

## Gibbs-non-Gibbs dynamical transitions. A large-deviation paradigm

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Ka
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Gibbs measures and their transformations

## The Gibbs – non-Gibbs saga

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#### Intuitively, $\mu$ Gibbs if

 $\mu \propto e^{-\beta H}$ 

#### This is, however, valid only on finite regions

To pass to the thermodynamic limit must introduce:

- Interactions
- ► Specifications

Gibbs measures designed to describe *equilibrium* No reason to expect them out of equilibrium, e.g. under evolutions

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Gibbs measures and their transformations

## **Examples of non-Gibsianness**

#### **Renormalization transformations**

- $\blacktriangleright$  Block-renormalization: blocks of spins  $\rightarrow$  effective spins
- ▶ Renormalized measure: coarser, blurred
- ▶ In many instances: renormalized measure non-Gibbsian
- ▶ Reason: hidden variables bringing info from infinite

### **Stochastic evolutions**

Gibbs measures subjected to Glauber dynamics

- ▶ Can loose Gibsianness at some finite time
- Gibbsianness recovered in some cases

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## Dynamic non-Gibbsianness

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## Original explanation

- ▶ Two-slice system: past acts as hidden variables for present
- $\blacktriangleright$  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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- Review of Gibbsianness
- ▶ Review of original proof of dynamical non-Gibbsianness

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- ▶ New paradigm for dynamical non-Gibbsianness
- Rigorous results for
  - Mean-field spin models
  - ► Kac models

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Definition						

#### Basic ingredients:

- ▶ Lattice L: e.g.  $\mathbb{Z}^d$
- Single-spin space S: e.g.  $\{-1, 1\}$
- Configuration space  $\Omega = S^{\mathbb{L}}$ Topology and  $\sigma$ -algebra  $\mathcal{F}$  generated by cylinders:

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

Interaction: Family of local functions (=local contributions)

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

**Hamiltonian** on region  $\Lambda$  given  $\sigma$  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  $\Pi^{\Phi} = \{\pi^{\Phi}_{\Lambda} : \Lambda \subset \mathbb{L}\}$  with

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

## $[\pi^{\Phi}_{\Lambda}(\,\cdot\mid\sigma)=\text{equilibrium in }\Lambda\text{ given }\sigma]$

**Gibbs measures:**  $\mu$  is Gibbs for  $\Phi$  if, equivalently,

•  $\mu$  is left invariant by  $\Pi^{\Phi}$ :

$$\int \pi^{\Phi}_{\Lambda}(C_{\omega_{\Lambda}}) \, \mu(d\omega) = \mu(C_{\omega_{\Lambda}})$$

 $[\mu = \text{equilibrium in } \mathbb{L} = \text{every } \Lambda \text{ in equilibrium}]$ 

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

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Mean field Kac End

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Gibbsianness test

## How to recognize Gibbsianness

#### Kozlov – Sullivan: $\mu$ is Gibbs iff it is

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

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

Physics in  $\Lambda$  does not depend on state of the Andromeda galaxy



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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

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Renormalization and non-Gibbsianness

## **Renormalization transformations General definition:** A (stochastic) RT is a map

$$\operatorname{Prob}(\Omega) \longrightarrow \operatorname{Prob}(\Omega')$$
$$\mu \longmapsto \mu'(\cdot) = \int K(\cdot \mid \omega) \,\mu(d\omega)$$

#### where K is a probability kernel

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

$$K(\,\cdot\mid\omega)=\delta_{\!_{f(\omega)}}(\,\cdot\,)$$

A  $block \operatorname{RT}$  is of the form

$$K(d\omega'\mid\omega)=\prod_{x'}K'_x(d\omega'_{x'}\mid\omega_{\!_{B_{x'}}})$$

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

Intro	Gibbs	Non-Gibbs
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Mean field Kac End

Renormalization and non-Gibbsianness

## Examples of block transformations

## **Deterministic transformations:**

- Decimation
- Majority (odd block)

#### Stochastic transformations:

- Majority (even block)
- ▶ Kadanoff:

$$K'_x(d\omega'_{x'} \mid \omega_{\scriptscriptstyle B_{x'}}) = \frac{\exp\Big\{p\,\omega'_{x'}\sum_{x\in B_{x'}}\omega_x\Big\}}{\operatorname{Norm}}\,d\omega'_{x'}$$

[weighted majority;  $\rightarrow$  majority as  $p \rightarrow \infty$ ]

Intro	Gibbs	Non-Gibbs
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Renormalization and non-Gibbsianness

## Hidden variables and non-quasilocality

Hidden-variables mechanism:

- ► Each fixed  $\omega'_{\Lambda^c}$  determines a constrained  $\Omega$  system
- ▶  $\omega'^{\rm sp}$  is s.t. the constrained system has a phase transition

•  $\xi'$  far away decides the phase  $\rightarrow$  info form  $\infty$ 

Two-slice point of view:

- $\Omega$  = original slice = hidden variables
- $\Omega'$  = 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

$$K_{\Lambda}(d\omega' \mid \omega) \,\mu_{\Lambda}(d\omega) \propto \exp \left\{ eta \left[ H_{\Lambda}^{\mathrm{Kad}}(\omega, \omega') + H_{\Lambda}(\omega) 
ight] 
ight\} d\omega'_{\Lambda} \, d\omega_{\Lambda}$$

where

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

acts on the original spins as an extra magnetic field Constrained internal spins have phase transition if

$$\frac{p}{\beta}\omega'_x$$
 compensates  $h$  in the average (\*)

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and  $\beta$  is large enough
## Single-site Kadanoff transformations $(B_{x'} = \{x'\})$

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

$$K_{\Lambda}(d\omega' \mid \omega) \,\mu_{\Lambda}(d\omega) \propto \exp \left\{ eta \left[ H_{\Lambda}^{\mathrm{Kad}}(\omega, \omega') + H_{\Lambda}(\omega) 
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ight\} d\omega'_{\Lambda} \, d\omega_{\Lambda}$$

where

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

acts on the original spins as an extra magnetic field

Constrained internal spins have phase transition if

$$\frac{p}{\beta}\omega'_x$$
 compensates  $h$  in the average (\*)

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Renormalization and non-Gibbsianness

# Single-site Kadanoff transformations (cont.)

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Let h be the original Ising field and fix  $\beta$  large enough s.t.

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

If h = 0, alternated  $\omega' \Longrightarrow (*)$  for  $p/\beta$  small enough Hence,  $\exists p_1 > p_2$  s.t.

- $\mu'$  is Gibbs for  $p > p_1$
- $\mu'$  is not Gibbs for  $_2 > p$

If  $h \neq 0, \exists \omega' \text{ s.t. } (*) \text{ only for a range of } p/\beta$ Hence,  $\exists p_1 \geq p_2 > p_3 \geq p_4$  s.t.

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Renormalization and non-Gibbsianness

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 Dynamics
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Renormalization and non-Gibbsianness

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 Dynamics
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Renormalization and non-Gibbsianness

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Renormalization and non-Gibbsianness

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Definition a	nd example					

 $\mu_t = S_t \mu$ 

with  $S_t$  = semigroup of operators. Dynamic G-non-G:  $\mu$  Gibbs but  $\mu_t$  non-Gibbs at some

Example:

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

$$S = \prod_{x} S_{\{x\}} \quad \text{with} \quad \begin{array}{l} S_x(\omega_x \mid \omega_x) = 1 - \epsilon \\ S_x(-\omega_x \mid \omega_x) = \epsilon \end{array}$$

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac
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Definition and example

#### Un-quenching G-non-G transitions

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In matrix form 
$$(S_x)_{\omega\omega'} \equiv S_x(\omega'_x \mid \omega_x)$$
  
 $S_x = \begin{pmatrix} 1-\epsilon & \epsilon \\ \epsilon & 1-\epsilon \end{pmatrix}$ ,  $S_x^n = \frac{1}{2} \begin{pmatrix} 1+a_n & 1-a_n \\ 1-a_n & 1+a_n \end{pmatrix}$   
with  $a_n = (1-2\epsilon)^n$ . Hence  
 $S_x^n(\omega' \mid \omega_x) = A_n e^{p_n \omega'_x \omega_x}$ ,  $p_n = \log\left(\frac{1+a_n}{2}\right)$ 

Kadanoff with  $p_n \xrightarrow{n \to 0} \infty$  and  $p_n \xrightarrow{n \to \infty} 0$ 

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac
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End

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Definition a	nd example					

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

Using previous results on Kadanoff-renormalized measures:

$$(h = 0) \qquad \underbrace{ \begin{array}{c} \text{Gibbs} \\ 0 \end{array}}_{n_1} \\ (h > 0) \qquad \underbrace{ \begin{array}{c} \text{Gibbs} \\ 0 \end{array}}_{n_1} \\ n_2 \end{array} \\ \underbrace{ \begin{array}{c} \text{Non-Gibbs} \\ n_3 \end{array}}_{n_3} \\ n_4 \end{array} \\ \underbrace{ \begin{array}{c} \text{Gibbs} \\ n_4 \end{array}}_{n_4} \\ \end{array} \\ \begin{array}{c} \text{Kon-Gibbs} \\ \text{Kon-Gibbs} \\ n_3 \\ n_4 \end{array} \\ \begin{array}{c} \text{Kon-Gibbs} \\ \ \text{Kon-Gibbs} \\ \text{Kon-Gibbs}$$

Mathematical mechanism: hidden variables (two-slice view) Physical mechanism?

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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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:
  - ▶  $\omega^{sp}$  nurtured, but  $\xi^{\pm}$  nature
  - ▶ Hence  $\xi^{\pm}$  determines original phase → discontinuity

▶ Long t: If  $h \neq 0$  only one phase  $\rightarrow$  no tilting mechanism

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Mechanism						

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* 

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- Single minimizer = Gibbsianness
- Multiple minimizers = non-Gibbsianness
   Perturbation of conditioning

   → discontinuous choice of trajectory

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Perturbation of conditioning

 $\rightarrow$  discontinuous choice of trajectory

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Mechanism						

#### Alternative paradigm: graphical summary

[h = 0]



One trajectory = Gibbs

Many trajectories = non-Gibbs

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Mean-field	models: Defin	nition				

## 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

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(i) Mean-field:

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

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Intro	Gibbs
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Non-Gibbs

Dynamics

Mean-field models: Definition

## Mean-field Ising model

N Ising spins  $(\omega_i \in \{-1, 1\})$ 

$$H^{N}(\sigma) = -\frac{J}{2N} \sum_{i,j=1}^{N} \sigma_{i}\sigma_{j} - h \sum_{i=1}^{N} \sigma_{i}$$
$$= N\overline{H}(m_{N}(\sigma))$$

where  $m_N$  is the *empirical magnetization* 

$$m_N(\sigma) = \frac{1}{N} \sum_{i=1}^{N} N\sigma_i$$

and, if  $m \in \mathcal{M}_N := \{-1, -1 + 2N^{-1}, \dots, +1 - 2N^{-1}, +1\},\$ 

$$\overline{H}(m) := -\frac{1}{2}Jm^2 - hm$$

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Non-Gibbs 000000 Dynamics 000000 Mean-field models: Definition

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N Ising spins  $(\omega_i \in \{-1, 1\})$ 

$$H^{N}(\sigma) = -\frac{J}{2N} \sum_{i,j=1}^{N} \sigma_{i}\sigma_{j} - h \sum_{i=1}^{N} \sigma_{i}$$
$$= N\overline{H}(m_{N}(\sigma))$$

where  $m_N$  is the empirical magnetization

$$m_N(\sigma) = \frac{1}{N} \sum_{i=1} N \sigma_i$$

and, if  $m \in \mathcal{M}_N := \{-1, -1 + 2N^{-1}, \dots, +1 - 2N^{-1}, +1\},\$ 

$$\overline{H}(m) := -\frac{1}{2}Jm^2 - hm$$

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Mean-field models: Definition

#### Mean-field measures and evolution

The  $H^N$ -Gibbs measure [ $\beta$  absorbed]

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

induces a measure on  $\mathcal{M}_N$ 

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

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

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Mean-field models: Definition

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Mean-field models: Definition

#### Mean-field evolution

Independent  $(T = \infty)$  dynamics on  $\Omega_N$  induces on  $\mathcal{M}_N$  a continuous-time Markov chain  $(m_t^N)_{t\geq 0}$  with generator

$$(\overline{L}_N f)(m) = \frac{1+m}{2} N [f(m-2N^{-1}) - f(m)]$$
  
 
$$+ \frac{1-m}{2} N [f(m+2N^{-1}) - f(m)]$$

This induces a dynamics on measures on  $\mathcal{M}_N$ 

$$\overline{\mu}_t^N(f) = \overline{\mu}^N \left( \mathrm{e}^{t\overline{L}_N} f \right)$$

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Mean-field models: Definition

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Intro	Gibbs
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Dynamics 000000 Mean field Kac End

Mean-field models: Non-Gibbsianness

#### Mean-field single-site specification (Külske and Le Ny)

Consider the single-spin conditional probabilities

$$\gamma_t^N(\sigma_1 \mid \alpha_{N-1}) := \mu_t^N(\sigma_1 \mid \sigma_{N-1}) ,$$

with

- $\sigma_1 \in \{-1, +1\},$
- $\bullet \ \alpha_{N-1} \in \mathcal{M}_{N-1},$

•  $\sigma_{N-1} \in \Omega_{N-1}$  any configuration s.t.  $m_{N-1}(\sigma_{N-1}) = \alpha_{N-1}$ [By permutation invariance RHS independ of choice of  $\sigma_{N-1}$ ]

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Mean-field models: Non-Gibbsianness

### Gibbs and Non-Gibbs mean-field models (Külske and Le Ny)

For fixed  $t \ge 0$ : (a) A magnetization  $\alpha \in [-1, 1]$  is good for  $\mu_t$  if

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

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

for  $\widetilde{\alpha}$  in a neighbourhood of  $\alpha$ 

(b) A magnetization α ∈ [-1, +1] is bad if it is not good
(c) μt is Gibbs if it has no bad magnetizations

Intro	Gibbs	Non-Gibbs
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Mean-field models: Non-Gibbsianness

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(c) μt is Gibbs if it has no bad magnetizations

Intro	Gibbs	Non-Gibbs
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Mean-field models: Large deviations

## Large-deviations: General definition Informally:

A family of measures  $(\nu^N)$  satisfies a large-deviation principle if

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

- $\blacktriangleright$  N is the LDP speed, I the rate function
- As a consequence,  $\operatorname{supp}(\nu^N) \to \operatorname{argmin}(I)$

#### Formally:

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

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

Intro	Gibbs	Non-Gibbs
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Mean-field models: Large deviations

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Mean-field models: Large deviations

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Mean-field models: Large deviations

### LDP for mean-field Ising:

#### "Static" part:

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

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

#### Independent evolutions:

Let  $P^N = \text{law of } (m_t^N)_{t \ge 0}$ Defined on the space of càdlàg trajectories: Skorohod

Intro	Gibbs	Non-Gibbs
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Mean-field models: Large deviations

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#### Independent evolutions:

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

Defined on the space of càdlàg trajectories; Skorohod topology

Intro	Gibbs	Non-Gibbs
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Mean field Kac End

Mean-field models: Large deviations

# LDP for mean-field evolutions

#### (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

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

Intro	Gibbs	Non-Gibbs
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Mean field Kac End

Mean-field models: Large deviations

## LDP for mean-field evolutions

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

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### LDP for conditioned mean-field evolutions

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The family of measures on trajectory space

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

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

$$I^{t,\alpha}(m\phi) = \begin{cases} I^t(\phi) & \text{if } \phi_t = \alpha \\ \infty & \text{otherwise} \end{cases}$$

Hence, conditioned optimal trajectories correspond to

 $\underset{\phi: \phi(t)=\alpha}{\operatorname{argmin}} I^t(\phi)$ 

Gibbs Non-Gibbs Dynamics Mean field Intro 

Mean-field models: Large deviations

### LDP for conditioned mean-field evolutions

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Hence, conditioned optimal trajectories correspond to

$$\operatorname{argmin}_{\phi: \phi(t)=\alpha} I^t(\phi)$$

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac
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Mean-field models: Results

# The mean-field computational advantage

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#### 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

$$\inf_{m \in [-1,+1]} C_{t,\alpha}(m) = \inf_{\phi: \phi(t) = \alpha} I^t(\phi)$$

Hence,

multiple conditioned trajectories  $\iff$  multiple global minima of  $C_{t,\alpha}(m)$ 

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Mean-field models: Results

# The mean-field computational advantage

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Intro	Gibbs	Non-Gibbs
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Mean-field models: Results

### I. Single optimal trajectory = Gibbsianness

- $\alpha \mapsto \gamma_t(\sigma \mid \alpha)$  is continuous at  $\alpha_0$  if and only if
  - ►  $I^t(\phi)$  has a unique minimizing path  $\widehat{\phi}$
  - ▶ or, equivalently,  $C_{t,\alpha_0}(m)$  has a unique minimizing m.

Furthermore, in this case, the specification kernel equals

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

 $p_t(\cdot, \cdot) =$  kernel of Markov jump process on  $\{-1, +1\}$  with  $\blacktriangleright$  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  $\gamma_t$  proven by Ermolaev and Külske

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Mean-field models: Results

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Mean-field models: Results

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Mean-field models: Results

### II. Short-term Gibbsianness

#### **Theorem** If $J \leq 1$ the evolved measures $\mu_t$ are Gibbs for all $t \geq 0$

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

Intro	Gibbs
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Mean-field models: Results

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#### Theorem

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- Note that  $1 = \beta_{cr}^{MF}$

Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End	
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Man-field models, Results							

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  $\mu_t$  is Gibbs
- $t > \Psi_c$ : Discontinuity at  $\alpha = 0$ ; two optimal trajectories  $\pm \phi$
- If  $\Lambda_{t,0}(J) = \text{cone between the trajectories } \pm \phi$ 
  - ▶ No trajectory can penetrate  $\Lambda_{t,0}(J)$
  - For  $J \leq 3/2$  the map  $t \mapsto \Lambda_{t,0}(J)$  is continuous
  - For J > 3/2 the map  $t \mapsto \Lambda_{t,0}(J)$  is continuous except at  $t = \Psi_c$  where it exhibits a right-continuous jump

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End		
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Moon-field models: Results								

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Moon-field models: Results								

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End		
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Moon-field models: Results								

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End	
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Moon-field models: Results							

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac
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Mean-field models: Results

### Graphic summary: $h = 0, \alpha = 0$

End



 $t < \Psi_c$   $t = \Psi_c$   $t > \Psi_c$ First row: Minimizing trajectories for (J,h) = (1.6,0)Second row: Corresponding plots of  $m \mapsto C_{t,0}(m)$ 

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End	
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#### Mean-field models: Results

### IV. Bad magnetizations as function of time



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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac
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#### Kac models: Definition

### Kac models: Basic definitions

- $\Delta_n^d := \mathbb{Z}^d / n \mathbb{Z}^d$  = the discrete torus of size n
- $\Omega_n := \{-1, +1\}^{\Delta_n^d} =$ Ising-spin configurations on  $\Delta_n^d$

► Kac-type Hamiltonian:

$$H^{n}(\sigma) := -\frac{1}{2n^{d}} \sum_{x, y \in \Delta_{n}^{d}} J\left(\frac{x-y}{n}\right) \, \sigma(x) \sigma(y) - \sum_{x \in \Delta_{n}^{d}} h(\frac{x}{n}) \, \sigma(x)$$

 $[J \ge 0 \text{ symmetric}]$ 

• Gibbs measure associated with  $H^n$ :

$$\mu^n(d\sigma) := \frac{e^{-\beta H^n(\sigma)}}{Z^n} \, d\sigma$$

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Kac models: Definition						

#### Continuum limit

•  $\mathbb{T}^d := \mathbb{R}^d / \mathbb{Z}^d$ , the *d*-dimensional unit torus

•  $\mathbb{T}_n^d := \frac{1}{n} \Delta_n^d = (1/n)$ -discretization of  $\mathbb{T}^d$ 

•  $\mathcal{M}(\mathbb{T}_n^d)$   $[\mathcal{M}(\mathbb{T}^d)]$  = signed measures on  $\mathbb{T}_n^d$   $[\mathbb{T}^d]$  (TV  $\leq 1$ )

The empirical density of  $\sigma \in \Omega_n$  inside  $\Lambda \subseteq \Delta_n^d$  is

 $\pi^n_\Lambda \colon \, \Omega_n o \mathcal{M}(\mathbb{T}^d_n) \subseteq \mathcal{M}(\mathbb{T}^d)$ 

$$\pi^n_{\Lambda}(\sigma) := \frac{1}{|\Lambda|} \sum_{x \in \Lambda} \sigma(x) \delta_{x/n}$$

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Intro	Gibbs	Non-Gibbs	Dynamics	Mean field	Kac	End
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Kac models	Definition					

### Induced objects. Profiles

Via  $\pi^n$  we define induced Gibbs measures on  $\mathcal{M}(\mathbb{T}_n^d)$ :

$$\check{\mu}^n = \mu^n \circ (\pi^n)^{-1}$$

and rewrite

$$H^n(\sigma) = n^d H(\pi^n(\sigma))$$

with

$$H(\nu) = -\left\langle \frac{1}{2}J * \nu + h, \nu \right\rangle$$

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A measure on  $\mathcal{M}(\mathbb{T}^d)$  of the form  $\alpha \lambda$  with

•  $\lambda$  Lebesgue measure

•  $\alpha \in B$  density function, with B=unit ball in  $L^{\infty}(\mathbb{T}^d)$ will be referred as a **profile**  $\alpha \in B$ 

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Kac models:	Definition					

### Single-site Kac specifications

Given

- A (continuum) site  $u \in \mathbb{T}^d$
- A probability measure  $\rho^n$  on  $\Omega_n$
- A measure  $\alpha_{n-1}^u \in \mathcal{M}(\mathbb{T}_n^d \setminus \lfloor nu \rfloor)$

The single-site conditional probability at site  $\lfloor nu \rfloor \in \mathbb{T}_n^d$  is

$$\gamma^{u,n}\big(\cdot \mid \alpha_{n-1}^u\big) := \rho^n\Big(\sigma(\lfloor nu \rfloor) = \cdot \mid \pi^{u,n}(\sigma) = \alpha_{n-1}^u\Big)$$

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#### Gibbs and no-Gibbs Kac measures

(a) A profile 
$$\alpha \in B$$
 is good for  $(\rho^n)$ 

$$\gamma^{u}(\cdot \mid \widetilde{\alpha}) := \lim_{n \to \infty} \gamma^{u,n}(\cdot \mid \alpha_{n-1}^{u})$$

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- ▶ exists and is independent of the sequence  $\alpha_{n-1}^u \to \tilde{\alpha}\lambda$
- it is continuous in α

for  $\widetilde{\alpha}$  in a neighbourhood of  $\alpha$ 

(b) A profile α ∈ B is bad if it is not good
(c) (ρ<sup>n</sup>) is Gibbs if it has no bad profiles

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Kac models: Definition

# LDP for ("static") Kac measures (Comets)

- $(\check{\mu}^n)$  satisfies an LDP with
  - ▶ speed  $n^d$
  - rate function  $I_S \inf_{\mathcal{M}(\mathbb{T}^d)} I_S$  with

$$I_{S}(\nu) := \begin{cases} -\beta \left\langle \frac{1}{2}J \ast \alpha + h, \alpha \lambda \right\rangle + \left\langle \Phi \circ \alpha, \lambda \right\rangle & \text{if } \nu = \alpha \lambda, \, \alpha \in B \\ \infty & \text{otherwise} \end{cases}$$

where  $\Phi$  is the relative entropy

$$\Phi(m) := \frac{1+m}{2} \log(1+m) + \frac{1-m}{2} \log(1-m), \qquad m \in [-1,+1].$$
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## LDP for conditioned Kac evolutions

Let

$$P^n := \text{ law of } (\pi^n_s)_{s \ge 0} \text{ conditional on } \pi^n_0 \sim \check{\mu}^n,$$

and

$$Q_{t,\alpha}^n(\cdot) := P^n\big((\pi_s^n)_{s \in [0,t]} \in \cdot \mid \pi_t^n = \alpha_n\big),$$

with  $\alpha_n \in \mathcal{M}^n$  the element closest to  $\alpha \lambda$ . Then

For  $t \ge 0$  and  $\alpha \in B$ ,  $(Q_{t,\alpha'}^n)_{n \in \mathbb{N}}$  satisfies an LDP with  $\blacktriangleright$  speed  $n^d$ 

• rate function 
$$I^{t,\alpha} - \inf_{D_{[0,t](\mathcal{M}(\mathbb{T}^d))}} I^{t,\alpha}$$
 with  
$$I^{t,\alpha}(\phi) := \begin{cases} I_S(\phi_0) + I_D^t(\phi) & \text{if } \phi_t \equiv \alpha\\ \infty, & \text{otherwise.} \end{cases}$$

with  $I_D^t(\phi)$  given by the integral of an explicit Lagrangian

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Kac models: Results

# **Results for Kac models**

### (A) For general Glauber dynamics:

#### Gibbsianness $\iff$ unique minimizing path

(B) For independent spin flips:
Let J := ∫<sub>T<sup>d</sup></sub> J(u)du, then
(i) Short-time Gibbs: ∃t<sub>0</sub> = t<sub>0</sub>(J, h) s.t. no bifurcation in [0, t<sub>0</sub>]
(ii) Mean-Field behaviour: If

$$h \equiv c \in [0, \infty)$$
 and  $\alpha' \equiv c' \in [-1, +1]$ 

then

bifurcation ~ MF with  $(J^{\rm MF}, h^{\rm MF}) = (\beta J, \beta c); \ \alpha = c'$ 

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Kac models: Results

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Kac models: Results

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Conclusions				5		

- ▶ New paradigm seems to work
- ▶ However: needs LDP in spaces of trajectories of measure

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▶ Practical consequences (numerics, other phenomena)?