Retrodiction in Stochastic Thermodynamics

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52nd Symposium on Mathematical Physics, Toruń (online), 17 Jun 2021

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New physics!!

Long-Awaited Muon Measurement Boosts Evidence for New Physics

Initial data from the Muon g-2 experiment have excited particle physicists searching for undiscovered subatomic particles and forces

أعرض هذا باللغة العربية By Daniel Garisto on April 7, 2021

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Muon g-2 magnetic storage ring, seen here at Brookhaven National Laboratory in New York State befor ts 2013 relocation to Fermi National Accelerator Laboratory in Illinois. Credit: Alamy

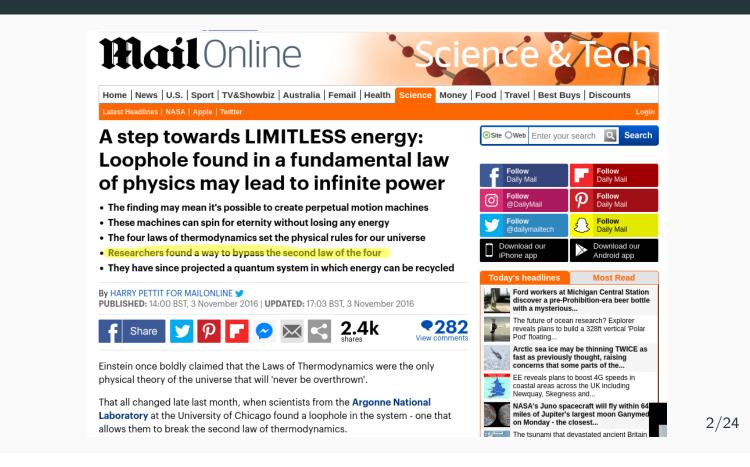
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New "physics"??



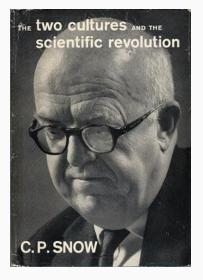
The Second Law is special

"The law that entropy always increases holds, I think, the supreme position among the laws of Nature. [...] If your theory is found to be against the Second Law of Thermodynamics I can give you no hope; there is nothing for it to collapse in deepest humiliation."

A.S. Eddington

"[...] the only physical theory of universal content concerning which I am convinced that, within the framework of the applicability of its basic concepts, it will never be overthrown." A. Einstein

Have you read a work of Shakespeare's?



"Once or twice I have been provoked and have asked the company how many of them could describe the Second Law of Thermodynamics. The response was cold: it was also negative. Yet I was asking something which is about the equivalent of: Have you read a work of Shakespeare's?" C.P. Snow

The "to be or not to be" of thermodynamics

Clausius Inequality

 $\langle \Delta S_{\rm tot} \rangle \ge 0$

Why should the above inequality be so "special"? What does it *really* say?

That is the question.

That is the question!



"*No one understands entropy very well.*"

von Neumann (apocryphal)

The Second Law "without entropy"

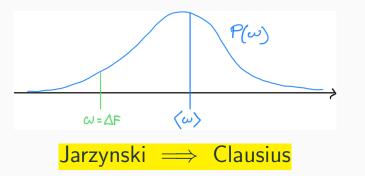


Clausius' inequality (1865):

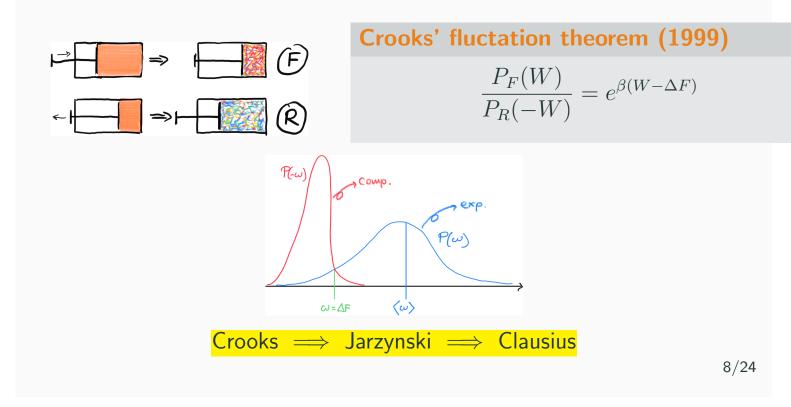
Jarzynski's equality (1997):

$$\langle W \rangle \ge \Delta F$$

$$\left\langle e^{-\beta W} \right\rangle = e^{-\beta \Delta F}$$



The Second Law and irreversibility



Usual explanation

Crooks' theorem, and hence Jarzynski's relation, and hence the Second Law, all rely on two assumptions satisfied at equilibrium:

- 1. thermal equilibrium: initial distribution is $P(\xi) \propto e^{-\beta \epsilon(\xi)}$
- microscopic reversibility: molecular processes and their reverses occur at the same rate

But, again: why does the Second Law feel so "special" then?

Is that because of some kind of "special" microscopic balancing mechanism?

A hint from Ed Jaynes

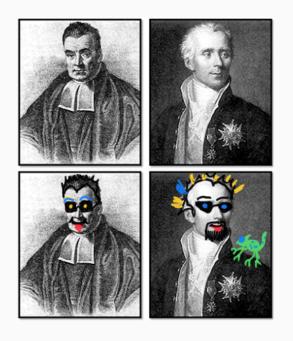


"To understand and like thermo we need to see it, not as an example of the *n*-body equations of motion, but as an example of the logic of scientific inference."

E.T. Jaynes (1984)

First idea: reverse process as Bayesian retrodiction

The Bayes-Laplace Rule



Inverse Probability Formula

 $\underbrace{\mathcal{P}(H|D)}_{\text{inv. prob.}} \propto \underbrace{\mathcal{P}(D|H)}_{\text{likelihood}} \underbrace{\mathcal{P}(H)}_{\text{prior}}$ where H is a hypothesis, D is the result of observation (i.e., evidence)

postmodern Bayesianism!

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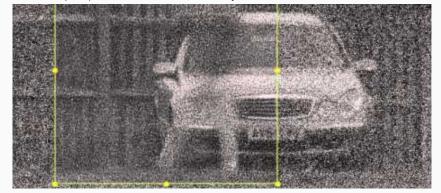
Meanings of the inverse probability

- it is the main *tool* of Bayesian statistics for problems like:
 - estimation (e.g.: how many red balls are in an urn?)
 - decision (e.g.: is ACME's stock a good investment? should I buy some?)
 - predictive inference (e.g.: weather forecasts)
 - retrodictive inference (e.g.: what kind of stellar event possibly caused the Crab Nebula?)
- it measures the degree of belief that a rational agent should have in one hypothesis, among other mutually exclusive ones, given the data

Noisy data and uncertain evidence

BUT! Bayes-Laplace Rule *does not* tell us how to update the prior in the face of uncertain data...

• suppose that a noisy observation suggests a probability distribution Q(D) for the data (e.g., the license plate no.)



• how should we update our prior $\mathcal{P}(H)$ given *uncertain evidence* in the from $\mathcal{Q}(D)$?

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Jeffrey's rule of probability kinematics

Vanilla Bayes:Extended Bayes:
$$\mathcal{P}(H|D) = \mathcal{P}(D|H)\mathcal{P}(H)/\mathcal{P}(D)$$
 $\mathcal{P}(H|\mathcal{Q}(D)) = ?$ Jeffrey's conditioning* (1965) $\mathcal{P}(H|\mathcal{Q}(D)) = \sum_{D} \underbrace{\mathcal{P}(H|D)}_{D} \mathcal{Q}(D)$
 $= \sum_{D} \underbrace{\mathcal{P}(D|H)\mathcal{P}(H)}_{D} \mathcal{Q}(D)$ * Jeffrey's rule was introduced *ad hoc*, but it can be proved from Bayes-Laplace Rule and

Pearl's method of virtual evidence (1988)

Construction of the reverse process

- physical setup:
 - \circ a stochastic transition rule: $\varphi(y|x)$
 - $\circ\,$ a steady (viz. invariant) state: $\sum_x \varphi(y|x) s(x) = s(y)$
- Bayesian inversion at the steady state:

$$s(y)\hat{\varphi}(x|y) := s(x)\varphi(y|x) \iff \frac{\varphi(y|x)}{\hat{\varphi}(x|y)} = \frac{s(y)}{s(x)}$$

- two priors:
 - predictor's prior: p(x)
 - \circ retrodictor's prior q(y)
- two processes:
 - forward process (prediction): $\mathcal{P}_F(x,y) = \varphi(y|x)p(x)$
 - reverse process (retrodiction): $\mathcal{P}_R(x,y) = \hat{\varphi}(x|y)q(y)$

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A picture

$$S \rightarrow \varphi(y|z) \rightarrow = \left\{ - \left(\hat{\varphi}(z|y) \right) - S \right\}$$
$$P \rightarrow \varphi(y|z) \rightarrow \neq \left\{ - \left(\hat{\varphi}(z|y) \right) - q \right\}$$

- at the steady state: prediction = retrodiction
- otherwise: asymmetry (irreversibility, *irretrodictability*)

Measures of statistical divergence

Second idea: fluctuation relations as measures of *statistical divergence* between $\mathcal{P}_F(x, y)$ and $\mathcal{P}_R(x, y)$

• relative entropy: $D(\mathcal{P}_F \| \mathcal{P}_R) := \left\langle -\ln \frac{\mathcal{P}_R(x,y)}{\mathcal{P}_F(x,y)} \right\rangle_F =: \left\langle -\ln r(x,y) \right\rangle_F$

 \rightsquigarrow more generally, one can use $D_f(\mathcal{P}_R \| \mathcal{P}_F) := \langle f(r(x,y)) \rangle_F$

- introduce probability density functions

From f-divergences to f-fluctuation theorems

• for $f : \mathbb{R}^+ \to \mathbb{R}$ invertible

f-Fluctuation Theorem $\mu_R(\omega) = f^{-1}(\omega)\mu_F(\omega) \implies \langle f^{-1}(\omega) \rangle_F = 1$

$$\rightsquigarrow$$
 for $f(u) = -\ln u$, we have $f^{-1}(v) = e^{-v}$, that is

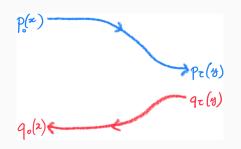
$$\frac{\mu_F(\omega)}{\mu_R(\omega)} = e^{\omega} \quad \Longrightarrow \quad \left\langle e^{-\omega} \right\rangle_F = 1$$

further discussions in arXiv:2009.02849 and arXiv:2106.08589

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Examples of known results recovered by retrodiction

Example: driven closed system evolution



• driving protocol: $H(0) \to H(t) \to H(\tau)$

•
$$H(0) = \sum_{x} \epsilon_x \pi_x$$
, $H(\tau) = \sum_{y} \eta_y \pi'_y$

•
$$\varphi(y|x) = \delta_{y,y(x)}$$
, i.e., one-to-one

•
$$s(x) = d^{-1} \implies \varphi(y|x) = \hat{\varphi}(x|y)$$

•
$$p_0(x) = e^{\beta(F - \epsilon_x)}$$
, $q_\tau(y) = e^{\beta(F' - \eta_y)}$

In this case, for the choice $f(u) = -\ln u$,

$$\Omega(x,y) = \ln \frac{\mathcal{P}_F(x,y)}{\mathcal{P}_R(x,y)} = \ln \frac{s(y)p(x)}{s(x)q(y)} = \ln \frac{p(x)}{q(y)}$$
$$= \beta(F - \epsilon_x + F' + \eta_y) = \beta(W - \Delta F)$$
$$\Rightarrow \frac{\mu_F(W)}{\mu_R(W)} = e^{\beta(W - \Delta F)}$$

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Example: nonequilibrium steady states

- stochastic process $\varphi(y|x)$ with non-thermal steady state s(x)
- thermal equilibrium priors: $p(x) = q(x) \propto e^{-\beta \epsilon_x}$
- fluctuation variable: $\omega = \ln \frac{\mathcal{P}_F(x,y)}{\mathcal{P}_R(x,y)} = \ln \frac{p(x)}{q(y)} \frac{s(y)}{s(x)} = \beta(\epsilon_y - \epsilon_x) + (\ln s(y) - \ln s(x))$
- nonequilibrium potential: $V(x) := -\frac{1}{\beta} \ln s(x)$ (e.g., Manzano&al 2015)
- $\left\langle e^{\beta(\Delta E \Delta V)} \right\rangle_F = 1$, but $\left\langle e^{\beta \Delta E} \right\rangle_F =$ "efficacy"
- — nonequilibrium potentials (usually introduced ad hoc) are understood here as remnants of Bayesian inversion

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But why known relations are compatible with Bayesian inversion?

Is that a necessity?

Sketch argument

•
$$D(\mathcal{P}_F \| \mathcal{P}_R) = \left\langle \ln \frac{\mathcal{P}_F(x,y)}{\mathcal{P}_R(x,y)} \right\rangle_H$$

- let us impose that the fluctuation variable is a microstate
 - **function**: $\ln \frac{\mathcal{P}_F(x,y)}{\mathcal{P}_R(x,y)} = \Omega(x,y) \stackrel{!}{=} G'(y) G(x)$

$$\implies \frac{\mathcal{P}_F(y|x)}{\mathcal{P}_B(x|y)} = \frac{H(y)}{H(x)}$$

•
$$\implies$$
 $H(x)\mathcal{P}_F(y|x) = H'(y)\mathcal{P}_R(x|y)$

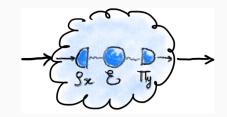
• sum over
$$x \implies H'(y) = \sum_x H(x) \mathcal{P}_F(y|x)$$

•
$$\implies \mathcal{P}_R(x|y) = \frac{1}{\sum_x H(x)\mathcal{P}_F(y|x)} H(x)\mathcal{P}_F(y|x)$$

Hence, a Bayesian inverse-like form for the reverse process is **inevitable** if we want the (total stochastic) entropy production to be a **microstate function**!

And what about the quantum case?

Quantum retrodiction and the Petz map



- assume $\varphi(y|x) = \operatorname{Tr}[\Pi_y \mathcal{E}(\rho_x)]$
- let s(x) be invariant distribution
- according to the formalism of *quantum retrodiction*:

$$\circ \Sigma := \sum_{x} s(x) \rho_{x}$$

$$\circ \hat{\rho}_{y} := \frac{1}{s(y)} \sqrt{\mathcal{E}(\Sigma)} \Pi_{y} \sqrt{\mathcal{E}(\Sigma)}$$

$$\circ \hat{\Pi}_{x} := s(x) \frac{1}{\sqrt{\Sigma}} \rho_{x} \frac{1}{\sqrt{\Sigma}}$$

$$\circ \hat{\mathcal{E}}(\cdot) := \sqrt{\Sigma} \left\{ \mathcal{E}^{\dagger} \left[\frac{1}{\sqrt{\mathcal{E}(\Sigma)}} (\cdot) \frac{1}{\sqrt{\mathcal{E}(\Sigma)}} \right] \right\} \sqrt{\Sigma}$$



• Bayesian inversion works seamlessly $\hat{\varphi}(x|y) = \operatorname{Tr}[\hat{\Pi}_x \ \hat{\mathcal{E}}(\hat{\rho}_y)]$

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Some remarks

- the Petz recovery map reduces to Bayes–Laplace rule when operators commute
- to a unique Bayes–Laplace rule there correspond infinite possible Petz maps ("rotated" Petz maps)
- retrodiction (both classical and quantum) depends on the choice of reference prior
- exceptions are unitary (i.e., "bilateral deterministic") channels, for which:
 - 1. there is a unique Petz reverse (the retrodiction is independent of the choice of prior, and all rotated Petz maps coincide)
 - 2. retrodiction and (linear) inversion coincide

Conclusion

Four messages

- the Second Law is special among the laws of physics, because it is in fact a law of logic
- retrodiction provides the logical foundations of fluctuation theorems
- a clear distinction between (mechanical) *reversibility* and (thermodynamic) *retrodictability* avoids unnecessary paradoxes
- quantum retrodiction and quantum fluctuation relations follow seamlessly using Petz recovery map

