

Random hard sphere packings and constraint satisfaction problems

Francesco Zamponi

Laboratoire de Physique Théorique, École Normale Supérieure,
24 Rue Lhomond, 75231 Paris Cedex 05, France

Lorentz Center, Leiden, April 19, 2012

Presentation of the talk

Overview of the statistical mechanics (mean field) approach to the random packing problem

- The packing problem is a constraint satisfaction problem
- Discrete random constraint satisfaction problems (R-CSP) have been studied in detail by the statistical mechanics community
Franz, Kirkpatrick, Krzakala, Kurchan, Marsili, Mézard, Monasson, Montanari, Parisi, Ricci-Tersenghi, Semerjian, Zdeborová, Zecchina, and many others
in a very fruitful exchange with the mathematicians
- Review of results for simple R-CSP: structure of solution and *local* algorithms
- Difficulties in extending these results to the packing problem
- Results from the replica method (briefly)
Cardenas, Franz, Kirkpatrick, Mézard, Monasson, Parisi, Thirumalai, Wolynes, FZ

1 Random Constraint Satisfaction Problems

- The coloring problem
- Statistical mechanics approach
- Phase transitions in R-CSP
- Consequences for *local* algorithms
- A bridge between coloring and packing problems

2 The sphere packing problem

- Mean field phase diagram
- Mean field and large d
- Mean field and $3d$
- Where are the soft modes?

Outline

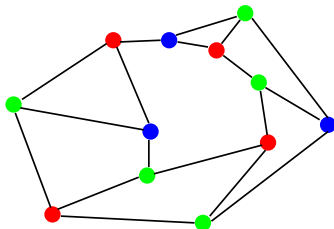
1 Random Constraint Satisfaction Problems

- The coloring problem
- Statistical mechanics approach
- Phase transitions in R-CSP
- Consequences for *local* algorithms
- A bridge between coloring and packing problems

2 The sphere packing problem

- Mean field phase diagram
- Mean field and large d
- Mean field and $3d$
- Where are the soft modes?

The coloring problem

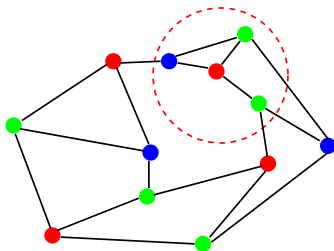


Input: A graph of N nodes and M links, and q colors.

Problem: Assign a color to each node in such a way that no link connects two nodes of different colors

A constraint satisfaction problem

The coloring problem as a packing problem



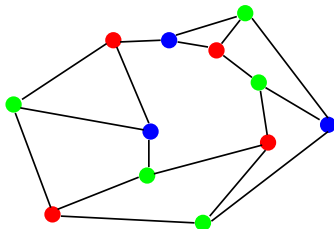
Input: A graph of N nodes and M links, and q types of particles.

Problem:

- On each site, put one particle (hard-core exclusion)
- If a site is occupied by a particle of color i , then adjacent sites cannot be occupied by particles of the same color (nearest-neighbor exclusion between particles of the same color)

A packing problem for a non-additive mixture of q types of particles

Coloring problem as a statistical mechanics problem



Input: A graph of N nodes and M links, and “Potts spins” $\sigma_i \in \{1, 2, \dots, q\}$.

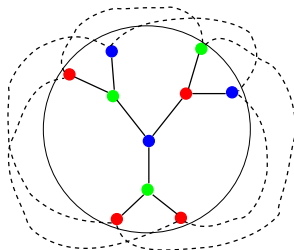
Problem:

- Define the Hamiltonian $H = \sum_{\langle i,j \rangle} \delta_{\sigma_i, \sigma_j}$
- Compute the partition function: $Z(\beta) = \sum_{\sigma_1, \dots, \sigma_N} e^{-\beta H}$
- The ground state energy is $E_{GS} = \lim_{\beta \rightarrow \infty} [-\beta^{-1} \log Z(\beta)]$

A coloring exists $\Leftrightarrow E_{GS} = 0$, and $Z(\beta = \infty) = \text{number of solutions}$

No coloring exist $\Leftrightarrow E_{GS} > 0$, and $E_{GS} = \text{minimal number of monochromatic links}$

Coloring of random graph as “spin glass” problem



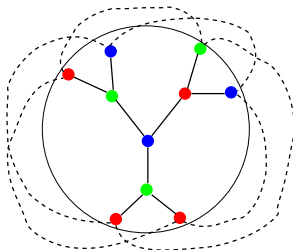
Solving the problem for each given graph is too hard

Introduce an ensemble of random graphs: e.g. Erdős-Rényi ensemble
Random Constraint Satisfaction Problem (R-CSP)

Key property of random graphs: locally tree-like for $N \rightarrow \infty$ and $M = cN/2$
Typical loops are long $\sim \log(N)$

The neighbors of a given spin are “far away” from each other
Hence, *they are decorrelated* (in absence of the central spin)
In physics, this is called a “mean field” model

Coloring of random graph as “spin glass” problem



Mean field statistical mechanics methods (replica and cavity methods) allow to compute exactly the partition function

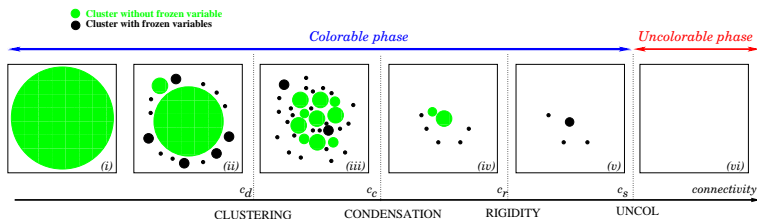
More precisely, the free energy averaged over the random graph:

$$f(\beta, c) = \lim_{N \rightarrow \infty, M = cN/2} \mathbb{E} \left[-\frac{1}{\beta N} \log Z(\beta) \right]$$

From this, all the information on the problem can be reconstructed: ground state energy, number of solutions, etc.

Phase transitions in R-CSP, at zero temperature

Structure of the set of the solutions: [Krzakala et al. PNAS 104, 10318 (2007)]

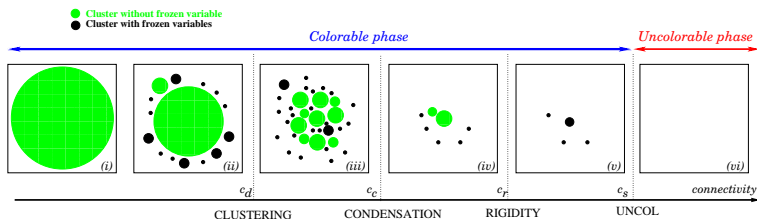


Phase transitions in Random CSP:

- $c < c_d$: Most of the solutions form a unique cluster
- $c_d < c < c_c$: The solutions form many ($\sim e^{N^\Sigma}$) clusters
- $c_c < c < c_s$: A small number of clusters dominate
- $c_r < c < c_s$: Frozen variables in a cluster
- $c > c_s$: No colorings, $E_{GS} > 0$ (UNSAT)

Phase transitions in R-CSP, at zero temperature

Structure of the set of the solutions: [Krzakala et al. PNAS 104, 10318 (2007)]



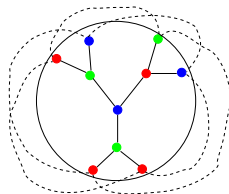
Condition for the clustering transition: $\forall c \leq c_d$ one has

$\mu(\cdot)$: uniform measure over solutions

σ_i : a randomly drawn spin

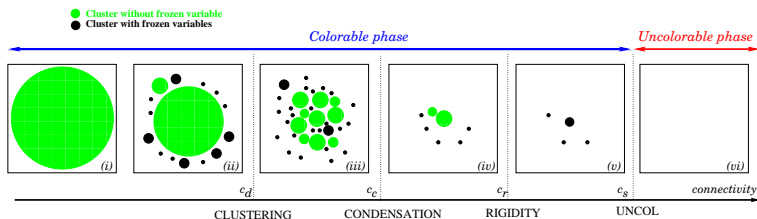
$\underline{\sigma}_\ell$: set of spins at distance ℓ from i

$$\lim_{\ell \rightarrow \infty} \lim_{N \rightarrow \infty} \mathbb{E} \sum_{\underline{\sigma}_\ell} \mu(\underline{\sigma}_\ell) \sum_{\sigma_i} |\mu(\sigma_i | \underline{\sigma}_\ell) - \mu(\sigma_i)| = 0$$



Phase transitions in R-CSP, at zero temperature

Structure of the set of the solutions: [Krzakala et al. PNAS 104, 10318 (2007)]



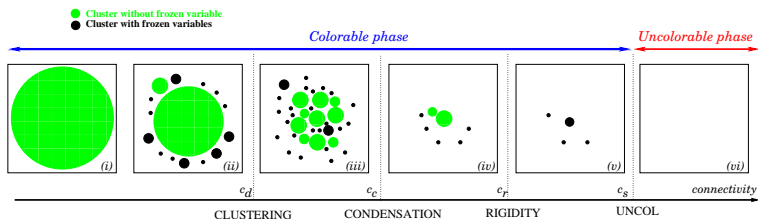
Condition for the condensation transition: draw n random spins i_1, \dots, i_n , then one has, $\forall n$ and $\forall c \leq c_c$,

$$\lim_{N \rightarrow \infty} \mathbb{E} \sum_{\sigma_{i_1}, \dots, \sigma_{i_n}} |\mu(\sigma_{i_1}, \dots, \sigma_{i_n}) - \mu(\sigma_{i_1}) \cdots \mu(\sigma_{i_n})| = 0$$

Hence c_c is really a “phase transition”!

Phase transitions in R-CSP, at zero temperature

Structure of the set of the solutions: [Krzakala et al. PNAS 104, 10318 (2007)]

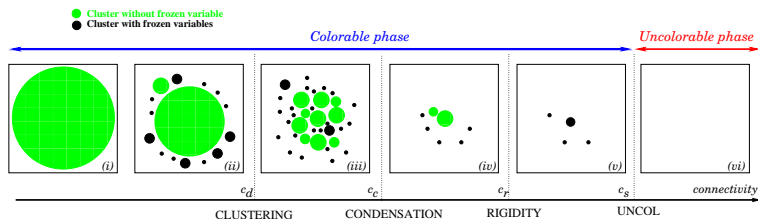


Condition for the satisfiability transition: $\forall c > c_s$, one has

$$\lim_{N \rightarrow \infty} P(Z > 0) = 0$$

Phase transitions in R-CSP, at zero temperature

Structure of the set of the solutions: [Krzakala et al. PNAS 104, 10318 (2007)]



In the limit of large q , one has

$$c_d \sim q(\log q + \log \log q + O(1))$$

[Theorem: Sly CMP 288, 943 (2009)]

Gerschenfeld, Montanari, IEEE FOCS'07, 194 (2007)]

$$c_c \sim 2q \log q - \log q - 2 \log 2 + o(1)$$

Existence of this transition proven in a similar model

[Coja-Oghlan, Zdeborova, arXiv:1107.2341]

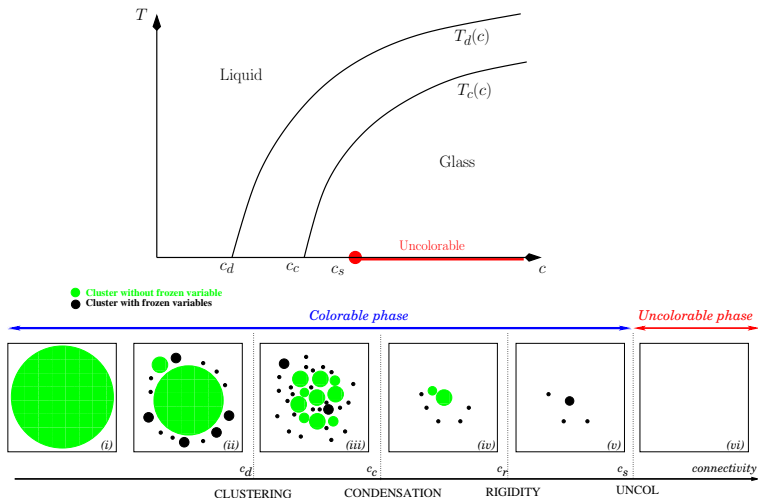
$$c_s \sim 2q \log q - \log q - 1 + o(1)$$

[Theorem ($2q \log q$): Achlioptas, Naor, Ann. Math. 162, 1335 (2005)]

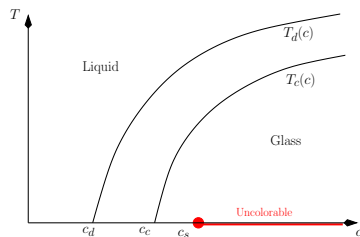
Phase transitions in R-CSP, at non-zero temperature

The temperature-connectivity phase diagram:

[Krzakala, Zdeborová, EPL 81, 57005 (2008)]

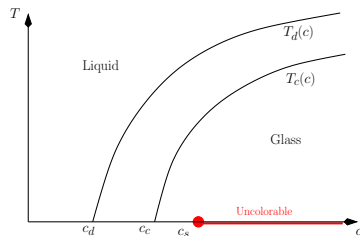


Consequences for *local* algorithms



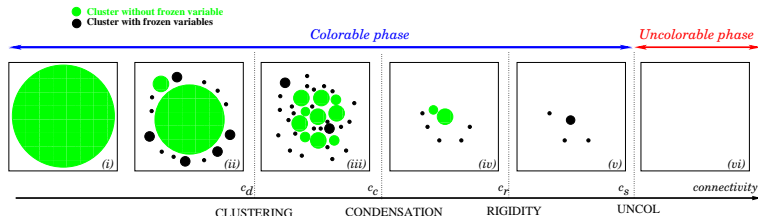
- Theorem: [Montanari, Semerjian, J. Stat. Phys. 125, 23 (2006)]
Local Monte-Carlo algorithms do not equilibrate for $T < T_d(c)$
- This means that sampling is hard in this region - it takes a time $\sim \exp(N)$
A glassy region!
- This means that dynamics becomes very algorithm-dependent in this region
- This *does not* mean that finding a solution is hard in this region

Consequences for *local* algorithms



- Sometimes, even thermal annealing is able to find a solution for $c > c_d$
- Several algorithms (WalkSAT, Belief Propagation, etc.) are able to find solutions quite close to c_s
- The point c_j where the algorithm fails is algorithm-dependent

Consequences for *local* algorithms



- Main message: whatever *local* algorithm will get stuck in one cluster
- There is a wide distribution of clusters and it is hard to predict in which cluster a given algorithm will get stuck
- Even being stuck in a cluster, one can find solutions
- In some cases, a more precise analysis can be done

[Krzakala, Zdeborová PRB 81, 224205 (2010)]

[Ricci-Tersenghi, Semerjian, J. Stat. Mech. P09001 (2009)]

A bridge between coloring and packing problems

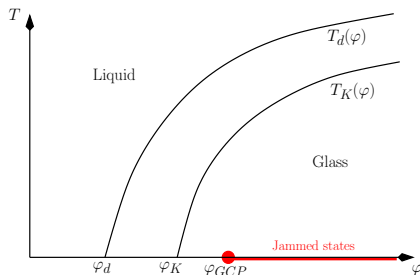
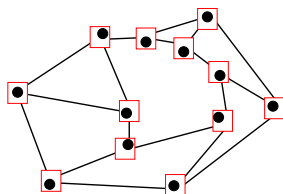
On each site, a variable $r_i \in [0, 1]^d$
"a continuous color"

[Mari, Krzakala, Kurchan,
 PRL 103, 025701 (2009)]

Hamiltonian $H = \sum_{\langle i,j \rangle} v(|r_i - r_j|)$
 with $v(r) = (D - r)^2 \theta(D - r)$

Random graphs as for the COL problem

Same qualitative phase diagram:



Outline

1 Random Constraint Satisfaction Problems

- The coloring problem
- Statistical mechanics approach
- Phase transitions in R-CSP
- Consequences for *local* algorithms
- A bridge between coloring and packing problems

2 The sphere packing problem

- Mean field phase diagram
- Mean field and large d
- Mean field and $3d$
- Where are the soft modes?

Mean field phase diagram for the sphere packing problem

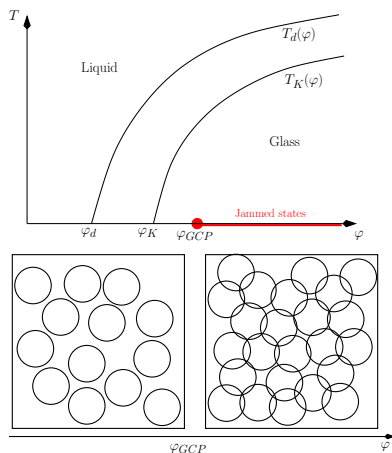
The temperature-packing fraction phase diagram:

N particles in a volume V
 d dimensions

$$H = \sum_{\langle i,j \rangle} v(|r_i - r_j|)$$

$$v(r) = (D - r)^2 \theta(D - r)$$

$$Z(\beta) = \int dr_1 \cdots dr_N e^{-\beta H}$$



Mean field phase diagram for the sphere packing problem

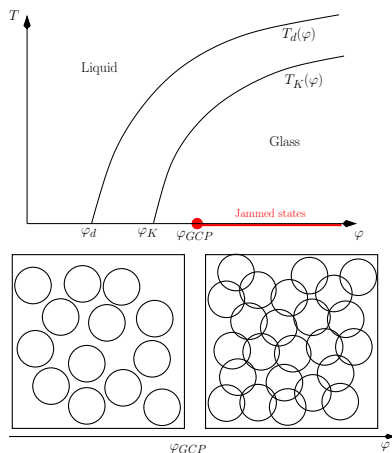
The temperature-packing fraction phase diagram:

New difficulties:

- Neighbors are correlated
- Geometry, crystallization

New interesting properties:

- Mechanical properties
- Soft modes
- Scaling laws
- A large correlation length



Mean field phase diagram for the sphere packing problem

The temperature-packing fraction phase diagram:

New difficulties:

- Geometry, crystallization

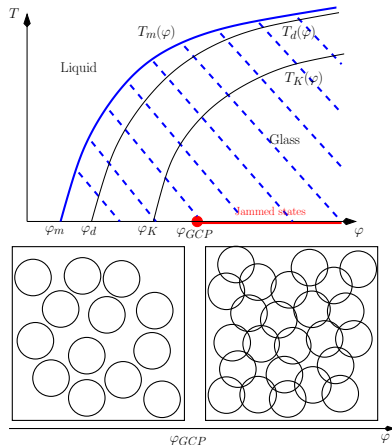
Because of crystallization, the amorphous region becomes metastable

Additional history dependence effects

In large dimensions, crystallization is suppressed

[Van Meel et al.,

PRE 80, 061110 (2009)]



Mean field phase diagram for the sphere packing problem

The temperature-packing fraction phase diagram:

New difficulties:

- Neighbors close to each others,
hence correlated

Mean field theory cannot be used

In large dimensions:

$z = 2d$ (isostaticity)

$z \geq e^d (-\log \sin(\pi/3))$ (individual kissing number)

[Wyner, Bell System Tech. J. 46, 2111 (1967)]

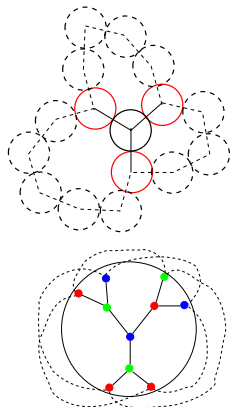
Neighbors are “far” from each others

The liquid lacks any short range structure

The Van der Waals equation is exact

Parisi, Slanina, PRE 62, 6554 (2000)

Skoge et al., PRE 74, 041127 (2006)



Mean field phase diagram for the sphere packing problem

Mean field theory of the glass transition in hard spheres has been formulated in many (almost) equivalent ways:

- Density functional theory

Singh, Stoessel, Wolynes, PRL 54, 1059 (1985)

- Random First Order Transition (RFOT) theory

Kirkpatrick, Thirumalai, Wolynes, PRA 40, 1045 (1989)

- Mode Coupling Theory

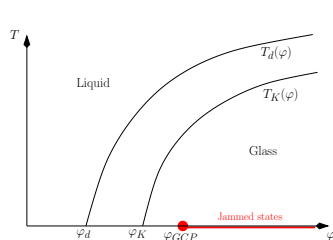
W. Gotze, Oxford University Press (2009)

- Replica Theory

Mézard, Parisi, PRL 82, 747 (1999); Parisi, FZ, RMP 82, 789 (2010)

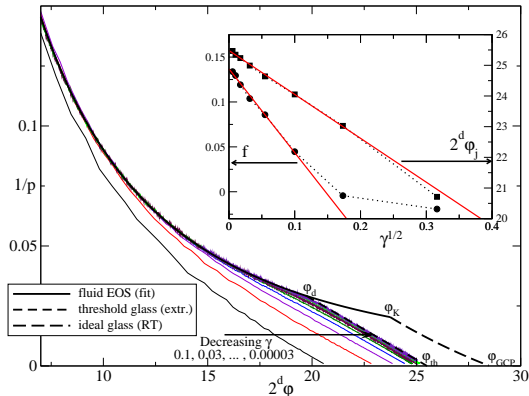
It might be exact for $d \rightarrow \infty$ and it is a reasonable approximation at finite d

Mean field theory should be exact in large d



$$p_{\text{fluid}}(\varphi) = 1 + 2^{d-1} \varphi \frac{1 - A_d \varphi}{(1 - \varphi)^d}$$

$$p_{\text{fv}}(\gamma, \varphi) = \frac{d \varphi_j(\gamma) [1 - f(\gamma)]}{\varphi_j(\gamma) - \varphi}$$

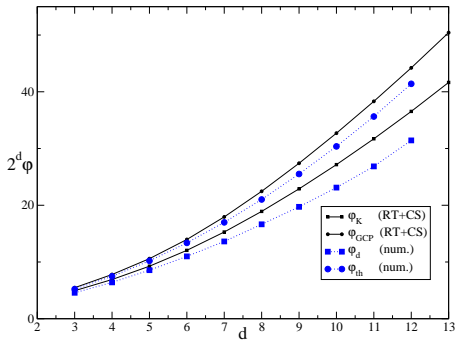
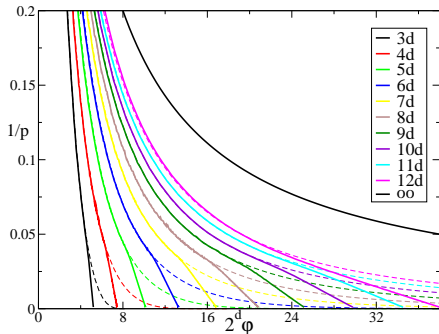


Compression of 9-dimensional hard spheres at constant rate
while doing local moves (event-driven collisions)

[Skoge, Donev, Stillinger, Torquato, PRE 74, 041127 (2006)]

[Charbonneau, Ikeda, Parisi, FZ, PRL 107, 185702 (2011)]

Mean field theory should be exact in large d



Large d scaling: $\phi_{GCP} \sim 2^{-d} d \log d$

The liquid equation of state approaches the mean field one upon increasing d

Replica theory captures well the trend with increasing dimension

Work in progress: an exact Replica theory formulation

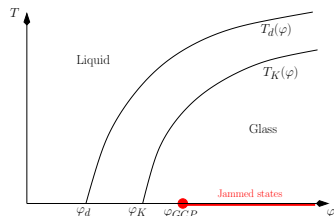
The results are inconsistent with Mode-Coupling Theory

More work is needed to understand the dynamics

[Charbonneau, Ikeda, Parisi, FZ, PRL 107, 185702 (2011)]

Mean field theory is not so bad in $d = 3$

A set of scaling laws around φ_{GCP} (or any φ_j):



Energy: $U(\varphi, T) = T \mathcal{U}\left(\frac{\varphi - \varphi_j}{\sqrt{T}}\right)$

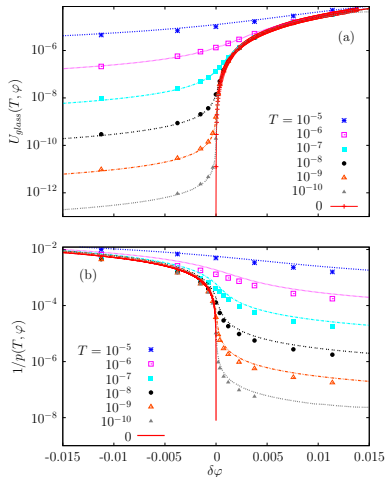
Pressure: $P(\varphi, T) = \sqrt{T} \mathcal{P}\left(\frac{\varphi - \varphi_j}{\sqrt{T}}\right)$

Mean square displacement:

$$A(\varphi, T) = \sqrt{T} \mathcal{A}\left(\frac{\varphi - \varphi_j}{\sqrt{T}}\right)$$

Effective temperature:

$$T_{\text{eff}}(\varphi, T) = \frac{T}{m(\varphi, T)} = \sqrt{T} \mathcal{T}\left(\frac{\varphi - \varphi_j}{\sqrt{T}}\right)$$



[Parisi, FZ, RMP 82, 789 (2010)]

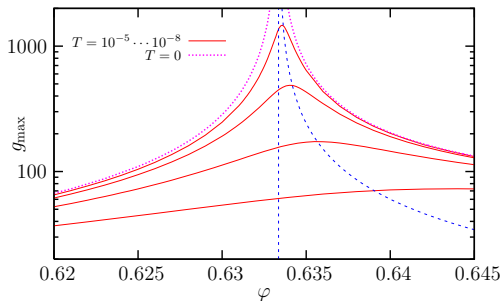
[Berthier, Jacquin and FZ, PRL 106, 135702 (2011) - PRE 84, 051103 (2011)]

[Otsuki, Hayakawa, arXiv:1111.1313]

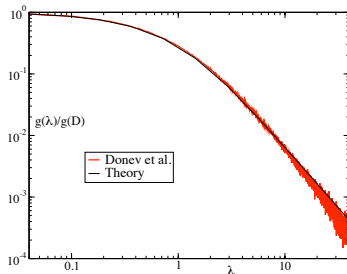
Mean field theory is not so bad in $d = 3$

A set of scaling laws around φ_{GCP} (or any φ_j):

shape of the contact peak of $g(r)$



$$p \sim g(D) \sim (\varphi_j - \varphi)^{-1}$$



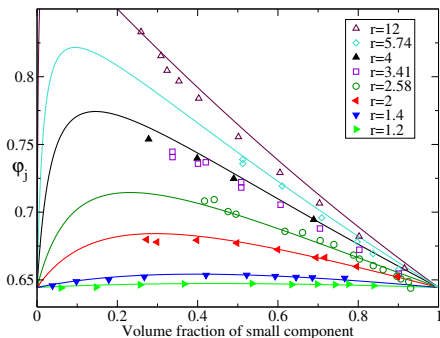
$$\frac{g(r)}{g(D)} = \Delta\left(\frac{r-D}{D}p\right) = \Delta(\lambda)$$

[Parisi, FZ, RMP 82, 789 (2010)]

[Berthier, Jacquin and FZ, PRL 106, 135702 (2011) - PRE 84, 051103 (2011)]

Mean field theory is not so bad in $d = 3$

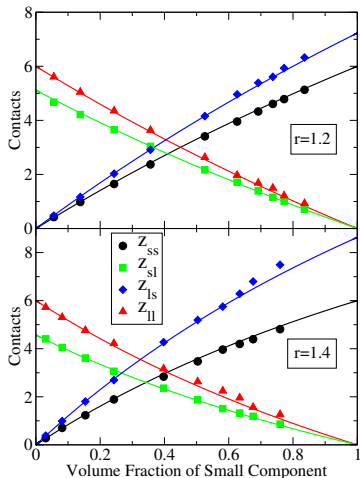
Binary mixtures: jamming density and interparticle contacts



$$r = D_A/D_B$$

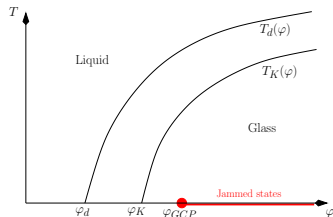
All packings are predicted to be globally isostatic.
 Partial contact numbers are almost independent of ϕ_j .

[Biazzo, Caltagirone, Parisi, FZ, PRL 102, 195701 (2009)]



Something is missing: where are the soft modes?

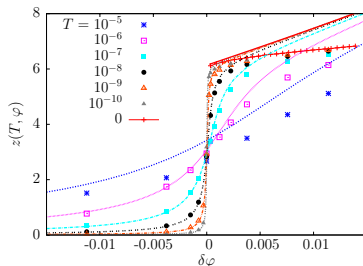
The scaling of some observables does not follow the prediction!



Replica theory: $A(\varphi, T) = \sqrt{T} \mathcal{A}\left(\frac{\varphi - \varphi_j}{\sqrt{T}}\right)$
 $\mathcal{A}(x \rightarrow -\infty) \propto |x| \Rightarrow A_{HS}(\varphi) \propto (\varphi_j - \varphi)$

Numerically: $A(\varphi, T) = T^{3/4} \mathcal{A}\left(\frac{\varphi - \varphi_j}{\sqrt{T}}\right)$
 $\mathcal{A}(x \rightarrow -\infty) \propto |x|^{3/2} \Rightarrow A_{HS}(\varphi) \propto (\varphi_j - \varphi)^{3/2}$

[Berthier, in preparation]



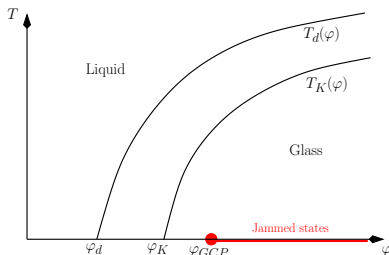
Replica theory: $z(\varphi, T = 0) \propto \varphi - \varphi_j$

Numerically: $z(\varphi, T = 0) \propto \sqrt{\varphi - \varphi_j}$

[Berthier, Jacquin and FZ, PRE 84, 051103 (2011)]

These “anomalous” scalings are related to the presence of soft modes around φ_j [Wyart]
 They persist for large d and on the Bethe lattice [Mari, Krzakala, Kurchan]

Conclusions



- A glass transition temperature $T_d(\varphi)$. Below it, history dependence and lack of equilibration.
- Jamming happens inside the glass region. The “random close packing” point depends on the algorithm used to create amorphous packings.
- Mean field theory predicts the existence of a “glass close packing”, that would be reached only through an infinitely slow annealing. This is an upper bound to any amorphous jammed state.
- In large d , $\varphi_{GCP} \sim 2^{-d} d \log d$. Amorphous packings seem to evolve smoothly with dimension.
- Mean field theory might be exact at large d , and it is a reasonable approximation in low d .
- Is it possible to include the soft modes physics in the mean field description?