$

 $x \in \mathcal{E}(y)$, $S(x) = h \Rightarrow x$ is MaxEnt dist. of $L^{x}(h)$

Thus one should try and determine the elements in $\mathcal{E}(y)$. Looking at it, you find that for the worlds determined by one of the interactions π_q , every x for which there exist constants α and β such that

$$\forall i \in \mathbb{A} : \kappa(x_i) = \alpha + \beta \kappa(y_i)$$

are in $\mathcal{E}(y)$. For q = 1 this leads to classical analysis (with partition function etc.) and even without the use of Lagrange multipliers. For the general case, you have to adjust constants so that

$$\sum_{i \in \mathbb{A}} \kappa^{-1} \left(\alpha + \beta \kappa(y_i) \right) = 1.$$



Figure shows family $(\kappa_q)_{q\geq 0}$ of descriptors

So, given a probability t, you can see what effort is needed, measured in nats (natural units), in order to describe events with probability t when you use the chosen descriptor.



Figure shows inverses of descriptors. "probability checkers"

You can use these functions (q-deformed exponentials) to check, for a chosen descriptor, how complicated events you can describe with a given number of nat's available, i.e. how low a probability an event can have and still be describable with the available number of nat's. This kind of consideration is important in order to carry out MaxEnt calculations indicated previously.