

Dynamic non-Gibbsianness

Original explanation

- ▶ Two-slice system: past acts as hidden variables for present
- ▶ Two-slice system \sim equilibrium duplicated variables

Alternative paradigm

- ▶ Intuitively: most probable history of an improbable state
- ▶ Formally: large deviations in trajectory space
- ▶ Non-Gibbs = multiple optimal trajectories \rightarrow discontinuity

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Plan and credits

In this talk:

- ▶ Review of Gibbsianness
- ▶ Review of original proof of dynamical non-Gibbsianness
- ▶ New paradigm for dynamical non-Gibbsianness
- ▶ Rigorous results for
 - ▶ Mean-field spin models
 - ▶ Kac models

Acknowledgements:

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Victor Ermolaev (Nedap)

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Lattice systems

Basic ingredients:

- ▶ *Lattice* \mathbb{L} : e.g. \mathbb{Z}^d
- ▶ *Single-spin space* S : e.g. $\{-1, 1\}$
- ▶ *Configuration space* $\Omega = S^{\mathbb{L}}$

Topology and σ -algebra \mathcal{F} generated by cylinders:

$$C_{\omega_{\Lambda}} = \{\omega \in \Omega : \omega_{\Lambda} = \sigma_{\Lambda}\}, \Lambda \subset\subset \mathbb{L} \quad [\omega_{\Lambda} = (\omega_x)_{x \in \Lambda}]$$

Interaction: Family of local functions (=local contributions)

$$\Phi = \{\phi_B : \Omega \rightarrow \mathbb{R}, \mathcal{F}_B\text{-measurable}\} \quad [\phi_B(\omega) = \phi_B(\sigma) \text{ if } \omega_B = \sigma_B]$$

Hamiltonian on region Λ given σ outside:

$$H_{\Lambda}(\omega \mid \sigma) = \sum_{B: B \cap \Lambda \neq \emptyset} \phi_B(\omega_{\Lambda} \sigma)$$

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(Lattice) Gibbs measures: formal definition

Gibbsian specification: Family $\Pi^\Phi = \{\pi_\Lambda^\Phi : \Lambda \subset \mathbb{L}\}$ with

$$\pi_\Lambda^\Phi(C_{\omega_\Lambda}) = \frac{e^{-\beta H_\Lambda(\omega|\sigma)}}{\text{Norm.}}$$

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Gibbs measures: μ is Gibbs for Φ if, equivalently,

- ▶ μ is left invariant by Π^Φ :

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- ▶ $\mu = w - \lim_{\Lambda \rightarrow \mathbb{L}} \pi_\Lambda^\Phi(\cdot | \sigma) + \text{convex combinations}$
[thermodynamic limit]

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How to recognize Gibbsianness

Kozlov – Sullivan: μ is Gibbs iff it is

- ▶ *Non-null*: $\mu(C_{\omega_\Lambda}) > 0$ for every cylinder C_{ω_Λ}
- ▶ *Quasilocal*: If $\Lambda \subset \Gamma \subset \mathbb{L}$,

$$\sup_{\sigma, \omega, \xi^\pm} \left| \mu(C_{\omega_\Lambda} \mid \sigma_\Gamma \xi^+) - \mu(C_{\omega_\Lambda} \mid \sigma_\Gamma \xi^-) \right| \xrightarrow{\gamma \rightarrow \mathbb{L}} 0$$

Physics in Λ does not depend on state of the Andromeda galaxy

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Essential non-quasilocality

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for *every* realisation of $\mu(C_{\omega_\Lambda} \mid \cdot)$

- ▶ Quasilocality = continuity w.r.t. external conditions
- ▶ Non-quasilocality = essential discontinuity w.r.t. external conditions

Interpretation: Info from ∞ despite frozen fluctuations

Possible explanation: hidden variables

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Renormalization transformations

General definition: A (stochastic) RT is a map

$$\begin{aligned} \text{Prob}(\Omega) &\longrightarrow \text{Prob}(\Omega') \\ \mu &\longmapsto \mu'(\cdot) = \int K(\cdot | \omega) \mu(d\omega) \end{aligned}$$

where K is a probability kernel

The transformation is *deterministic* if $\exists f : \Omega \rightarrow \Omega'$ s.t.

$$K(\cdot | \omega) = \delta_{f(\omega)}(\cdot)$$

A *block* RT is of the form

$$K(dw' | \omega) = \prod_{x'} K'_x(dw'_{x'} | \omega_{B_{x'}})$$

each $B_{x'} \subset \subset \mathbb{L}$ is the block associated to x'

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Hidden variables and non-quasilocality

Hidden-variables mechanism:

- ▶ Each fixed ω'_{Λ^c} determines a constrained Ω system
- ▶ ω'^{sp} is s.t. the constrained system has a phase transition
- ▶ ξ' far away decides the phase \rightarrow info from ∞

Two-slice point of view:

- ▶ Ω = original slice = hidden variables
- ▶ Ω' = present slice = observed variables

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Single-site Kadanoff transformations ($B_{x'} = \{x'\}$)

On finite volumes, the two-slice measures are of the form

$$K_{\Lambda}(d\omega' | \omega) \mu_{\Lambda}(d\omega) \propto \exp\left\{\beta[H_{\Lambda}^{\text{Kad}}(\omega, \omega') + H_{\Lambda}(\omega)]\right\} d\omega'_{\Lambda} d\omega_{\Lambda}$$

where

$$H_{\Lambda}^{\text{Kad}}(\omega, \omega') = \sum_{x'} \left\{ \frac{p}{\beta} \omega'_{x'} \omega_x - \frac{1}{\beta} \log[2 \cosh(p \omega_x)] \right\}$$

acts on the original spins as an extra magnetic field

Constrained internal spins have phase transition if

$$\frac{p}{\beta} \omega'_{x'} \text{ compensates } h \text{ in the average} \quad (*)$$

and β is large enough

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Single-site Kadanoff transformations (cont.)

Let h be the original Ising field and fix β large enough s.t.

- ▶ Original model with $h = 0$ has phase transition
- ▶ Pirogov-Sinai theory holds

If $h = 0$, alternated $\omega' \implies (*)$ for p/β small enough

Hence, $\exists p_1 > p_2$ s.t.

- ▶ μ' is Gibbs for $p > p_1$
- ▶ μ' is not Gibbs for $p_2 > p$

If $h \neq 0$, $\exists \omega'$ s.t. $(*)$ *only* for a range of p/β

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Dynamic Gibbs – non-Gibbs transitions

Simulations, particle systems, PCA:

$$\mu_t = S_t \mu$$

with $S_t =$ semigroup of operators.

Dynamic G–non-G: μ Gibbs but μ_t non-Gibbs at some t

Example:

- ▶ $\mu =$ low- T Ising model
- ▶ $S_t = S^n$ infinite- T discrete-time Glauber

$$S = \prod_x S_{\{x\}} \quad \text{with} \quad \begin{aligned} S_x(\omega_x | \omega_x) &= 1 - \epsilon \\ S_x(-\omega_x | \omega_x) &= \epsilon \end{aligned}$$

[invariant measure = product measure = infinite- T Gibbs]

Unquenching: heating up a low- T Ising model

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Definition and example

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Un-quenching G–non-G transitions

In matrix form $(S_x)_{\omega\omega'} \equiv S_x(\omega'_x | \omega_x)$

$$S_x = \begin{pmatrix} 1 - \epsilon & \epsilon \\ \epsilon & 1 - \epsilon \end{pmatrix}, \quad S_x^n = \frac{1}{2} \begin{pmatrix} 1 + a_n & 1 - a_n \\ 1 - a_n & 1 + a_n \end{pmatrix}$$

with $a_n = (1 - 2\epsilon)^n$. Hence

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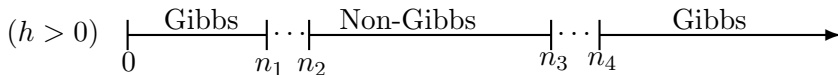
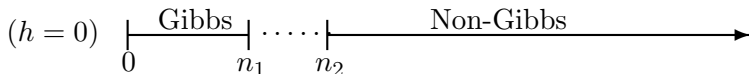
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Un-quenching G–non-G transitions (cont.)

Using previous results on Kadanoff-renormalized measures:

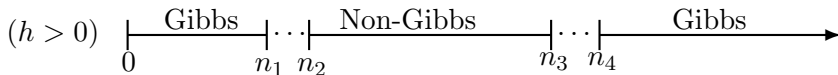
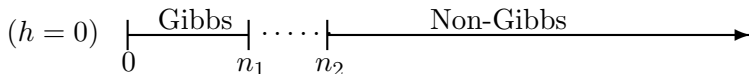


Mathematical mechanism: hidden variables (two-slice view)

Physical mechanism?

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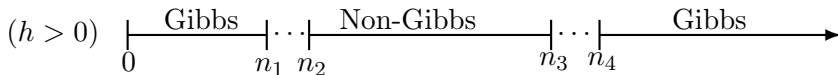
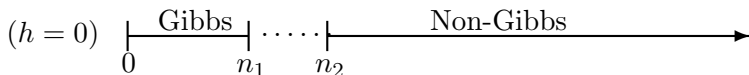


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Alternative paradigm: Heuristic version

A. van Enter: *most probable history of an improbable state*

Given a large improbable droplet. How did it get there?

- ▶ **Nurture:** Created by the dynamics (cost exp-volume)
- ▶ **Nature:** Present at $t = 0$ and survived
To compete: typical of the other phase (cost exp-perimeter)

Heuristic version:

- ▶ Short t : Only nature, no time to change much
- ▶ Mid t :
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 - ▶ Hence ξ^\pm determines original phase \rightarrow discontinuity
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- ▶ Short t : Only nature, no time to change much
- ▶ Mid t :
 - ▶ ω^{sp} nurtured, but ξ^{\pm} nature
 - ▶ Hence ξ^{\pm} determines original phase \rightarrow discontinuity
- ▶ Long t : If $h \neq 0$ only one phase \rightarrow no tilting mechanism

Alternative paradigm: Heuristic version

A. van Enter: *most probable history of an improbable state*

Given a large improbable droplet. How did it get there?

- ▶ **Nurture:** Created by the dynamics (cost exp-volume)
- ▶ **Nature:** Present at $t = 0$ and survived
To compete: typical of the other phase (cost exp-perimeter)

Heuristic version:

- ▶ Short t : Only nature, no time to change much
- ▶ Mid t :
 - ▶ ω^{SP} nurtured, but ξ^\pm nature
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Alternative paradigm: rigorous version

most probable history = most probable trajectory

most probable = minimizer of the large-deviation rate

Paradigm: Establish a large-deviation principle for trajectories of *measures conditioned* to a given final *empirical measure*

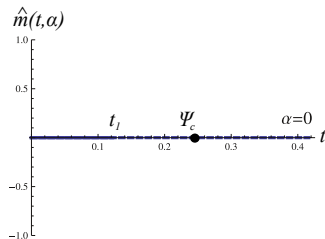
- ▶ Single minimizer = Gibbsianness
- ▶ Multiple minimizers = non-Gibbsianness

Perturbation of conditioning

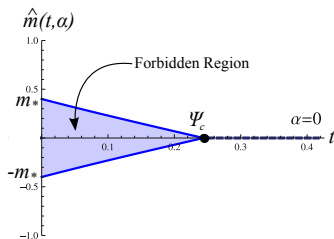
→ discontinuous choice of trajectory

Alternative paradigm: graphical summary

$$[h = 0]$$



One trajectory = Gibbs



Many trajectories = non-Gibbs

The program

Prove rigorously the previous paradigm.

Steps:

- (i) Mean-field models
- (ii) Kac models
- (iii) Finite-range models

At present: (i) and (ii) for Ising under independent dynamics

(i) Mean-field:

- ▶ No geometry – no notion of neighbourhood
- ▶ Everything in terms of empirical magnetization

Mean-field measures and evolution

The H^N -Gibbs measure [β absorbed]

$$\mu^N(d\sigma) = \frac{e^{-H^N(\sigma)}}{Z^N} d\sigma$$

induces a measure on \mathcal{M}_N

$$\bar{\mu}^N(dm) := \binom{N}{\frac{1+m}{2} \quad N} \frac{e^{-N\bar{H}(m)}}{Z^N} dm$$

$[\mu^N \longleftrightarrow \bar{\mu}^N + \text{permutation invariance}]$

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Gibbs and Non-Gibbs mean-field models

(Külske and Le Ny)

For fixed $t \geq 0$:

(a) A magnetization $\alpha \in [-1, 1]$ is **good** for μ_t if

$$\gamma_t(\cdot \mid \tilde{\alpha}) := \lim_{\substack{N \rightarrow \infty \\ \alpha_N \rightarrow \tilde{\alpha}}} \gamma_t^N(\cdot \mid \alpha_{N-1}),$$

- ▶ exists and is independent of the sequence $\alpha_N \rightarrow \tilde{\alpha}$
- ▶ it is continuous in $\tilde{\alpha}$

for $\tilde{\alpha}$ in a neighbourhood of α

(b) A magnetization $\alpha \in [-1, +1]$ is bad if it is not good

(c) μ_t is **Gibbs** if it has no bad magnetizations

Large-deviations: General definition

Informally:

A family of measures (ν^N) satisfies a *large-deviation principle* if

$$\nu^N(A) \sim e^{-N I(A)}$$

- ▶ N is the LDP *speed*, I the *rate function*
- ▶ As a consequence, $\text{supp}(\nu^N) \rightarrow \text{argmin}(I)$

Formally:

(ν^N) on a Borel space satisf. LDP with rate fcn I and speed N if

$$\liminf_{N \rightarrow \infty} \frac{1}{N} \log \nu^N(A) \geq - \inf_{x \in A} I(x) \quad \text{for } A \text{ open}$$

$$\limsup_{N \rightarrow \infty} \frac{1}{N} \log \nu^N(A) \leq - \sup_{x \in A} I(x) \quad \text{for } A \text{ closed}$$

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LDP for mean-field Ising:

“Static” part:

The family $(\bar{\mu}^N)$ satisfies a LDP with speed N and rate $I_S - \inf(I_S)$ with

$$I_S(m) := \bar{H}(m) + \frac{1+m}{2} \log(1+m) + \frac{1-m}{2} \log(1-m).$$

Independent evolutions:

Let $P^N = \text{law of } (m_t^N)_{t \geq 0}$

Defined on the space of càdlàg trajectories; Skorohod topology

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(Ermolaev and Külske)

(P^N) restricted to $[0, T]$ satisfies LDP with speed N and rate $I^T - \inf(I^T)$ given by

$$I^T(\phi) := I_S(\phi(0)) + I_D^T(\phi),$$

where

$$I_D^T(\phi) := \begin{cases} \int_0^T L(\phi(s), \dot{\phi}(s)) ds & \text{if } \dot{\phi} \text{ exists} \\ \infty & \text{otherwise} \end{cases}$$

is the action integral with Lagrangian

$$L(m, \dot{m}) = -\frac{1}{2} \sqrt{4(1-m^2) + \dot{m}^2} + \frac{1}{2} \dot{m} \log \left(\frac{\sqrt{4(1-m^2) + \dot{m}^2} + \dot{m}}{2(1-m)} \right) + 1$$

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LDP for conditioned mean-field evolutions

The family of measures on trajectory space

$$Q_{t,\alpha}^N(\cdot) := P^N((m_N(s))_{0 \leq s \leq t} = \cdot \mid m_N(t) = \alpha)$$

satisfies LDP with speed N and rate $I^{t,\alpha} - \inf(I^{t,\alpha})$, with

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Hence, conditioned optimal trajectories correspond to

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The mean-field computational advantage

Simplifying feature:

- ▶ There is an explicit expression for

$$C_{t,\alpha}(m) := \inf_{\substack{\phi: \phi(0)=m, \\ \phi(t)=\alpha}} I^t(\phi)$$

- ▶ We have the identity

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I. Single optimal trajectory = Gibbsianness

$\alpha \mapsto \gamma_t(\sigma \mid \alpha)$ is continuous at α_0 if and only if

- ▶ $I^t(\phi)$ has a unique minimizing path $\hat{\phi}$
- ▶ or, equivalently, $C_{t,\alpha_0}(m)$ has a unique minimizing m .

Furthermore, in this case, the specification kernel equals

$$\gamma_t(z \mid \alpha) = \frac{\sum_{x \in \{-1, +1\}} e^{x[J\hat{\phi}(0)+h]} p_t(x, z)}{\sum_{x, y \in \{-1, +1\}} e^{x[J\hat{\phi}(0)+h]} p_t(x, y)}$$

$p_t(\cdot, \cdot) =$ kernel of Markov jump process on $\{-1, +1\}$ with

- ▶ jumping rate 1
- ▶ jump probabilities $p_t(i, \pm i) = e^{-t} \begin{cases} \cosh(t) \\ \sinh(t) \end{cases}$

“If” part and for of γ_t proven by Ermolaev and Külske

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II. Short-term Gibbsianness

Theorem

If $J \leq 1$ the evolved measures μ_t are Gibbs for all $t \geq 0$

- ▶ Proven by Külske and Le Ny and Külske and Ermolaev
- ▶ Note that $1 = \beta_{\text{cr}}^{\text{MF}}$

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III. Case $J > 1$, $h = 0$

Consider the critical time

$$\Psi_c(J) := \begin{cases} \frac{1}{2} \operatorname{acoth}(2J - 1) & \text{if } 1 < J \leq \frac{3}{2}, \\ t_*(J) \text{ implicitly calculable} & \text{if } J > \frac{3}{2}, \end{cases}$$

Then:

- ▶ $t < \Psi_c$: Evolved measure μ_t is Gibbs
- ▶ $t > \Psi_c$: Discontinuity at $\alpha = 0$; two optimal trajectories $\pm \tilde{\phi}$
- ▶ If $\Lambda_{t,0}(J) = \text{cone}$ between the trajectories $\pm \hat{\phi}$
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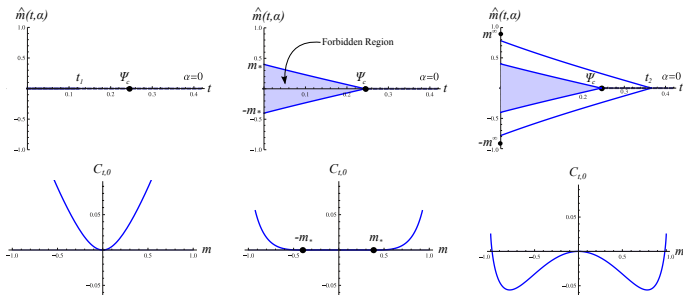
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Graphic summary: $h = 0, \alpha = 0$



$$t < \Psi_c$$

$$t = \Psi_c$$

$$t > \Psi_c$$

First row: Minimizing trajectories for $(J, h) = (1.6, 0)$

Second row: Corresponding plots of $m \mapsto C_{t,0}(m)$

IV. Bad magnetizations as function of time

