
Dynamics of Glassy Systems :

an overview

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Experiments

Equilibrium

The compounds

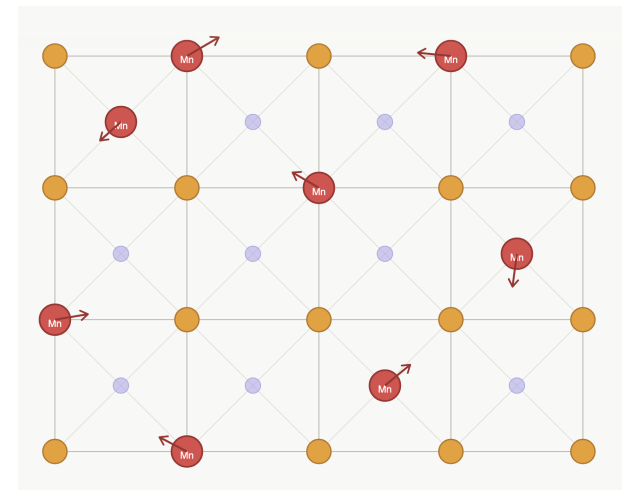
Dilute magnetic impurities in some non-magnetic host

Metallic spin glasses

- $\text{Cu}_{1-x}\text{Mn}_x$ the archetypal one
- $\text{Au}_{1-x}\text{Fe}_x$ extensively studied
- $\text{Ag}_{1-x}\text{Mn}_x$ similar to CuMn

Insulating spin glasses

- $\text{Eu}_x\text{Sr}_{1-x}$ a semiconductor
- $\text{Fe}_{1-x}\text{Mn}_x\text{TiO}_3$ a frustrated insulator



How much spin-glass material is there in physics labs ? Grams.

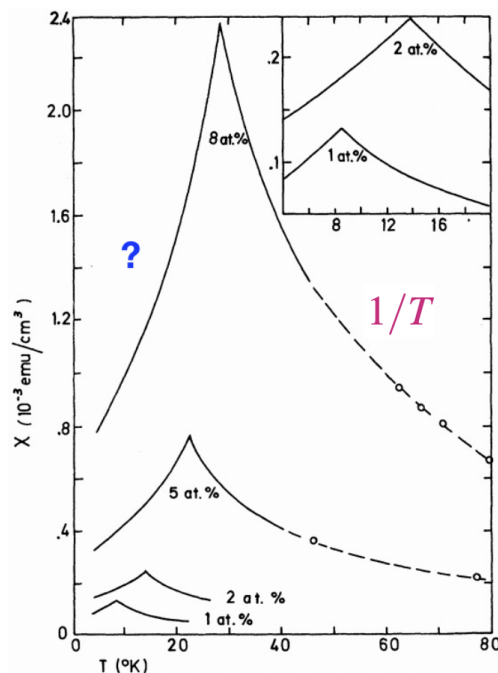
Are they useful materials ? **No**

Have they been inspirational ? **So much so - Fundamental research**

The anomaly

The experiment & the phase transition - 70s

Magnetic susceptibility



Temperature

Canella & Mydosh 1972

The cusp indicates a phase transition towards **which phase?**

Hypothesis : **canonical equilibrium**

for simplicity, Ising spins $s_i = \pm 1$ or \uparrow, \downarrow

partition function $Z = \sum_{\text{conf}} e^{-\beta \mathcal{H}}$

linear susceptibility $\chi = \left. \frac{\partial m_h}{\partial h} \right|_{h=0}$

magnetization density $m_h = N^{-1} \sum_i \langle s_i \rangle_h$

average with Z_h and $\mathcal{H}_h = \mathcal{H} - h \sum_i s_i$

$\langle s_i \rangle_h = \sum_{\text{conf}} s_i e^{-\beta \mathcal{H}_h} / Z_h$

Taking the derivative explicitly & $h \rightarrow 0$:

Fluctuation-dissipation theorem :

$$\chi = \beta (N^{-1} \sum_i \langle s_i^2 \rangle - (N^{-1} \sum_i \langle s_i \rangle)^2)$$

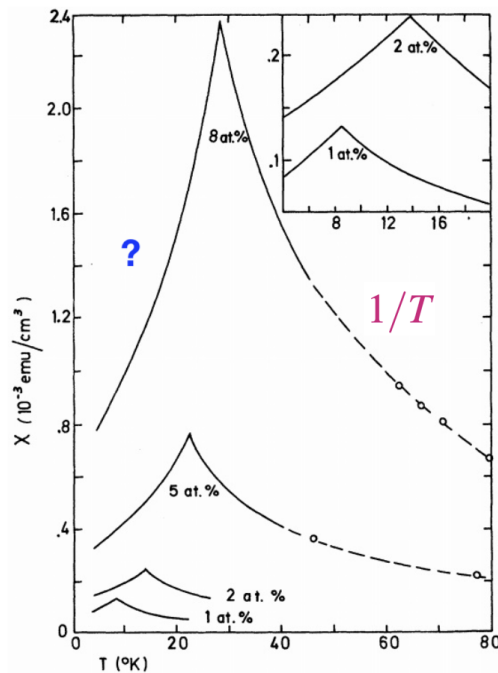
Thus, if $\langle s_i \rangle = 0$ and $s_i = \pm 1$

$$\chi = \beta = 1/T \quad \text{Curie law}$$

The anomaly

The experiment & the phase transition - 70s

Magnetic susceptibility



Temperature

Canella & Mydosh 1972

The cusp indicates a phase transition towards **which phase ?**

Hypothesis : **canonical equilibrium**

$$\text{FDT } \chi = \beta(1 - N^{-1} \sum_i \langle s_i \rangle^2)$$

The low T phase cannot be ferro since

$$m = N^{-1} \sum_{i=1}^N \langle s_i \rangle = 0 \text{ measured}$$

but $\langle s_i \rangle$ could be $\neq 0$ for each i

decrease in χ implies

$$q_{\text{EA}} \equiv \frac{1}{N} \sum_i \langle s_i \rangle^2 \begin{cases} = 0 & T \geq T_c \\ \neq 0 & T < T_c \end{cases}$$

EA order parameter

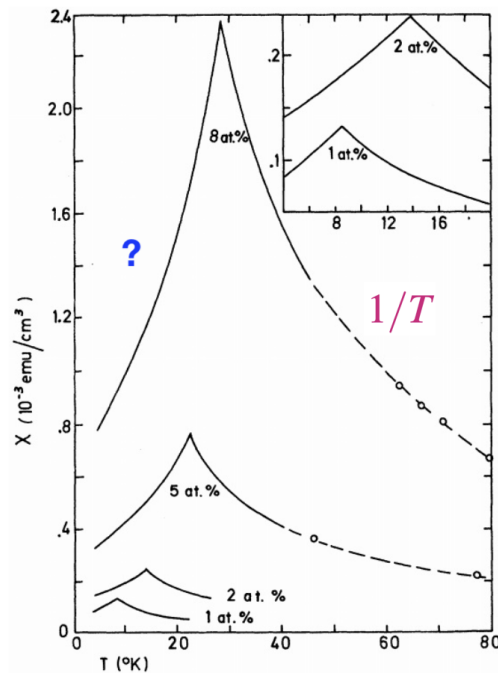
Theory

Models & analysis

Edwards-Anderson Model

The experiment, the paper & the authors - 70s

Magnetic susceptibility



Temperature

Canella & Mydosh 1972

The cusp indicates a phase transition towards **which phase?**

J. Phys. F: Metal Phys., Vol. 5, May 1975. Printed in Great Britain. © 1975.

Theory of spin glasses

S F Edwards[†] and P W Anderson[‡]
Cavendish Laboratory, Cambridge, UK

Received 14 October 1974, in final form 13 February 1975



Edwards-Anderson Model

Disordered low temperature equilibrium

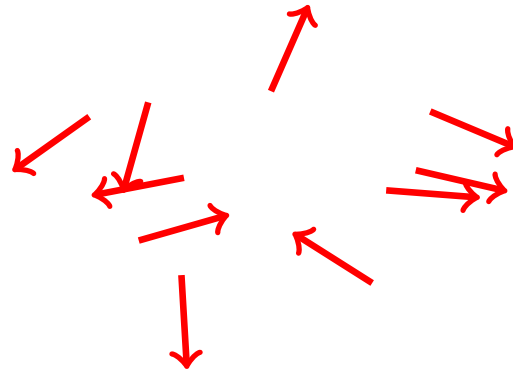
Theory of spin glasses

S F Edwards† and P W Anderson‡

Cavendish Laboratory, Cambridge, UK

Received 14 October 1974, in final form 13 February 1975

Abstract. A new theory of the class of dilute magnetic alloys, called the spin glasses, is proposed which offers a simple explanation of the cusps found experimentally in the susceptibility. The argument is that because the interaction between the spins dissolved in the matrix oscillates in sign according to distance, there will be no mean ferro- or antiferromagnetism, but there will be a ground state with the spins aligned in definite directions, even if these directions appear to be at random. At the critical temperature, the existence of these preferred directions affects the orientation of the spins, leading to a cusp in the susceptibility. This cusp is smoothed by an external field. If the potential



$$\mathcal{H} = -\frac{1}{2} \sum_{i \neq j} J_{ij} \vec{s}_i \cdot \vec{s}_j$$

with

$$J_{ij} = \frac{F(2k_F r_{ij})}{r_{ij}^3} \quad \text{RKKY}$$

Edwards-Anderson Model

A minimal model

$$\mathcal{H} = -\frac{1}{2} \sum_{\langle ij \rangle} J_{ij} s_i s_j$$

$s_i = \pm 1$ Ising spins - the variables

from magnetic moments to bimodal variables

$\langle ij \rangle$ nearest-neighbours on a cubic lattice

from random positions to the vertices of a regular lattice

J_{ij} quenched random, taken from a probability distribution - the couplings

$P[\{J_{ij}\}]$ bimodal, Gaussian, or other with no fat tails, with

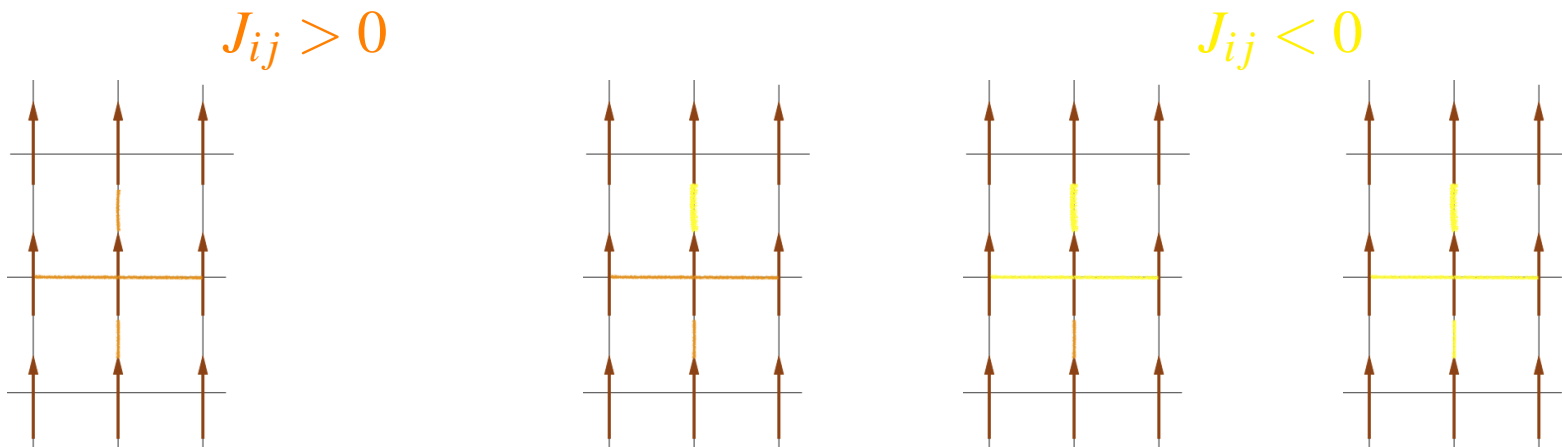
$$[J_{ij}] = 0 \text{ and } [J_{ij}^2] = J^2$$

from RKKY to first neighbour random interactions

Properties

Heterogeneity

Each variable, spin or other, feels a different local field, $h_i = \sum_{j=1}^z J_{ij} s_j$, contrary to what happens in a ferromagnetic sample, for instance.



Homogeneous

$$h_i = 4J \quad \forall i$$

Heterogeneous

$$h_j = 2J \quad h_k = -2J \quad h_l = -4J.$$

Each sample is *a priori* different but,

do they all have a different thermodynamic and dynamic behavior ?

Low temperature phases

Phenomenology : homogeneity vs inhomogeneity

In a **ferromagnet in equilibrium** at temperature $T < T_c$, $\langle s_i \rangle = m(T) \forall i$ or $\langle s_i \rangle = -m(T) \forall i$ in the two homogeneous, symmetric and degenerate equilibrium states

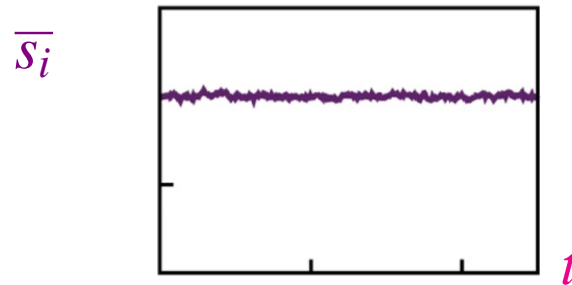
Low temperature phases

Phenomenology : homogeneity vs inhomogeneity

In a **ferromagnet in equilibrium** at temperature $T < T_c$, $\langle s_i \rangle = m(T) \forall i$ or $\langle s_i \rangle = -m(T) \forall i$ in the two homogeneous, symmetric and degenerate equilibrium states

If one were to follow the time evolution of each spin in one of the two equilibrium states at $T < T_c$, one would see $\overline{s_i}(t) = m(T) + \delta_i(t)$ with $\delta_i(t)$ a small time-dependent fluctuation and the overline for a running time average

$$\overline{s_i}(t) = \tau^{-1} \int_t^{t+\tau} dt' s_i(t')$$



Low temperature phases

Phenomenology : homogeneity vs inhomogeneity

In a **spin-glass in equilibrium** at temperature $T < T_c$, one expects $\langle s_i \rangle = m_i(T)$, with a different value for each i , in each inhomogeneous and degenerate equilibrium state.

There may be many different ensembles $\{m_i(T)\}$ that are equilibrium states (degeneracy, similar to what we saw in the frustrated magnets for the ground states but here in the full low T phase)

There is also the up-down symmetry $\{m_i(T)\} \mapsto \{-m_i(T)\}$

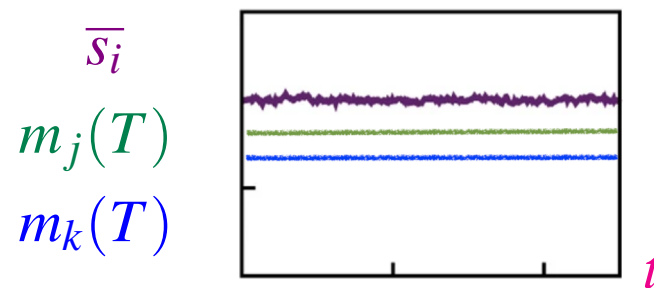
Low temperature phases

Phenomenology : homogeneity vs inhomogeneity

In a **spin-glass in equilibrium** at temperature $T < T_c$, one expects $\langle s_i \rangle = m_i(T)$, with a different value for each i , in each inhomogeneous and degenerate equilibrium state.

If one were to follow the time evolution of each spin in one of the possibly many equilibrium states at $T < T_c$, one would see $\bar{s}_i(t) = m_i(T) + \delta_i(t)$ with $\delta_i(t)$ a small time-dependent fluctuation and the overline, a running time average

$$\bar{s}_i(t) = \tau^{-1} \int_t^{t+\tau} dt' s_i(t')$$



Edwards-Anderson Model

Dynamic order parameter

Theory of spin glasses

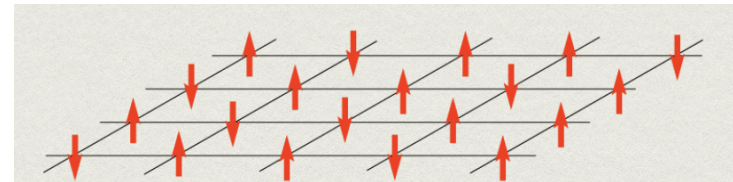
S F Edward† and P W Anderson‡
Cavendish Laboratory, Cambridge, UK

Received 14 October 1974, in final form 13 February 1975

viewed later will be the *same* random coil. Thus what we must argue is that if on one observation a particular spin is $s_i^{(1)}$ then if it is studied again a long time later, there is a nonvanishing probability that $s_i^{(2)}$ will point in the same direction, ie

$$q = \langle s_i^{(1)} \cdot s_i^{(2)} \rangle \neq 0.$$

$$q_{EA} = \lim_{t \gg t'} \lim_{N \rightarrow \infty} \langle s_i(t) s_i(t') \rangle \neq 0$$



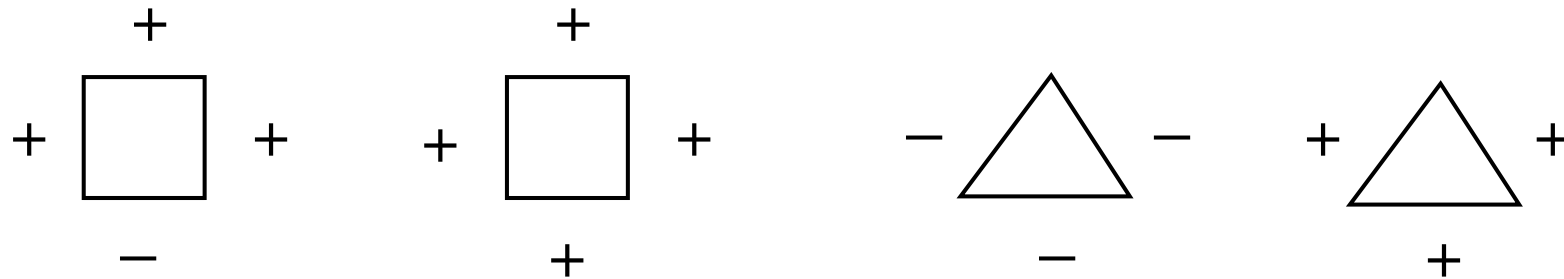
Spatially disordered but dynamically frozen, on average

$$\chi = \beta(1 - q_{EA})$$

Properties

Frustration

$$H_J[\{s\}] = -\sum_{\langle ij \rangle} J_{ij} s_i s_j \quad \text{Ising model}$$



Disordered

Geometric

$$E_{GS}^{\text{frust}} > E_{GS}^{\text{FM}} \quad \text{and} \quad S_{GS}^{\text{frust}} > S_{GS}^{\text{FM}}$$

Frustration enhances the **ground-state** energy and degeneracy (entropy)

One can expect to have many **metastable states** too

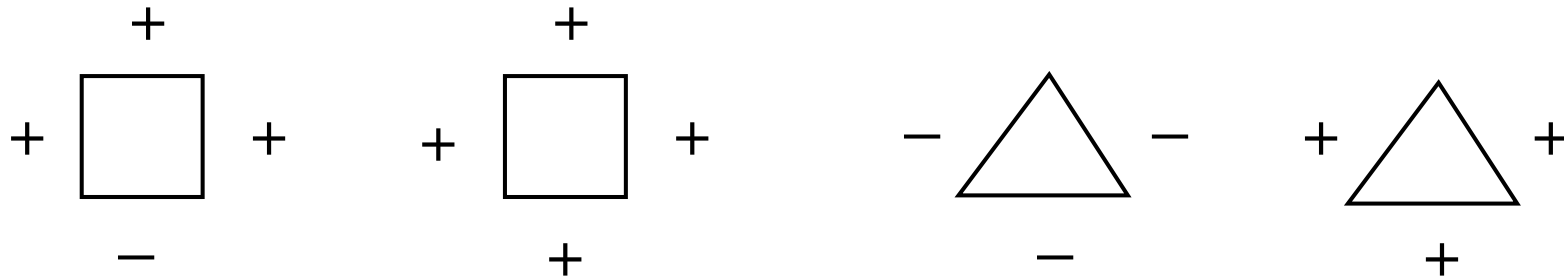
One cannot satisfy all couplings simultaneously if

$$\prod_{\text{loop}} J_{ij} < 0$$

Properties

Frustration

$$H_J[\{s\}] = -\sum_{\langle ij \rangle} J_{ij} s_i s_j \quad \text{Ising model}$$



Disordered

$$E_{GS}^{\text{frust}} > E_{GS}^{\text{FM}}$$

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Geometric

$$S_{GS}^{\text{frust}} > S_{GS}^{\text{FM}}$$

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One cannot satisfy all couplings simultaneously if

$$\prod_{\text{loop}} J_{ij} < 0$$

Properties

Self-averageness

Given a quantity A_J , which depends on the quenched randomness J_{ij} , it is distributed according to

$$P(A) = \int \prod_{ij} dJ_{ij} P(\{J_{ij}\}) \delta(A - A_J)$$

For “additive” quantities, $P(A)$ expected to be more peaked as $N \rightarrow \infty$

Therefore, one will observe $A = A_{\text{typ}}$ such that $P(A_{\text{typ}}) = \max_A P(A)$

However, it is difficult to calculate A_{typ}

What about calculating $[A] = \int \prod_{ij} dJ_{ij} P(\{J_{ij}\}) A$?

Self-averageness

$$[A] = A_{\text{typ}}$$

Sherrington-Kirkpatrick Model

The mean-field extension - all to all couplings - 70s

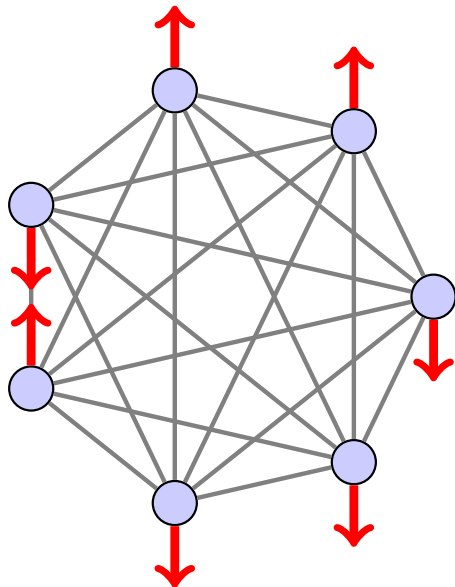
Solvable Model of a Spin-Glass

David Sherrington* and Scott Kirkpatrick

IBM Thomas J. Watson Research Center, Yorktown Heights, New York 10598

(Received 16 October 1975)

We consider an Ising model in which the spins are coupled by infinite-ranged random interactions independently distributed with a Gaussian probability density. Both “spin-glass” and ferromagnetic phases occur. The competition between the phases and the type of order present in each are studied.



$$\mathcal{H} = -\frac{1}{2} \sum_{i \neq j} J_{ij} s_i s_j$$

$$[J_{ij}] = 0 \quad [J_{ij}^2] = \frac{J^2}{N}$$

Sherrington-Kirkpatrick Model

Self-averageness & the Replica Method

Take a fully-connected ising spin $\{s_i\}$, $i = 1, \dots, N$ model with quenched random interactions J_{ij} drawn from a probability distribution $P(\{J_{ij}\})$

$$\mathcal{H} = - \sum_{i \neq j} J_{ij} s_i s_j$$

In the $N \rightarrow \infty$ limit, disorder averaged & typical free-energy densities, coincide

$$f \stackrel{N \rightarrow \infty}{=} [f] = - \frac{k_B T}{N} [\ln Z]$$

self-averageness

The disorder average can be evaluated with the help of the **replica trick** which

uses the identity $x^n = \exp(n \ln x)$ Taylor expanded around $n = 0$

$$x^n \stackrel{n \rightarrow 0}{=} 1 + n \ln x + O(n^2) \quad \Rightarrow$$

$$[\ln Z] \stackrel{n \rightarrow 0}{=} \frac{[Z^n] - 1}{n}$$

Sherrington-Kirkpatrick Model

The replica method - a sketch

$$-\beta[f] = \lim_{N \rightarrow \infty} \frac{[\ln Z]}{N} = \lim_{N \rightarrow \infty} \lim_{n \rightarrow 0} \frac{[Z^n] - 1}{Nn}$$

Z^n is the partition function of n independent copies of the system: the **replicas**

Gaussian $P(\{J_{ij}\})$ average over disorder $\Rightarrow a, b$ **replica coupling**

$$\sum_a \sum_{i \neq j} J_{ij} s_i^a s_j^a \Rightarrow \sum_{i \neq j} \sum_{ab} s_i^a s_j^a s_i^b s_j^b$$

Quadratic decoupling with the Hubbard-Stratonovich (Gaussian) trick

$$Q_{ab} \sum_i s_i^a s_i^b + \frac{1}{2} Q_{ab}^2$$

The elements of Q_{ab} can be evaluated by saddle-point if one exchanges

$$\lim_{N \rightarrow \infty} \lim_{n \rightarrow 0} \dots \mapsto \lim_{n \rightarrow 0} \lim_{N \rightarrow \infty} \dots$$

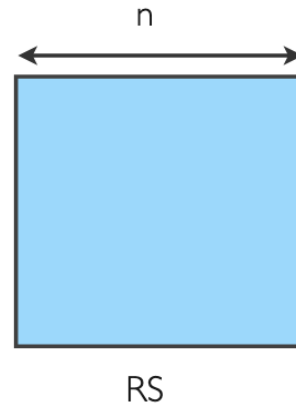
Q_{ab}^* is a 0×0 matrix but it admits an interpretation in terms of **overlaps**

$$NQ_{ab}^* = \sum_i [\langle s_i^a s_i^b \rangle]$$

Sherrington-Kirkpatrick Model

The structure of the matrix $Q_{ab}^* = [\langle s_a s_b \rangle]$

Replica Symmetry



Sherrington & Kirkpatrick, *Solvable model of a spin-glass*, PRL 35, 1792 (1975)

but $S(T = 0) < 0$ and the saddle-point $Q_{ab}^* = q$ is not stable

de Almeida & Thouless, *Stability of the Sherrington-Kirkpatrick solution of a spin glass model*, J. Phys. A : Math. Gen. **11**, 983 (1978)

Fig. from **Morone, Caltagirone, Harrison & Parisi**, *Replica Theory and Spin Glasses*,
Les Houches 2013

Sherrington-Kirkpatrick Model

Parisi's solution - Replica Symmetry Breaking Nobel 2021

J. Phys. A: Math. Gen. **13** (1980) L115-L121. Printed in Great Britain

LETTER TO THE EDITOR

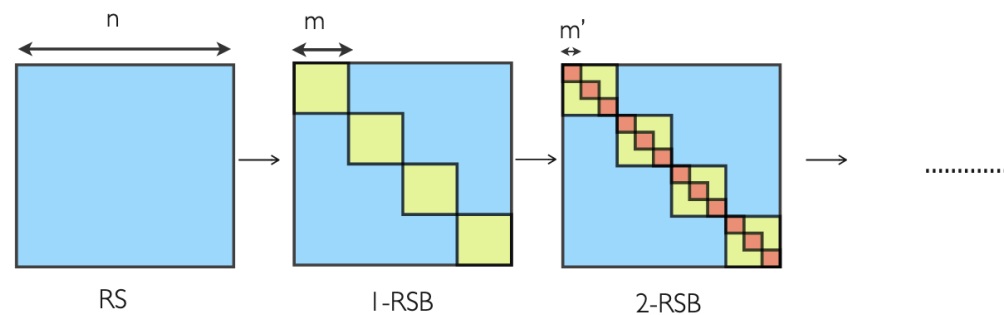
A sequence of approximated solutions to the S-K model for spin glasses

G Parisi

Istituto Nazionale de Fisica Nucleare, Laboratori Nazionali di Frascati, Casella Postale 13, 0004 Frascati, Roma, Italy

Received 4 January 1980

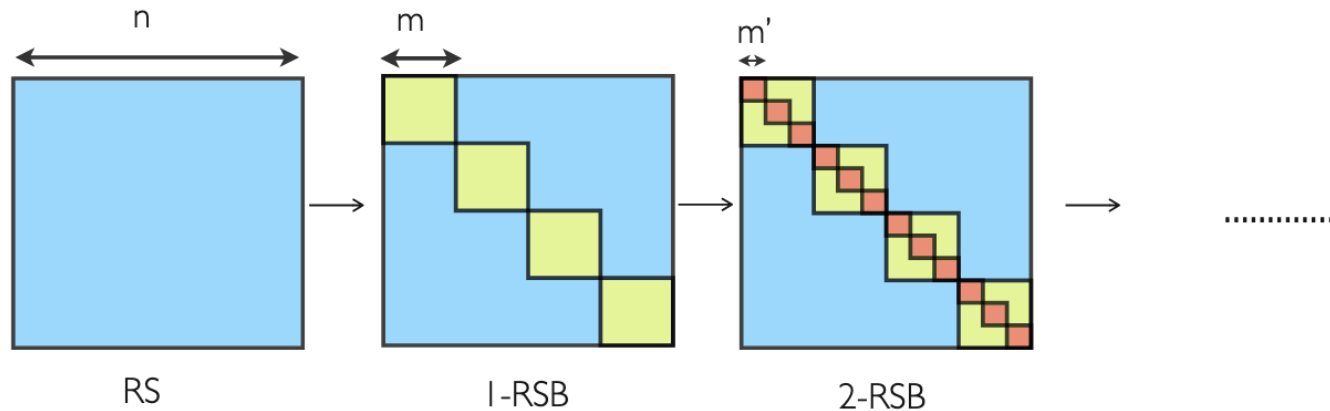
Abstract. In the framework of the new version of the replica theory, we compute a sequence of approximated solutions to the Sherrington-Kirkpatrick model of spin glasses.



Sherrington-Kirkpatrick Model

The structure of the overlap matrix $Q_{ab}^* = [\langle s_a s_b \rangle]$

Replica Symmetry Breaking



Now $S(T = 0) = 0$ and the saddle-point Q_{ab} is stable

Sherrington-Kirkpatrick Model

Multiplicity of equilibrium states

Sample the equilibrium states

at fixed realization of J_{ij}

Get the configurations $\{s_i^a\}$

calculate the **overlap/correlation**

$$q = N^{-1} \sum_i \langle s_i^{(a)} s_i^{(b)} \rangle$$

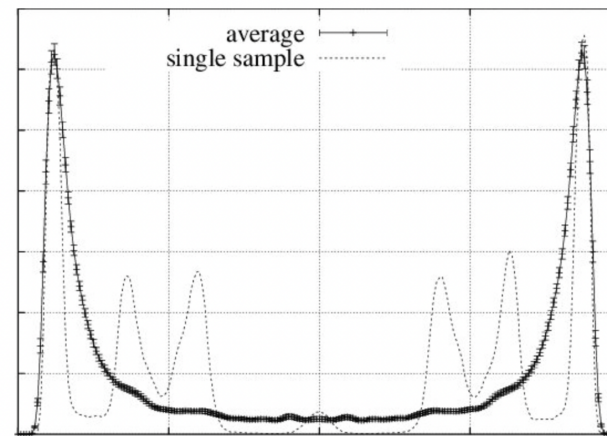
Result : $\pm q_{EA}$:

same state or reversed found

Result : $q \neq q_{EA}$

two different states sampled

P_J



q

Marinari, Martin & Zuliani, 2001

Real replicas

Bold curve, $[P_J](q)$

Sherrington-Kirkpatrick Model

Multiplicity of equilibrium states & ultrametricity

VOLUME 52, NUMBER 13

PHYSICAL REVIEW LETTERS

26 MARCH 1984

Nature of the Spin-Glass Phase

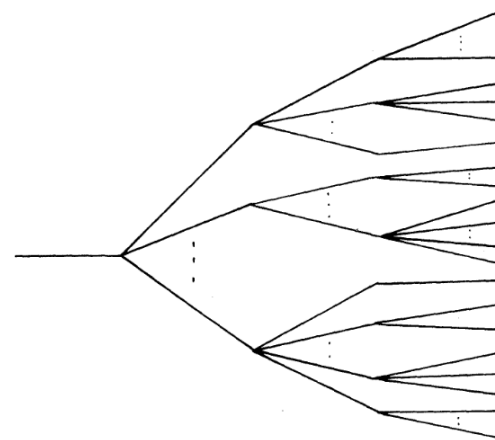
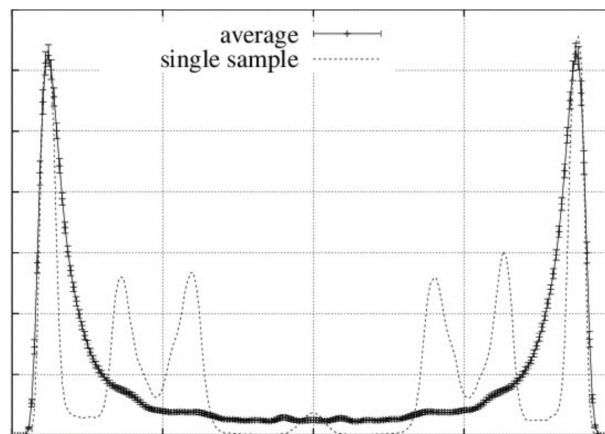
M. Mézard, G. Parisi,^(a) N. Sourlas, G. Toulouse,^(b) and M. Virasoro^(c)

Laboratoire de Physique Théorique de l'Ecole Normale Supérieure, F-75231 Paris-Cedex, France

(Received 12 December 1983)

A probability distribution has been proposed recently by one of us as an order parameter for spin-glasses. We show that this probability depends on the particular realization of the couplings even in the thermodynamic limit, and we study its distribution. We also show that the space of states has an ultrametric topology.

P_J



$$q = N^{-1} \sum_i \langle s_i^{(a)} s_i^{(b)} \rangle$$

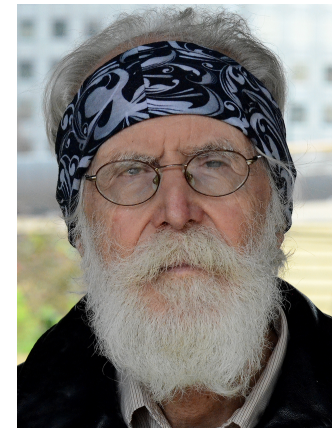
Weird organization of the equilibrium states

Sherrington-Kirkpatrick Model

Mathematics entered the game



Francesco Guerra 1942 - 2025



Michel Talagrand - Prix Abel 2024



$$f \geq f_{\text{FRSB}} \geq f \Rightarrow f = f_{\text{FRSB}}$$

Guerra 2003 Talagrand 2006

A recent picture in Paris

Talagrand & Parisi



Free-energy landscapes

Thouless, Anderson & Palmer - working at fixed disorder

PHILOSOPHICAL MAGAZINE, 1977, VOL. 35, No. 3, 593–601

Solution of ‘Solvable model of a spin glass’

By D. J. THOULESS

Department of Mathematical Physics, University of Birmingham,
Birmingham, England

and P. W. ANDERSON^{†‡} and R. G. PALMER

Department of Physics, Princeton University,
Princeton, New Jersey 08540, U.S.A.[†]

[Received 12 October 1976]

ABSTRACT

The Sherrington–Kirkpatrick model of a spin glass is solved by a mean field technique which is probably exact in the limit of infinite range interactions. At and above T_c the solution is identical to that obtained by Sherrington and Kirkpatrick (1975) using the $n \rightarrow 0$ replica method, but below T_c the new result exhibits several differences and remains physical down to $T = 0$.

$$\langle s_i \rangle = m_i = \tanh\{\beta \sum_{j(\neq i)} [J_{ij} m_j - \beta J_{ij}^2 (1 - m_j^2) m_i]\}$$

TAP equations

Orders of magnitude

The Thouless-Anderson-Palmer (TAP) equations read

$$m_i = \tanh \left\{ \sum_{j(\neq i)} [\beta J_{ij} m_j - \beta^2 m_i J_{ij}^2 (1 - m_j^2)] \right\}$$

Recall that $m_i = \langle s_i \rangle$

The first term in the rhs $\sum_{j(\neq i)} J_{ij} m_j \simeq \frac{1}{\sqrt{N}} \sqrt{N} = O(1)$ because of the central limit theorem.

The second term $\sum_{j(\neq i)} J_{ij}^2 (1 - m_j^2) \simeq \frac{1}{N} N = O(1)$ because all terms in the sum are positive definite ($m_j \leq 1 \forall j$)

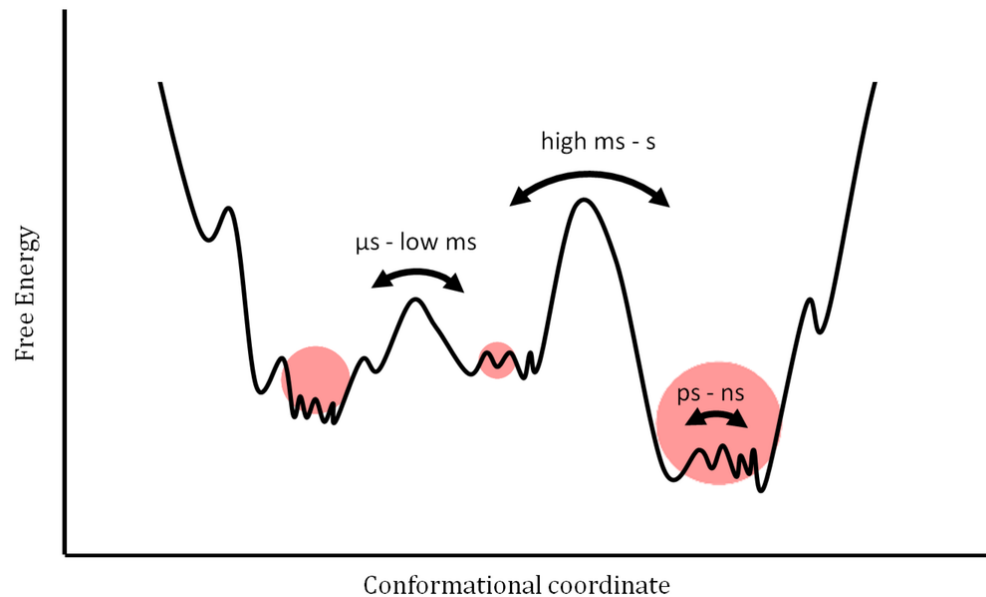
Onsager reaction term eliminates back reaction

Free-energy landscapes

Free-energy density at fixed randomness

The TAP equations are the extremization conditions on the *TAP free-energy*

$$F_J^{\text{tap}}(\{m_i\}) = -\frac{1}{2} \sum_{i \neq j} J_{ij} m_i m_j - \frac{\beta}{4} \sum_{i \neq j} J_{ij}^2 (1 - m_i^2)(1 - m_j^2) + T \sum_{i=1}^N \left[\frac{1 + m_i}{2} \ln \frac{1 + m_i}{2} + \frac{1 - m_i}{2} \ln \frac{1 - m_i}{2} \right]$$



At low temperatures

$\{m_i\}$

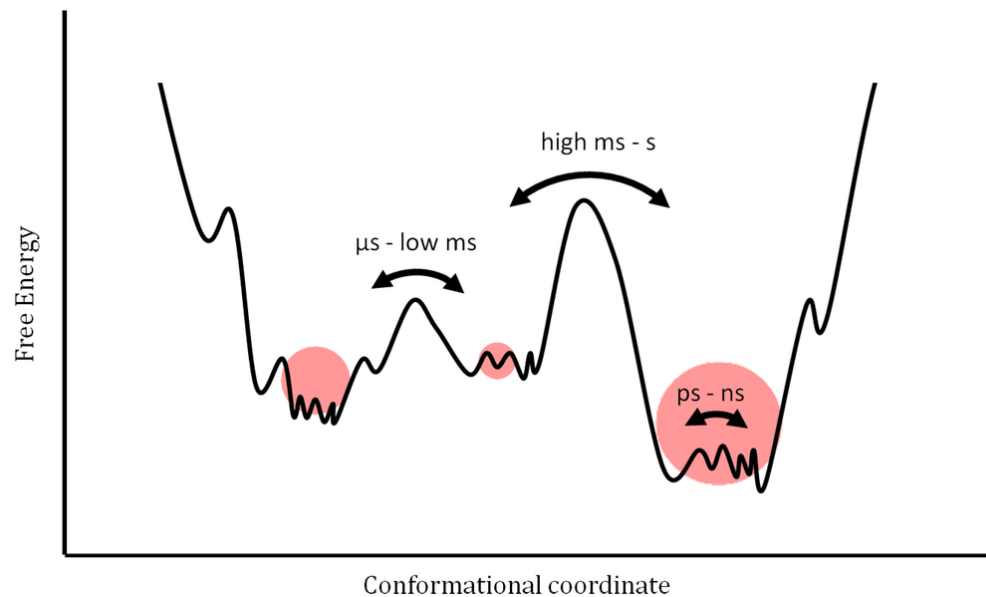
Free-energy landscapes

Free-energy density at fixed randomness

The TAP equations are the extremization conditions on the *TAP free-energy*

$$\frac{\delta F_J^{\text{tap}}(\{m_i\})}{\delta m_j} = 0$$

The stability of the solutions is determined by the Hessian $\frac{\delta^2 F_J^{\text{tap}}(\{m_i\})}{\delta m_j \delta m_k}$



At low temperatures

$\{m_i\}$

Free-energy landscapes

Free-energy density at fixed randomness

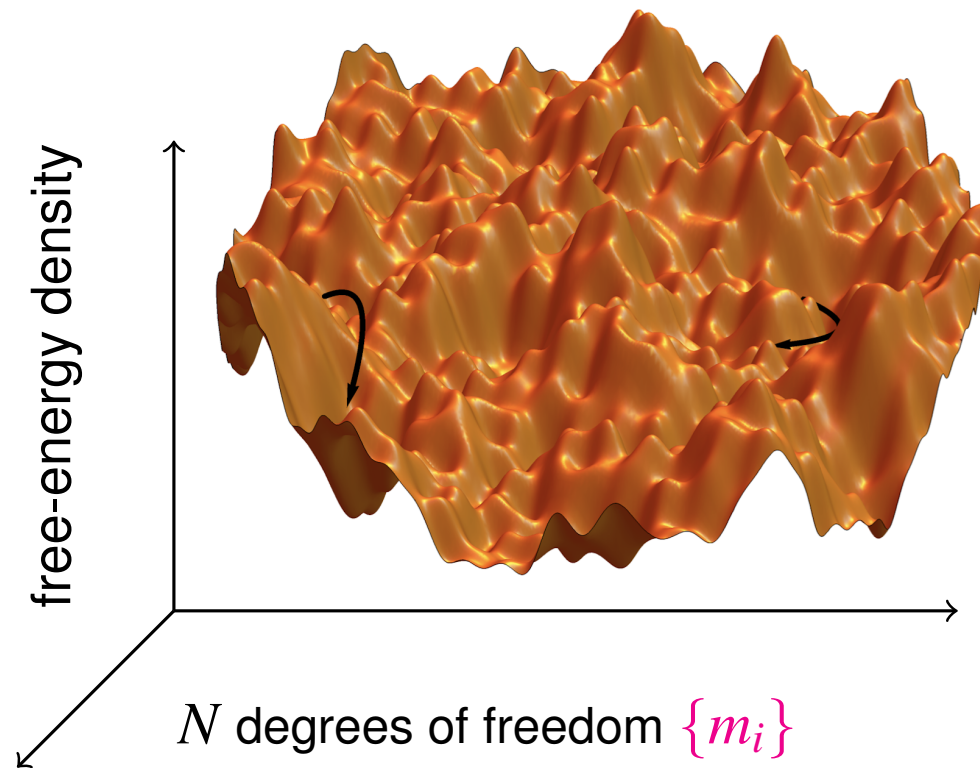


Figure adapted from a picture by **C. Cammarota**

For each $\{J_{ij}\}$ realization the landscape will be different
but the **statistical properties** will be the same

Bray & Moore, Cavagna, Giardinà & Parisi, Fyodorov & Ros, Kent-Dobias, etc.

Free-energy landscapes

Free-energy density at fixed randomness

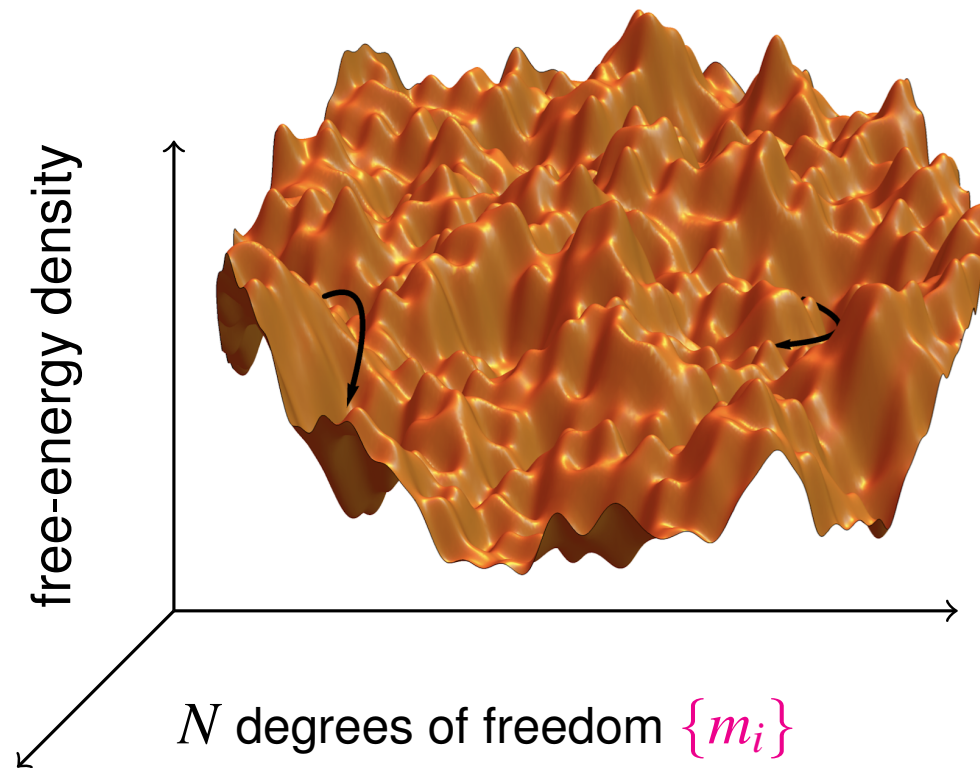


Figure adapted from a picture by **C. Cammarota**

Topography of the landscape on the N -dimensional substrate made by the N order parameters ?

Mean-field classes

According to equilibrium and metastability

$$\mathcal{H} = - \sum_{i_1 \neq \dots \neq i_p} J_{i_1 \dots i_p} s_{i_1} \dots s_{i_p} \quad p \text{ integer}$$

quenched random couplings $J_{i_1 \dots i_p}$ drawn from a Gaussian $P[\{J_{i_1 \dots i_p}\}]$
with $[J_{i_1 \dots i_p}] = 0$ and $[J_{i_1 \dots i_p}^2] = J^2 / N^{p-1}$

Choice of spin variables

Ising $s_i = \pm 1$

Globally spherical $\sum_{i=1}^N s_i^2 = N$ and $\mathcal{H} \rightarrow \mathcal{H} + \frac{z}{2} \left(\sum_{i=1}^N s_i^2 - N \right)$

$p = 2$ Ising Sherrington-Kirkpatrick **spin-glass** Model

FRSB

$p = 2$ spherical : ferromagnetic **coarsening** Berlin-Kac 52

RS

$p \geq 3$ Ising or spherical RFOT modelling of **glasses**

1RSB

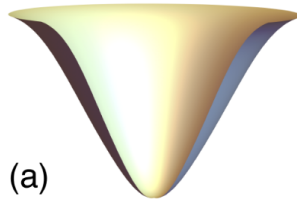
Kirkpatrick, Thirumalai & Wolynes 87-89

Mean-field classes

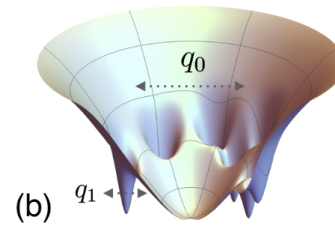
According to Replica Theory and TAP analysis

Equilibrium states

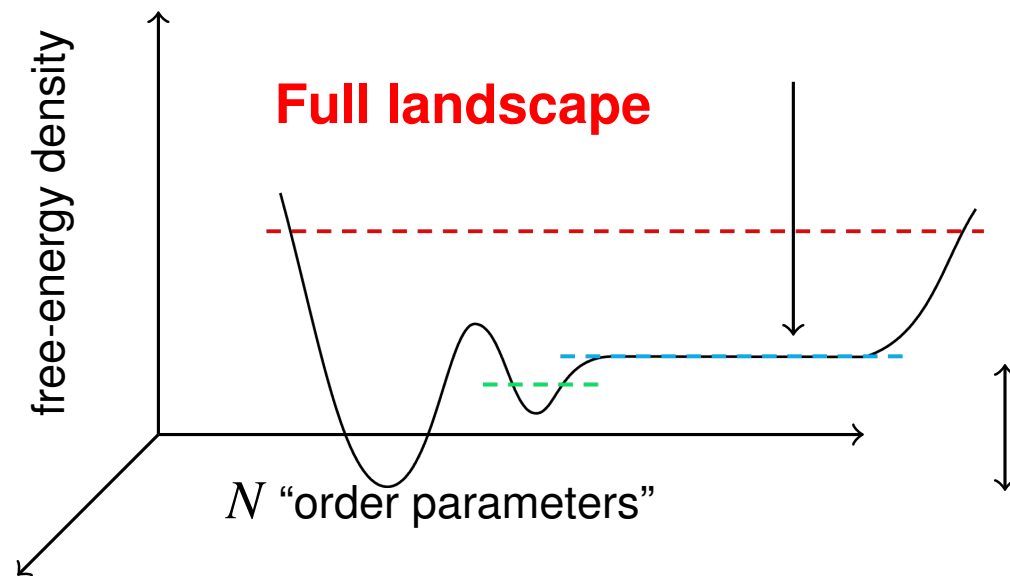
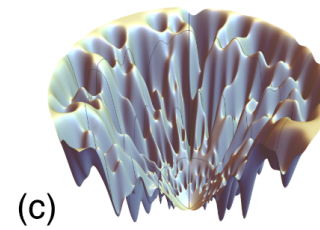
RS



1RSB



FRSB



Finite dimensions

Scaling Theory

Two ground states related by spin reversal \implies no replica symmetry breaking.

Compact clusters (droplets) in excitation above the ground state.

The surface of the droplets is fractal, with dimension $d_f > d - 1$.

A droplet of size L costs free-energy proportional to L^θ ,
where θ is the stiffness exponent.

Unstable in any magnetic field (no AT-line in finite dimensions).

The spin glass is like a magnet with a rough energy landscape but a single ordered phase – excitations cost more and more energy as they grow, unlike the RSB picture where infinitely many competing states exist at no cost.

The alternative view of **Bray & Moore, Fisher & Huse** late 80s

Issue not settled

More history

Texts & Interviews

“For posterity and future study, we are conducting a series of interviews with the original scientific participants and deposit them at the Centre d’Archives de Philosophie, d’Histoire et d’Édition des Sciences (CAPHÉS) of École normale supérieure de Paris (ENS)

<https://caphes.ens.fr/history-of-replica-symmetry-breaking-in-physics/>

The interviews are also assembled on a dedicated site

<https://historyrsb.nakala.fr/en>

Interestingly, two recent Nobel prize winners fall within the scope of this project :
Giorgio Parisi (Physics, 2021) and John Hopfield (Physics, 2024).”

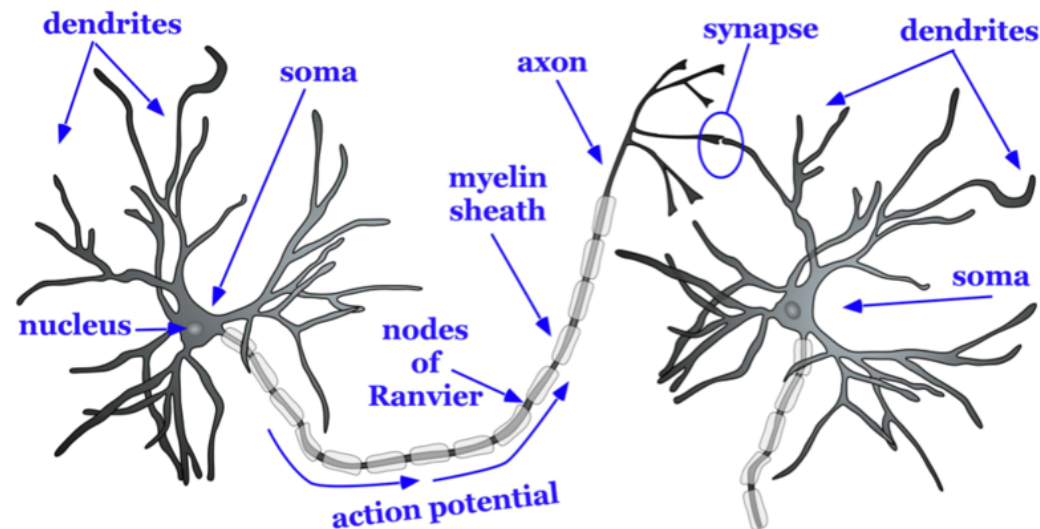
Charbonneau & Zamponi

Theory

Beyond : neural networks

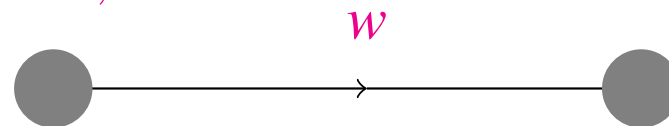
Neural networks

From biology to physics : simplest possible modelling



$$a = 1,0$$

$$a = 1,0$$

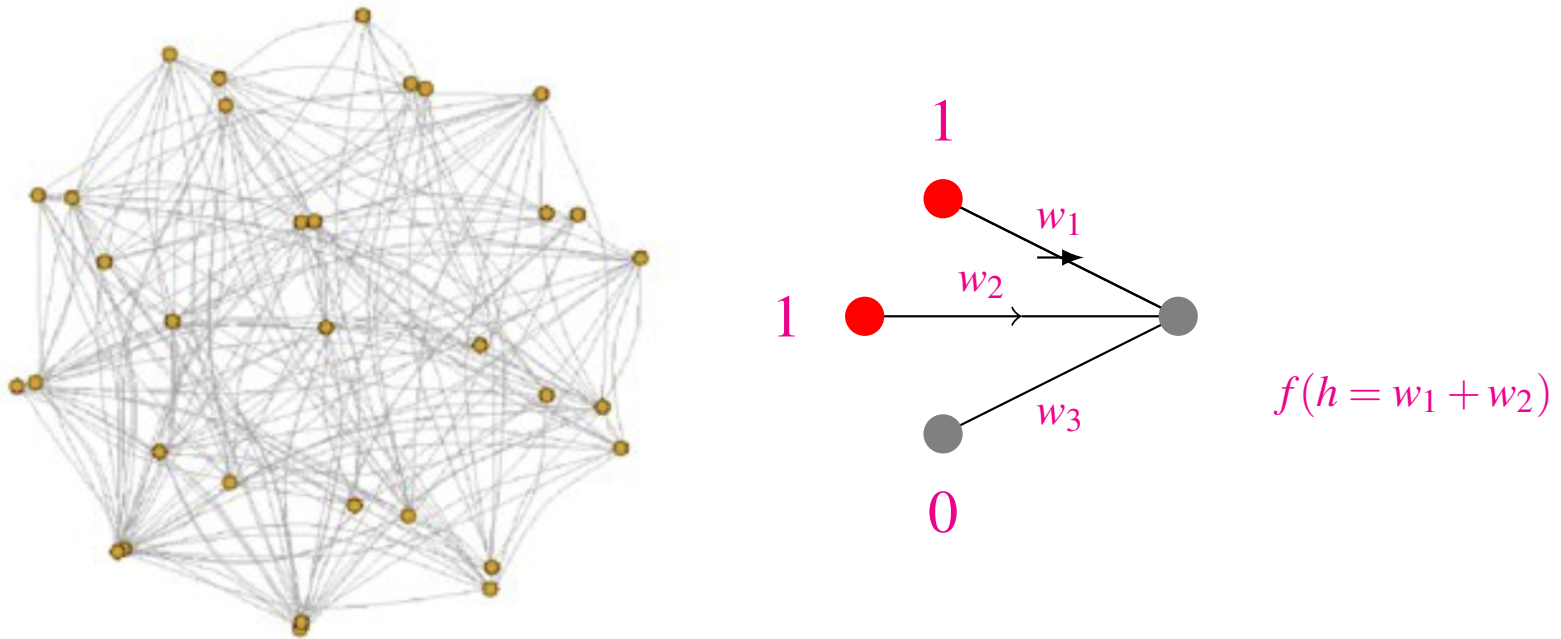


Ramón y Cajal, ca. 1890 **Nobel 1906** W. S. McCulloch & W. Pitts 43

(father of neuroscience working in Madrid, neurophysiologist, autodidact math & logic at Chicago)

Neural networks

Random graphs with weighted connections



At each instant, each neuron calculates the sum of the entries sent but its neighbors weighted by the strength of the synapses, $h_i = \sum_j w_{ij} a_j$, and it applies a function, f . if the result overcomes a threshold, $f(h_i) > \theta$, the receptor neuron spikes, $a_i = 1$.

So on and so forth on the whole network

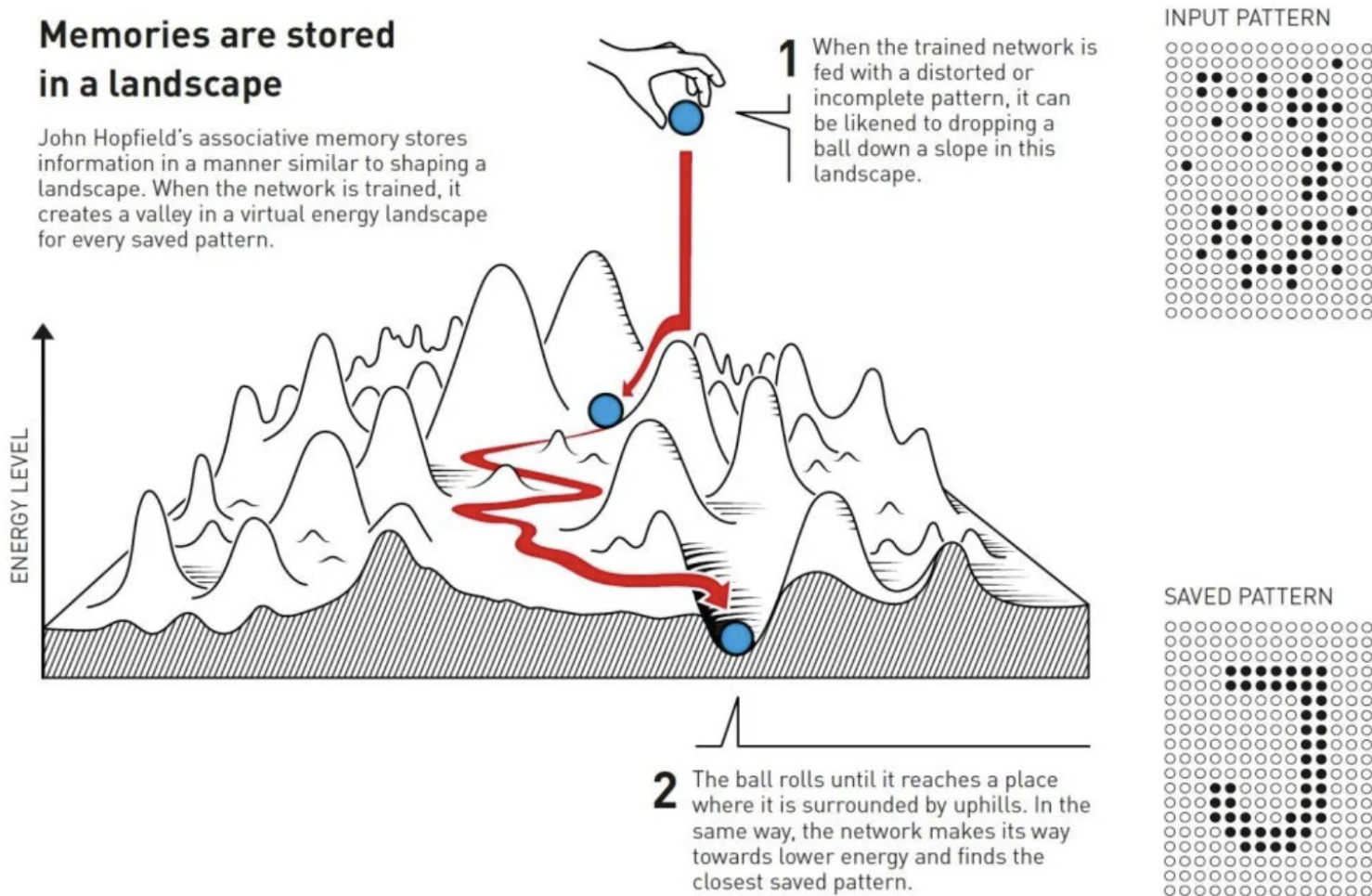
Details to make precise, e.g. parallel or random sequential dynamics

Hopfield Model

Associative memories : one chooses the synaptic weights w_{ij}

Memories are stored in a landscape

John Hopfield's associative memory stores information in a manner similar to shaping a landscape. When the network is trained, it creates a valley in a virtual energy landscape for every saved pattern.



Hopfield Model

Properties

- The **synapses** $\{w_{ij}\}$ shape the landscape & encode the memories **Hebb 49**
- **Basins of attraction** : input objects must sufficiently resemble the network's learned templates for successful recognition.
- The network has a limiting **capacity** : beyond a threshold

$$\left(\frac{\text{number of learnt objects}}{\text{number of neurons}} \right)_c$$

it fails **Phase Transition : order (recall possible) - disorder (failure)**

- The network builds **illegitimate valleys** which do not correspond to learnt objects

Amit, Gutfreund & Sompolinsky 85 maximal capacity of Hopfield-like networks

Elizabeth Gardner 88 Space of interactions in neural network models : all possible connection weights that could successfully store a given set of memories

with replica methods as developed by Parisi **Nobel 2021**

More history

Personal views

A field with early contributions from many Argentine physicists, in part thanks to **Miguel Virasoro's** (Roma Sapienza, ICTP & Buenos Aires) influence.

Néstor Parga (Bariloche \mapsto Madrid)

Roberto Perazzo (Buenos Aires)

Edgardo Ferrán (Buenos Aires \mapsto bioinformatics at Sanofi & others, France)

Walter Theumann (Porto Alegre)

Tato Moscato (La Plata \mapsto Newcastle, Australia)

After the early 90s : Neural networks & AI winter

Some people stayed **Geoffrey Hinton Nobel 2024**, **Yann LeCun**, etc.

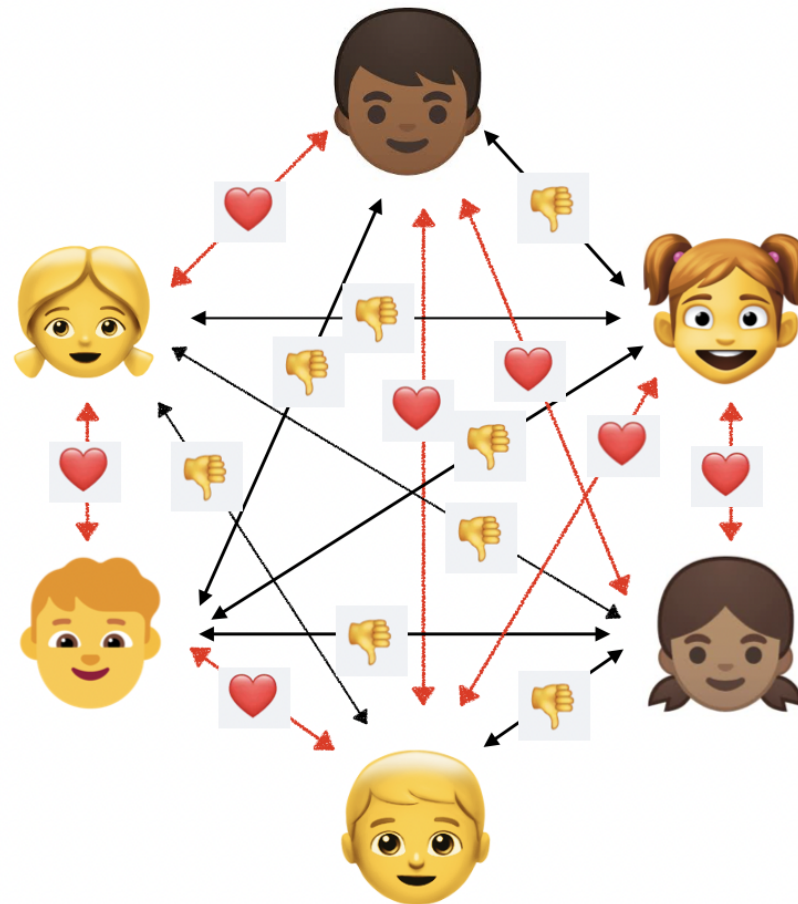
Huge amount of very important work by **Sara Solla** (Chicago, USA)

Theory

Beyond : random optimization

Random optimization

How do we split the group in two equal sized parts





and minimize frustration ?

Random optimization

Mapping to a statistical physics model

In the **graph partitioning - group splitting** example

- $i, j = 1, \dots, N$ label the persons.
- Predetermined $J_{ij} = -1$ for  love or $J_{ij} = 1$ for  hate feelings and $J_{ij} = 0$ for strangers
- $s_i = 1$ if i is in group A or $s_i = -1$ if i is in group B

find the assignment of all the s_i so that they **add up to zero** ($\sum_{i=1}^N s_i = 0$) & the

cost function



$\mathcal{H} =$ sum over all pairs of the love/hate values in the same group

is **minimized**

Random optimization

Mapping to a statistical physics model

In the **graph partitioning - group splitting** example

- $i, j = 1, \dots, N$ label the persons.
- Predetermined $J_{ij} = -1$ for  love or $J_{ij} = 1$ for  hate feelings
- $s_i = 1$ if i is in group A or $s_i = -1$ if i is in group B

find the assignment of all the s_i so that they **add up to zero** ($\sum_{i=1}^N s_i = 0$) & the

Cost function should be minimised like finding a ground state

$$\mathcal{H} = \underbrace{\sum_{i \neq j}}_{\text{sum over all pairs}} \underbrace{J_{ij}}_{\text{quenched love/hate}} \underbrace{\left(\frac{1 + s_i s_j}{2} \right)}_{\substack{\text{selects pairs in same group} \\ \text{vanishes if } i, j \text{ in different groups}}}$$

Random optimization

Mapping to a statistical physics model

Some successes of the statistical physics approach

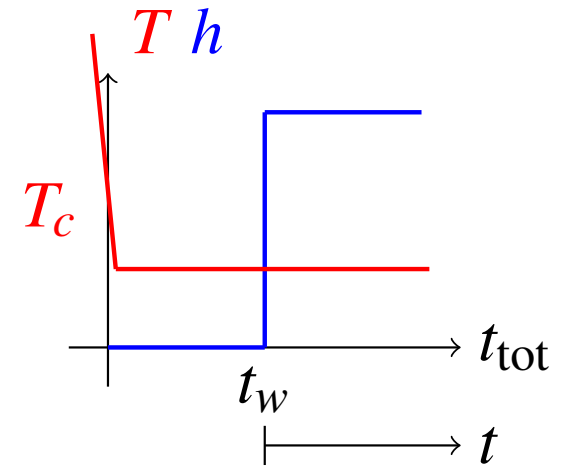
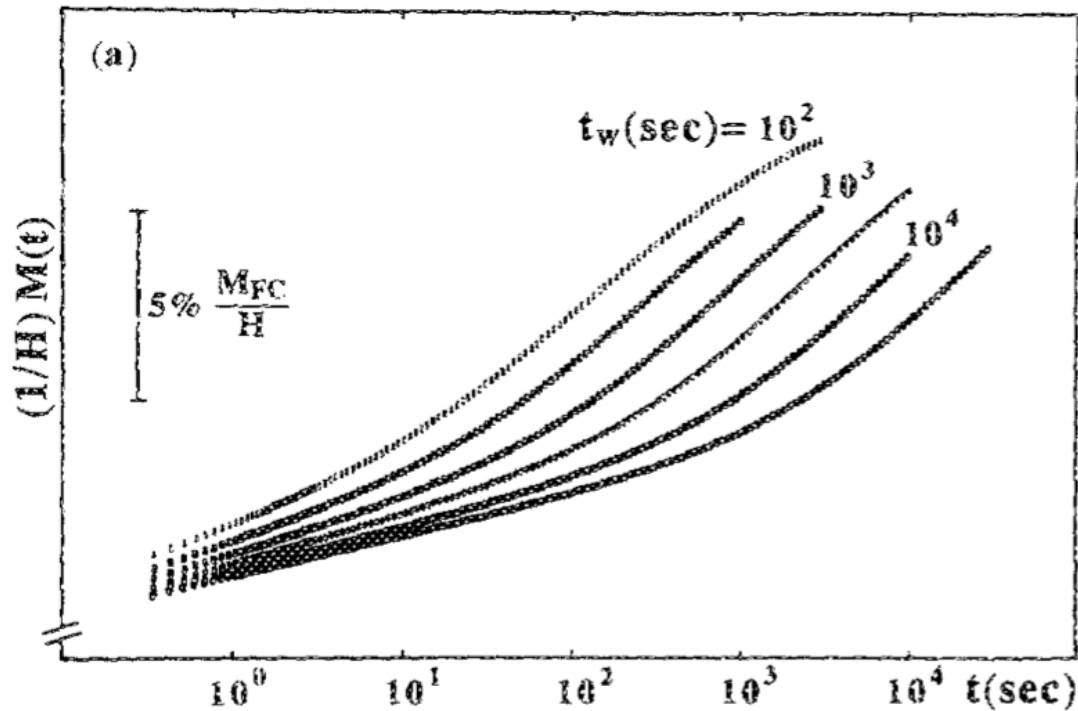
- **S. Kirkpatrick, C. D. Gelatt Jr and M. P. Vecchi**, “Optimization by **Simulated Annealing**”. Science 220, 671 (1983). **Dynamics**
- **R. Monasson, R. Zecchina, S. Kirkpatrick, B. Selman and L. Troyansky**, “Determining Computational Complexity from Characteristic **Phase Transitions**” Nature 400, 133 (1999)
- **J. S. Yedidia, W. T. Freeman and Y. Weiss**, “Understanding **Belief Propagation** and its Generalizations” Exploring artificial intelligence in the new millennium 8 (236-239), 0018-9448 (2003) **TAP**
- **M. Mézard, G. Parisi and R. Zecchina**, Science 297, 812 (2002) “Analytic and Algorithmic Solution of Random Satisfiability Problems” **TAP++**
Onsager Prize, APS, 2016

Back to SG experiments

Dynamics

Quenches

Zero field cooled magnetization - 80s

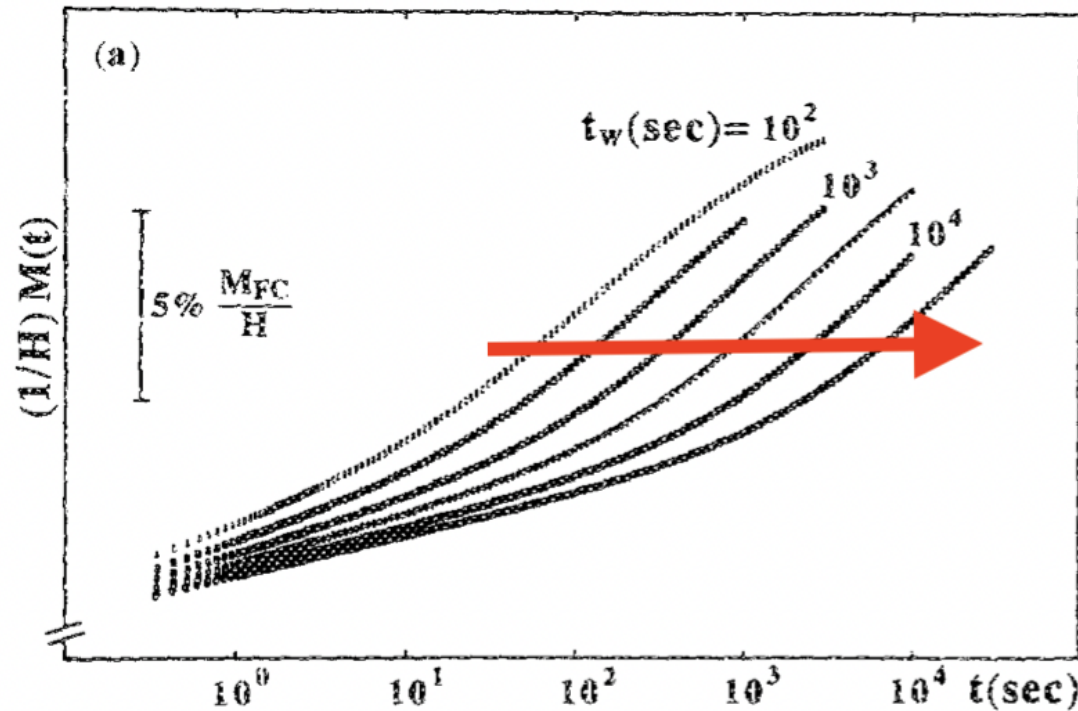


t_w are the waiting times, before the magnetic field is applied.

Aging

Breakdown of stationarity : out of equilibrium

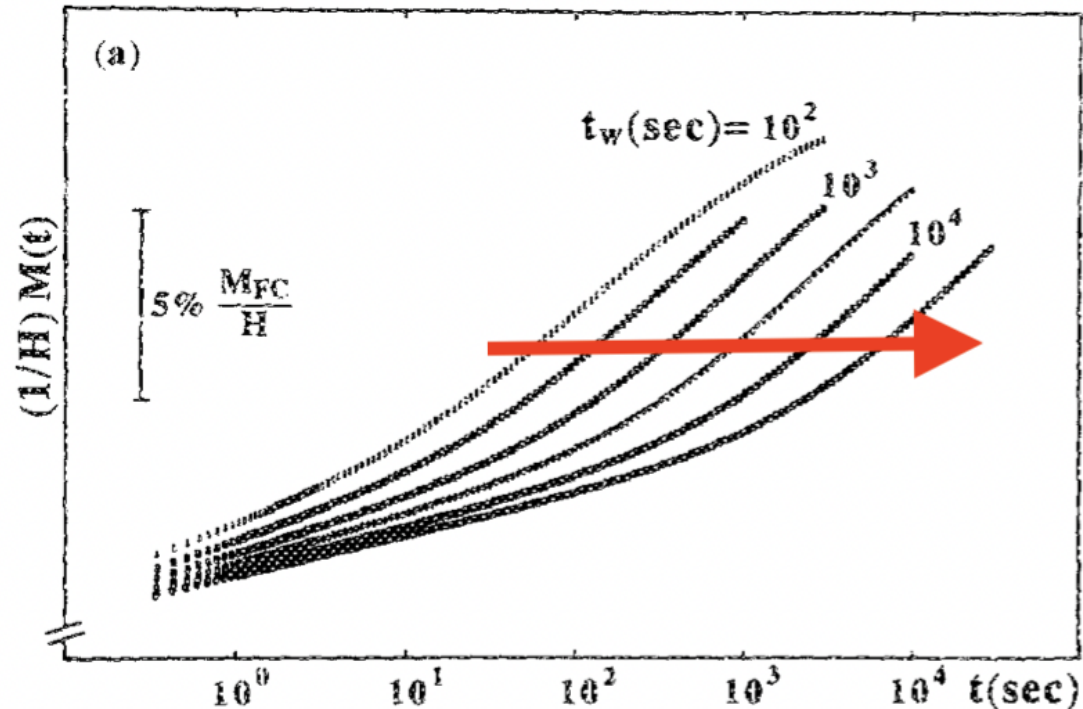
$$m(t, t_w) \neq m(t - t_w)$$



Older systems (longer t_w) relax more slowly than younger ones

Aging

Breakdown of stationarity : out of equilibrium



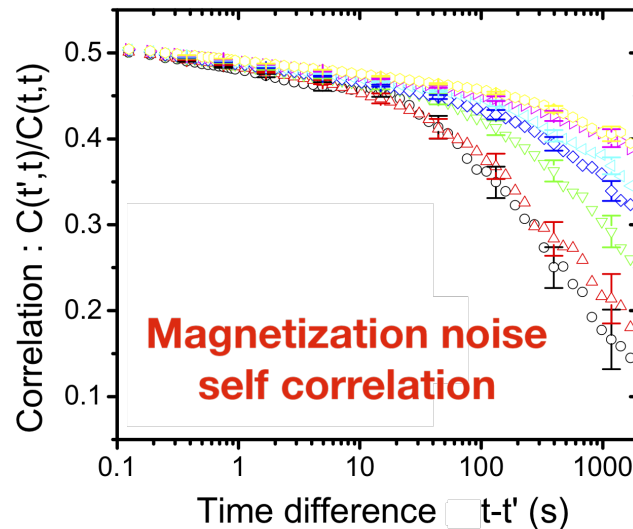
Older systems (longer t_w) relax more slowly than younger ones

Similar to previous measurements in all kinds of **polymer glasses**

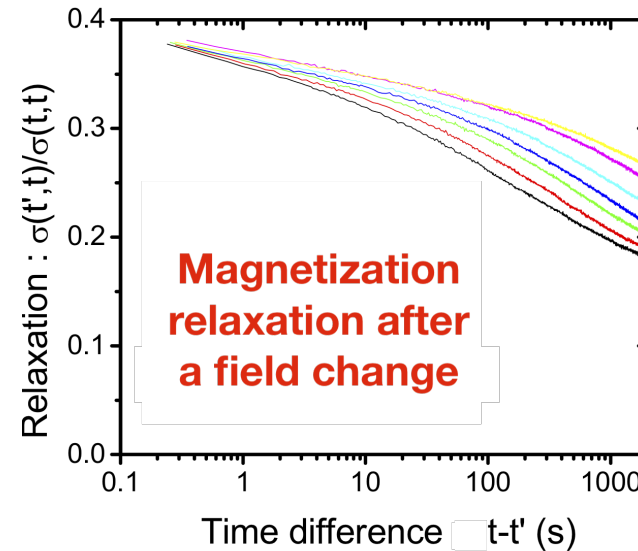
Kovacs, Struik, etc. 60s

Induced & spontaneous

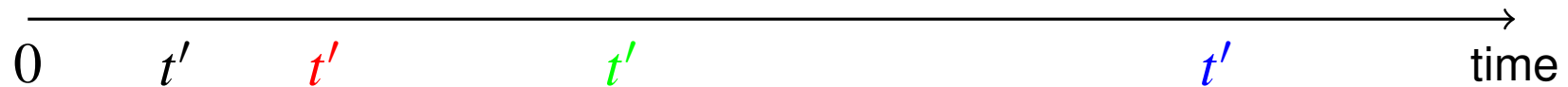
slow relaxation & loss of stationarity (aging)



Correlation



Linear response



Different curves are measured after log-spaced reference times t' after the quench : **breakdown of stationarity** \implies far from equilibrium

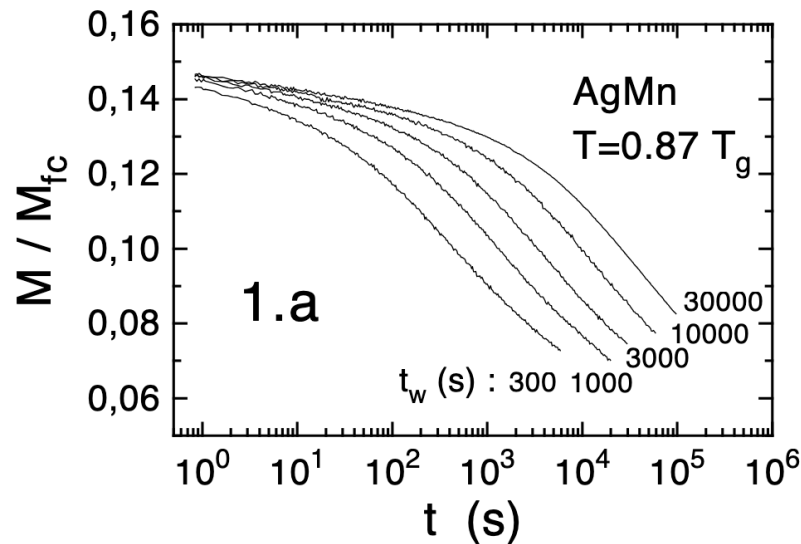
No identifiable growing length $\mathcal{R}(t)$: **glassy microscopic mechanisms ?**

Induced & spontaneous

thermo-remanent magnetization decay & self correlation

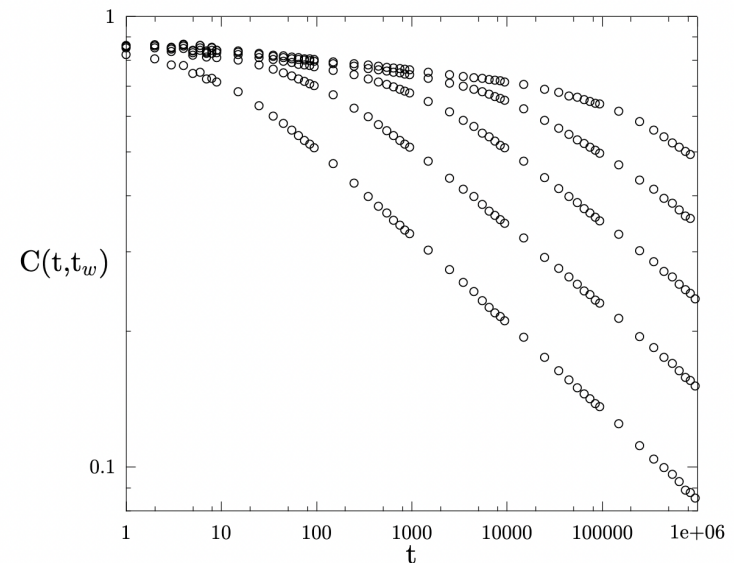
$$m(t_{tot}, t_w)$$

Experiments



$$C(t_{tot}, t_w) = N^{-1} \sum_i S_i(t_{tot}) S_i(t_w)$$

Monte Carlo 3D EA



In these plots, t_w are the waiting times when the magnetic field is switched off or the reference spin configuration is stored ; t is the time-difference

The equilibration time of macroscopic spin-glasses below T_c goes beyond the experimentally accessible time window t_{exp}

$$\lim_{N \gg 1} t_{\text{eq}}(N, T) \gg t_{\text{exp}} \quad T < T_c$$

t_{eq} grows with the system size N at low temperatures

They show **aging** both in magnetization (induced) and in correlation (spontaneous) measurements

Above T_c , in the paramagnetic phase, they equilibrate,

$$\lim_{N \gg 1} t_{\text{eq}}(N, T) < t_{\text{exp}} \quad T > T_c$$

The equilibration time of macroscopic

coarsening & glassy systems in a wide range of parameters,

e.g. low temperatures,

goes beyond the experimental window t_{exp}

$$\lim_{N \gg 1} t_{\text{eq}}(N, \text{parameters}) \gg t_{\text{exp}}$$

t_{eq} grows with the system size N at low temperatures

and they also show **aging** though with some differences, to be discussed

Theory

Explanations

Kinetic Ising Model

Archetypical example for classical magnetic systems

$$H = -J \sum_{\langle ij \rangle} s_i s_j$$

$s_i = \pm 1$ Ising spins.

$\langle ij \rangle$ sum over nearest-neighbours on the lattice.

$J > 0$ ferromagnetic coupling constant.

critical temperature $T_c > 0$ for $d > 1$.

Evolution, coupling to bath

Monte Carlo rule $s_i \rightarrow -s_i$ accepted with

$$p = 1 \quad \text{if } \Delta E < 0$$
$$p = e^{-\beta \Delta E} \quad \text{if } \Delta E > 0$$
$$p = 1/2 \quad \text{if } \Delta \mathcal{E} = 0$$

Non-conserved order parameter dynamics [$\uparrow\downarrow$ towards $\uparrow\uparrow$] etc. allowed.

[$m = 0$ to $m = 2$]

Scalar Field Theory

Summary : discrete vs. continuous

Ising spin models

$$H = -J \sum_{\langle ij \rangle} s_i s_j$$

NCOP [$\uparrow\downarrow \mapsto \uparrow\uparrow$]

COP [$\uparrow\downarrow \mapsto \downarrow\uparrow$]

Scalar field theories

$$\mathcal{F}[\phi] = \int d^d r \left[\frac{1}{2} (\nabla\phi)^2 - \frac{\mu}{2} \phi^2 + \frac{g}{4} \phi^4 \right]$$

$$\partial_t \phi(r, t) = -\delta_{\phi(r, t)} \mathcal{F}[\phi] + \xi(r, t)$$

$$\partial_t \phi(r, t) = -\nabla^2 \delta_{\phi(r, t)} \mathcal{F}[\phi] + \eta(r, t)$$

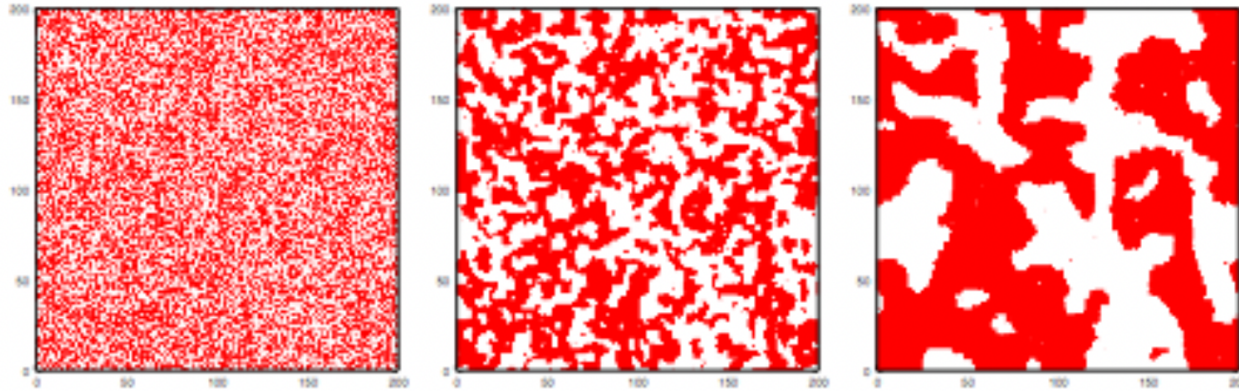
Overdamped limit is fine ; rescaling of time to eliminate γ_0

In the COP case $\langle \eta(x, t) \eta(y, t') \rangle = 2k_B T \nabla^2 \delta(x - y) \delta(t - t')$

Generalisations for vector cases. **Quenched disorder** can be introduced by taking the J_{ij} or the parameters in the field theory, e.g. μ , from a pdf.

Aging

One possible explanation : domain growth



Growth of the two degenerate ground states slowed down by quenched randomness

Dynamic scaling

Thermal activation arguments for a growing length $\mathcal{R} \sim (\ln t)^{1/\psi}$

Fisher & Huse, Bray & Moore, late 80s, for spin-glasses, general review Bray 94

Dynamic Scaling

Growth of 'red and white' patches separated by
interfaces which surround such geometric **domains**

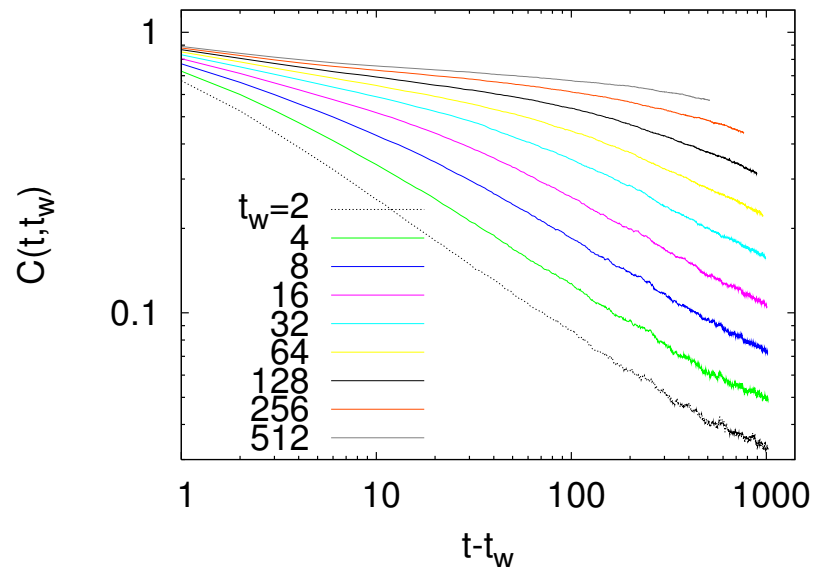
Spatial regions of local equilibrium (with non-vanishing order parameter, at $T < T_c$) grow in time

A single **growing length** $\mathcal{R}(t, T)$ can be identified and will be at the heart of dynamic scaling.

Here and in the following we measure T in units of J

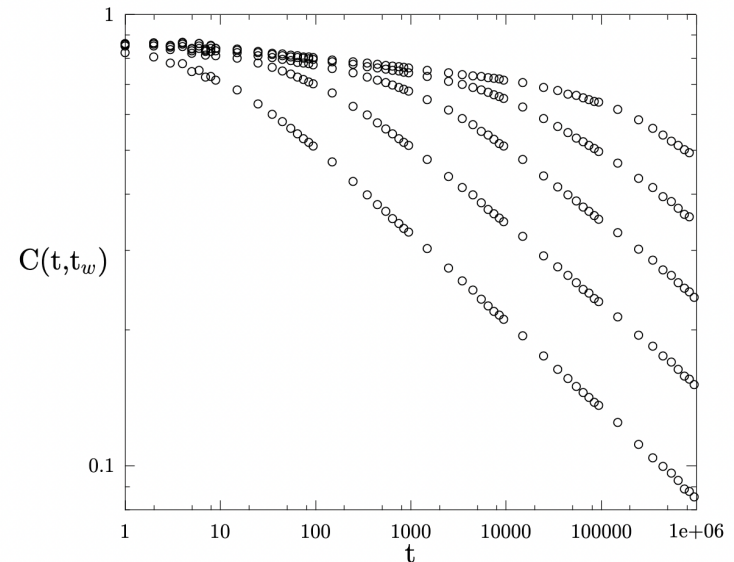
Ferromagnet vs spin-glass

Not so different as long as correlations are concerned



2D FM Ising model - spin-spin

Sicilia *et al.* 07



3D Edwards Anderson Model

Rieger 93

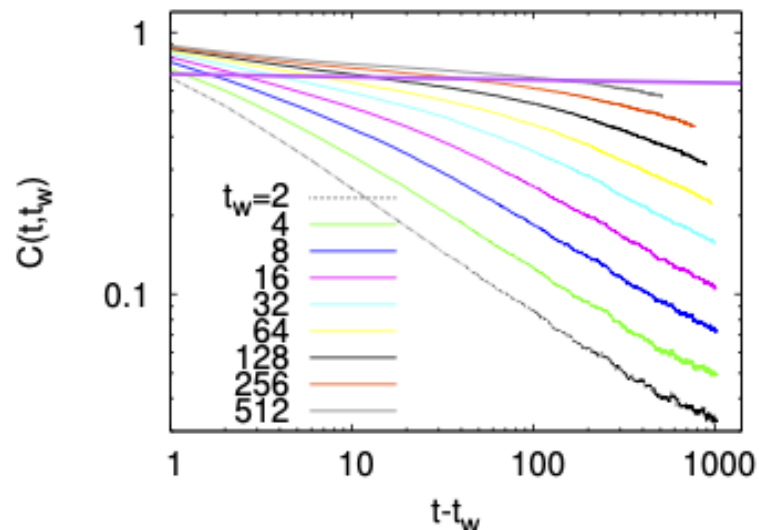
One correlation can exhibit stationary and non stationary relaxation

in different two-time regimes

Two-time self-correlation

e.g., MC simulation of the $2dIM$ at $T < T_c$

$$C(t, t_w) = N^{-1} \sum_{i=1}^N \langle s_i(t) s_i(t_w) \rangle$$



Stationary relaxation

$$m_{eq}^2$$

Aging decay

Separation of time-scales : stationary – aging

$$C(t, t_w) = C_{st}(t - t_w) + m_{eq}^2 f\left(\frac{\mathcal{R}(t, T)}{\mathcal{R}(t_w, T)}\right)$$

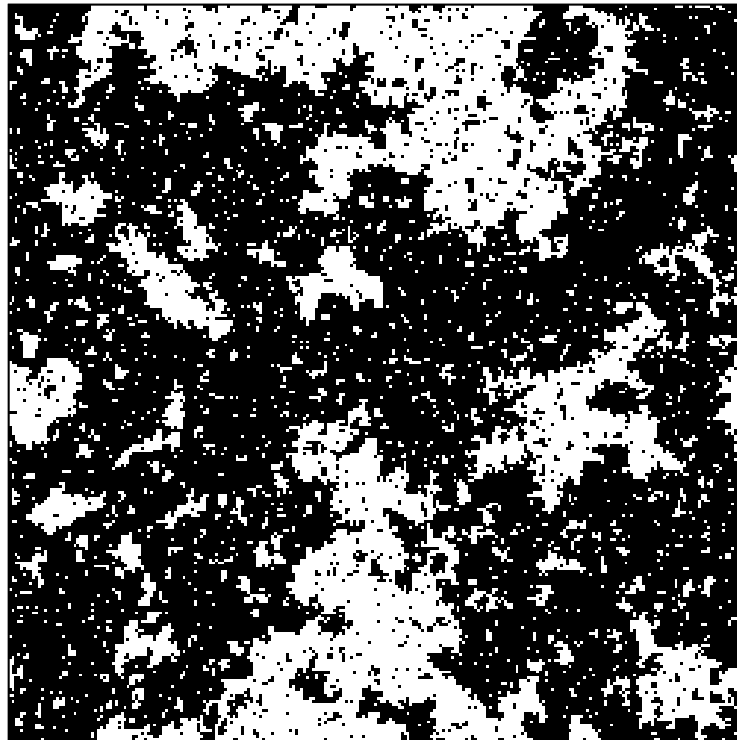
$$C_{st}(0) = 1 - m_{eq}^2, \quad \lim_{x \rightarrow \infty} C_{st}(x) = 0, \quad f(1) = 1, \quad \lim_{x \rightarrow \infty} f(x) = 0$$

Interpretation

Thermal fluctuations within domains & domain-wall motion

$$C_{\text{st}}(t - t_w)$$

$$m_{\text{eq}}^2 f(\mathcal{R}(t, T) / \mathcal{R}(t_w, T))$$

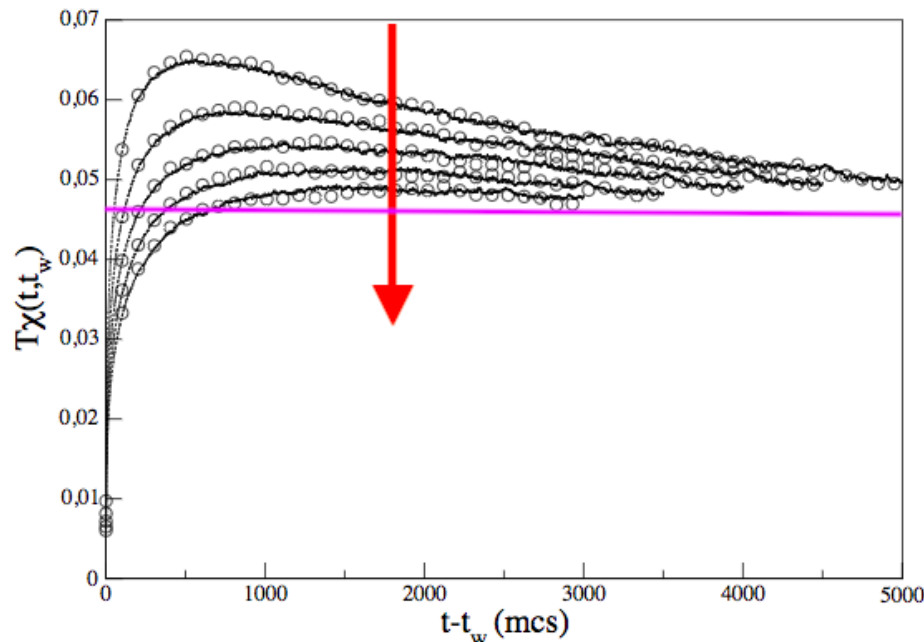


Thermo-remmanent magnetisation

Critical and sub-critical coarsening

$$m(t, t_w) = m_{eq}(t - t_w) + [\mathcal{R}(t_w, T)]^{-a_\chi} g \left(\frac{\mathcal{R}(t, T)}{\mathcal{R}(t_w, T)} \right)$$

$$m_{eq}(t - t_w) = -(k_B T)^{-1} dC_{eq}(t - t_w) / d(t - t_w)$$



$$1 - m_{eq}^2$$

The equilibration time of a coarsening system scales as

$$\mathcal{R}(t_{\text{eq}}) \sim L$$

$$\lim_{N \gg 1} t_{\text{eq}}(N, \text{parameters}) \gg t_{\text{exp}}$$

t_{eq} grows with the system size $N \propto L^D$

Correlations **age** similarly to the ones in spin-glasses

Integrated responses, e.g. thermo-remanent magnetisation, do not

Characterisation of the Collective Relaxation

when there is no “visible” length \mathcal{R}

Global Observables

two-time correlations and linear responses

Two-time dependencies

Self displacement & correlation – integrated linear response

$$\Delta^2(t, t') \equiv \frac{1}{N} \sum_i [\langle (x_i(t) - x_i(t'))^2 \rangle]$$

Displacement

$$C(t, t') \equiv \frac{1}{N} \sum_i [\langle x_i(t) x_i(t') \rangle]$$

Correlation

}
Unperturbed

$$\chi(t, t') \equiv \frac{1}{N} \sum_i \int_{t'}^t dt'' R_i(t, t'') = \frac{1}{N} \sum_i \int_{t'}^t dt'' \left[\frac{\delta \langle x_i(t) \rangle_h}{\delta h_i(t'')} \Big|_{h=0} \right]$$

Extend the notion of order parameter

They are not related by FDT out of equilibrium

The averages are thermal (and over initial conditions) $\langle \dots \rangle$

and over quenched randomness $[\dots]$ (if present)

t' “waiting-time” and t “measuring-time” after preparation

Mean-Field Modelling

**Usual Curie-Weiss for PM-FM
Unusual for Glasses**

Mean-Field Modelling

Classical p -spin Spherical Models

Potential energy

$$\mathcal{V} = - \sum_{i_1 \neq \dots \neq i_p} J_{i_1 \dots i_p} s_{i_1} \dots s_{i_p} \quad p \text{ integer}$$

quenched random couplings $J_{i_1 \dots i_p}$ drawn from a Gaussian $P[\{J_{i_1 \dots i_p}\}]$

(over-damped) **Langevin dynamics** for continuous spins $s_i \in \mathbb{R}$

coupled to a white bath $\langle \xi_i(t) \rangle = 0$ and $\langle \xi_i(t) \xi_j(t') \rangle = 2\gamma k_B T \delta_{ij} \delta(t - t')$

$$\gamma \frac{ds_i}{dt} = - \frac{\delta \mathcal{V}}{\delta s_i} + z_t s_i + \xi_i$$

z_t is a Lagrange multiplier that fixes the spherical constraint $\sum_{i=1}^N s_i^2 = N$

$p = 2$ mean-field **coarsening**

$p \geq 3$ **RFOT** modelling of **glasses**

Kirkpatrick, Thirumalai & Wolynes 87-89

Dynamic equations

Integro-differential eqs. on the correlation and linear response

In the $N \rightarrow \infty$ limit exact and closed causal Schwinger-Dyson equations

(Average over randomness, random initial conditions and thermal noise)

$$\begin{aligned}(\gamma\partial_t - z_t)C(t, t') &= \int dt'' [\Sigma(t, t'')C(t'', t') + D(t, t'')R(t', t'')] \\ &\quad + 2\gamma k_B T R(t', t) \\ (\gamma\partial_t - z_t)R(t, t') &= \int dt'' \Sigma(t, t'')R(t'', t') + \delta(t - t')\end{aligned}$$

where Σ and D are the self-energy and vertex, which for p spin models read

$$D(t, t') = \frac{p}{2} C^{p-1}(t, t') \quad \Sigma(t, t') = \frac{p(p-1)}{2} C^{p-2}(t, t') R(t, t')$$

z_t is fixed by $C(t, t) = 1$

Sompolinsky & Zippelius 82, LFC & Kurchan 93

Similar to **Mode-Coupling Theory** for liquids Götze et al 80s or **DMFT** for quantum systems Georges & Kotliar 90s, but not necessarily in equilibrium

How to solve these equations ?

Input from numerical solutions \implies

Asymptotic Ansatz

Weak ergodicity breaking

$$\lim_{t-t' \rightarrow \infty} \lim_{t' \rightarrow \infty} C(t, t') = q_{\text{EA}}$$

$$\lim_{t \gg t'} C(t, t') = 0$$

Bouchaud 92

Weak long-term memory

$$\lim_{t-t' \rightarrow \infty} \lim_{t' \rightarrow \infty} R(t, t') \simeq 0$$

but

$$\sigma(t, t') = \int_0^{t'} dt'' R(t, t'') \longrightarrow f_{\sigma}(C(t, t')) = \text{finite}$$

LFC & Kurchan 93

allow us to solve the integro-differential eqs. asymptotically

Weak ergodicity breaking

$$\lim_{t-t' \rightarrow \infty} \lim_{t' \rightarrow \infty} C(t, t') = q_{\text{EA}}$$

$$\lim_{t \gg t'} C(t, t') = 0$$

Bouchaud 92

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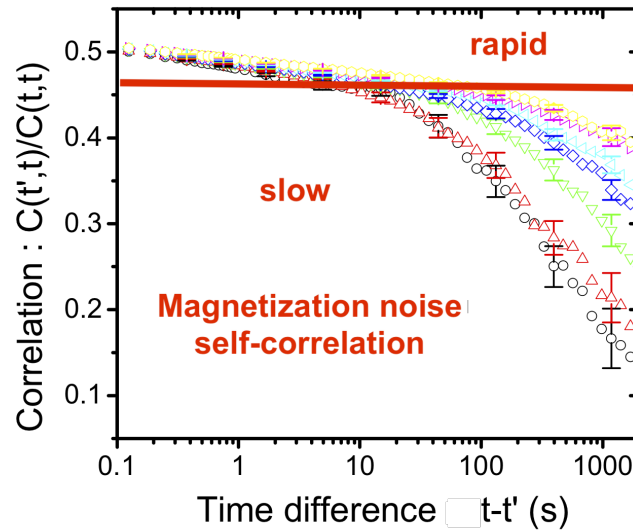
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LFC & Kurchan 93

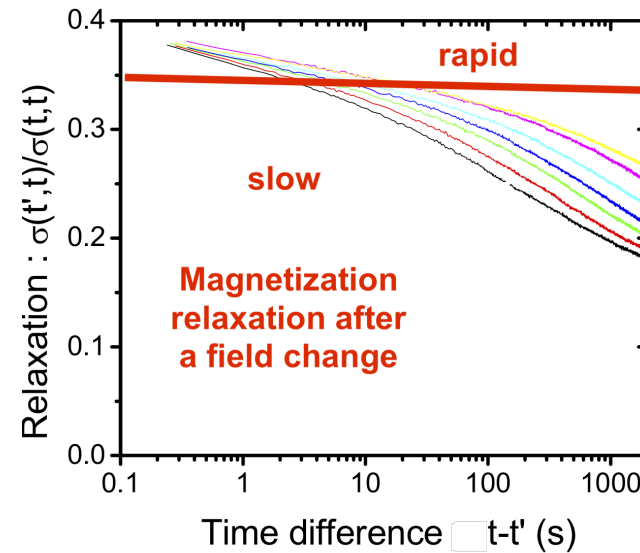
and capture aging

Slow Relaxation & Aging

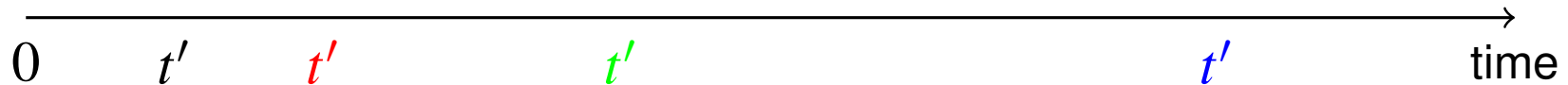
Separation of time scales & loss of stationarity



Correlation



Integrated linear response



Different curves are measured after log-spaced times t' after the quench

breakdown of stationarity \implies **aging**, far from equilibrium

microscopic mechanisms ?

Physical Aging

in words & scaling

Older systems (more time elapsed after the quench, longer t')
relax **more slowly** than younger ones

Breakdown of stationarity of correlation & integrated response

$$C(t, t') \neq C(t - t') \quad \sigma(t, t') \neq \sigma(t - t')$$

In each regime – $\underbrace{\text{rapid } (r)}_{\text{above plateau}}$ and $\underbrace{\text{slow } (s)}_{\text{below plateau}}$ – there is scaling*

$$C(t, t') = C_{r,s} \left(\frac{\mathcal{R}_{r,s}(t)}{\mathcal{R}_{r,s}(t')} \right) \quad \sigma(t, t') = \sigma_{r,s} \left(\frac{\mathcal{R}_{r,s}(t)}{\mathcal{R}_{r,s}(t')} \right)$$

* proven from general properties of temporal correlation functions

LFC & Kurchan 94

but no “visual” interpretation of $\mathcal{R}(t)$ in glassy systems

Physical Aging & Memory

the rate of relaxation

Older systems (more time elapsed after the quench, longer t')
relax **more slowly** than younger ones

Time variation - rate of change

$$\partial_t C_{r,s} \left(\frac{\mathcal{R}_{r,s}(t)}{\mathcal{R}_{r,s}(t')} \right) = C'_{r,s} \left(\frac{\mathcal{R}_{r,s}(t)}{\mathcal{R}_{r,s}(t')} \right) \frac{\dot{\mathcal{R}}_{r,s}(t)}{\mathcal{R}_{r,s}(t')} = \bar{C}_{r,s} (C_{r,s}(t, t')) \frac{\dot{\mathcal{R}}_{r,s}(t)}{\mathcal{R}_{r,s}(t)}$$

difference between **rapid** and **slow**, depending on $\mathcal{R}(t)$

$$\underbrace{\partial_t C_r(t, t') \simeq C_r(t, t')}_{\text{e.g. } \mathcal{R} \sim e^{t/\tau} \text{ exponential}}$$

$$\underbrace{\partial_t C_s(t, t') \ll C_s(t, t')}_{\text{e.g. } \mathcal{R} \sim t \text{ algebraic}}$$

Interpretation

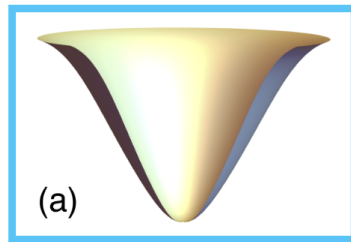
**Complex Landscapes
Beyond Ginzburg-Landau**

Mean-field classes

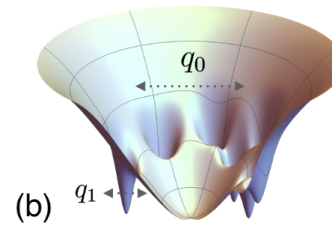
Domain growth

Equilibrium states

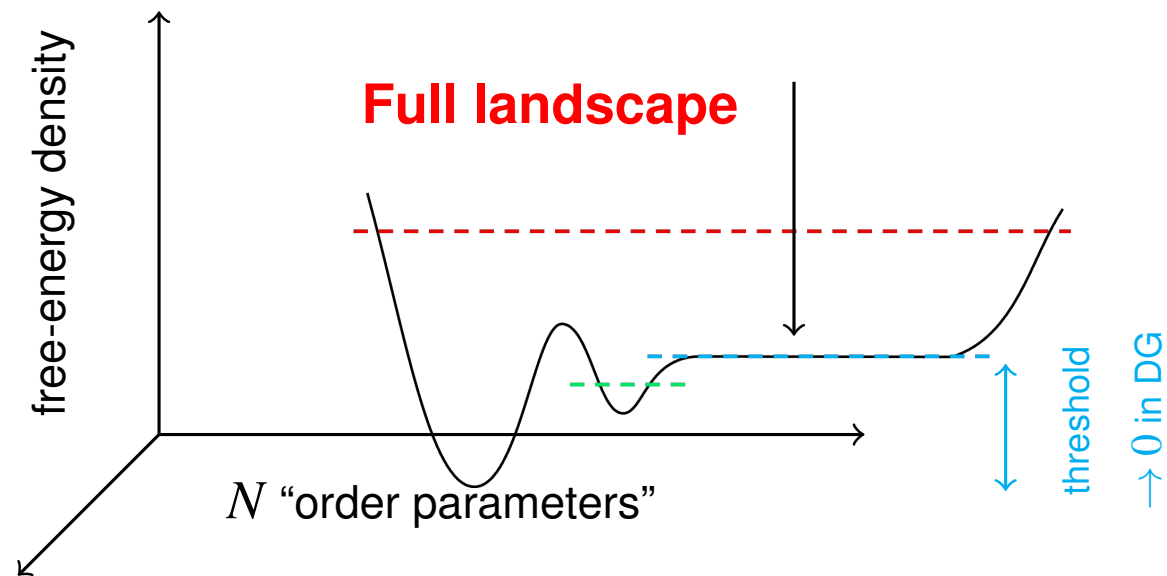
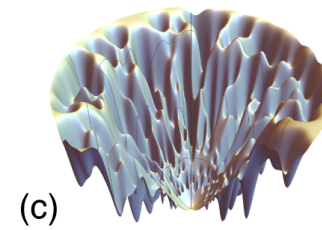
RS



1RSB



FRSB

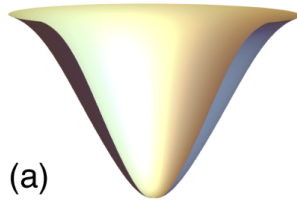


Mean-field classes

Sherrington-Kirkpatrick

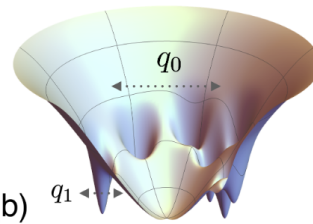
Equilibrium states

RS



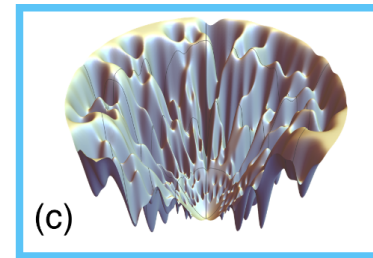
(a)

1RSB

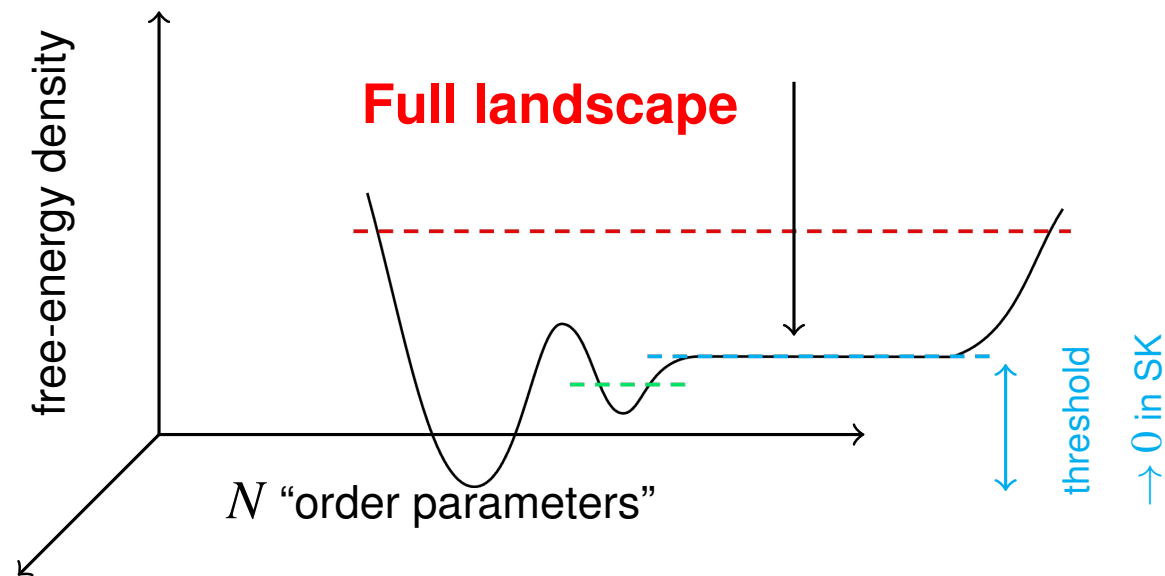


(b)

FRSB



(c)

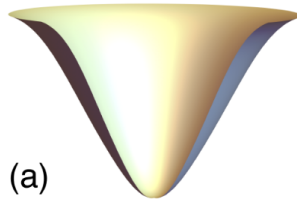


Mean-field classes

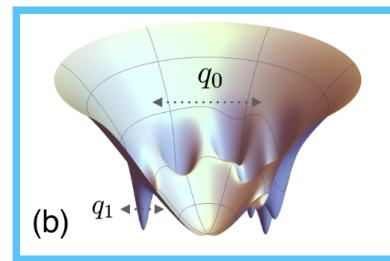
$p \geq 3$ spherical and Ising models

Equilibrium states

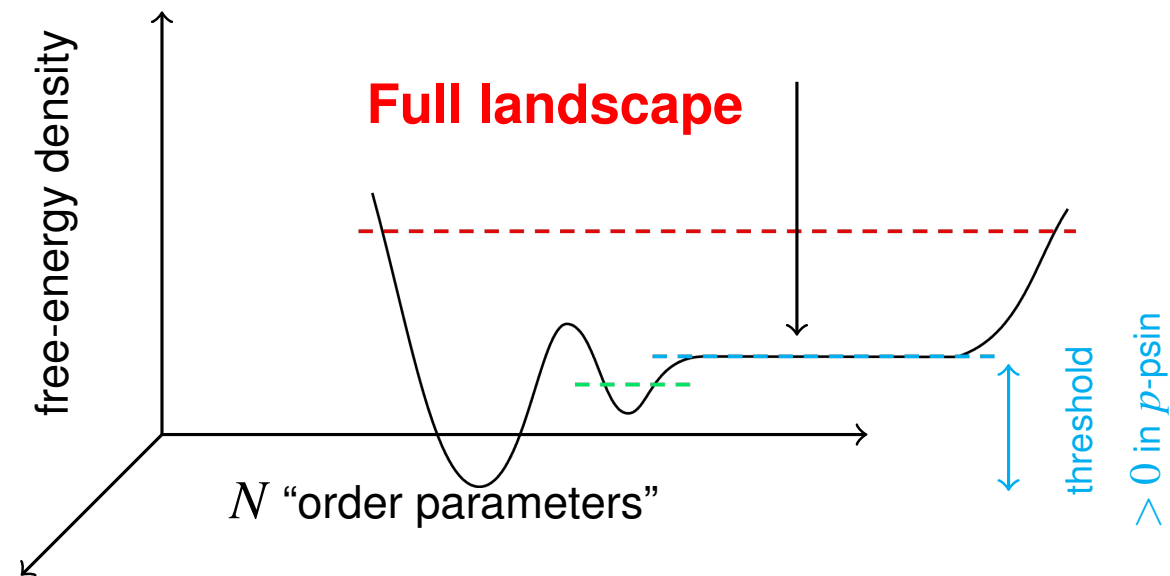
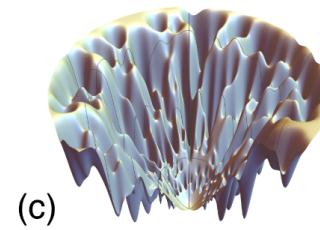
RS



1RSB

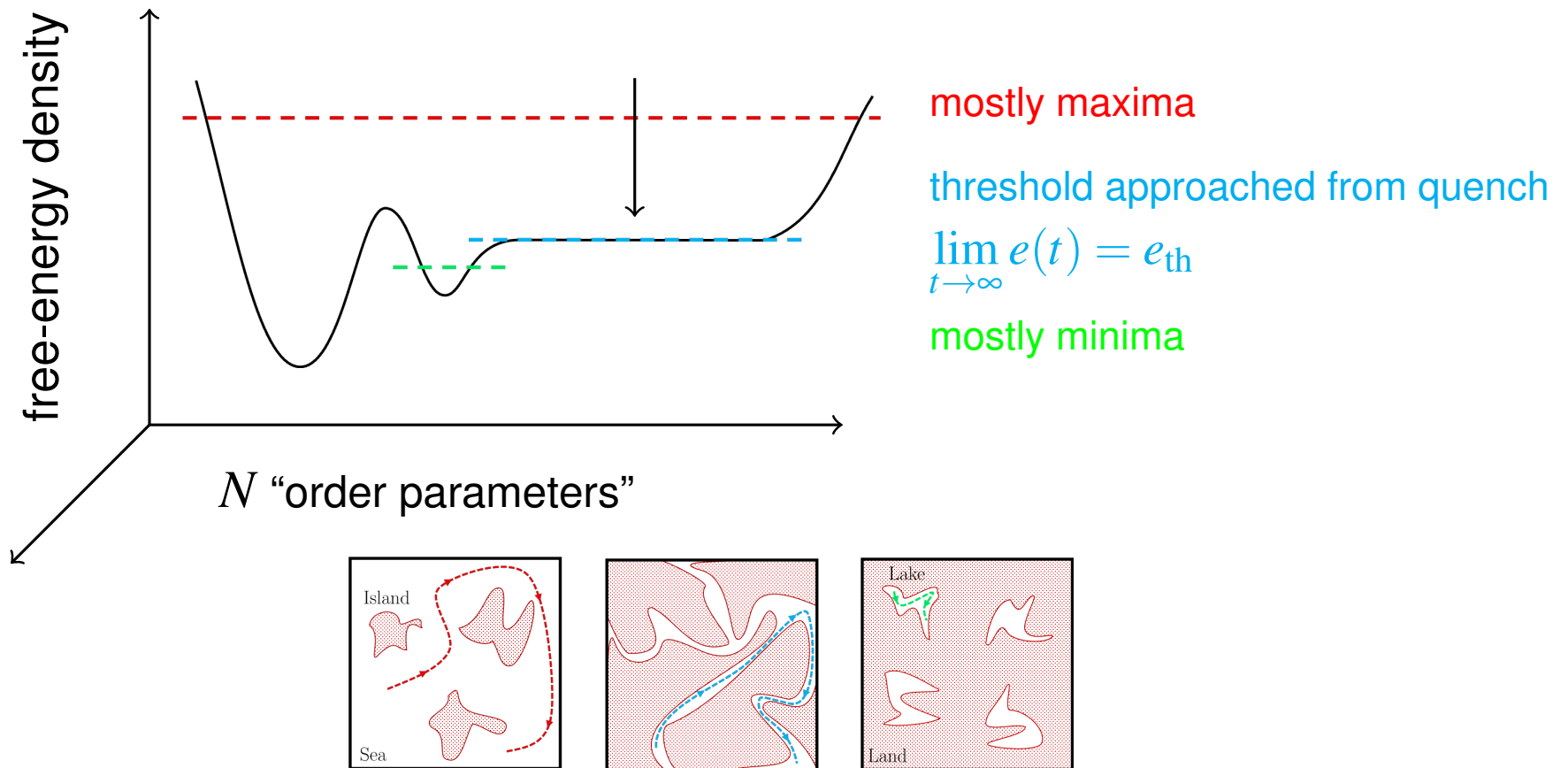


FRSB



$p \geq 3$ spin models

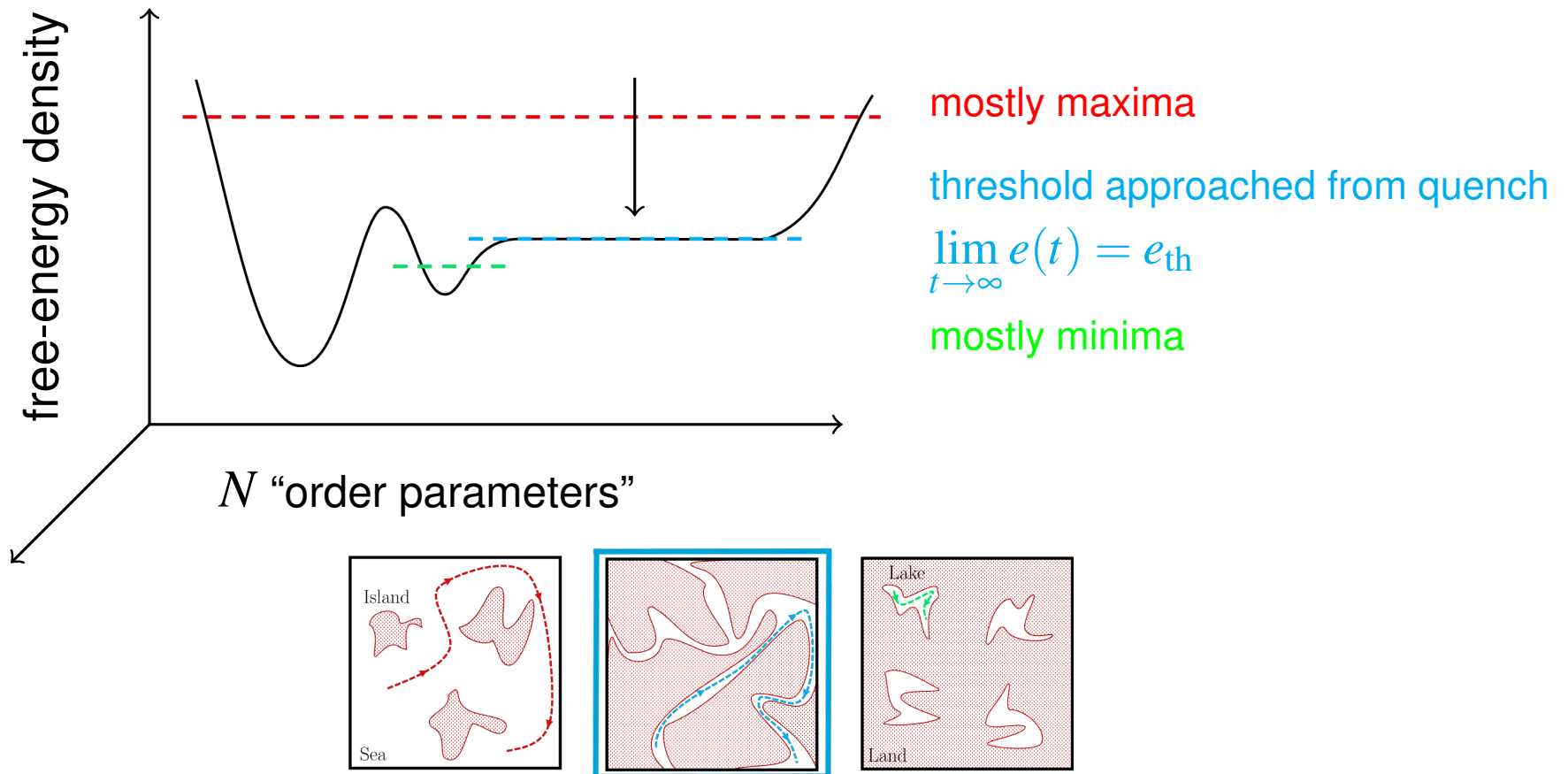
Glassy class



The dynamics is linked to the topography of the landscape

$p \geq 3$ spin models

Glassy class



Flat threshold as an attractor for the p -spin relaxation

Both for **physical** and **algorithmic** dynamic rules

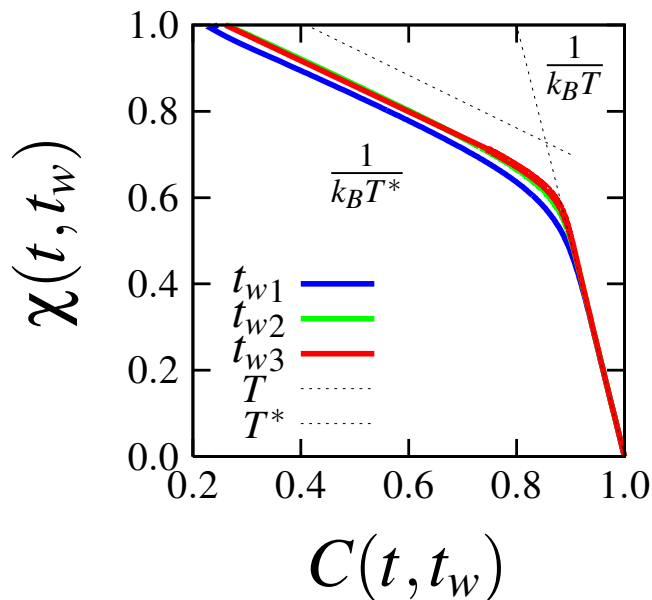
Some surprising predictions

with physical consequences

Fluctuation-dissipation

Induced vs. spontaneous fluctuations in glasses

A quench from $T_0 \rightarrow \infty$ to $T < T_c$



parametric construction

t_w fixed

$$t_{w1} < t_{w2} < t_{w3}$$

$t - t_w : 0 \rightarrow \infty$

used as a parameter

$$T^* > T$$

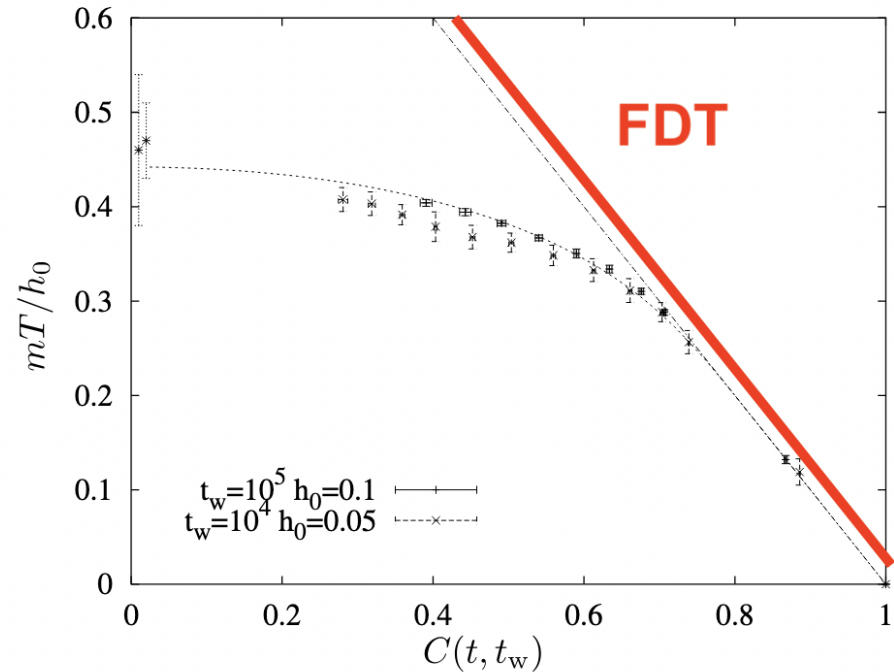
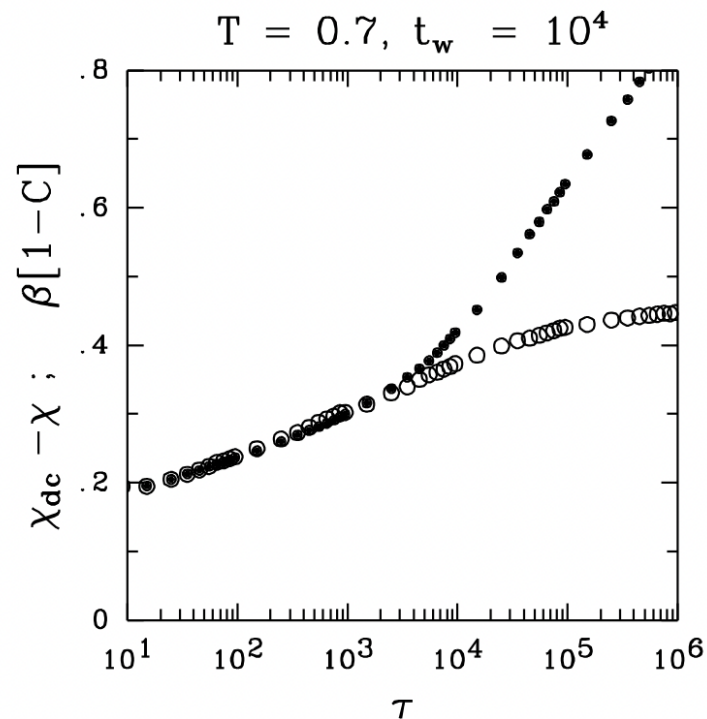
Breakdown of the equilibrium FDT $k_B T \chi = C$

Convergence to $k_B T \chi(C)$, two linear relations for $C \lesssim q_{EA}$

Mean-field **LFC & Kurchan 93** & effective temperature interpretation **LFC, Kurchan & Peliti 97**

Fluctuation-dissipation

Induced vs. spontaneous fluctuations in 3D EA

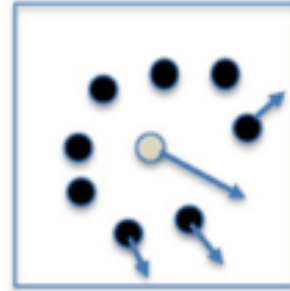


Franz & Rieger 95

Marinari, Parisi, Ricci-Tersenghi & Ruiz Lorenzo 98

Fluctuation-dissipation

Interpretation



Short-scale re-arrangements ruled by the **equilibrium external bath** & **local properties of landscape**

The fluctuation-dissipation relation holds with the bath temperature T

Large-scale re-arrangements follow the systems' **internal dynamics** & **large scale props of landscape**

The fluctuation-dissipation relation holds with another temperature T^*

After **cooling** from equilibrium at $T_0 > T_d$, hotter $T^* > T$

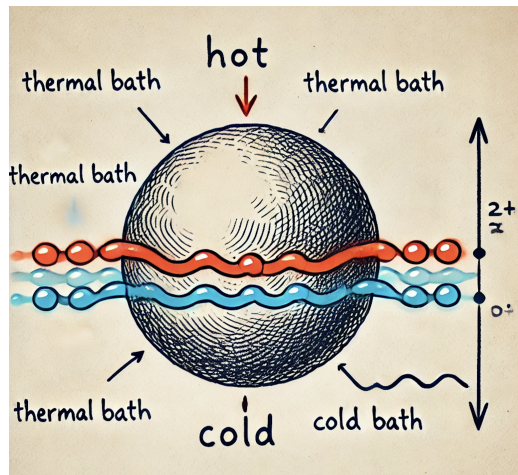
After **heating** from equilibrium at $T_0 < T_d$, colder $T^* < T$

Support for this interpretation :

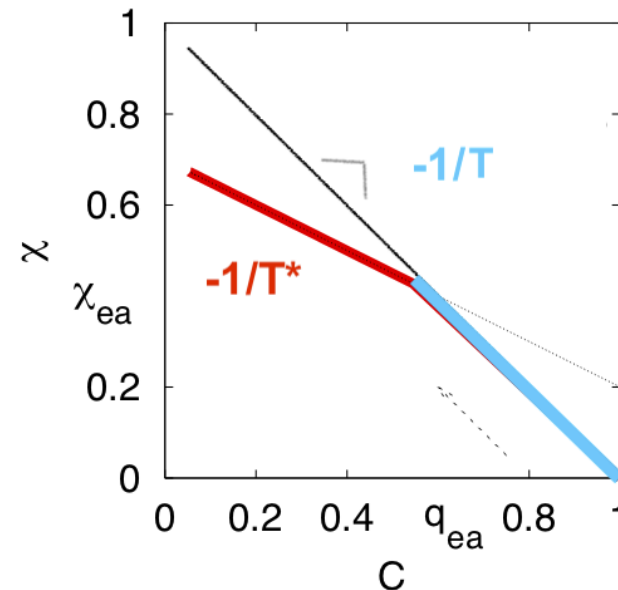
Effective temperatures

Induced by two (or more) baths

A tracer in contact with a complex bath



Sketch created by ChatGPT



$$\Gamma = \Gamma_{\text{cold}} + \Gamma_{\text{hot}}$$

$$\Gamma_{\text{cold}}(t - t') = 2\gamma\delta(t - t') \text{ and } T$$

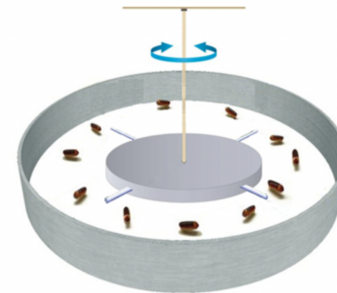
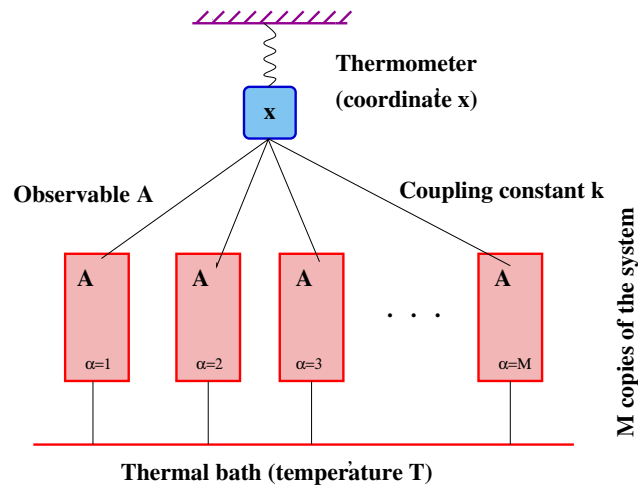
$$\Gamma_{\text{hot}}(t - t') = \gamma_{\text{hot}} e^{-(t-t')/\tau} \text{ and } T^*$$

LFC & Kurchan 00, Ilg & Barrat 07, etc., cfr. tracer in pasive & active bath

Effective temperatures

Measurement with thermometers

LFC, Kurchan & Peliti 97



Grigera & Israeloff 99 - **glassy**

D'Anna, Mayor, Barrat, Loreto & Nori 03 - **granular**

Boudet, Jagielka, Guerin, Barois, Pistoiesi & Kellay 24
artificial active matter - robots

- **Short internal time scale** fast dynamics is tested and T is recorded.
- **Long internal time scale** slow dynamics is tested and T^* is recorded.

Related to the phenomenological *fictive temperatures* of Tool 46, Gardon & Narayanaswamy 70, Moynihan et al 76, etc. but measurable & with a thermodynamic interpretation

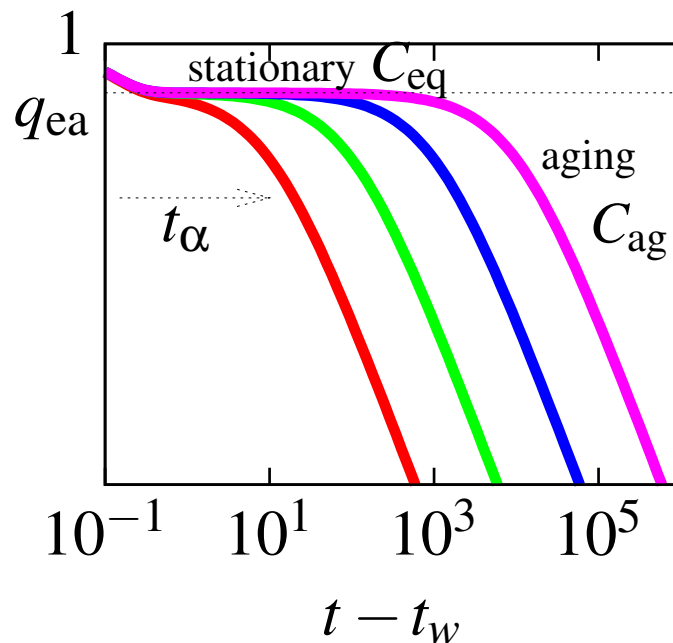
Also appearing in stochastic thermodynamic relations Zamponi et al 05

Time reparametrization invariance and fluctuations

Time reparametrization invariance

In the long t_w limit

Fast $t - t_w \ll t_w$



The aging part is **slow**

$$\mathcal{R}(t)/\mathcal{R}(t_w) = O(1)$$

$$C_{ag}(t, t_w) \sim f_{ag} \left(\frac{\mathcal{R}(t)}{\mathcal{R}(t_w)} \right)$$

$$\partial_t C_{ag}(t, t_w) \propto \frac{\dot{\mathcal{R}}(t)}{\mathcal{R}(t)} \xrightarrow{t \rightarrow \infty} 0$$

$$\partial_t C_{ag}(t, t_w) \ll C_{ag}(t, t_w)$$

Eqs. for the slow relaxation $C_{ag} < q_{EA}$ are invariant under

$$t \rightarrow h(t) \quad C(t, t_w) \rightarrow C(h(t), h(t_w)) \quad R(t, t_w) \rightarrow \dot{h}(t')/h(t_w) R(h(t), h(t_w))$$

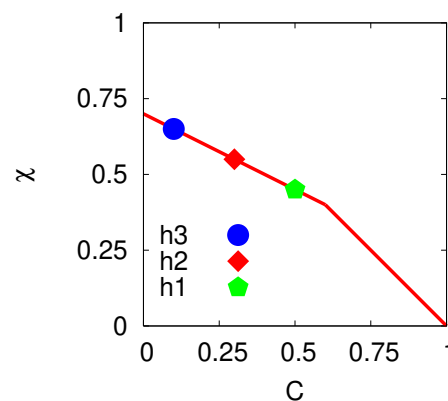
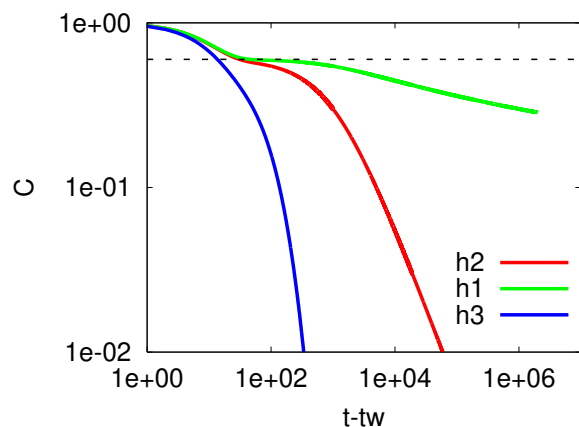
Leading fluctuations

Global to local correlations & linear responses

$$C_{ag}(t, t_w) \approx f_{ag} \left(\frac{\mathcal{R}(t)}{\mathcal{R}(t_w)} \right) \quad \text{global correlation}$$

Global time-reparametrization invariance $\Rightarrow C_{\vec{r}}^{ag}(t, t_w) \sim f_{ag} \left(\frac{h_{\vec{r}}(t)}{h_{\vec{r}}(t_w)} \right)$

Ex. $h_{\vec{r}_1} = \frac{t}{t_0}$, $h_{\vec{r}_2} = \ln \left(\frac{t}{t_0} \right)$, $h_{\vec{r}_3} = e^{\ln^{a>1} \left(\frac{t}{t_0} \right)}$ in different spatial regions



Castillo, Chamon, LFC, Iguain & Kennett 02, 03

Chamon, Charbonneau, LFC, Reichman & Sellitto 04

Jaubert, Chamon, LFC & Picco 07

Avila, Castillo & Parsaeian 12

Dyre 22-25

More recent perspective : time-reparametrization invariance in SYK models

Kitaev 15, Maldacena & Stanford 16

Triangular relations

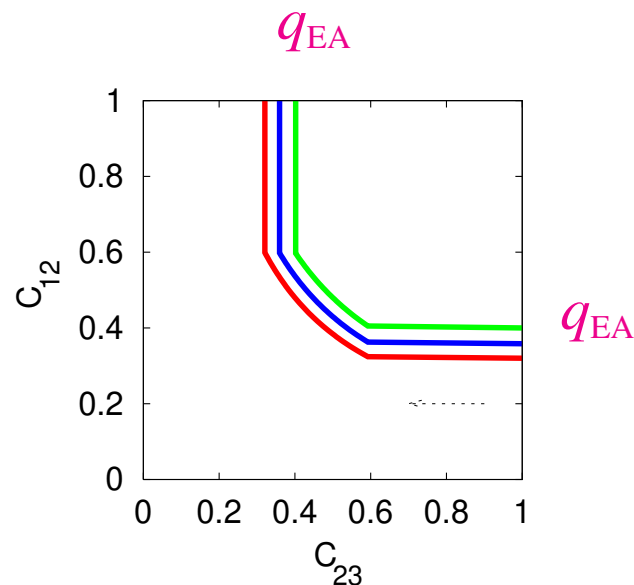
Scaling of the aging global correlation

Take three times $t_1 \geq t_2 \geq t_3$ and compute the three global correlations

$$C(t_1, t_2), C(t_2, t_3), C(t_1, t_3)$$

If, in the aging regime $C_{\text{ag}}^{ij} \equiv C_{\text{ag}}(t_i, t_j) = f_{\text{ag}} \left(\frac{\mathcal{R}(t_i)}{\mathcal{R}(t_j)} \right)$ with $t_i \geq t_j \Rightarrow$

$$C_{\text{ag}}^{12} = f_{\text{ag}} \left(\frac{\mathcal{R}(t_1)}{\mathcal{R}(t_3)} \frac{\mathcal{R}(t_3)}{\mathcal{R}(t_2)} \right) = f_{\text{ag}} \left(\frac{f_{\text{ag}}^{-1}(C_{\text{ag}}^{13})}{f_{\text{ag}}^{-1}(C_{\text{ag}}^{23})} \right)$$



choose t_3 and t_1 so that $C^{13} = 0.3$

the arrow shows the t_2 'flow' from t_3 to t_1

e.g. $C_{\text{ag}}^{12} = q_{\text{EA}} C_{\text{ag}}^{13} / C_{\text{ag}}^{23}$

in the spherical p spin model

Triangular relations

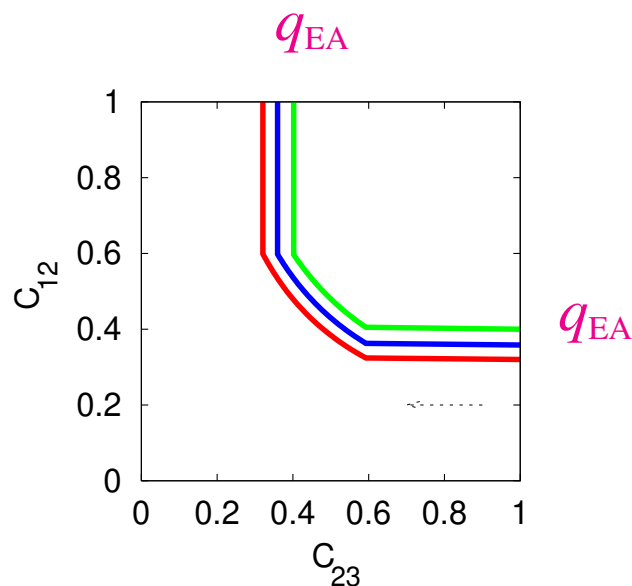
Scaling of the slow part of the global correlation

Take three times $t_1 \geq t_2 \geq t_3$ and compute the three local correlations

$$C_{\vec{r}}(t_1, t_2), C_{\vec{r}}(t_2, t_3), C_{\vec{r}}(t_1, t_3)$$

If, in the aging regime $C_{\vec{r}}^{ij} \equiv C_{\vec{r}}(t_i, t_j) = f_{\text{ag}} \left(\frac{h_{\vec{r}}(t_i)}{h_{\vec{r}}(t_j)} \right)$ with $t_i \geq t_j \Rightarrow$

$$C_{\vec{r}}^{12} = f_{\text{ag}} \left(\frac{f_{\text{ag}}^{-1}(C_{\vec{r}}^{13})}{f_{\text{ag}}^{-1}(C_{\vec{r}}^{23})} \right)$$



choose t_3 and t_1 so that $C^{13} = 0.3$

the arrow shows the t_2 'flow' from t_3 to t_1

e.g. $C_{\vec{r}}^{12} = q_{\text{EA}} C_{\vec{r}}^{13} / C_{\vec{r}}^{23}$

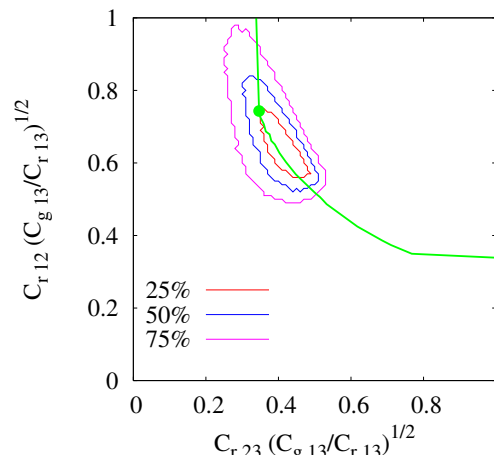
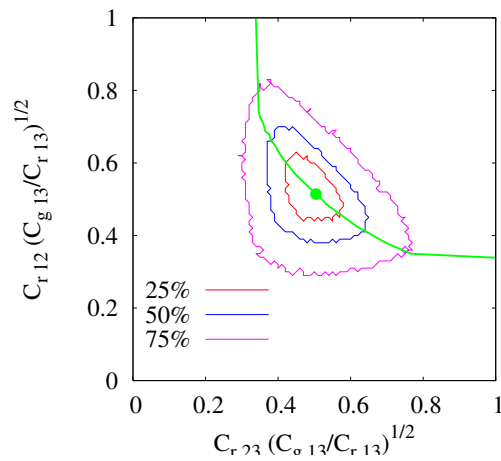
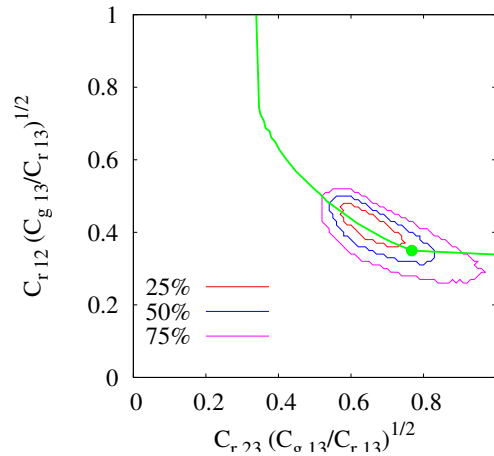
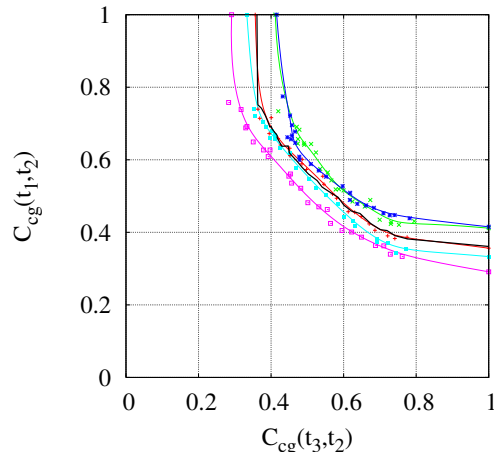
in the spherical p spin model

Triangular relations

3D Edwards-Anderson model

Black : global, other : five regions with size ℓ

Green : global ; other : 2d projections of pdf

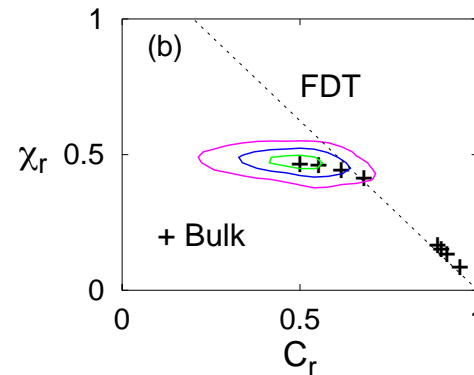
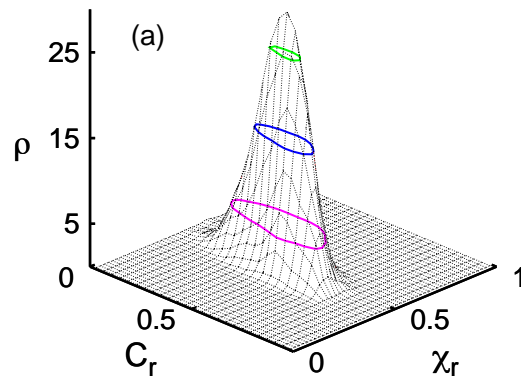


Local FDTs

3D Edwards-Anderson

$$C_{\vec{r}}(t, t_w) \equiv \frac{1}{V_{\vec{r}}} \sum_{i \in V_{\vec{r}}} s_i(t) s_i(t_w)$$

$$\chi_{\vec{r}}(t, t_w) \equiv \frac{1}{V_{\vec{r}}} \sum_{i \in V_{\vec{r}}} \int_{t_w}^t dt' \left. \frac{\delta s_i(t)}{\delta h_i(t')} \right|_{h=0}$$



+ Bulk : Parametric plot $\chi(t, t_w)$ vs $C(t, t_w)$ for t_w fixed and $7 t (> t_w)$

ρ corresponds to the maximum t yielding the smallest global C (left-most +)

Short Overview of Glassy Dynamics

Statics

& a bit of Relaxation Dynamics