Games, Entropy and Composability

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Goal

Operational definitions of entropy and related quantities covering the classical as well as non-extensive settings, thereby understanding which entropy measures are relevant for physics

Announcement: Workshop: "Facets of Entropy", Copenhagen, October 24-26, 2007 (if interested, ask FT or Robert Niven).

Overview

Entropy without games:

- Overall setting
- Listing some properties
- More specifics on structure
- Results for "*f*-entropies", especially on composability

Entropy with games:

- Complexity
- Defining entropy two types of entropy!
- Defining divergence
- MaxEnt via robustness ····
- Not primary focus on entropy -

complexity is what matters!

Conclusions:

• Which entropy ? – e.g. Tsallis or Rényi? *But:* does the question make sense?

Entropy without games

Overall setting: probabilities on discrete spaces!

Properties to consider include:

- minimal (0) on δ_i 's (deterministic), max on uniform
- continuous (lower semi-cont. in infinite case)
- concave : $H(mixture) \ge mixture \text{ of } H$
- datareduction inequality : H(coarse) ≤ H(fine)

• MaxEnt-principle should make sense for "natural" preparations (models) – and the nature of the entropy function should facilitate MaxEnt-calculations

• consistency: no feasible state is ignored under inference when you use the MaxEnt principle

- composable: $H(P \otimes Q) = g(H(P), H(Q))$
- – or even additive): $H(P \otimes Q) = H(P) + H(Q)$
- acceptable, physically significant interpretation!

"*f*-entropy": Based on generator f which is assumed to be nice convex and satisfy

$$f(0) = f(1) = 0, f'(1) = 1.$$

$$H_f(P) = -\sum f(p_i) \text{ or } H_f(P) = -\sum p_i \tilde{f}(\frac{1}{p_i})$$

- with \tilde{f} the Csiszár dual of f: $\tilde{f}(x) = xf(\frac{1}{x})$. BGS (classical): $f(x) = x \ln(x)$, $\tilde{f}(x) = \ln(\frac{1}{x})$, Tsallis family: (via "deformed logarithms"):

$$\mathsf{H}_q^T(P) = \frac{1}{1-q} \left(\sum p_i^q - 1 \right).$$

Well-known: H_f continuous, concave, satisfies datareduction principle – and MaxEnt? Wait!

Key result: Among *f*-entropies, only Tsallis entropies are composable. For these:

$$\mathsf{H}_{q}^{T}(P \otimes Q) = \mathsf{H}_{q}^{T}(P) + \mathsf{H}_{q}^{T}(Q)$$

+ (1 - q) $\mathsf{H}_{q}^{T}(P) \cdot \mathsf{H}_{q}^{T}(Q) .$

Entropy with games



Natures side: *P* Observers side (you!): *Q* connected by complexity function $\Phi = \Phi(P, Q)$.

Assumptions Minimal on diagonal: $\Phi(P,Q) \ge \Phi(P,P)$. Vanishes on deterministic dist.: $\Phi(\delta_i, \delta_i) = 0$. Examples:

 $\Phi^{BGS} = \sum p_i \ln \frac{1}{q_i}: BGS$ $\Phi^R_q = \frac{1}{1-q} \ln \frac{\sum p_i^q}{\sum p_i^q q_i^{1-q}}: Rényi$

 $\Phi_q^T = \frac{1}{1-q} \left(\frac{\sum p_i^q}{\sum p_i^q q_i^{1-q}} - 1 \right): \text{Tsallis}$

... continued: Entropy, Divergence, MaxEnt

Entropy = minimal complexity: $H(P) = \min_Q \Phi(P, Q)$. Divergence = actual – minimal complexity: $D(P,Q) = \Phi(P,Q) - H(P) (= \Phi(P,Q) - \Phi(P,P)).$

Dual entropy anticipates unknown but deterministic distribution: $\hat{H}(Q) = \sum q_i \Phi(\delta_i, Q) = \sum q_i D(\delta_i, Q)$.

MaxEnt-problem : given a preparation \mathcal{P} , to determine the MaxEnt-distribution and the corresponding MaxEntvalue: $H_{max} = H_{max}(\mathcal{P}) = \max_{P \in \mathcal{P}} H(P)$.

A highly useful, trivial, but neglected criterion:

If $Q \in \mathcal{P}$ is robust: $\Phi(P,Q)$ independent of $P \in \mathcal{P}$, say $\forall P \in \mathcal{P}$: $\Phi(P,Q) = h$, then Q is the MaxEntdistribution and $H_{max}(\mathcal{P}) = h$.

Proof. Firstly: $H(Q) = \Phi(Q, Q) = h$. Secondly: if $P \neq Q$ and $P \in \mathcal{P}$, then $H(P) < H(P) + D(P,Q) = \Phi(P,Q) = h$.

The examples (only Φ and H)

name	complexity	function of
BGS	$\sum p_i \ln rac{1}{q_i}$	$\langle \ln \frac{1}{Q}, P \rangle$
q-Rényi	$\frac{1}{1-q} \ln \frac{\sum p_i^q}{\sum p_i^q q_i^{1-q}}$	$\langle Q^{1-q}, P^{(q)} \rangle$
<i>q</i> -Tsallis¹	$rac{1}{1-q} \Bigl(rac{\sum p_i^q}{\sum p_i^q q_i^{1-q}} - 1 \Bigr)$	$\langle Q^{1-q}, P^{(q)} \rangle$
<i>q</i> -Tsallis ²	$rac{1}{1-q}\sum p_i^q(1-q_i^{1-q})$	$\langle 1 - Q^{1-q}, P^q \rangle$
q-Tsallis ³	$\sum \left(q_i^q - \frac{p_i(1 - qq_i^{q-1})}{1 - q} \right)$	$\sum q_i^q, \langle Q^{q-1}, P \rangle$

 $P^{(q)}$: the *q*-escort distribution: $i \curvearrowright p_i^q / \sum p_i^q$. P^q : the (non-normalized) measure $i \curvearrowright p_i^q$.

name	entropy	dual entropy
BGS	H^{BGS}	H^{BGS}
q-Rényi	H_q^R	H^{BGS}
<i>q</i> -Tsallis¹	H_q^T	H_q^T
<i>q</i> -Tsallis ²	H_q^T	H_{2-q}^{T}
<i>q</i> -Tsallis ³	H_q^T	H_q^T

The entropies: BGS: $-\sum p_i \ln p_i$, Rényi: $\frac{1}{1-q} \ln \sum p_i^q$, Tsallis: $\frac{1}{1-q} (\sum p_i^q - 1)$.

Property	Rényi	Tsallis
consistent inf.	q < 1 only	q < 1 only
concave	q < 1, few other	all q
composable	all q	all q
additive	all q	no q
interpretation	hmmm	hmmm
experimental evidence	hmmm	hmmm

Φ -exponential families etc.

To simplify, assume structure as in Tsallis³ (the Bregman case):

 $\Phi(P,Q) = \text{fct. of } Q + \langle \hat{Q}, P \rangle$

 \hat{Q} : a certain transform of Q. (Problem: Interpretation?)

For functions $\mathbf{f} = (f_1, \dots, f_k)$, define the Φ -exponential family \mathcal{E} as the set of distributions Q for which there exist constants λ_0 and $\lambda_1, \dots, \lambda_k$ such that:

$$\widehat{Q} = \lambda_0 + (\lambda_1 f_1 + \dots + \lambda_k f_k)$$

The natural preparations are those of the form

$$\mathcal{P}_{\mathbf{a}} = \{ P | \langle f_1, P \rangle = a_1, \cdots, \langle f_k, P \rangle = a_k \}.$$

¿From robustness criterion we find immediately:

Theorem If $Q \in \mathcal{E} \cap \mathcal{P}_{\mathbf{a}}$, then Q is the Φ -MaxEnt distribution.

Conclusions

Recall the key technical result (due to HDP):

Among *f*-entropies, only Tsallis entropies are composable.

This also covers entropies (like Rényi entropy) which are monotone functions of f-entropies.

Apart from the key result we conclude with some insights gained during the investigations:

(*i*) Never more use Lagrange multipliers!

- unless you deal with "ad hoc problem" or use these multipliers as a guide in preliminary investigations.

(*ii*) Be aware of the two types of entropies!

(iii) Never consider entropy measures alone! – you must supply with other considerations, at best:

take as point of departure a suitable complexity measure!