

# Structural Signatures of Mobility in Jammed and Glassy Systems

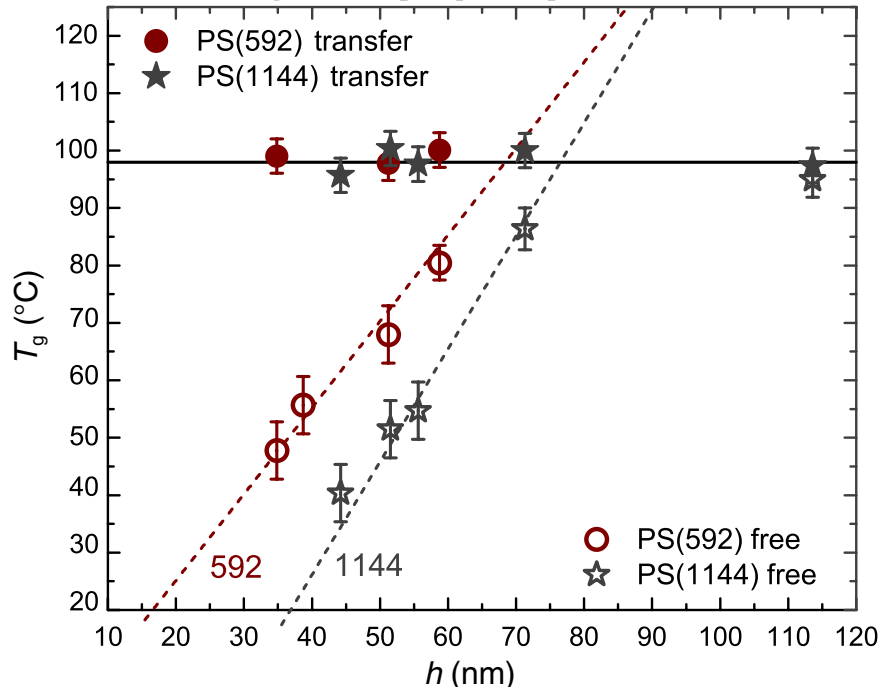
Andrea J. Liu

Department of Physics & Astronomy  
University of Pennsylvania

Lisa Manning	Syracuse
Sam Schoenholz	UPenn
Ekin Dogus Cubuk	Harvard
Brad Malone	Harvard
Tim Kaxiras	Harvard
Joerg Rottler	U British Columbia
Rob Riggleman	UPenn
Jennifer Rieser	UPenn
Doug Durian	UPenn
Daniel Sussman	UPenn
Carl Goodrich	UPenn
Sid Nagel	UChicago

# Free Surfaces

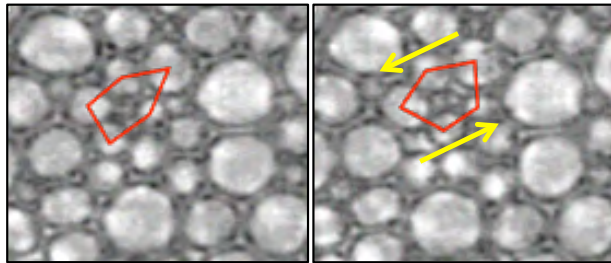
## thin glassy polymer films



- experiments see thin mobile layer of about 5-10 particle diameters
- no structural quantity has been identified with gradients on this length scale

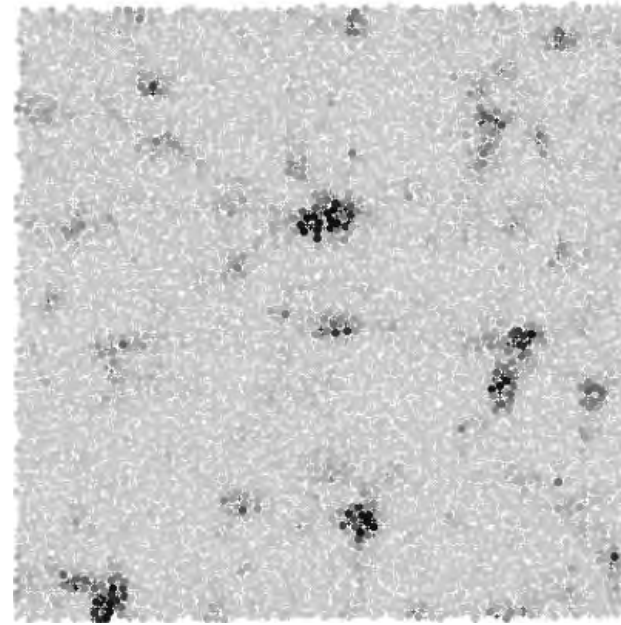
Courtesy of O. Baumchen et al. PRL (2012)

# Localized Rearrangements in Disordered Solids



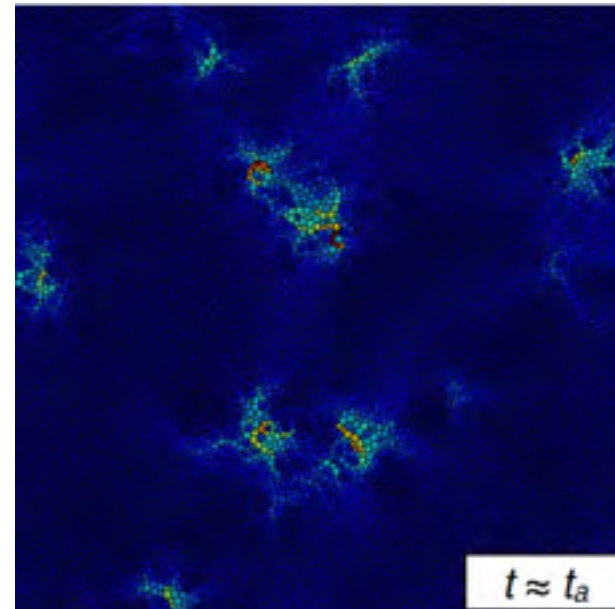
shaving  
cream

Courtesy of  
DJ Durian



2D binary  
Lennard-Jones

sheared glass



supercooled  
liquid

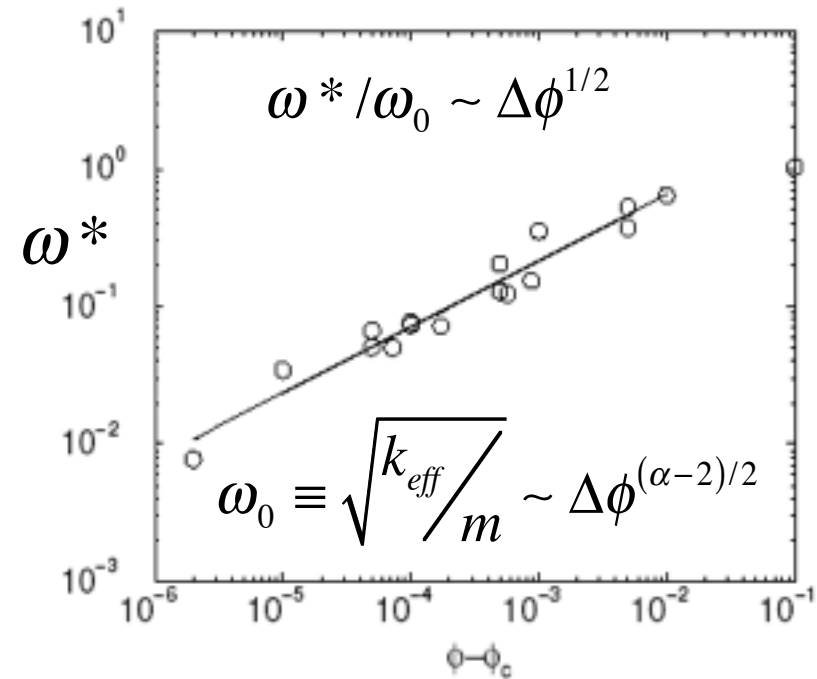
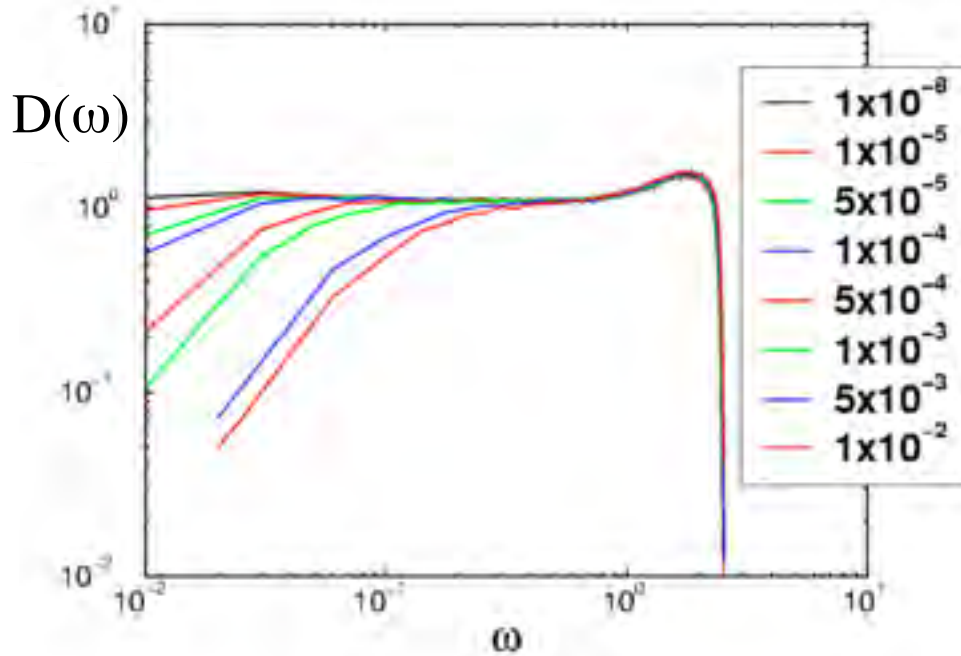
Keys, et al  
PRX (2011)

- Aim: identify population of “**flow defects**,” analogous to dislocations in crystals, where rearrangements are likely to occur

- Standard structural quantities fail to predict these

# Starting Point: Vibrations in Sphere Packings

L. E. Silbert, A. J. Liu, S. R. Nagel, PRL **95**, 098301 ('05)



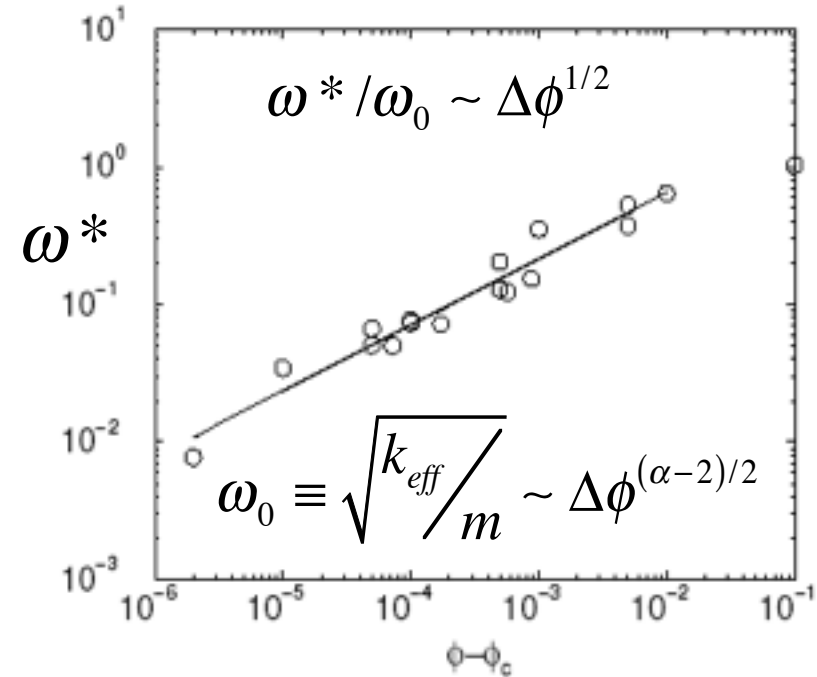
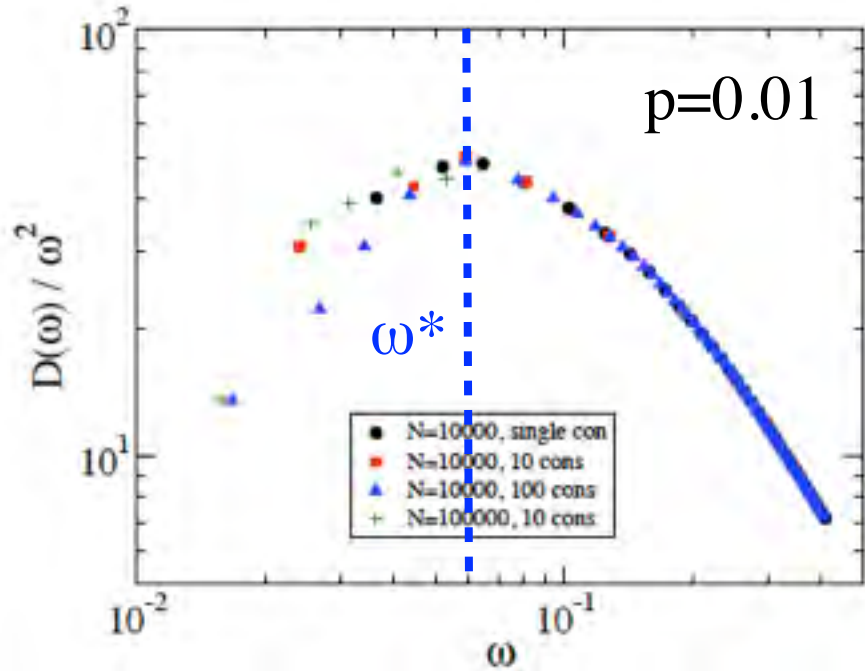
- New class of excitations originates from soft modes at Point J

M. Wyart, S.R. Nagel, T.A. Witten, EPL **72**, 486 (05)

- Related to diverging length scales  $\ell_L \approx c_L/\omega^*$   
 $\ell_T \approx c_T/\omega^*$

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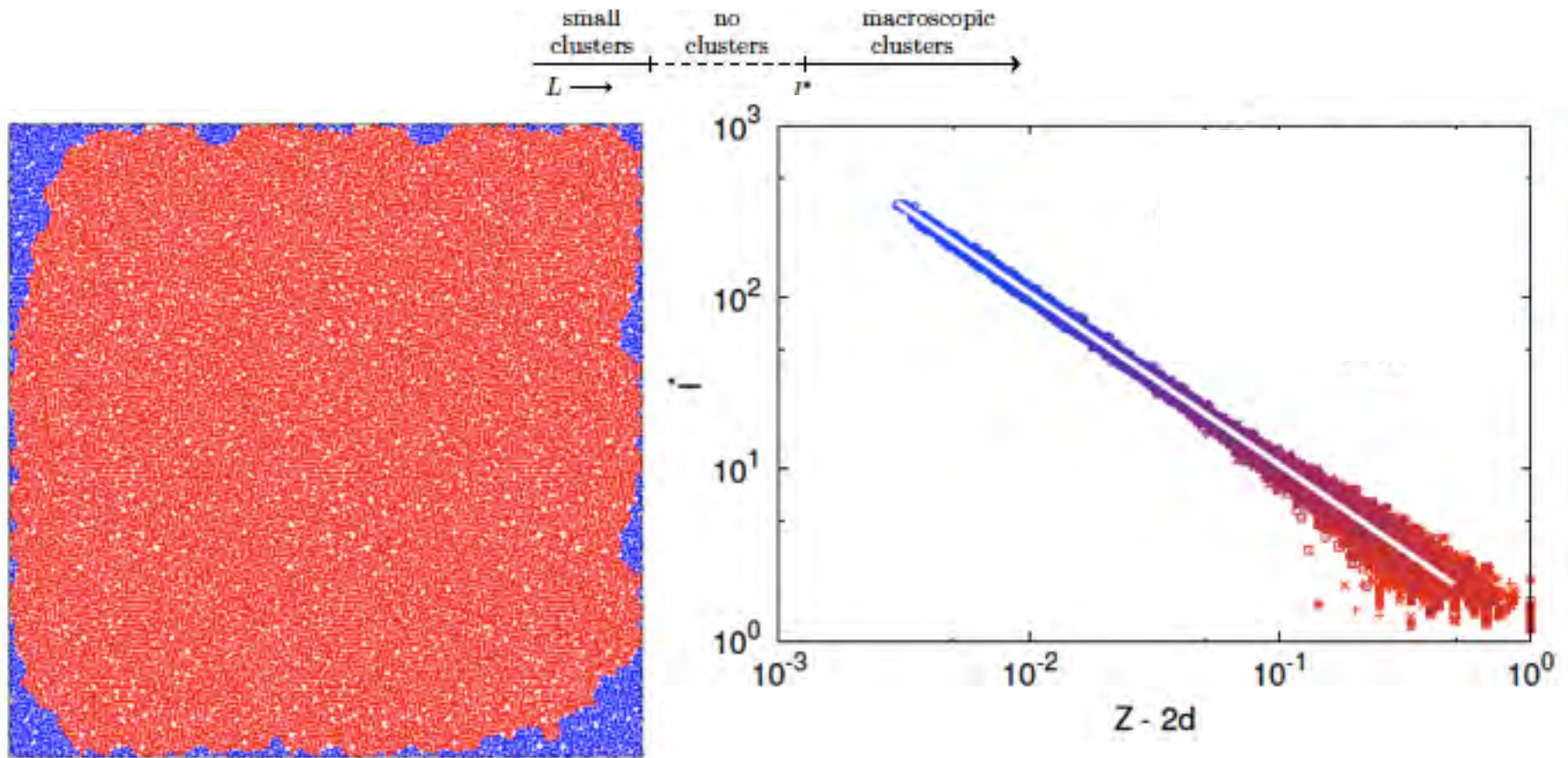
- Related to diverging length scales  $\ell_L \approx c_L / \omega^*$   
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# Stability to Boundary Cutting: $\ell_L$

M. Wyart, S.R. Nagel, T.A. Witten, EPL **72**, 486 (05)

Size of **smallest** macroscopic **rigid cluster** for system with a free boundary of any shape or size

Goodrich, Ellenbroek, Liu Soft Matter (2013)



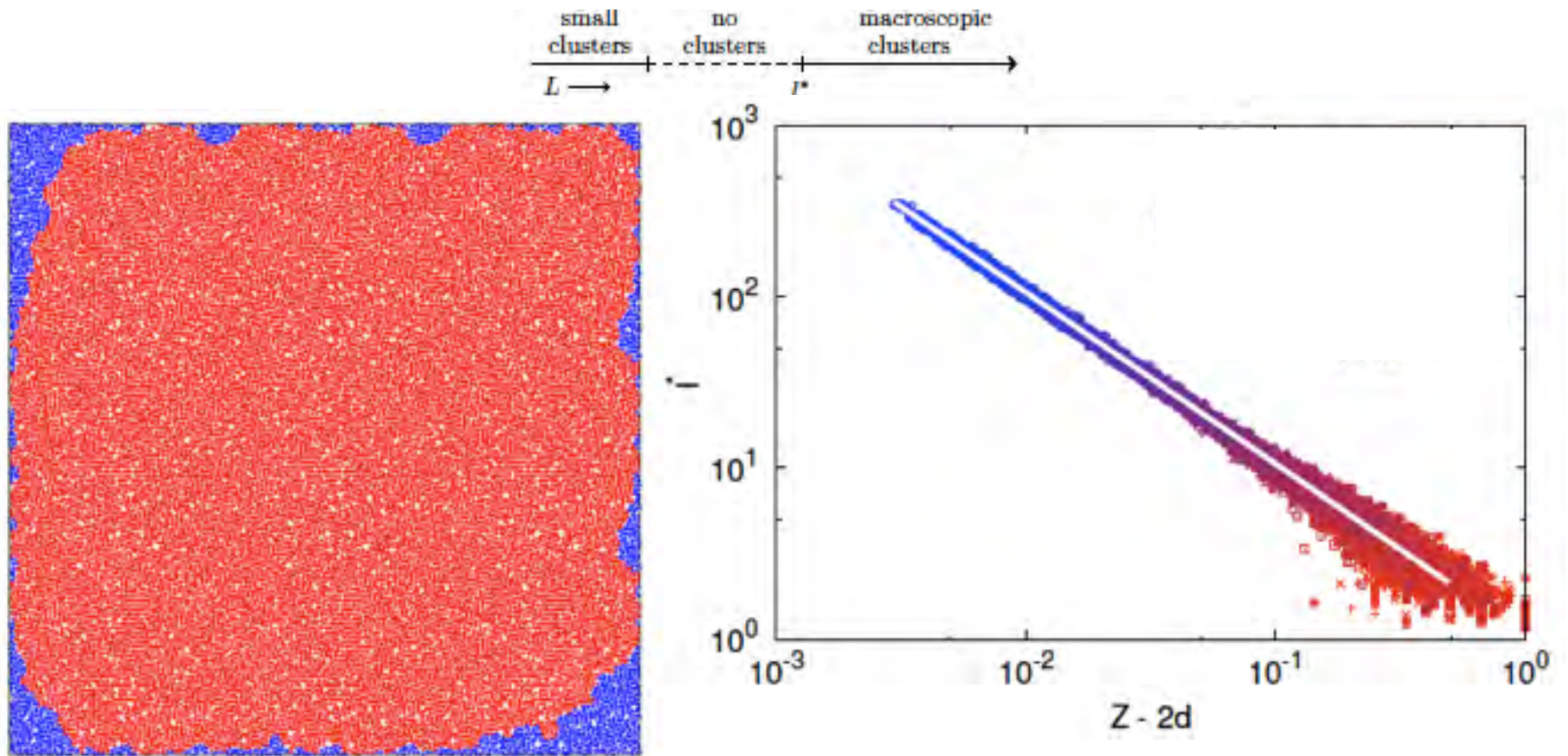
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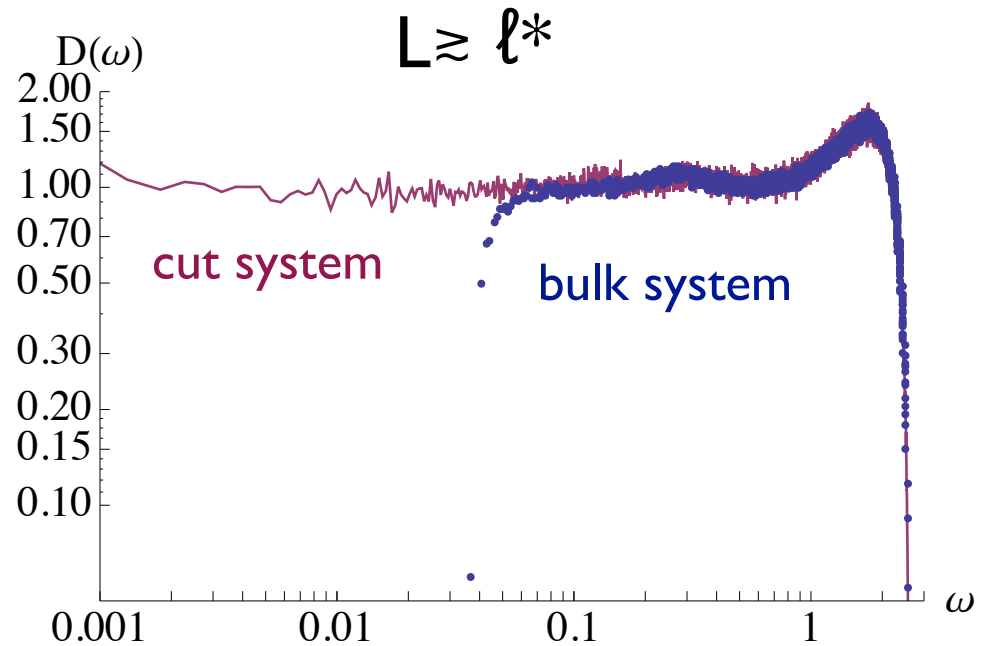
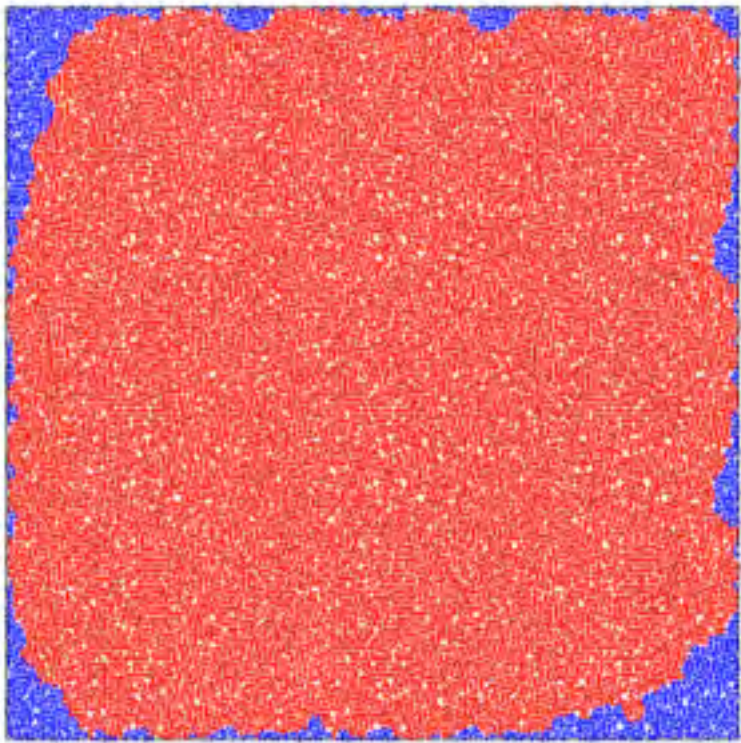
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# Free Surfaces

Goodrich, Liu, Nagel, Soft Matter (2013)



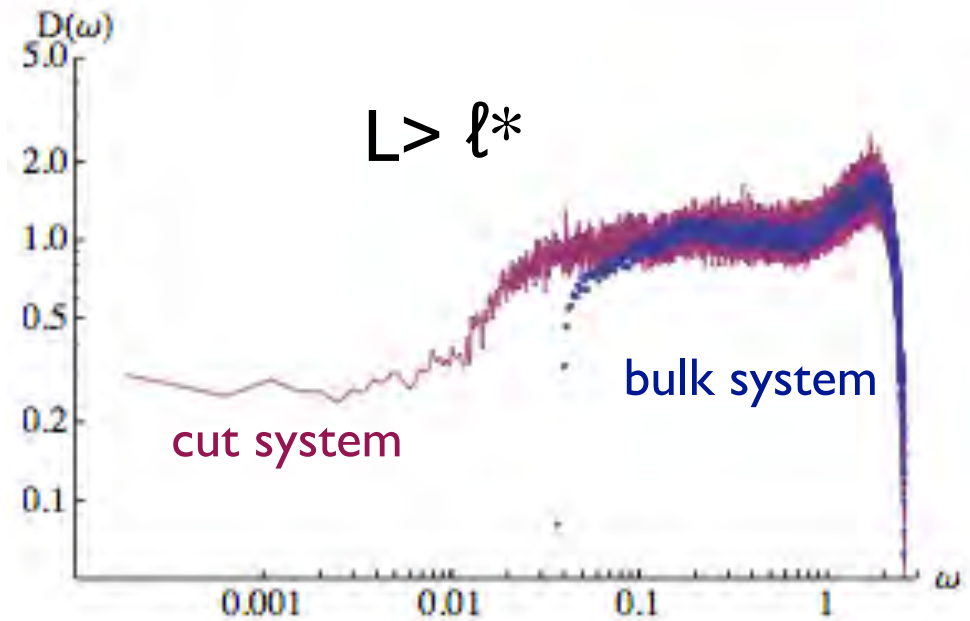
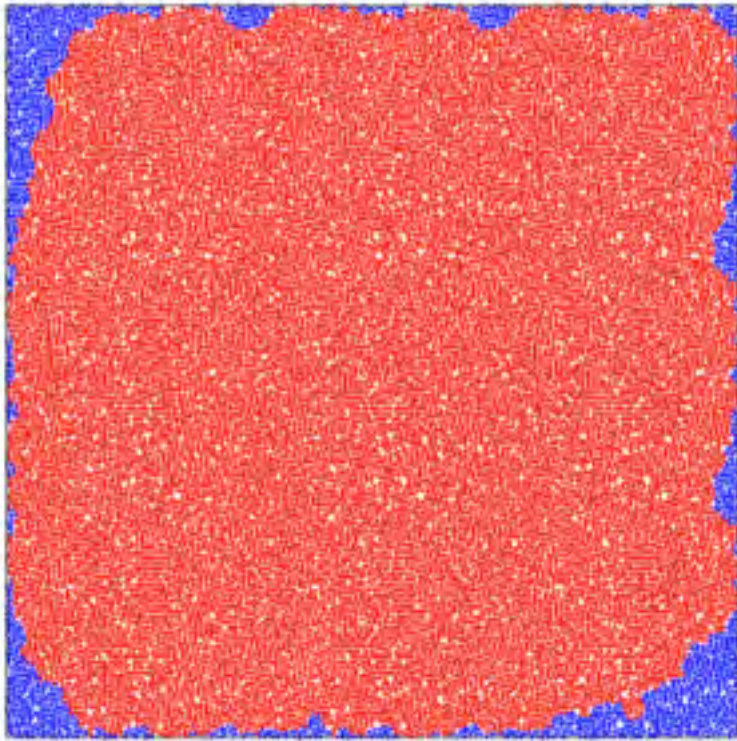
Daniel Sussman, Carl Goodrich

- There are zero frequency modes localized to the surface to within particle diameter
- There are also extra low frequency modes in excess of surface plane wave modes (Rayleigh waves, etc)



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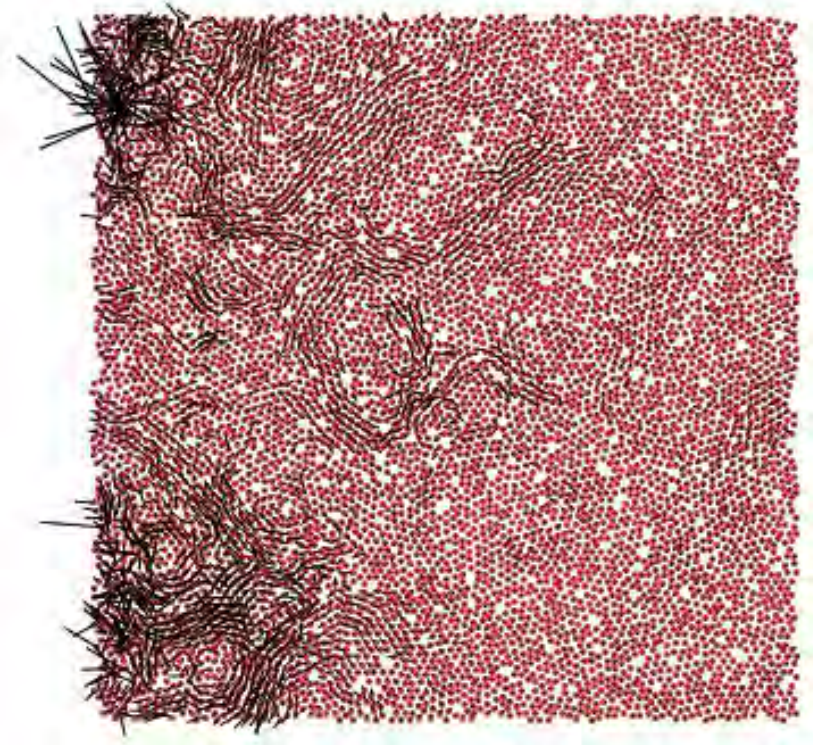
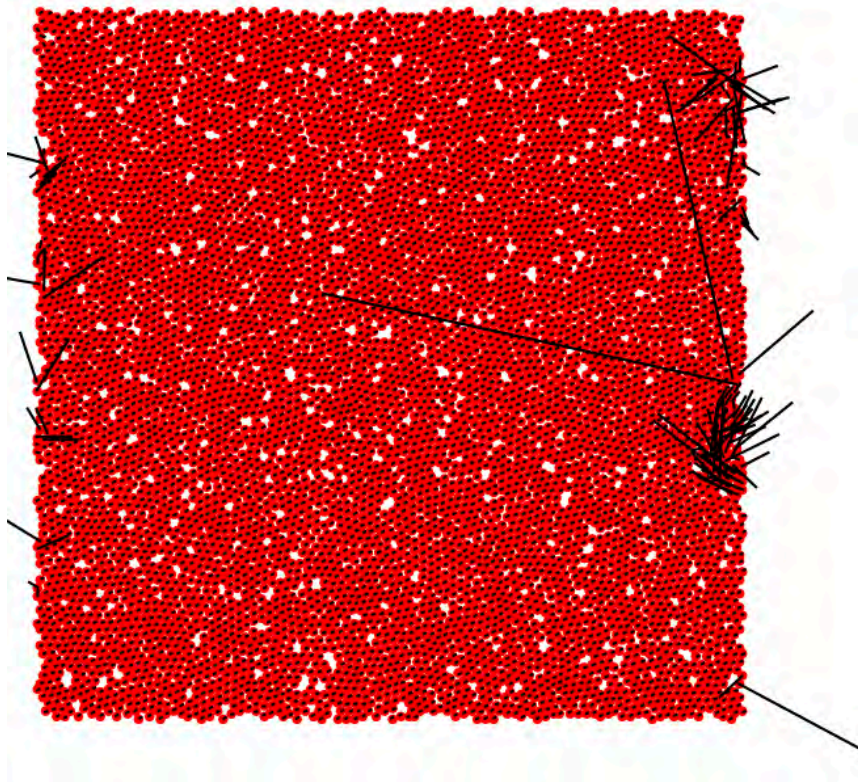


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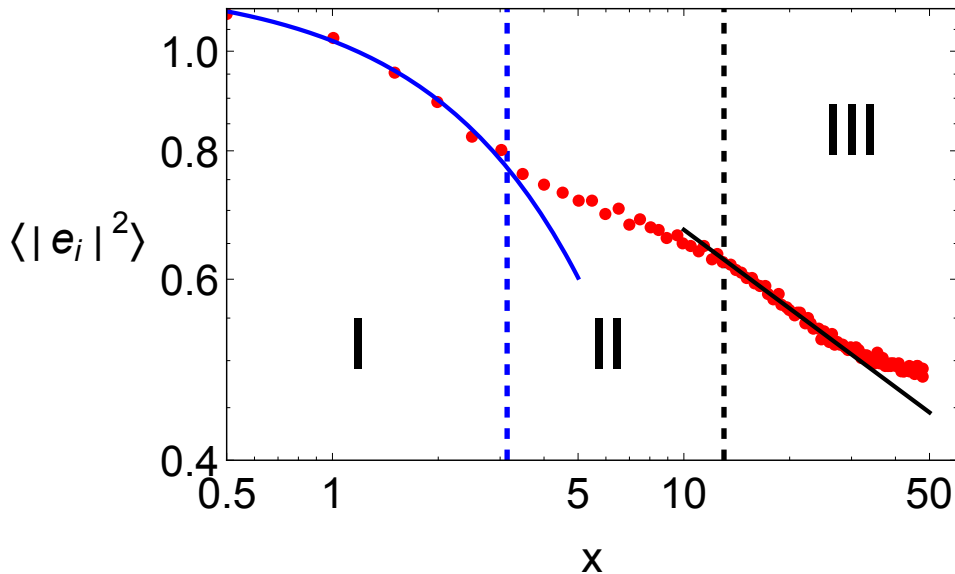
# Low-Frequency Surface Modes are Localized to Surface

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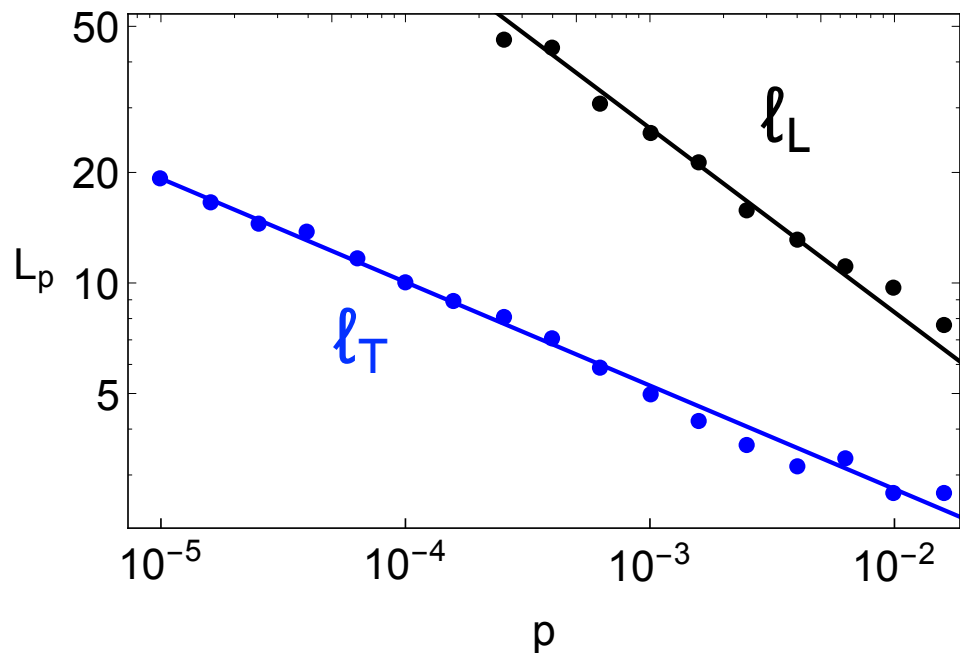
- Measure polarization vector magnitude vs distance from surface

# Penetration Depth of Surface Localized Modes



- Three regimes for decay of polarization vector magnitude
- Regime I consistent with exponential decay
- Regime III independent of pressure
- Extract crossover lengths separating Regimes I & II and II&III

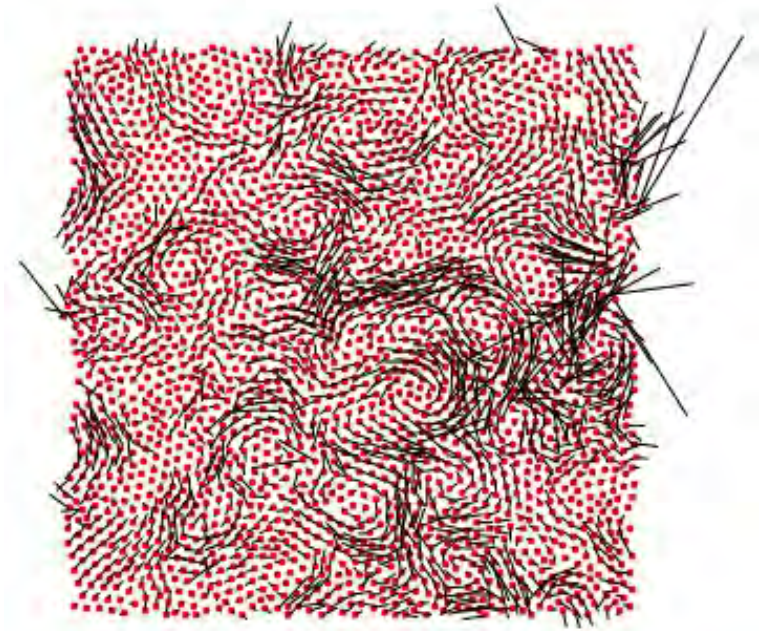
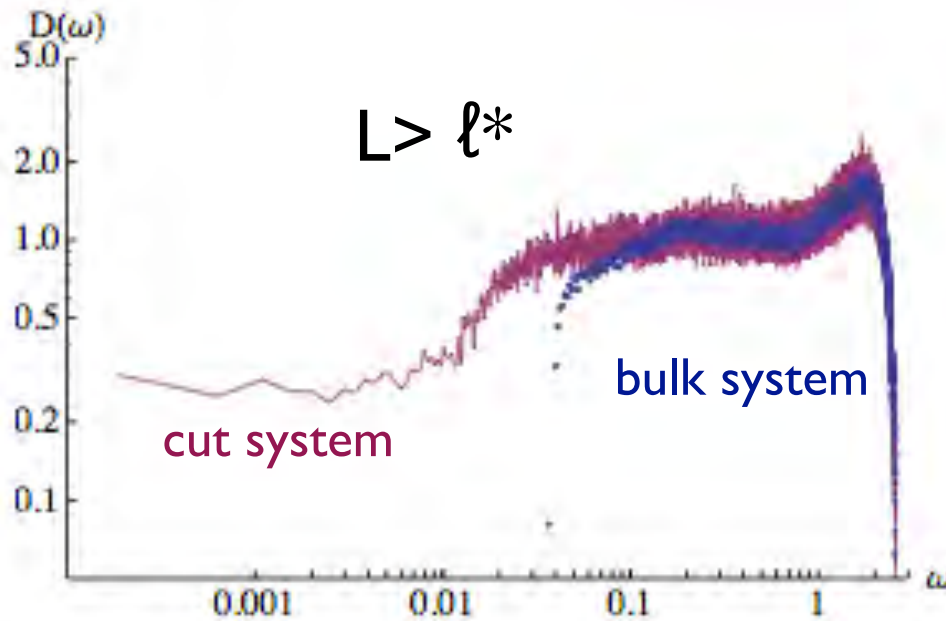
# Penetration Depth of Surface Localized Modes



- Regime I/II crossover scales as  $l_T$  which diverges at jamming transition as  $p^{-1/4}$
- Regime II/III crossover scales as  $l_L$  which diverges at jamming transition as  $p^{-1/2}$
- Consistent with response to local bond perturbations (Lerner, During, Wyart, *Soft Matter* (2014))

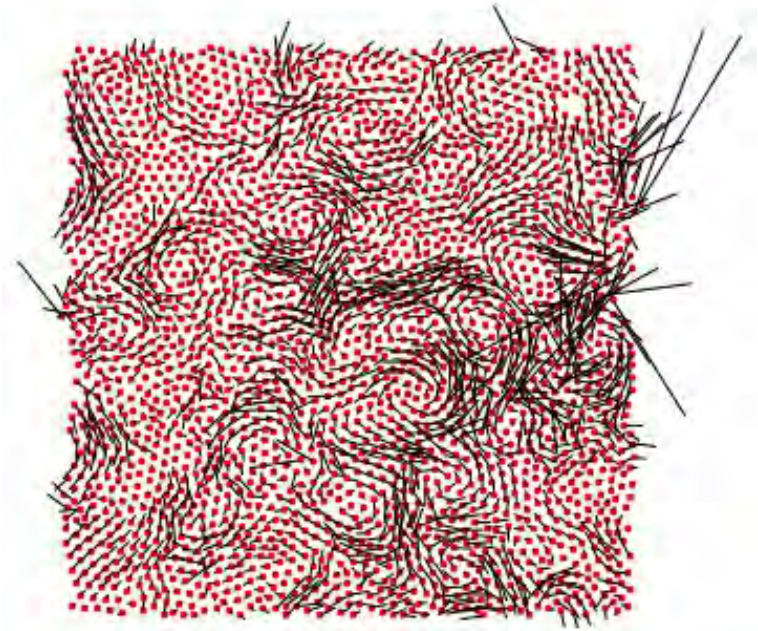
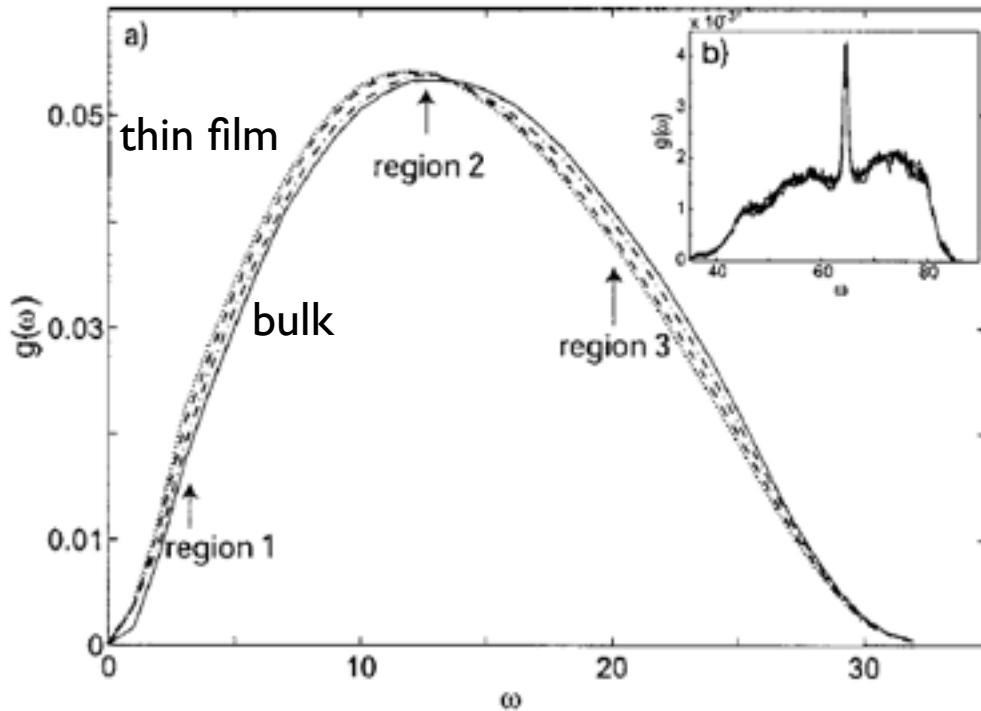
Low frequency surface modes penetrate into system much further than particle diameter. Can this explain glassy thin films?

# Problem:



- In polymer thin films (or Lennard-Jones glasses), the surface localized modes lie in the same frequency range as the bulk modes so no clean separation of surface from bulk modes
- But low frequency modes still show high polarization near surface

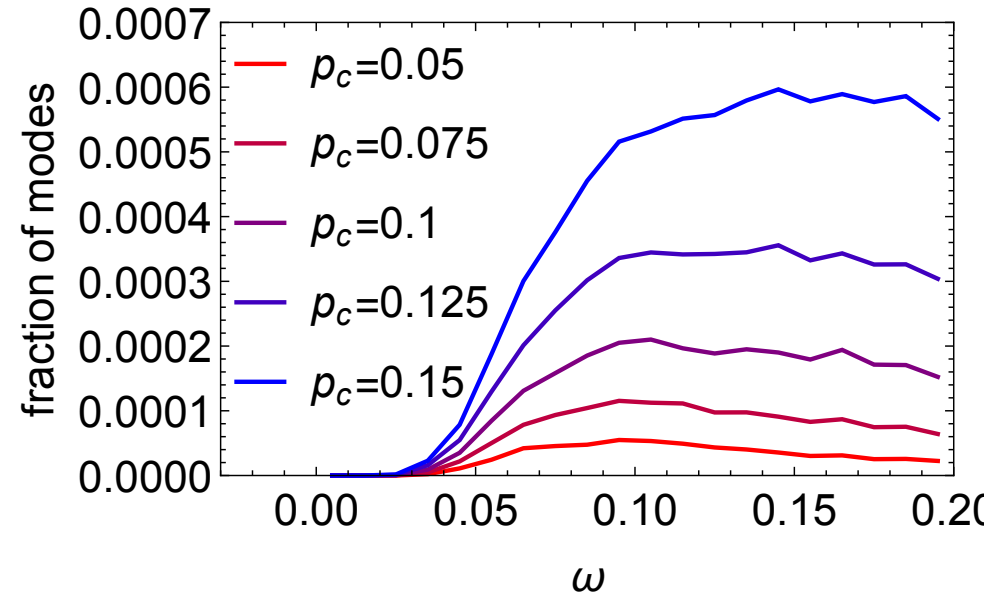
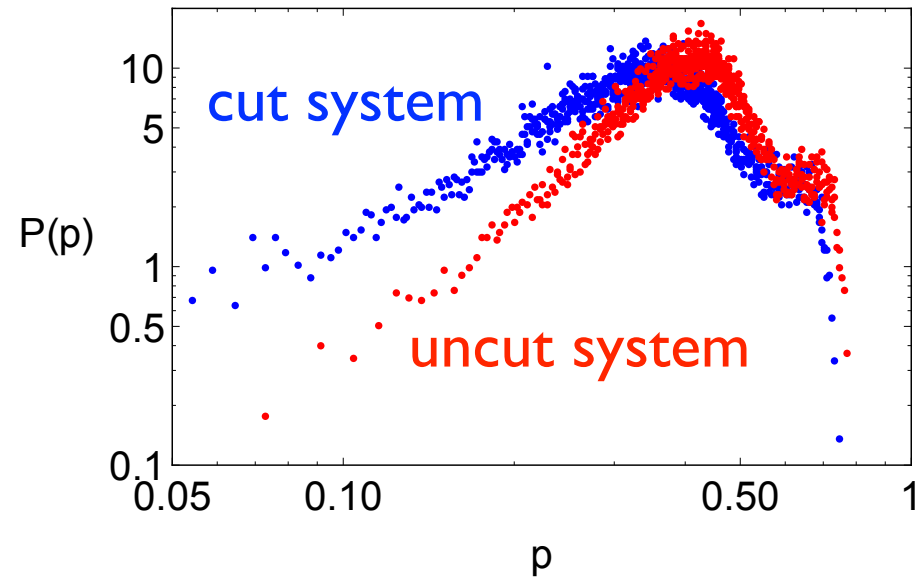
# Problem:



Jain and de Pablo, J Chem Phys (2004)

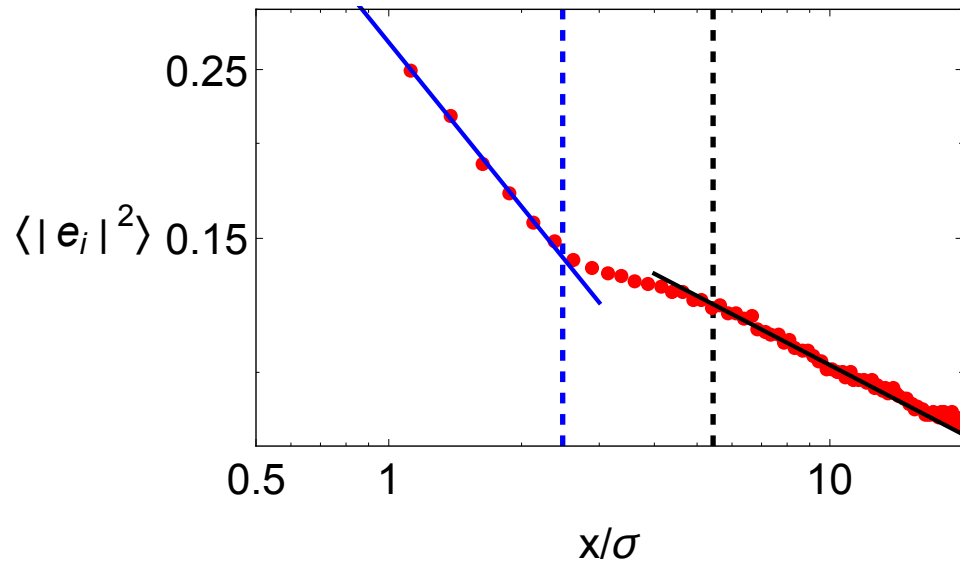
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# Focus on low-participation-ratio modes near $\omega^*$



- Cut system has more low participation-ratio modes than uncut system
- Low participation ratio modes most prevalent near  $\omega^*$
- So look at these modes

# Polarization Profile in Lennard-Jones Glass

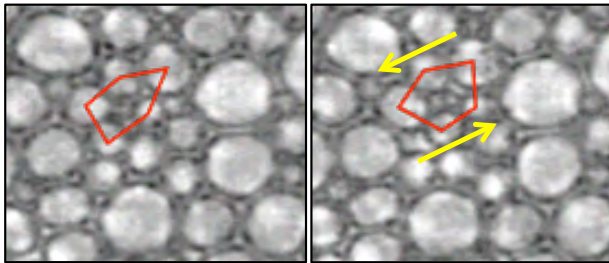


- Regime III has same form as for harmonically-repulsive spheres
- Regime I/II crossover consistent with  $\ell_T \approx c_T/\omega^*$
- Regime II/III crossover consistent with  $\ell_L \approx c_L/\omega^*$

- Polarization vector magnitude decays on scale of 5-10 particle diameters
- Consistent with observed thickness of mobile layer in thin glassy films
- High amplitude regions of low-frequency quasi localized modes are structural signature of mobile surface layer

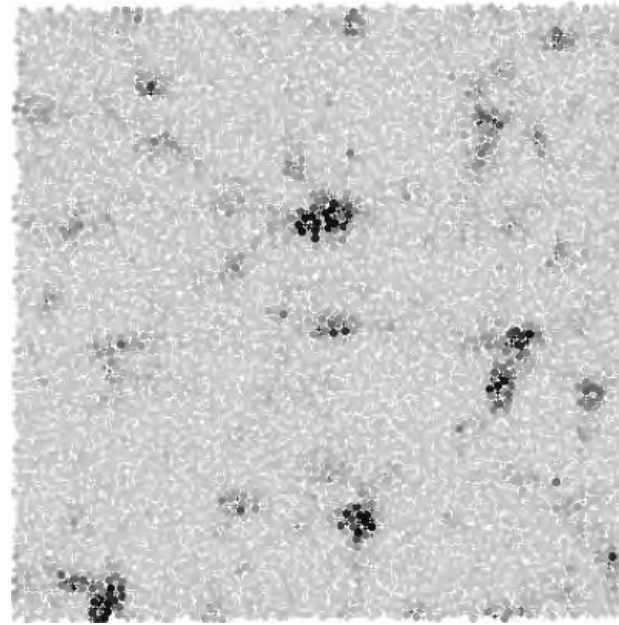


# Localized Rearrangements in Glasses and Supercooled Liquids



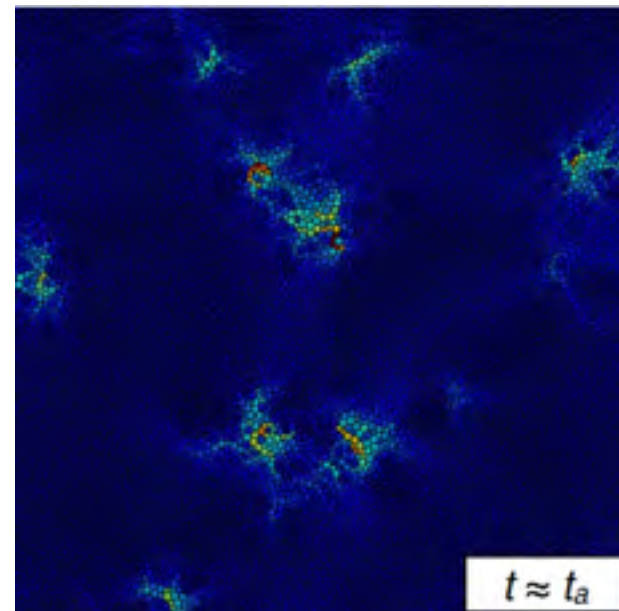
shaving  
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Courtesy of  
DJ Durian



2D binary  
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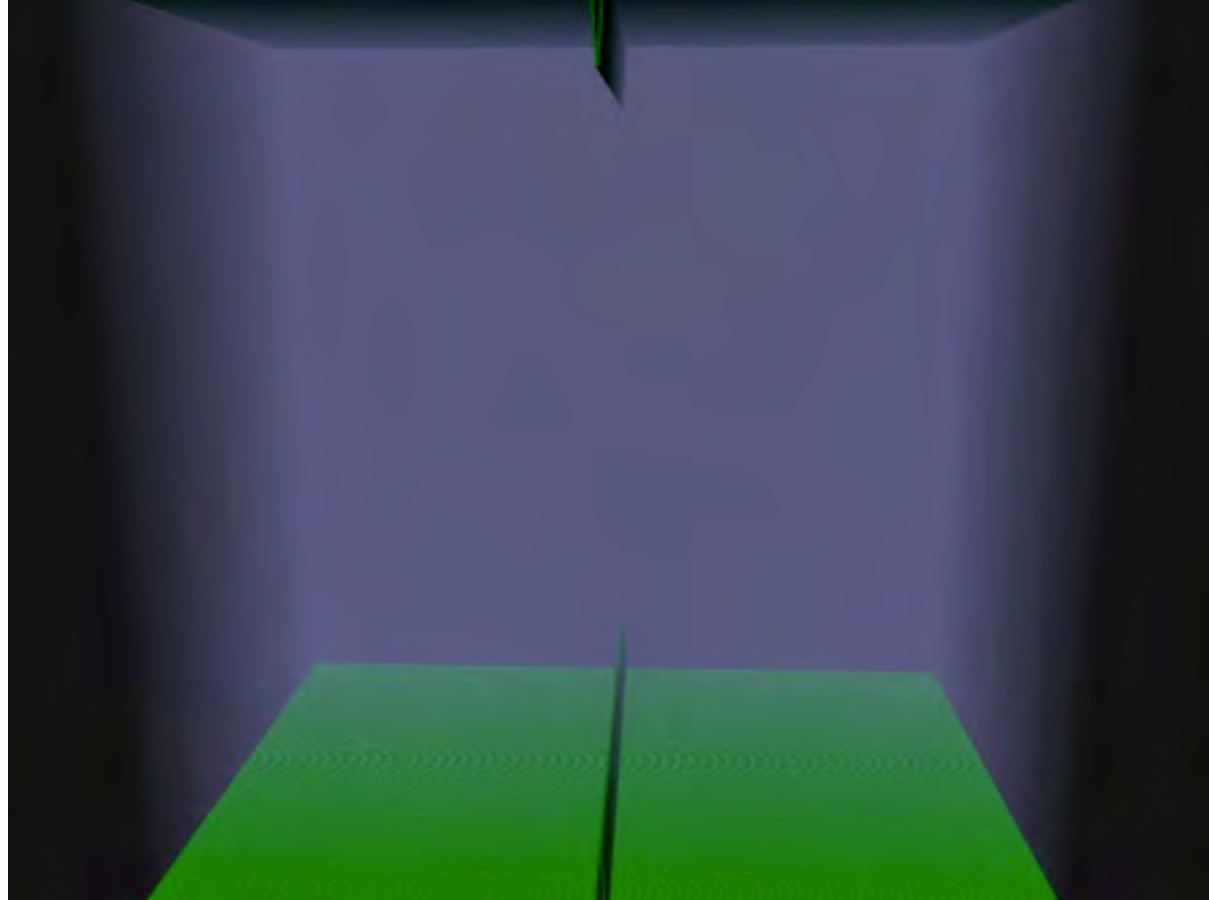
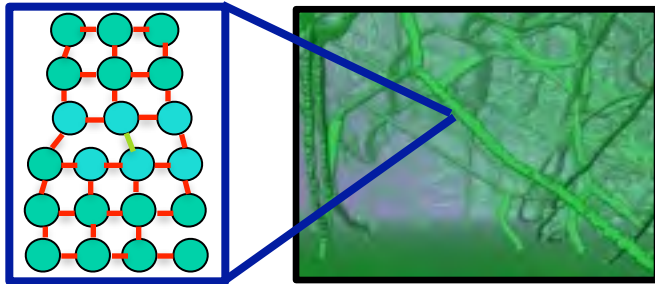
supercooled  
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Keys, et al  
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- Aim: identify population of “**flow defects**,” analogous to dislocations in crystals, where rearrangements are likely to occur
- Standard structural quantities fail to predict these

# How Crystals Flow

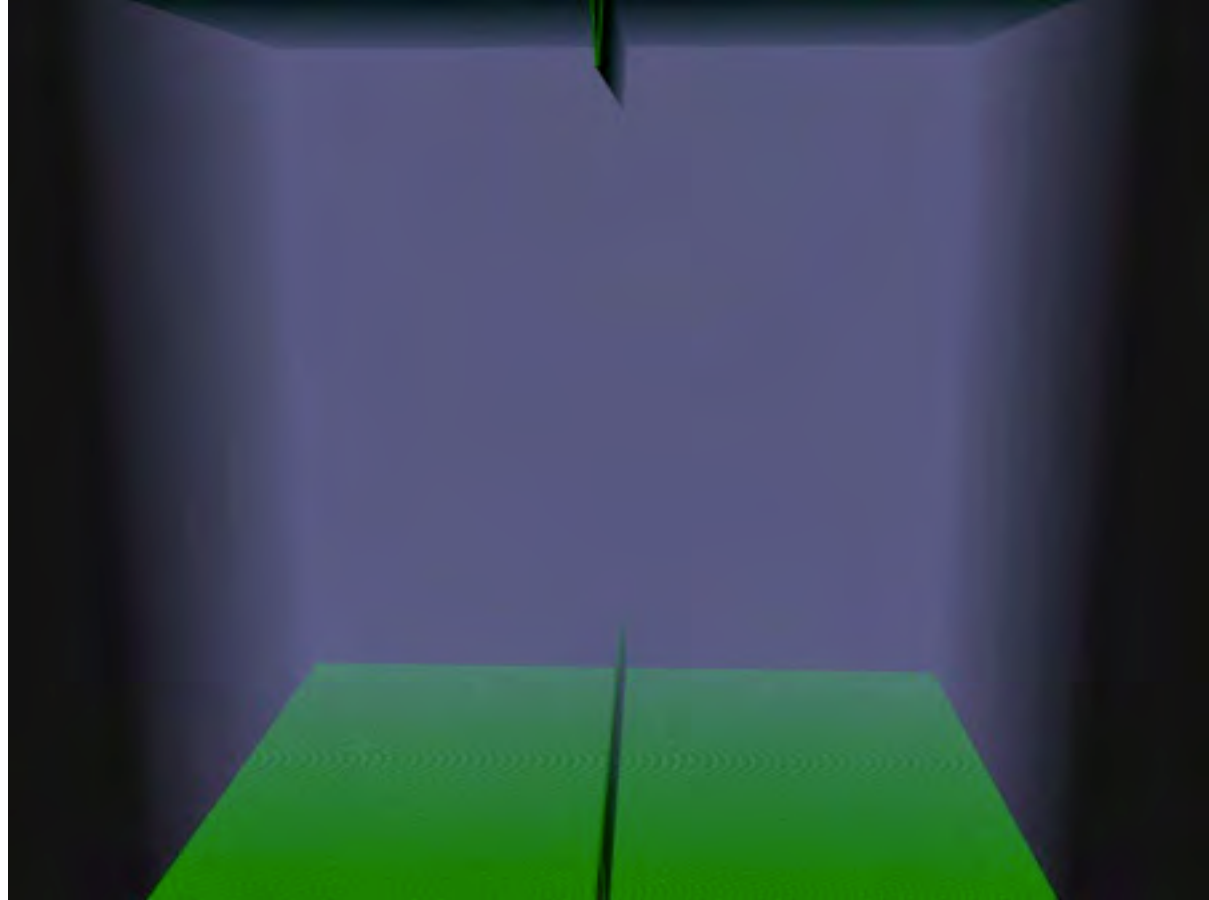
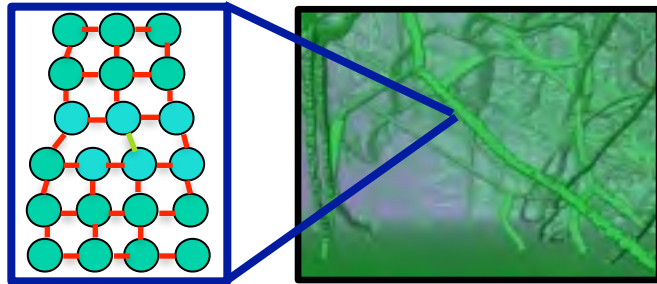
In crystalline packings, localized rearrangements occur at dislocations



Courtesy of F Abraham

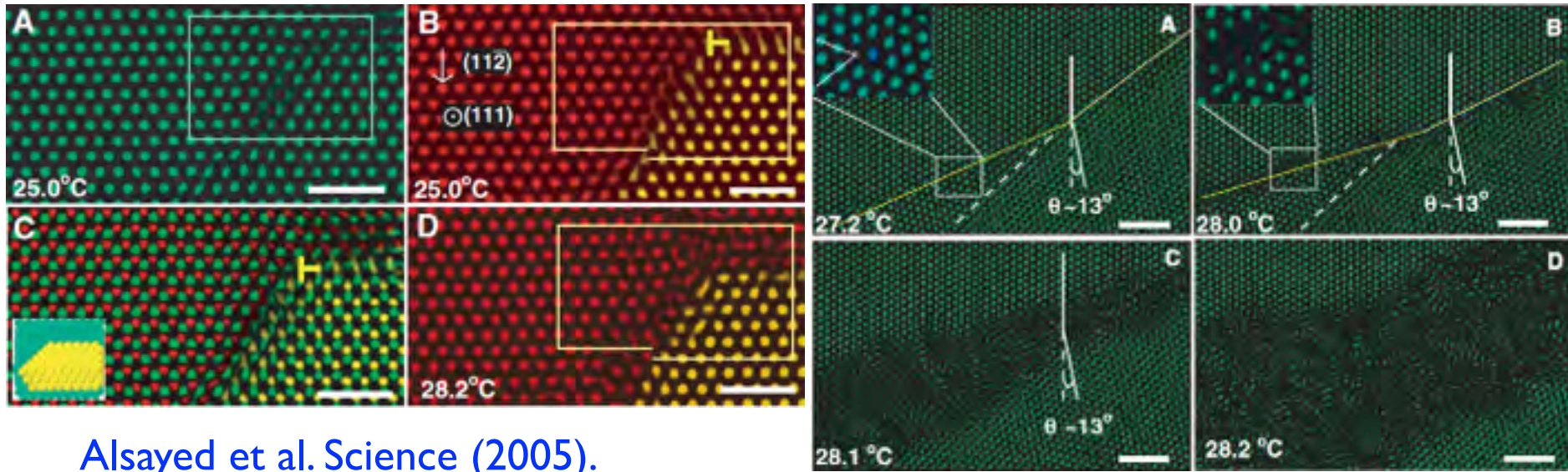
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# How Crystals Pre-Melt

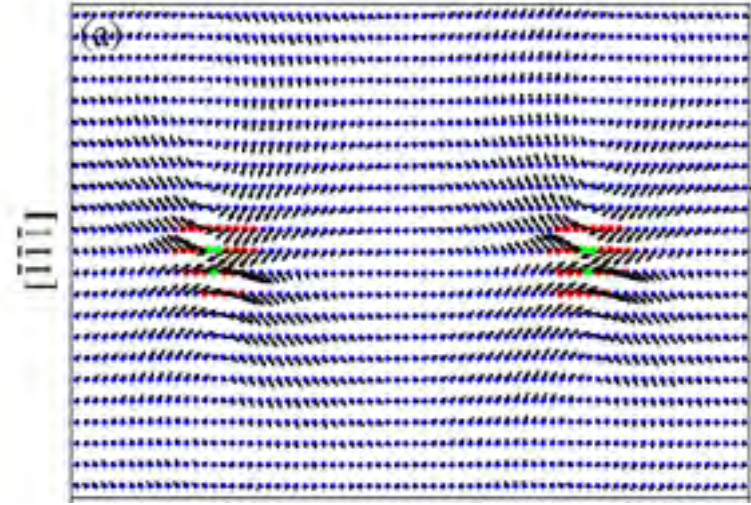


Alsayed et al. Science (2005).

- Premelting also occurs at dislocations and grain boundaries
- Dislocations vulnerable to rearrangement under mechanical load **or** temperature

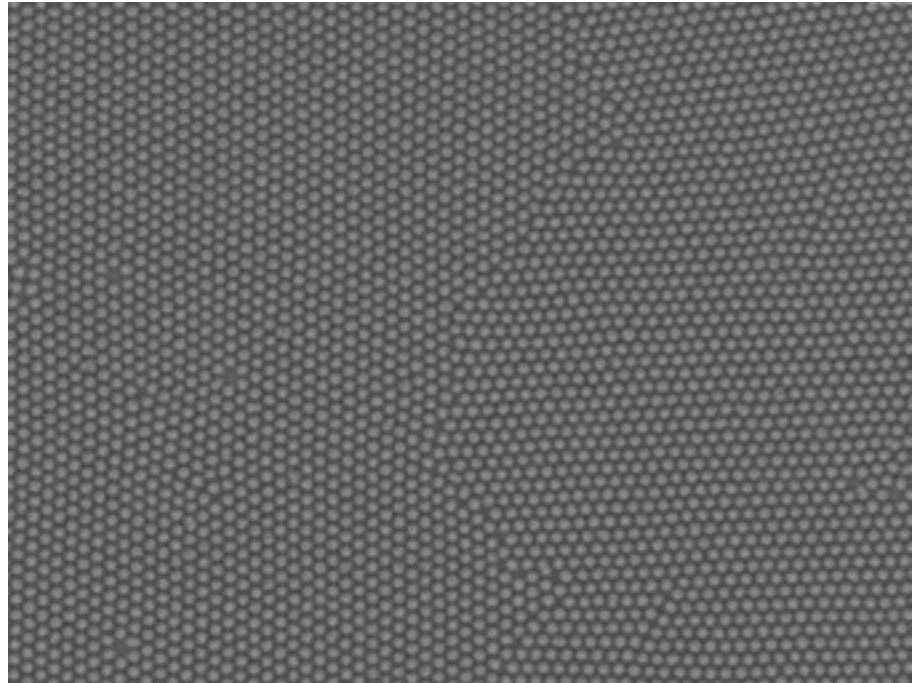
# Look at Quasilocalized Modes in Crystals

- Quasilocalized modes localize to **flow defects** (dislocations and grain boundaries but not vacancies) because these scatter sound most effectively
- look at quasi localized modes in disordered systems



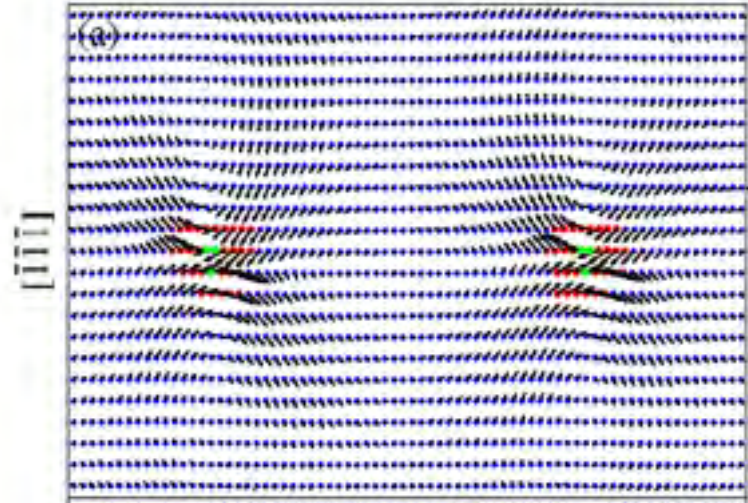
[110] Rottler, Schoenholz, Liu PRE (2014)

Chen, Still, Schoenholz,  
Aptowicz, Schindler, Maggs,  
Liu, Yodh, PRE 88, 022315  
(2013)

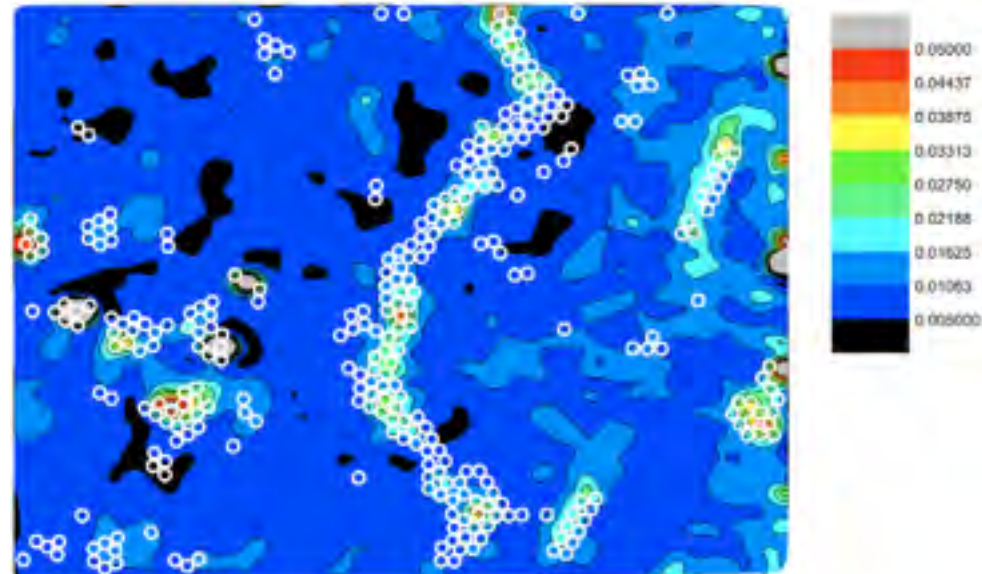


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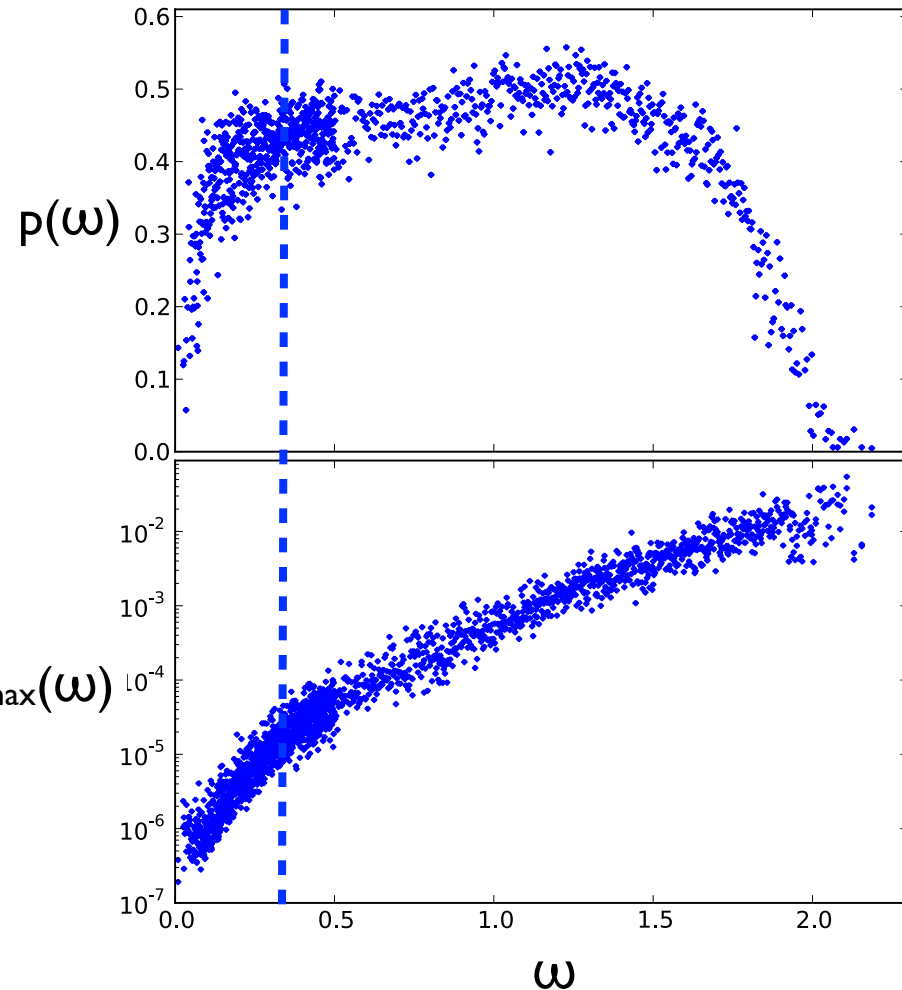


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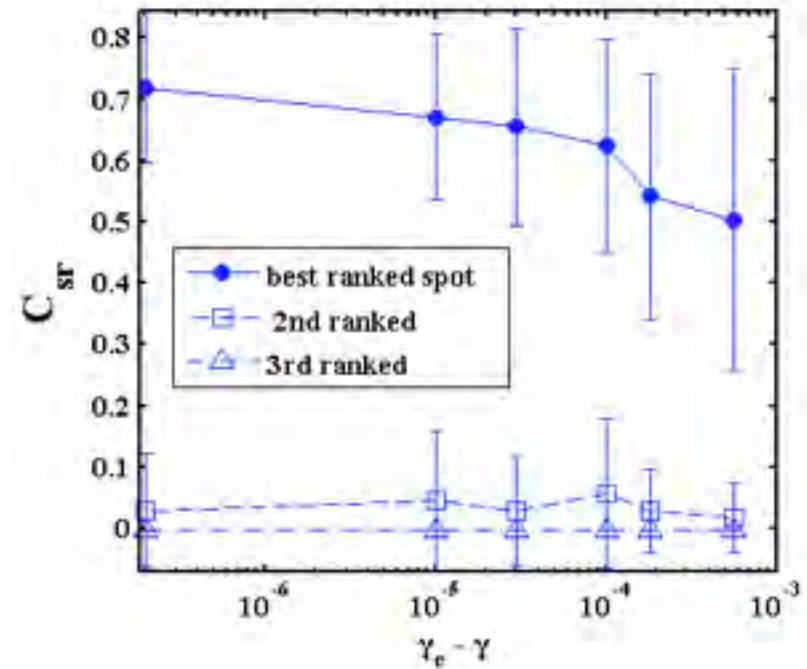
# Anharmonicity

N. Xu, Vi Vitelli, A. J. Liu, S. R. Nagel, EPL 90, 56001 (2010).

- Low-frequency quasi-localized modes have the **lowest** energy barriers to rearrangement
  - Barriers are likely to be lower if rearrangements are localized
  - These are the modes most likely to go **unstable** due to **thermal fluctuations** or **mechanical load**
- So QLM likely to contain information about **flow defects**

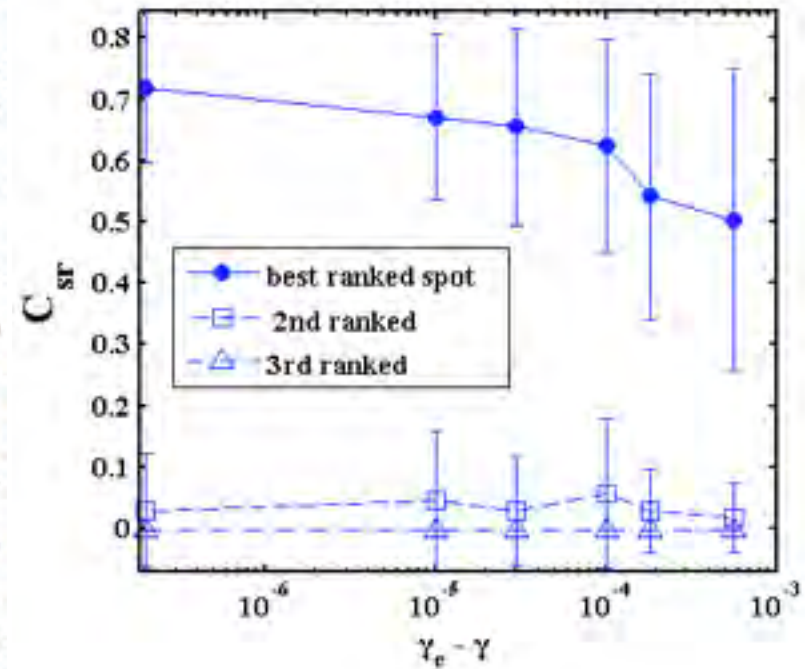
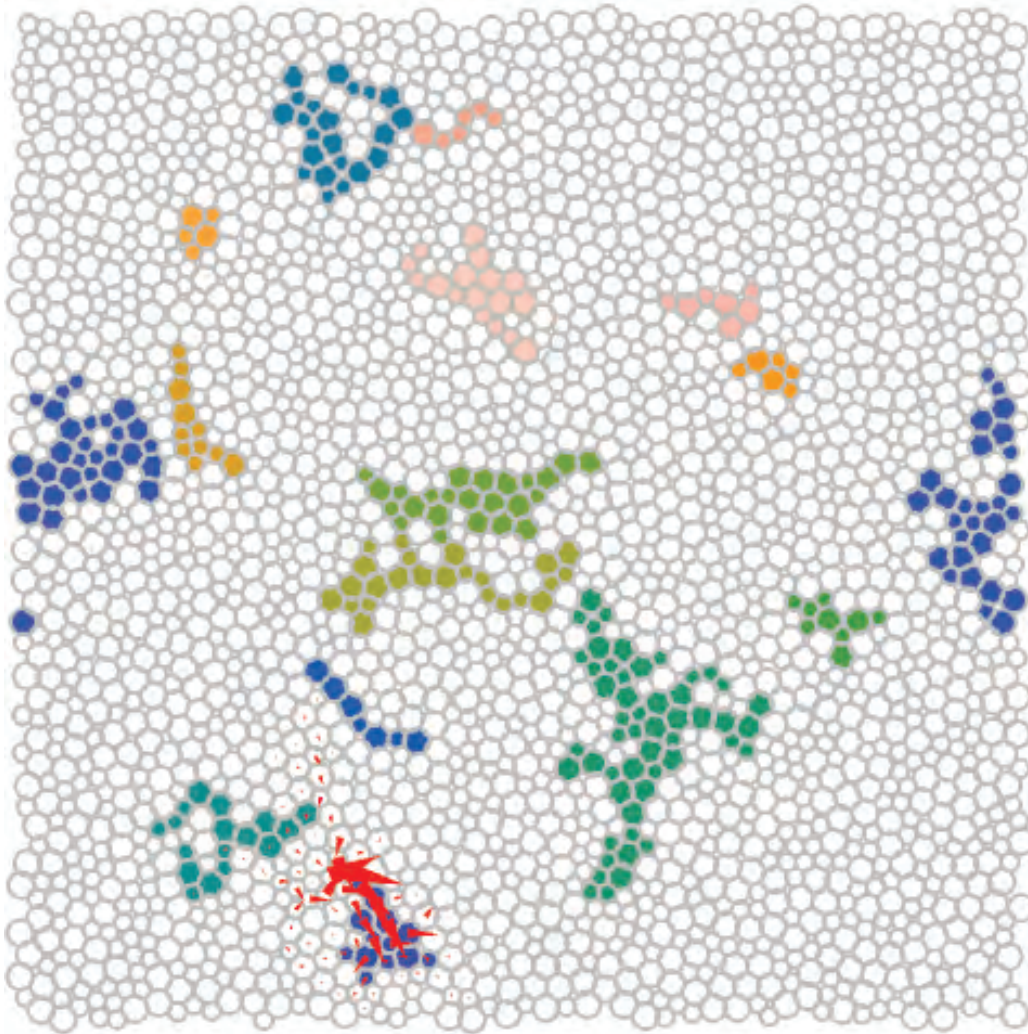


# QLM Method of Identifying Flow Defects





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M. L. Manning, A. J. Liu, *Phys. Rev. Lett.* 107, 108302 (2011)

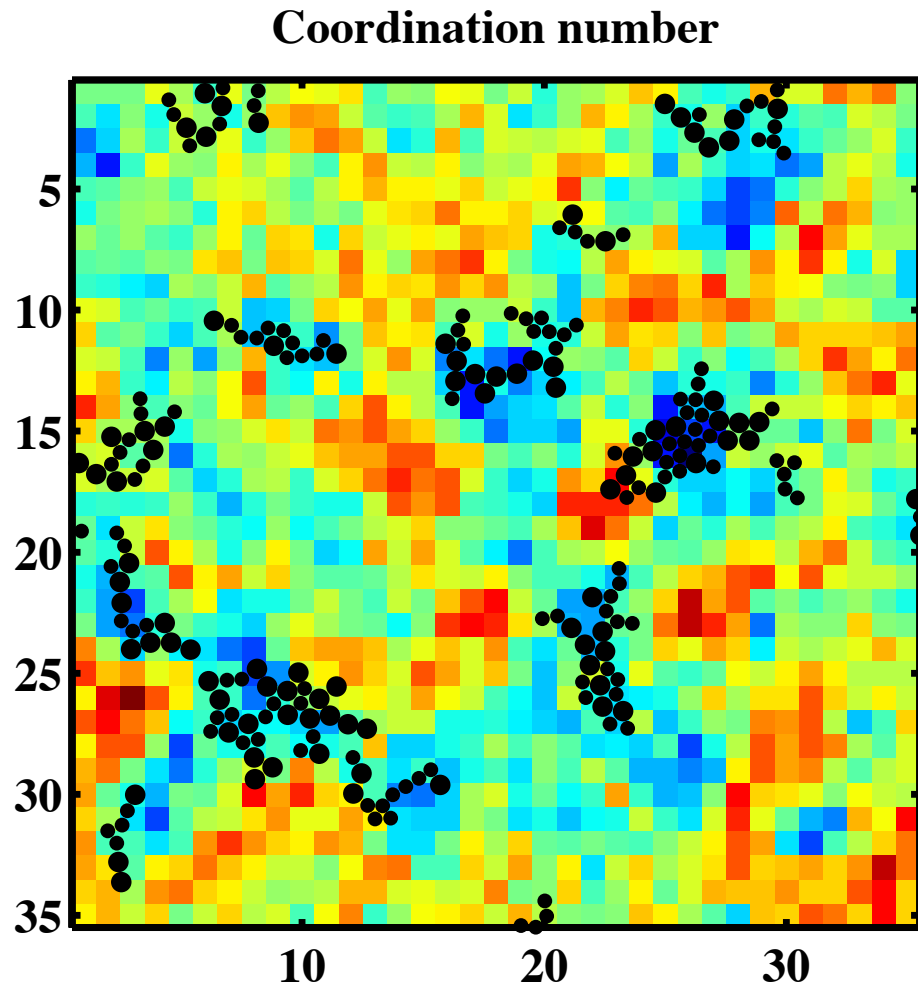
# QLM Soft Spots Are Promising, But...

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- To identify soft spots, need to
  - know interactions between particles precisely
  - quench system to obtain inherent structures
  - diagonalize dynamical matrix ( $O(N^3)$ )
- So this method cannot be applied to experimental data and is very slow even for simulation data
- We want a method for identifying regions vulnerable to rearrangement that relies on **local structure** alone
- Problem: all previous attempts have failed

# Standard Structural Quantities Don't Tell Us Where Soft Spots Are

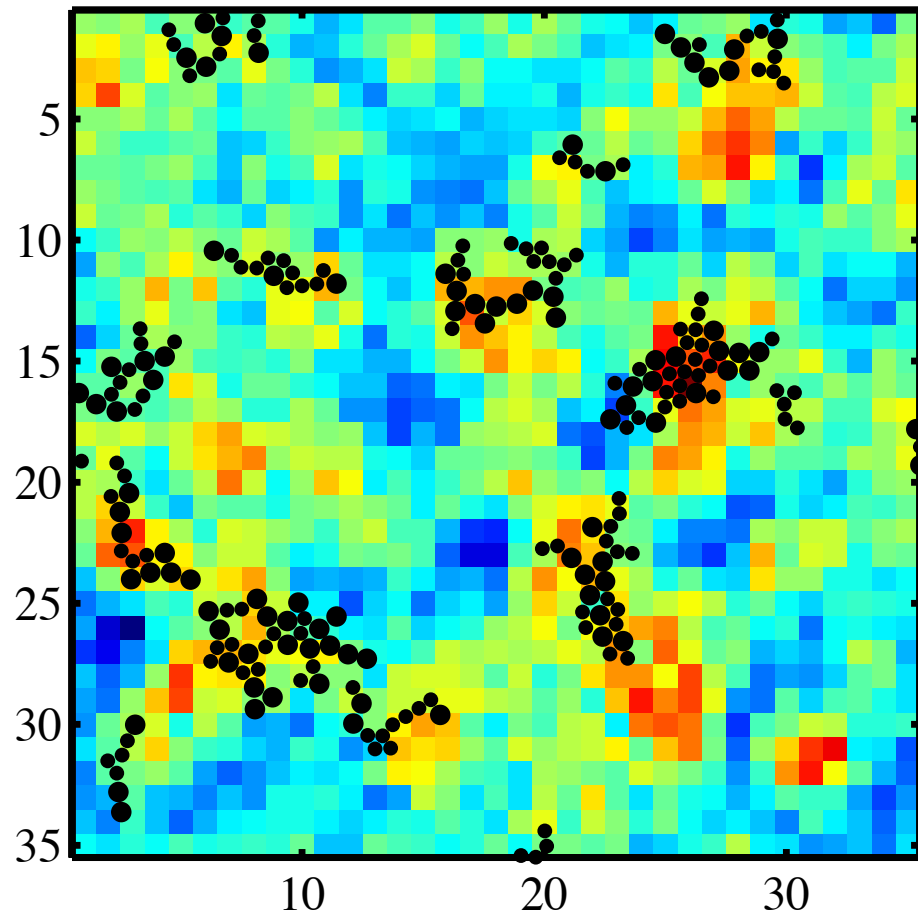
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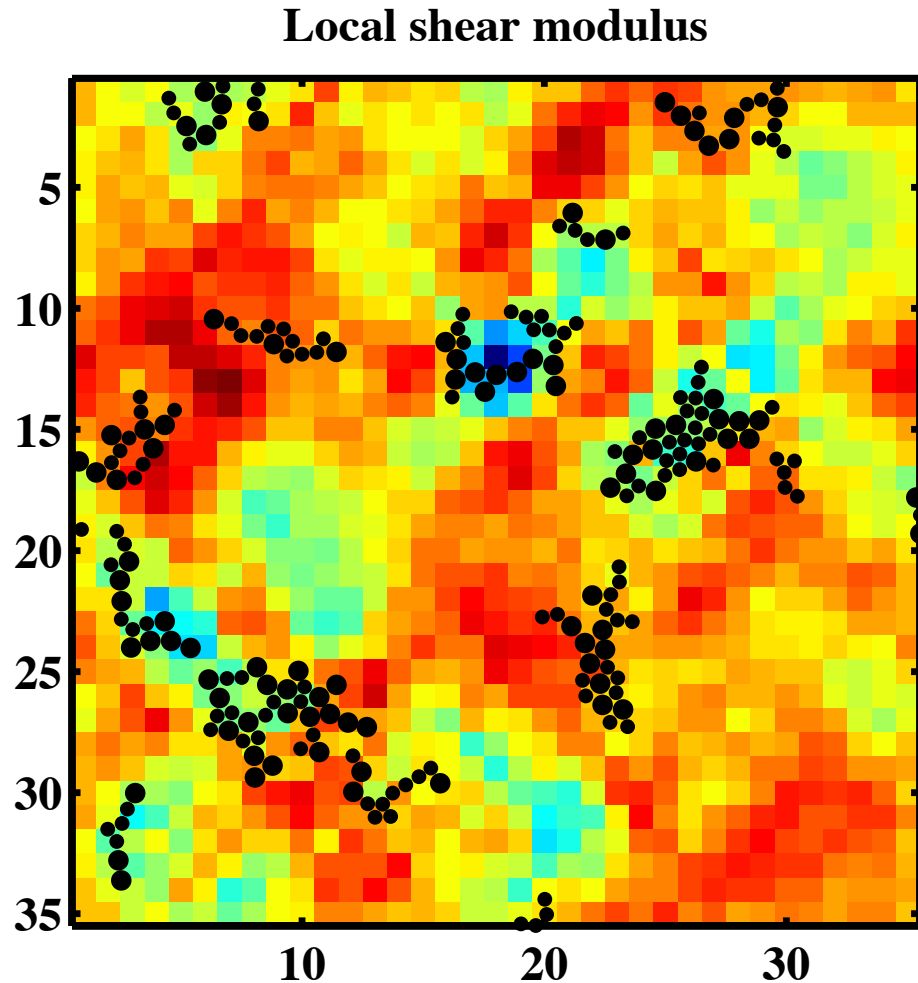
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Deviation in BOO from hexagonal



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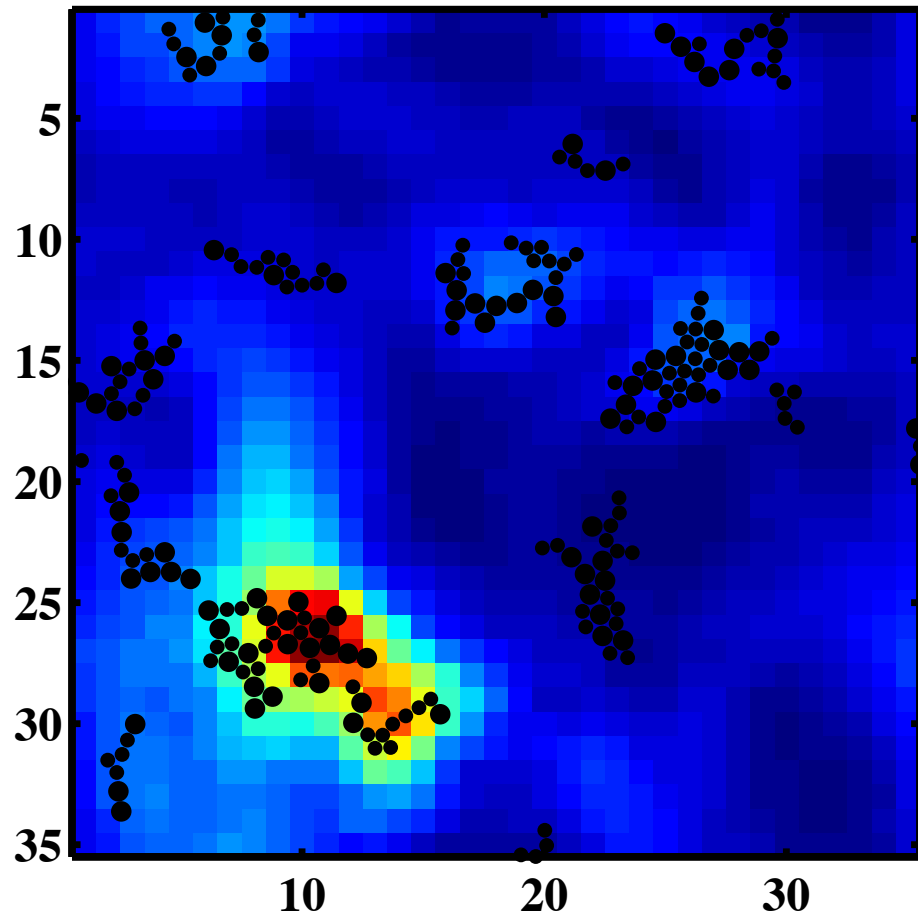
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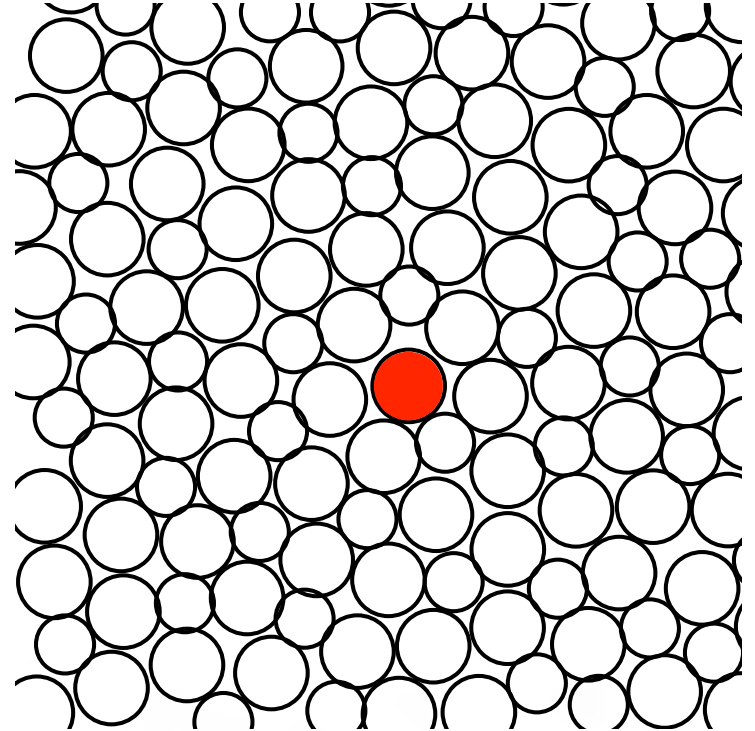
$\text{Log}(\Delta \text{ strain})$  is  $-3.3054$



# Solution

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- Don't just use one quantity to characterize structure
- Use MANY
- Introduce two families of functions



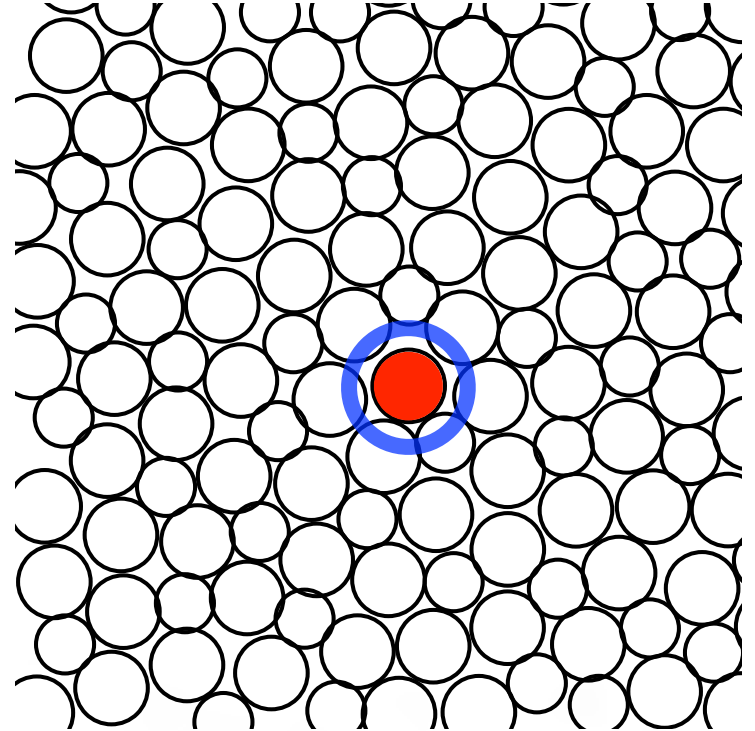
$$S_Y^X(i; \mu) = \sum_j e^{-(R_{ij} - \mu)^2 / L^2}$$

$$Q_{YZ}^X(i; \xi, \lambda, \zeta) = \sum_j \sum_k e^{-(R_{ij}^2 + R_{ik}^2 + R_{jk}^2) / \xi^2} (1 + \lambda \cos \theta_{ijk}) \zeta$$

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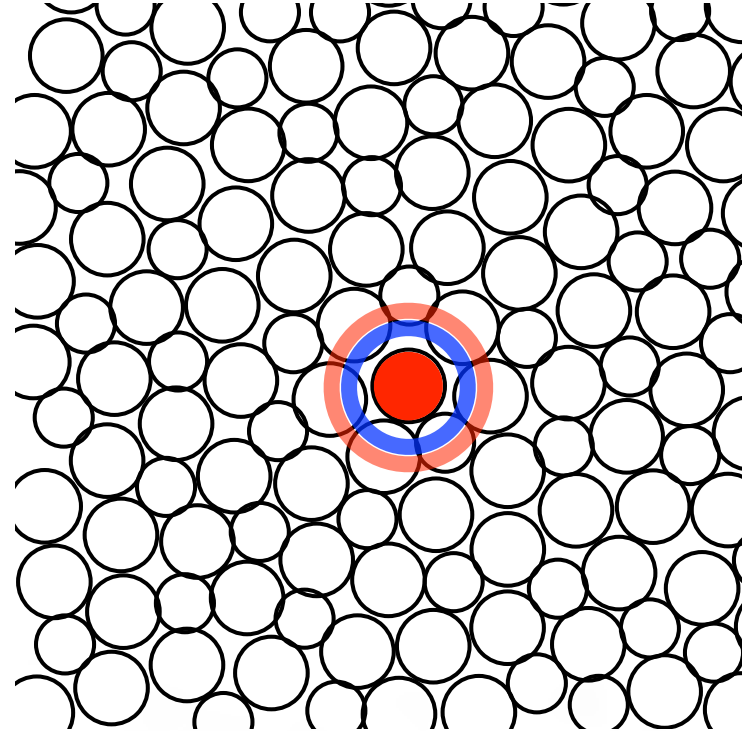
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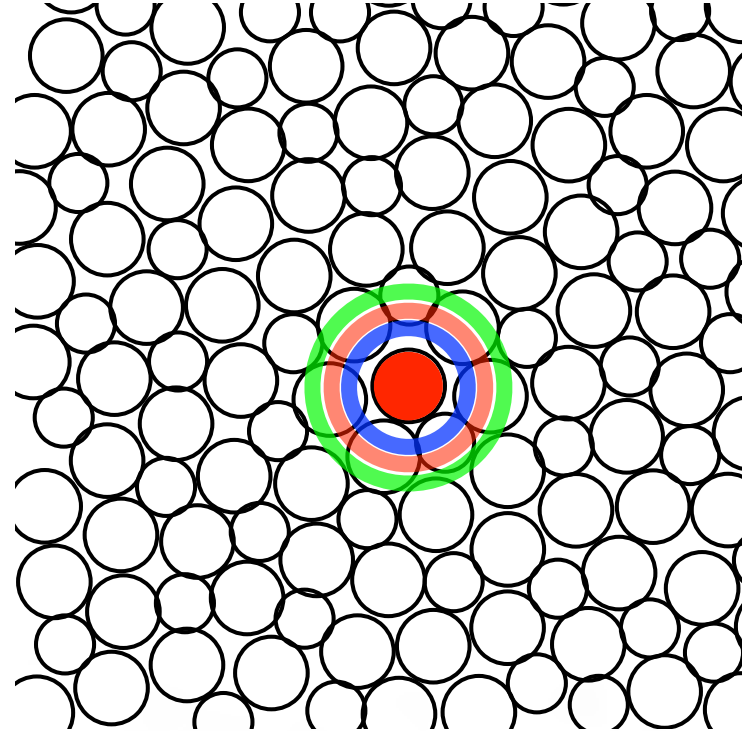
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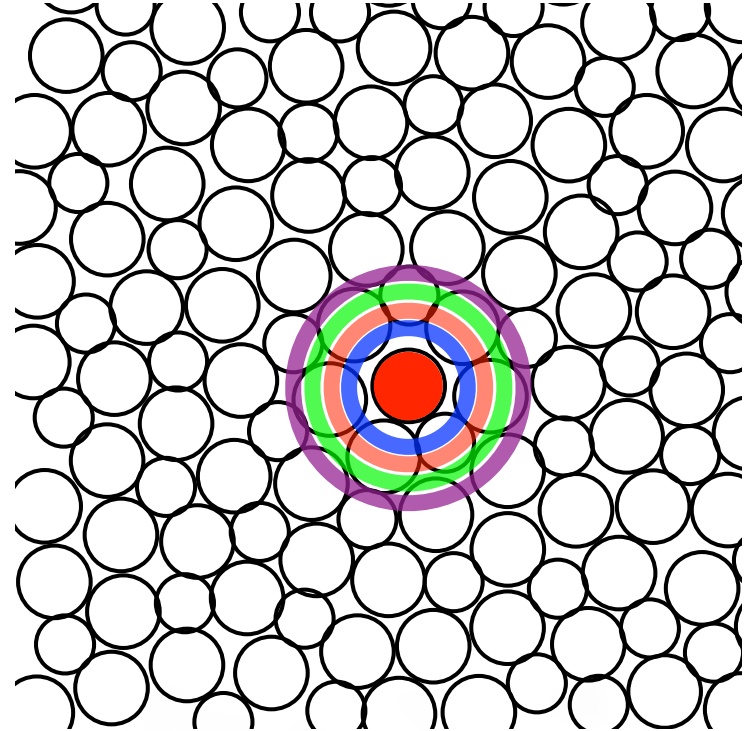
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# Physical Meaning of Structure Functions

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- First family measures density at radius  $r$

$$S_Y^X(i; \mu) = \sum_j e^{-(R_{ij} - \mu)^2 / L^2}$$

Radius to probe

Particle separation

Width of window

- Second family measures bond anticorrelation at radius  $r$

$$Q_{YZ}^X(i; \xi, \lambda, \zeta) = \sum_j \sum_k e^{-(R_{ij}^2 + R_{ik}^2 + R_{jk}^2) / \xi^2} (1 + \lambda \cos \theta_{ijk})^\zeta$$

Radius

Alignment/antialignment

Angular Resolution

# Classify “softness”

---

- Have a set of structural variables for each particle  $i$  (values of  $S$  and  $Q$  at different values of  $\mu, \xi, \lambda, \zeta$ )

$$S_Y^X(i; \mu) = \sum_j e^{-(R_{ij} - \mu)^2 / L^2}$$

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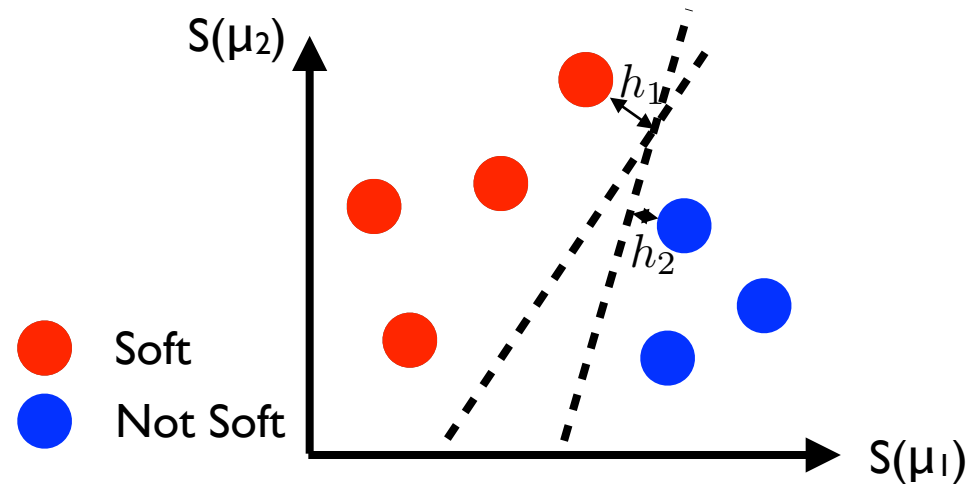
- Want to categorize each particle as **soft** (susceptible to rearrangement) or **hard**
- Use Support Vector Machines (SVM)
- Requires a “**training set**” with known classification

# Classifying “softness”

Use  $D_{\min}^2$  to find rearrangements

$$D_{\min}^2(i, \Delta t) = \sum_j [\mathbf{R}_{ij}(t + \Delta t) - \Lambda_{\min} \mathbf{R}_{ij}(t)]$$

- Particle is “soft” if  $D_{\min}^2 > D_{\min,0}^2 \approx \sigma_{AA}^2$
- Embed in space where each dimension is a structural variable
- Find dividing plane with maximum margin  $h$
- Classify new data with dividing plane

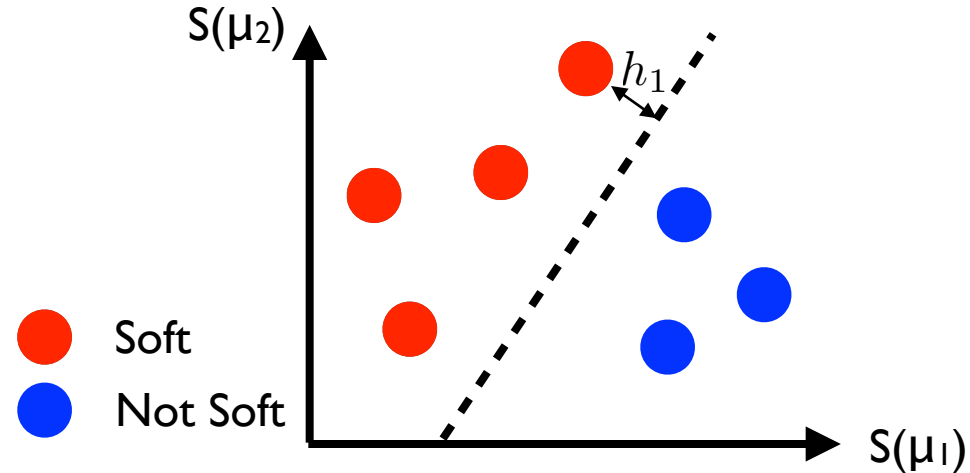


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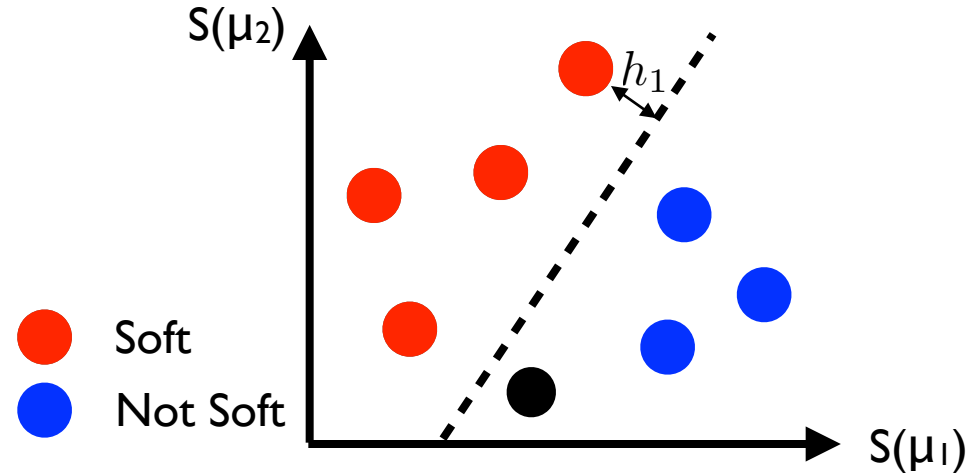


# Classifying “softness”

Use  $D_{\min}^2$  to find rearrangements

$$D_{\min}^2(i, \Delta t) = \sum_j [\mathbf{R}_{ij}(t + \Delta t) - \Lambda_{\min} \mathbf{R}_{ij}(t)]$$

- Particle is “soft” if  $D_{\min}^2 > D_{\min,0}^2 \approx \sigma_{AA}^2$
- Embed in space where each dimension is a structural variable
- Find dividing plane with maximum margin  $h$
- Classify new data with dividing plane





# SVM Method

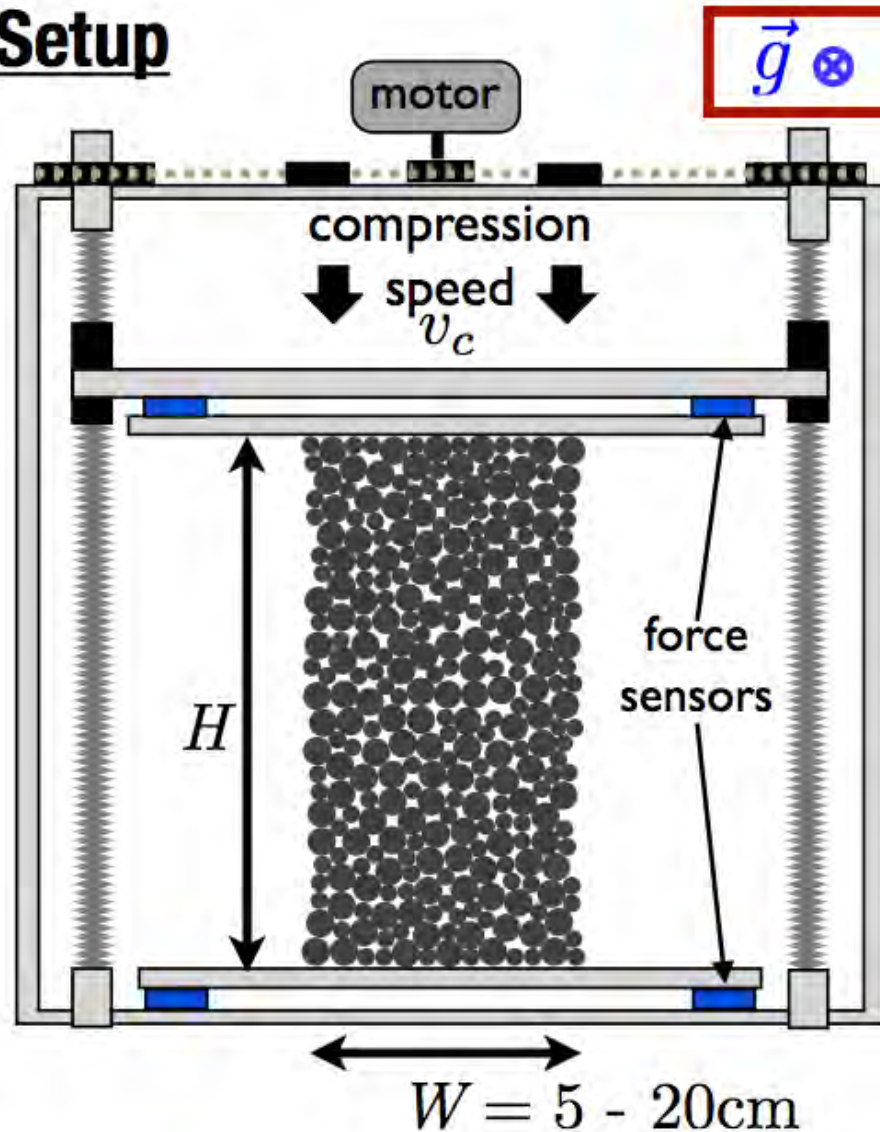
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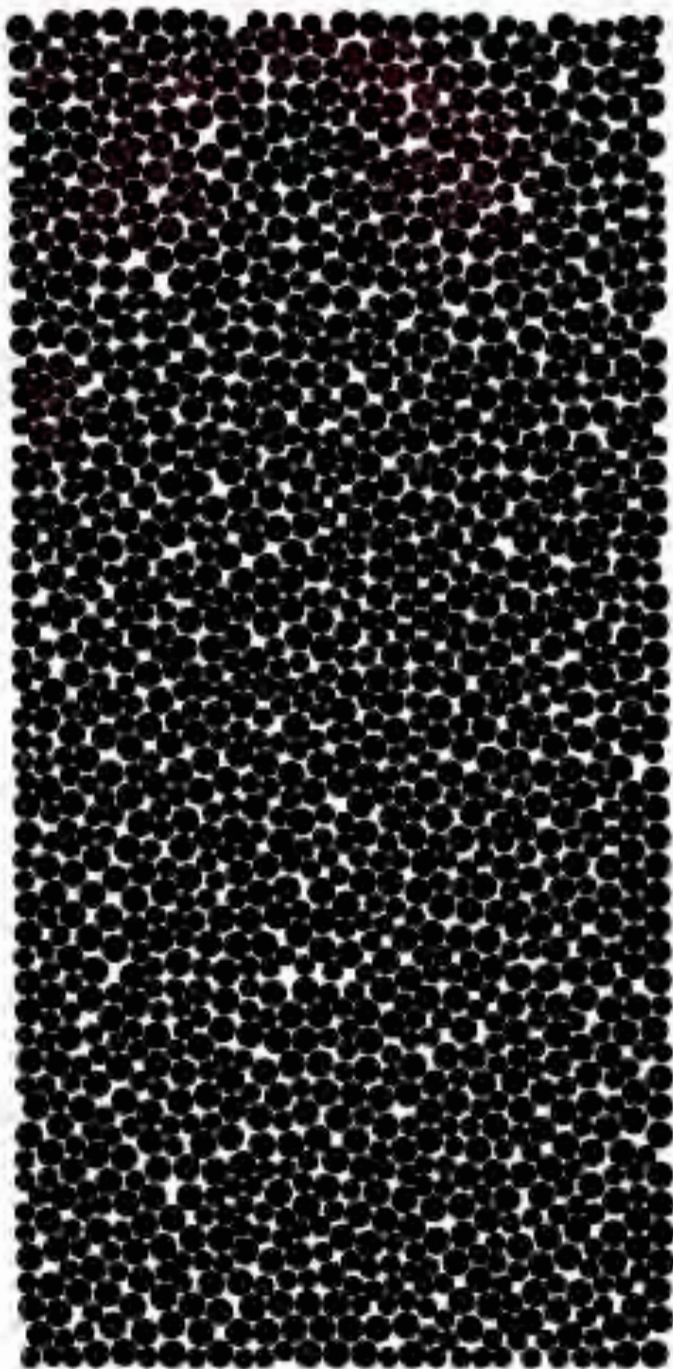
- Works better than QLM method
- Don't need to
  - quench system to obtain inherent structures
    - in fact, method works better because it can be applied to instantaneous snapshots
  - diagonalize dynamical matrix ( $O(N^3)$ )
    - method is  $O(N)$
    - know interactions between particles precisely
- Apply to experimental system

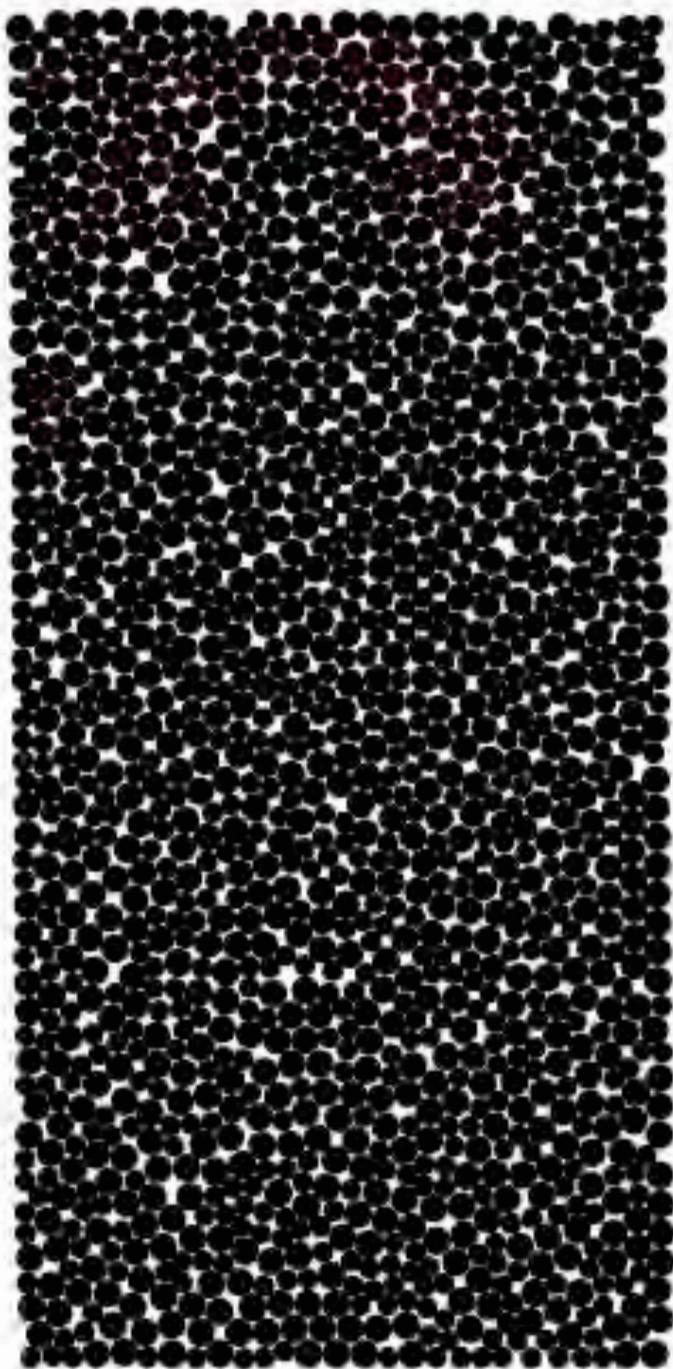
# Compressed Granular Pillars

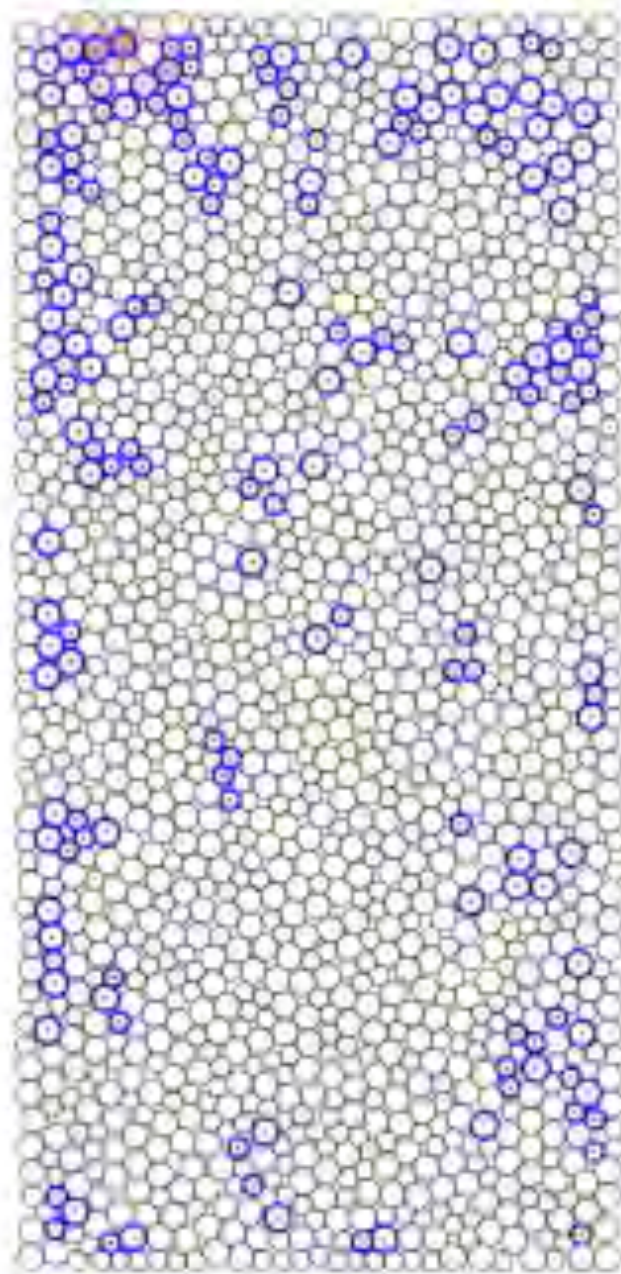
Jennifer Rieser, Doug Durian

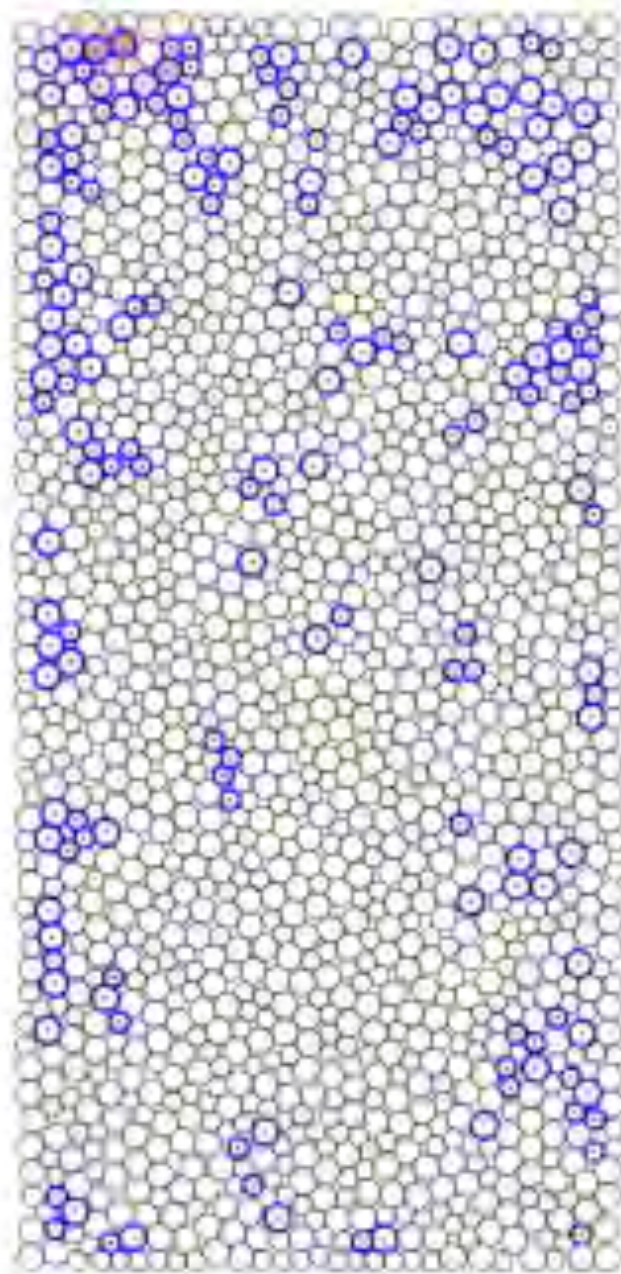
## Setup



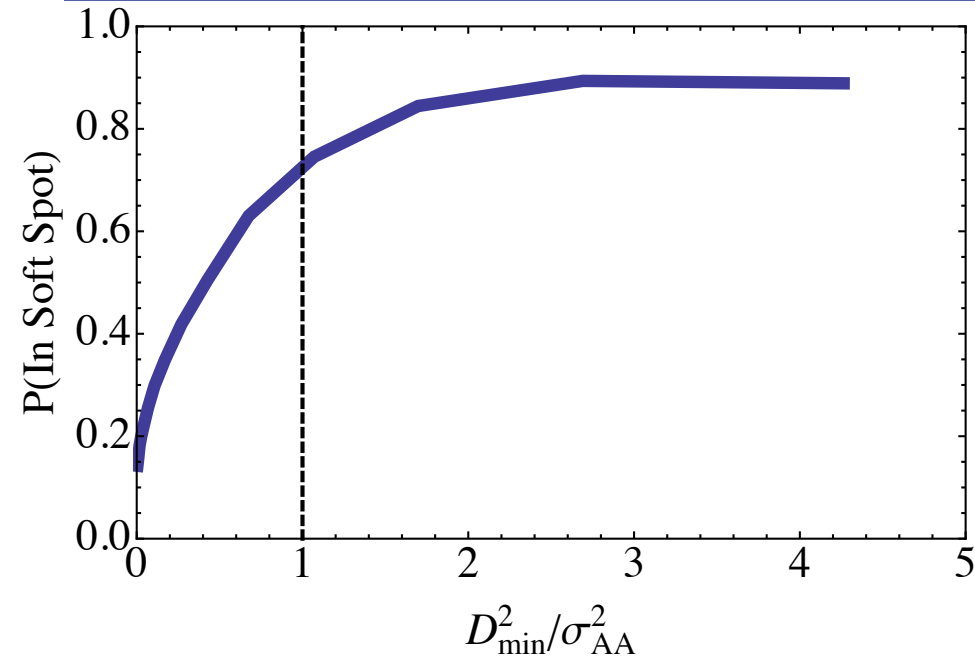








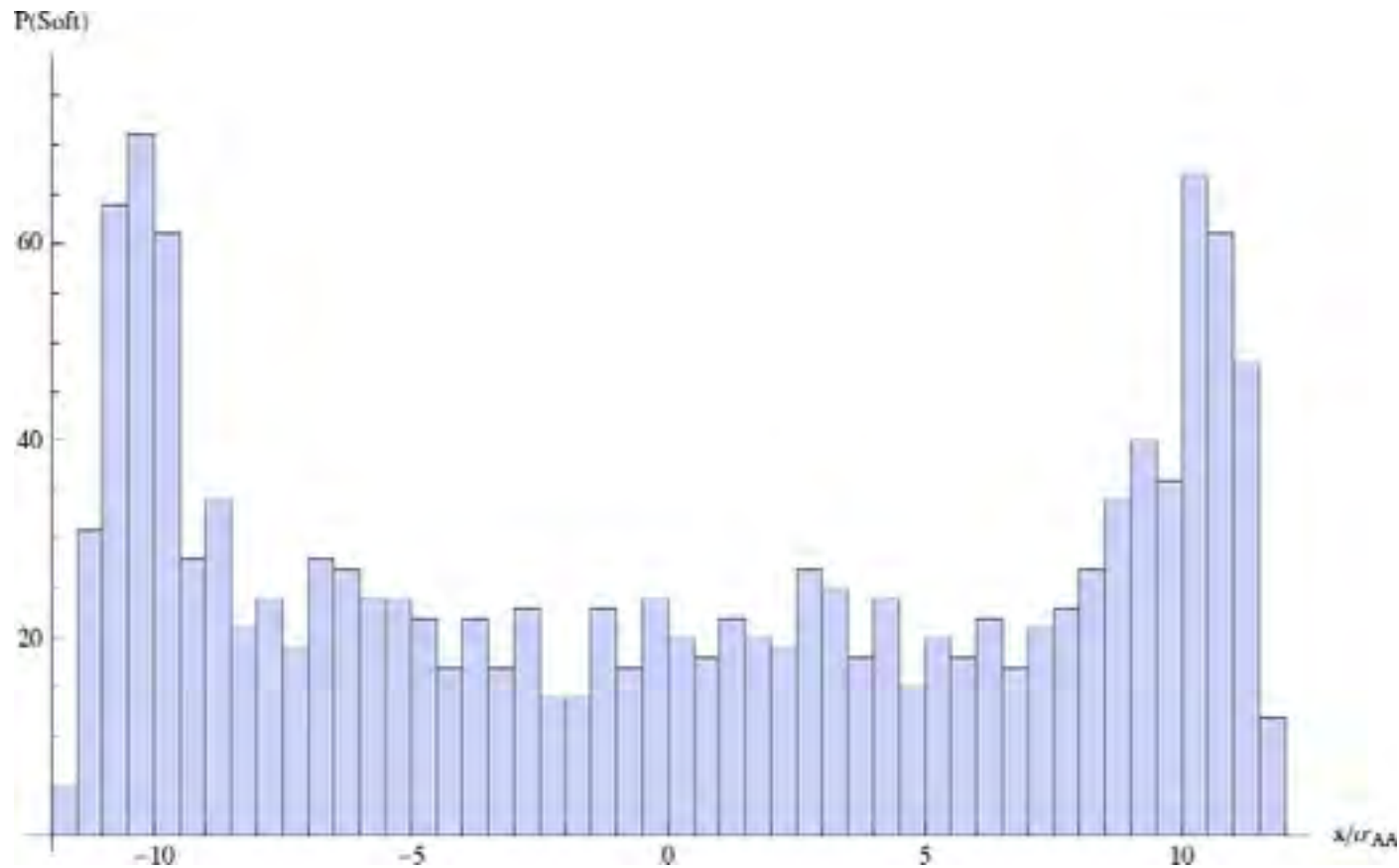
# Compressed Granular Pillars



- 21% of system captures 80% of rearranging particles

# Pillar Has Free Surfaces—Are There More Soft Particles There?

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- Yes, number of soft particles is higher at surface
- Soft particles enhanced over range of  $\sim 5$  particle diameters
- Granular pillars have “mobile layer” similar to that of glassy films



# Tests of SVM method

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(I) Sheared, thermal 2D Lennard-Jones glass

65:35 Kob-Andersen Lennard-Jones mixture

$$U(r) = 4\epsilon \left[ \left( \frac{\sigma}{r} \right)^{12} - \left( \frac{\sigma}{r} \right)^6 \right]$$

$$\sigma_{AA} = 1.0 \quad \sigma_{AB} = 0.88 \quad \sigma_{BB} = 0.8.$$

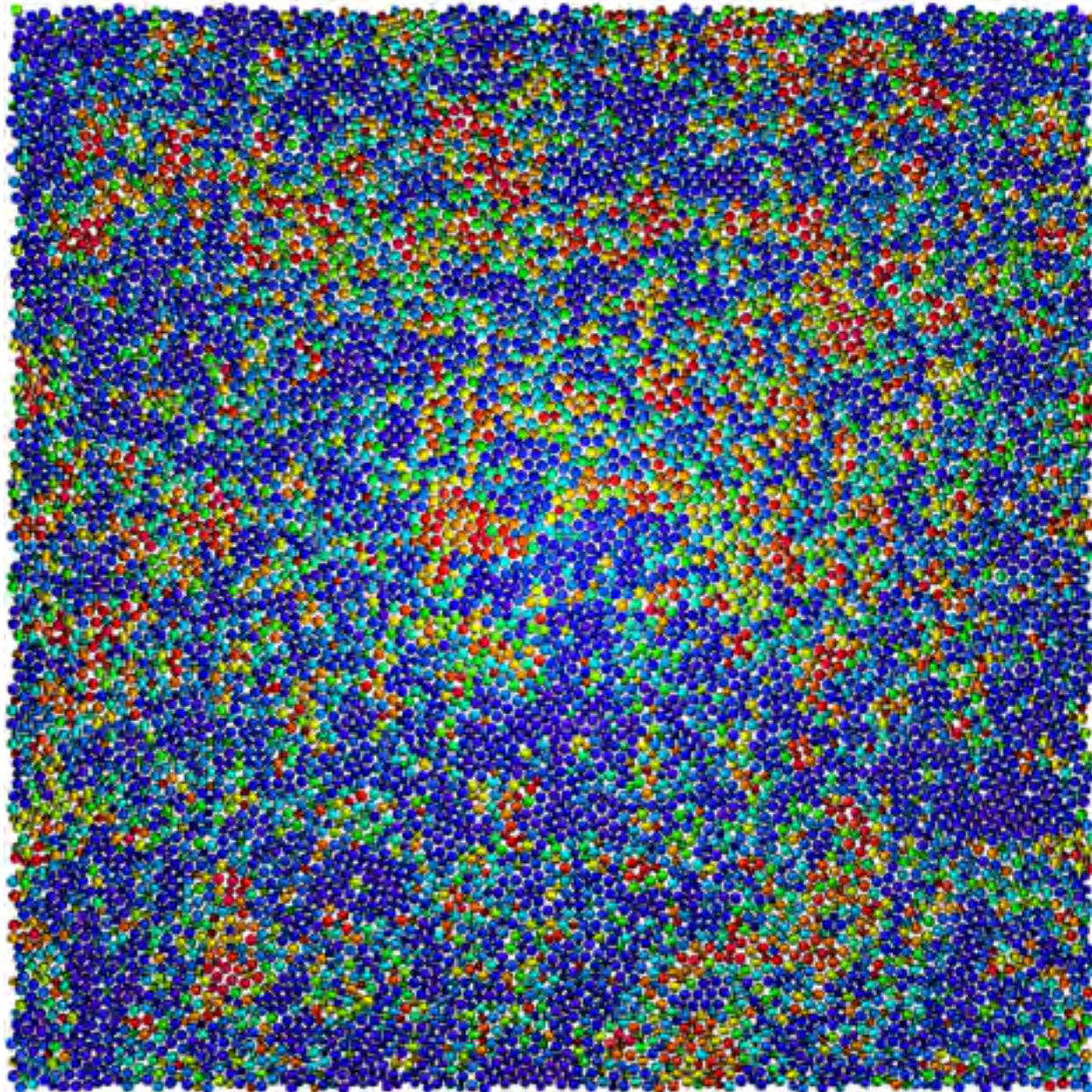
$$\epsilon_{AA} = 1.0, \quad \epsilon_{AB} = 1.5 \quad \epsilon_{BB} = 0.5.$$

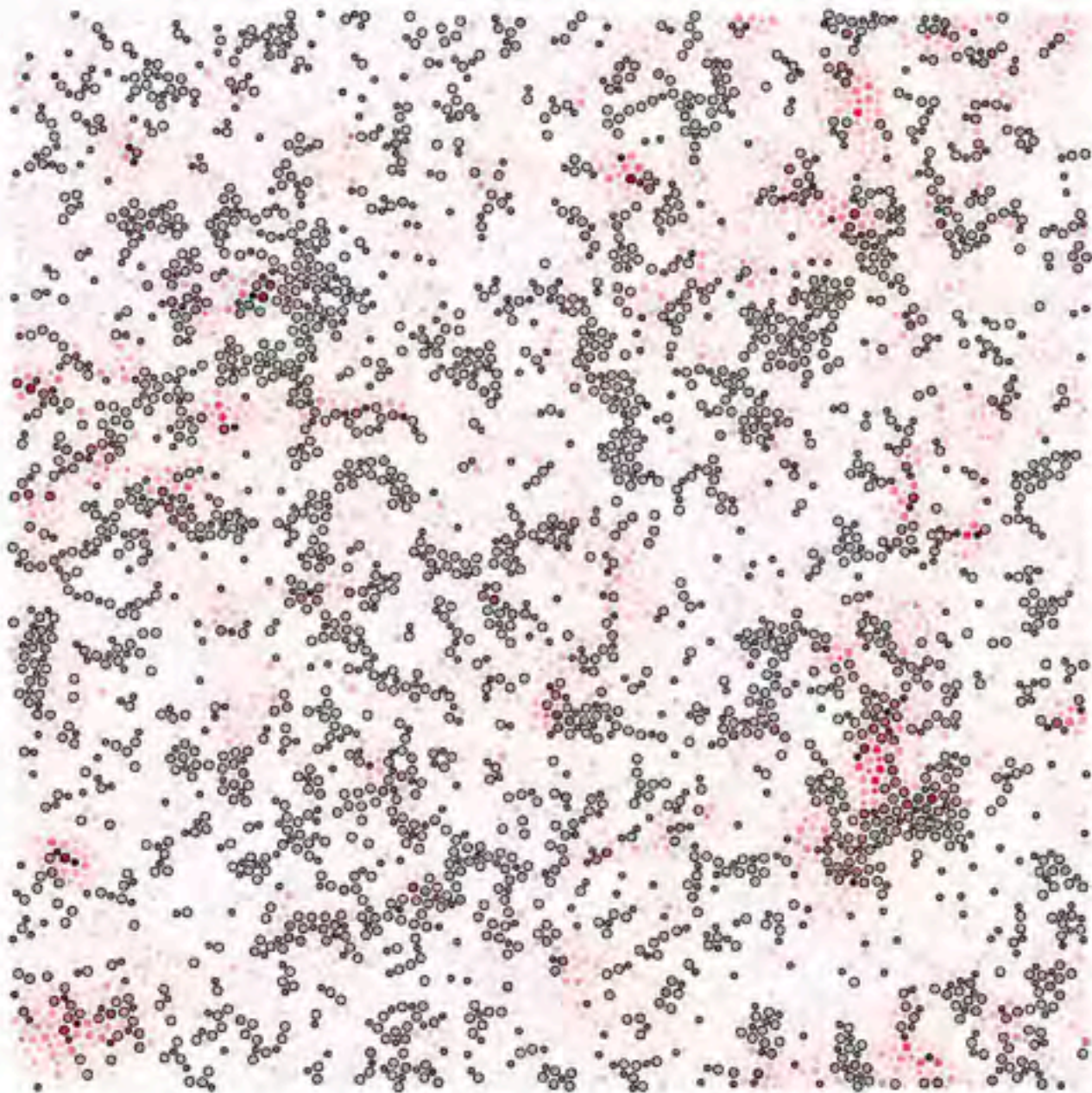
Characterized by Brüning *et al.* with  $T_g=0.33$

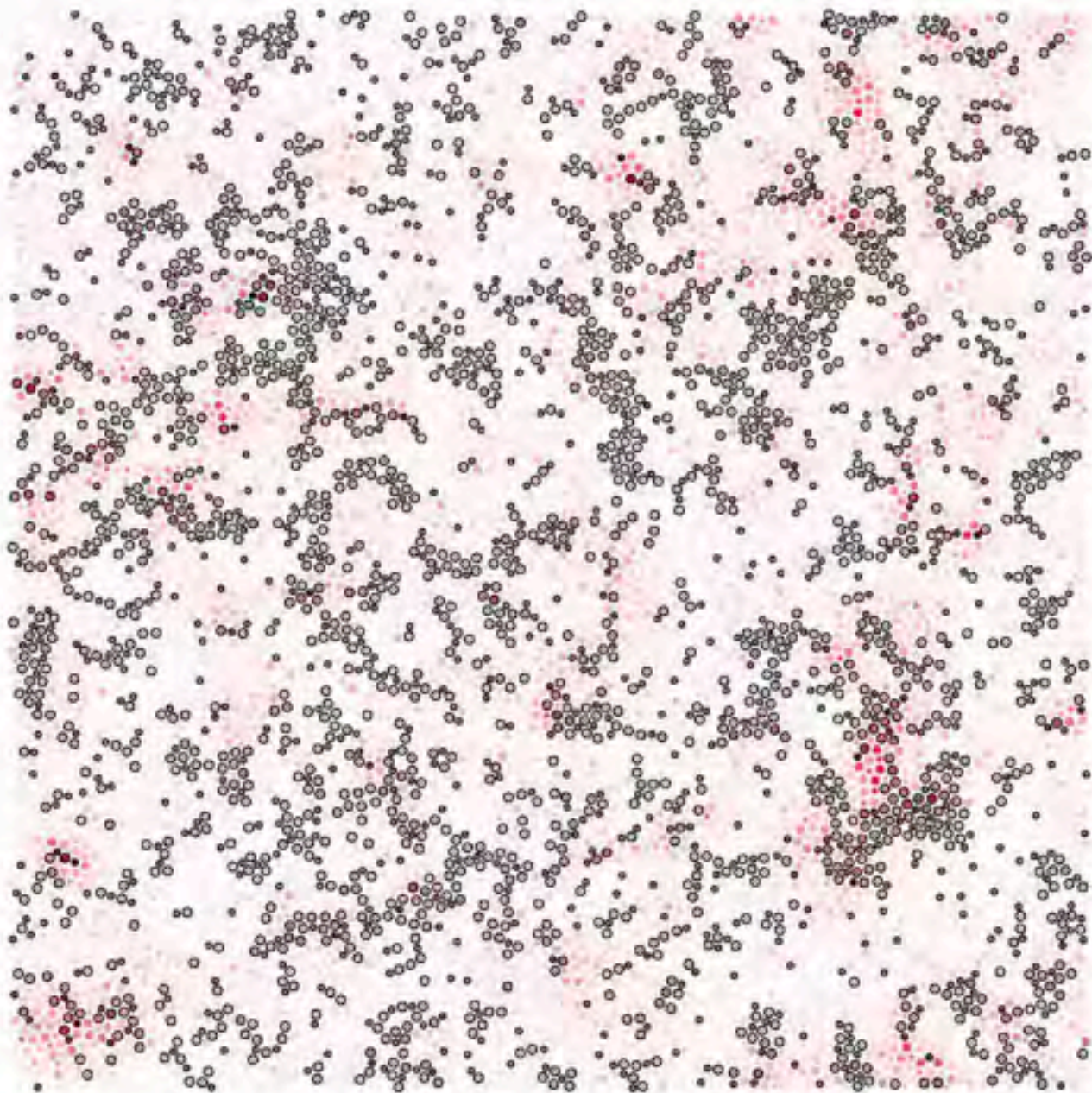
$$T = 0.1, 0.2, 0.3, 0.4 \quad \dot{\gamma} = 10^{-4}$$

# Softness Field

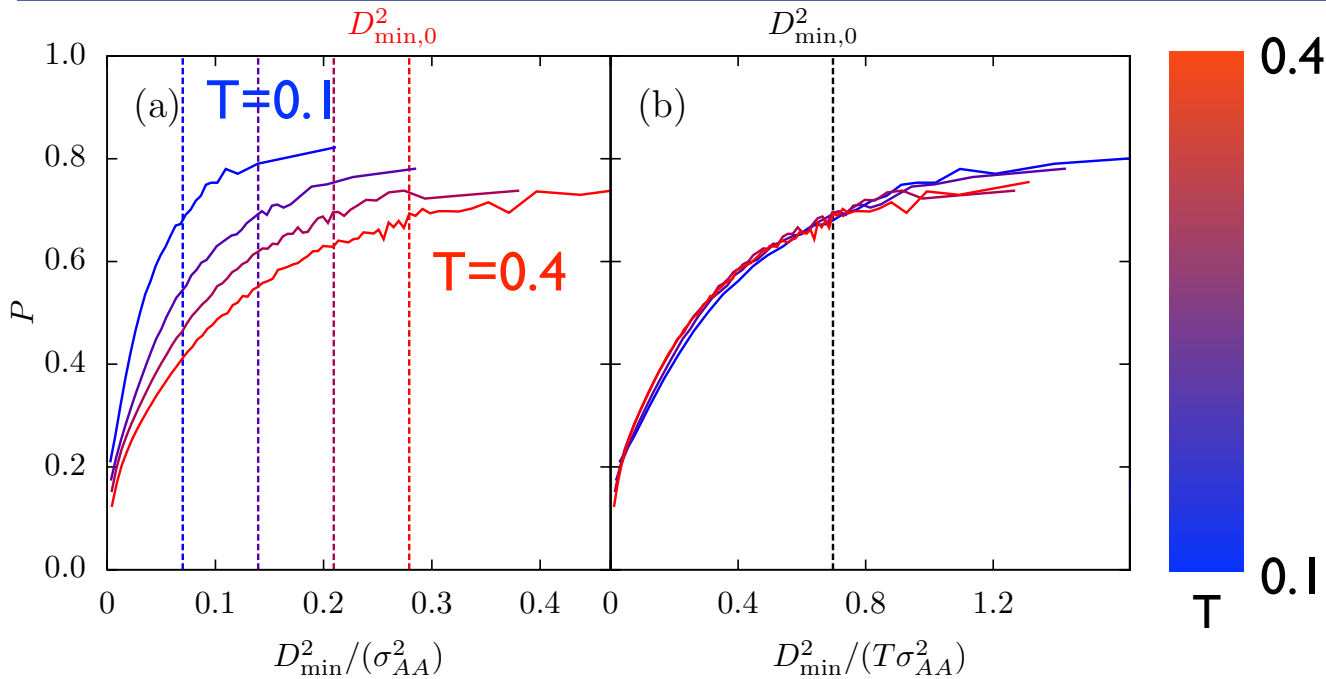
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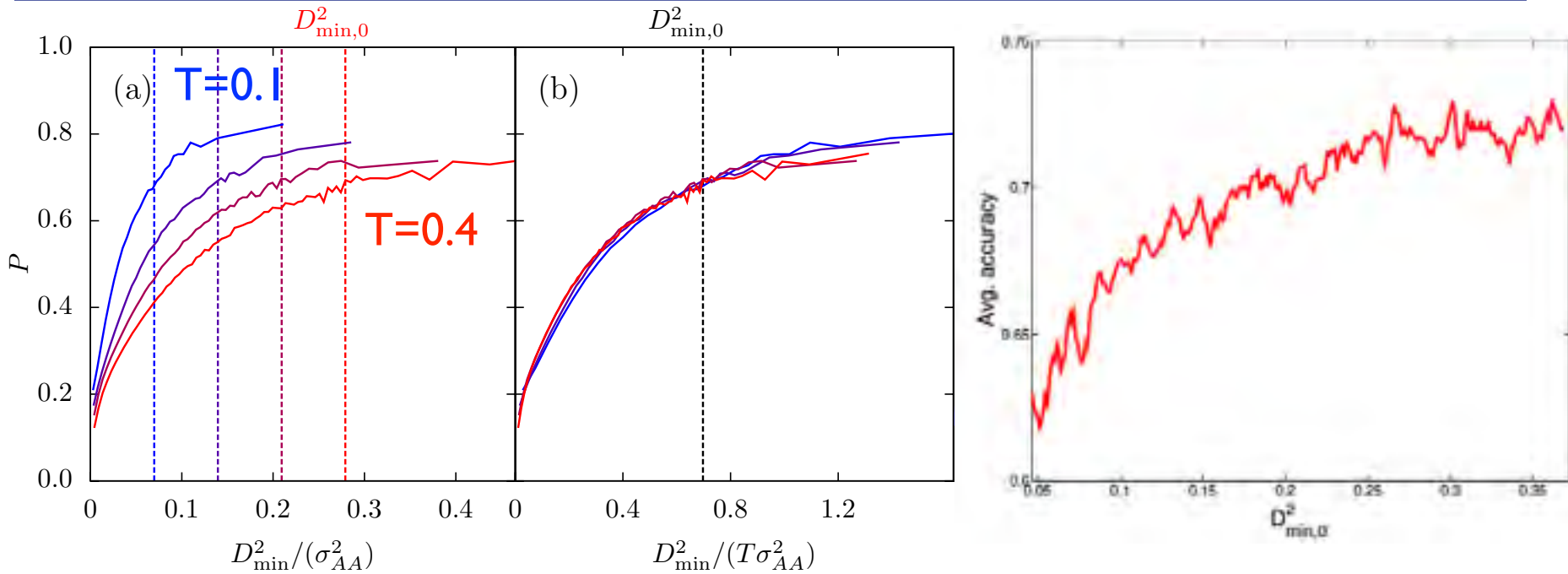


# Sheared 2D Lennard-Jones glass



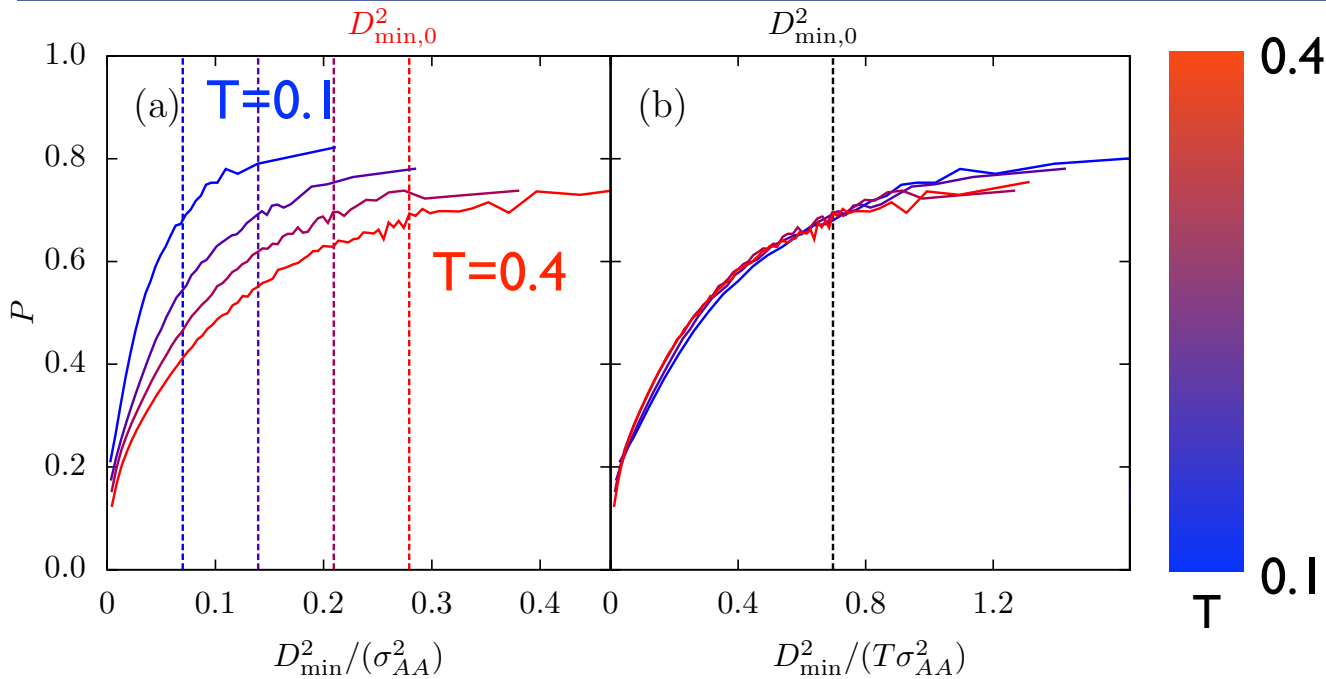
- 26% of system captures 73% of rearranging particles
- Works equally well at all temperatures, even above glass transition
- Accuracy insensitive to threshold

# Sheared 2D Lennard-Jones glass



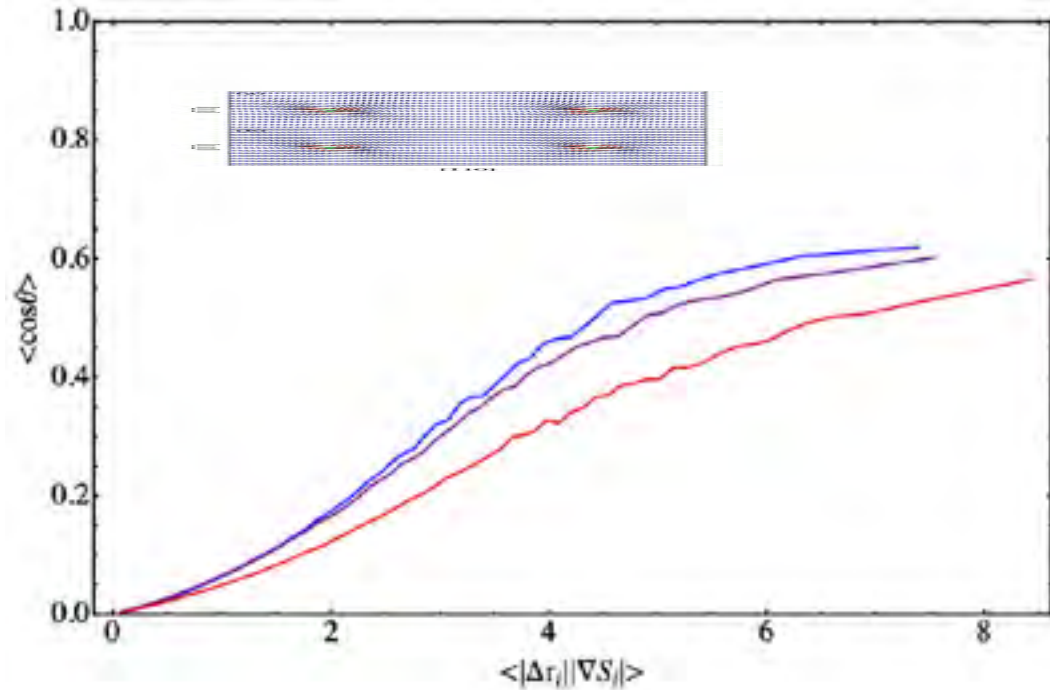
- 26% of system captures 73% of rearranging particles
- Works equally well at all temperatures, even above glass transition
- Accuracy insensitive to threshold

# Sheared 2D Lennard-Jones glass

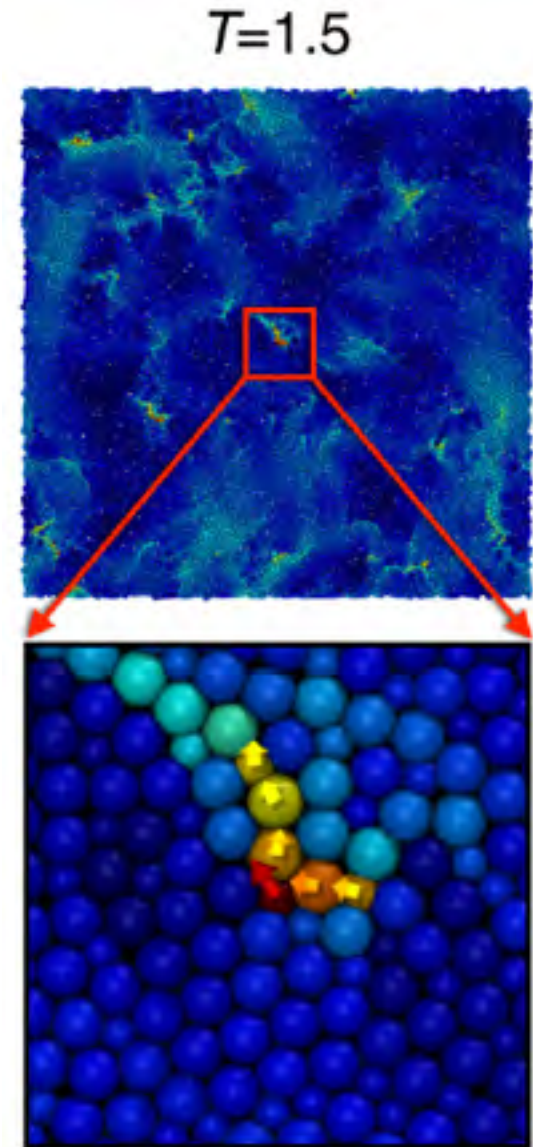


- 26% of system captures 73% of rearranging particles
- Works equally well at all temperatures, even above glass transition
- Accuracy insensitive to threshold

# Particle Displacements correlate with softness gradient



- Rearranging particles tend to move in direction of increasing softness
- Promising as structural signature of facilitation



Keys, et al PRX (2011)



# Tests of SVM method

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(2) Quiescent, thermal 3D Lennard-Jones glass

65:35 Kob-Andersen Lennard-Jones mixture

$$U(r) = 4\epsilon \left[ \left( \frac{\sigma}{r} \right)^{12} - \left( \frac{\sigma}{r} \right)^6 \right]$$

$$\sigma_{AA} = 1.0 \quad \sigma_{AB} = 0.88 \quad \sigma_{BB} = 0.8.$$

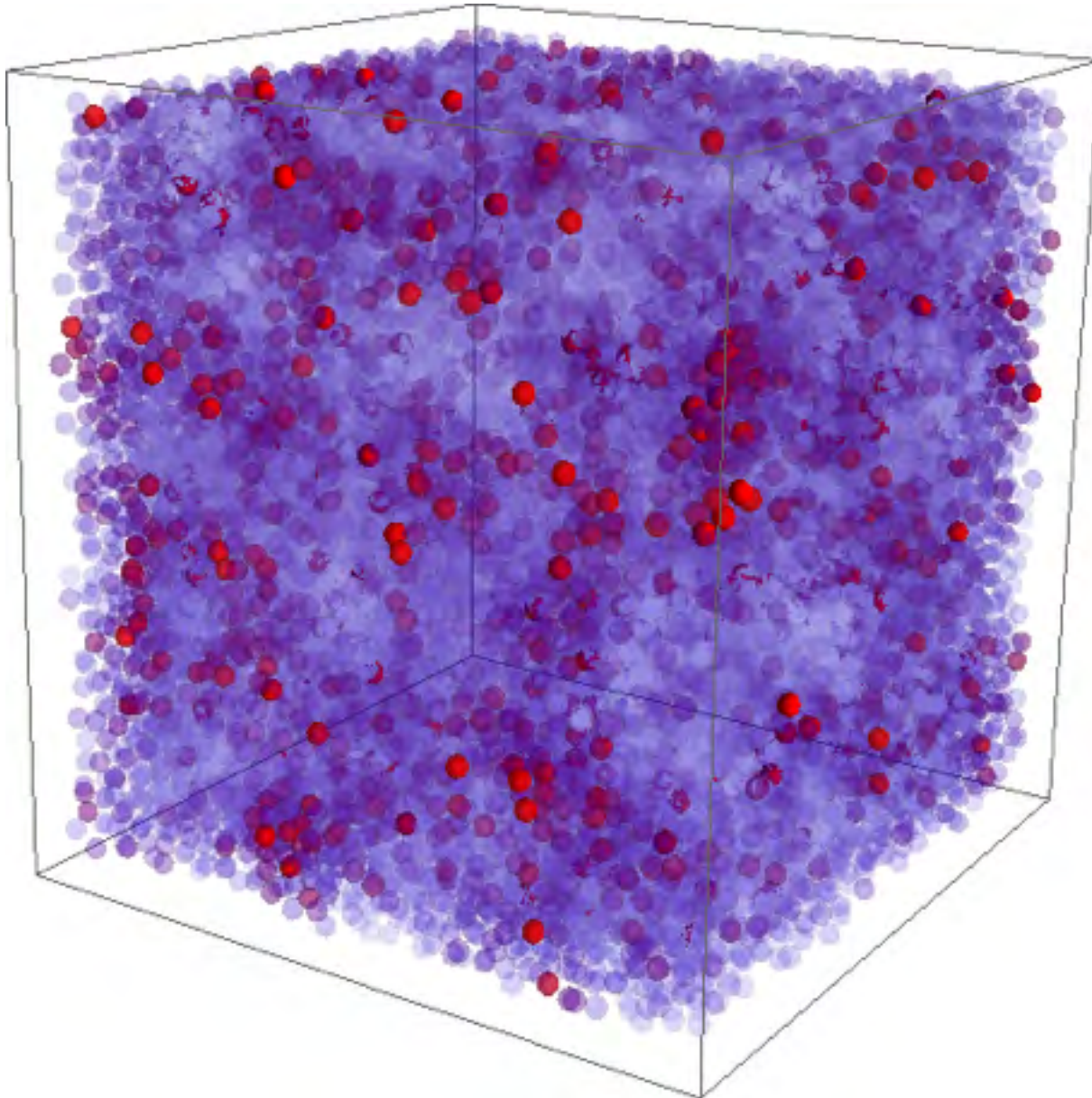
$$\epsilon_{AA} = 1.0, \quad \epsilon_{AB} = 1.5 \quad \epsilon_{BB} = 0.5.$$

Characterized by Brüning *et al.* with  $T_g=0.58$

$$T = 0.4, 0.5, 0.6$$

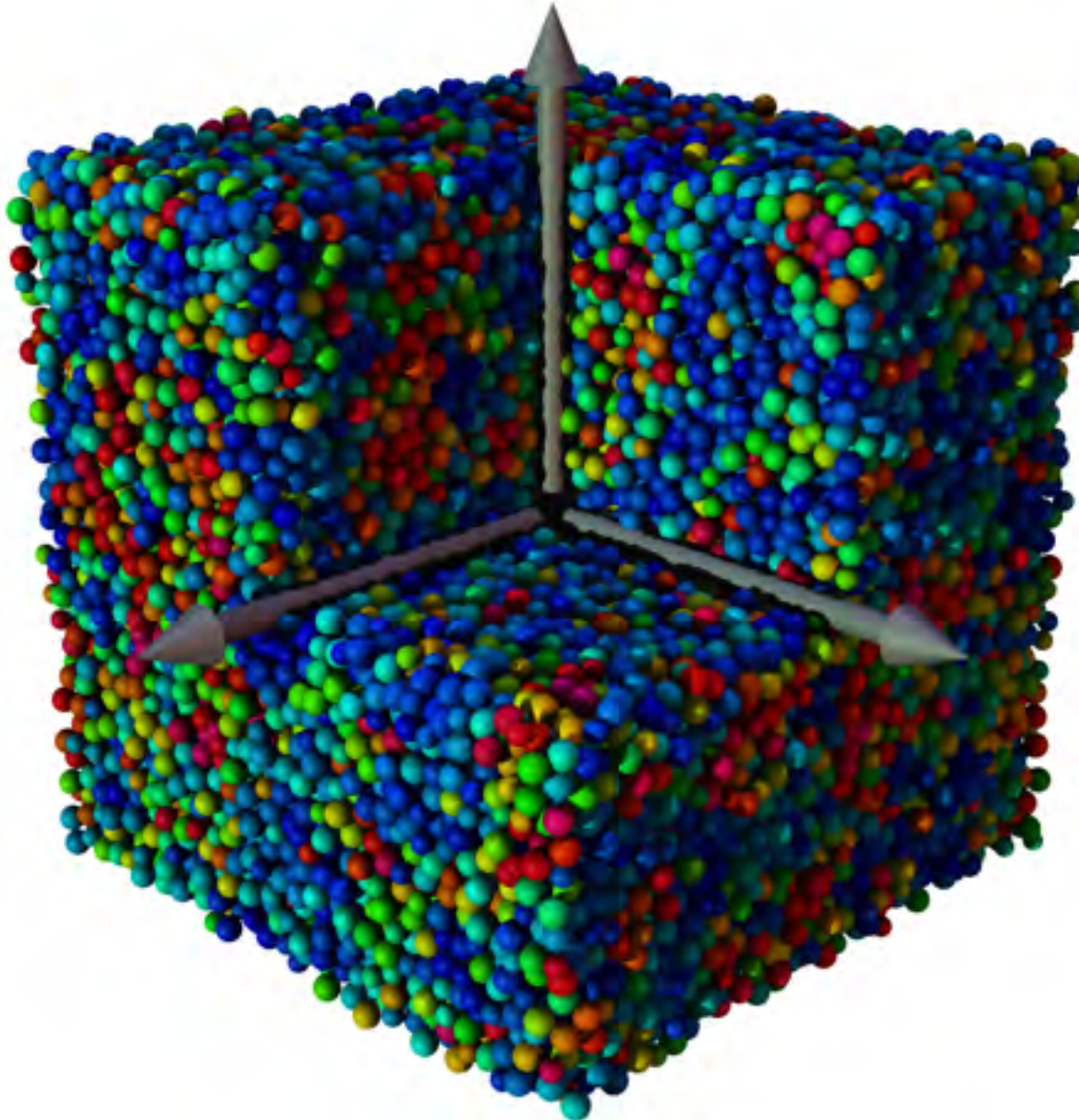
# Rearrangements are localized

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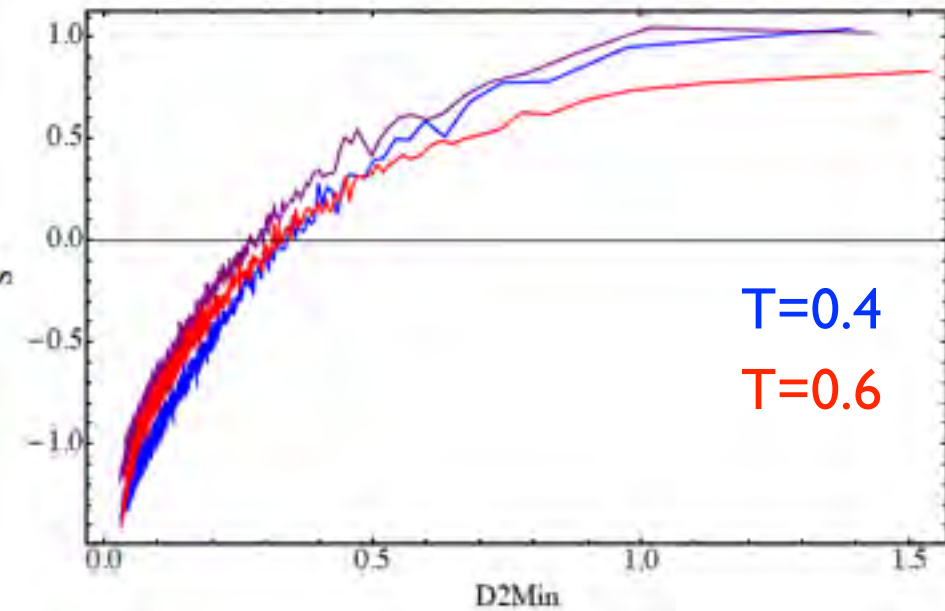
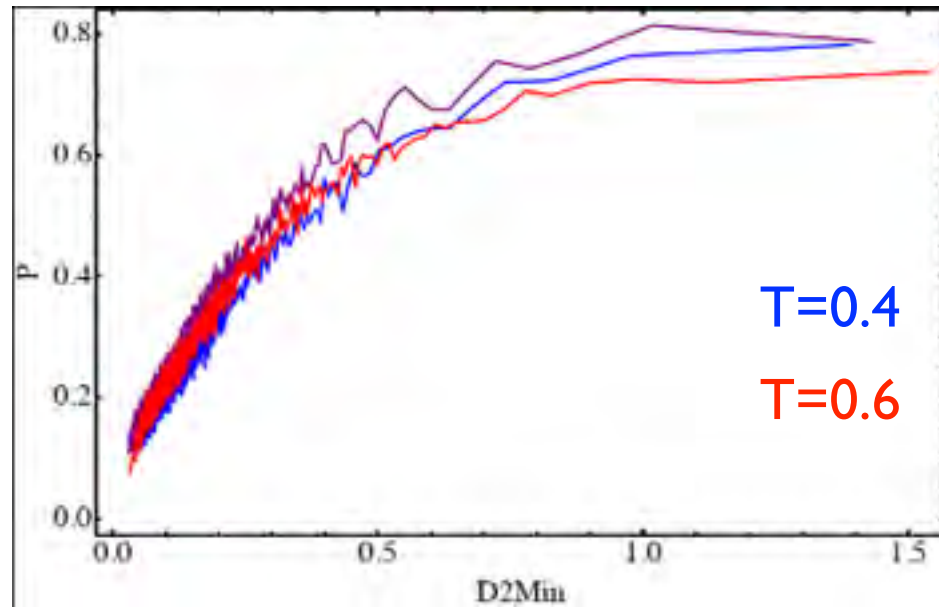


# Softness Field

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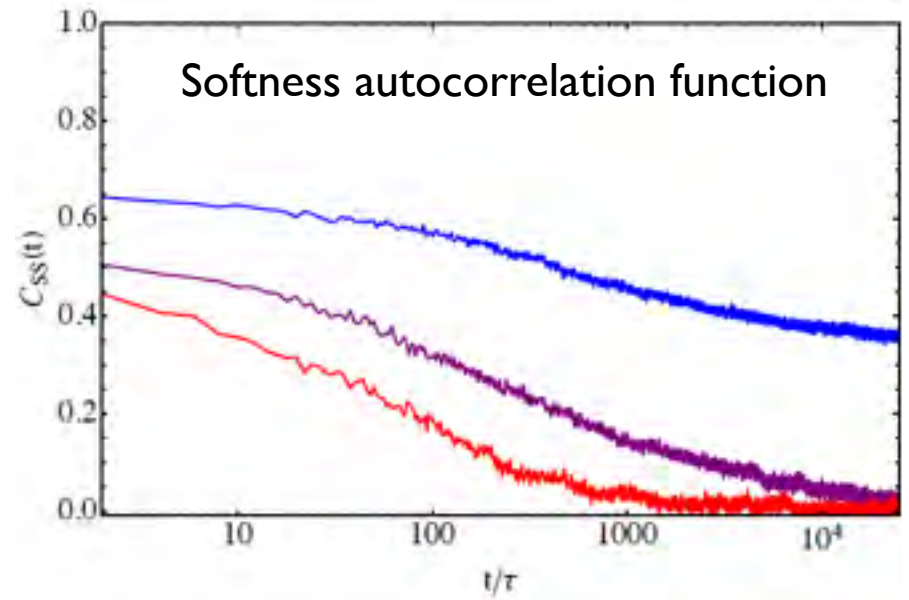
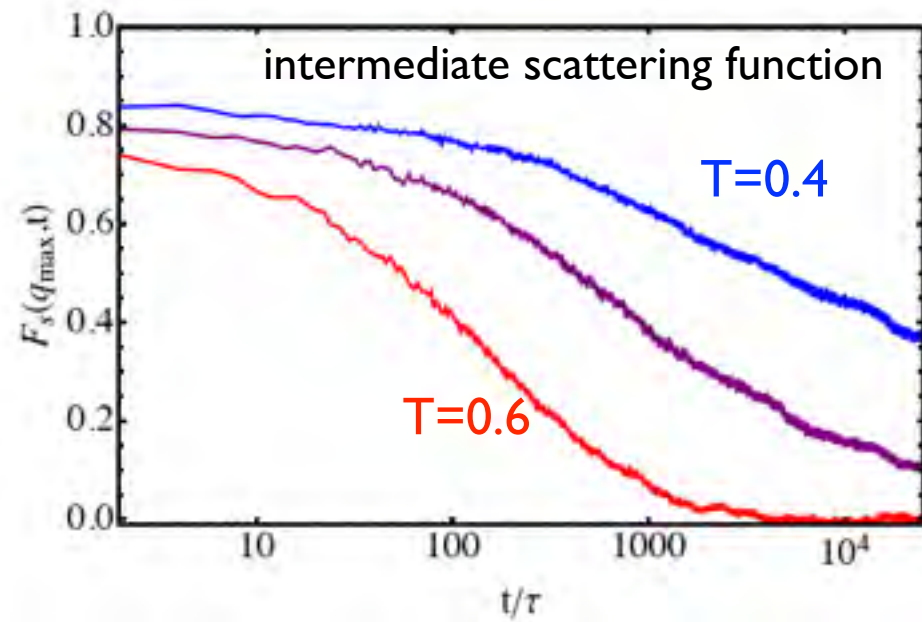


# Rearranging Particles are Soft



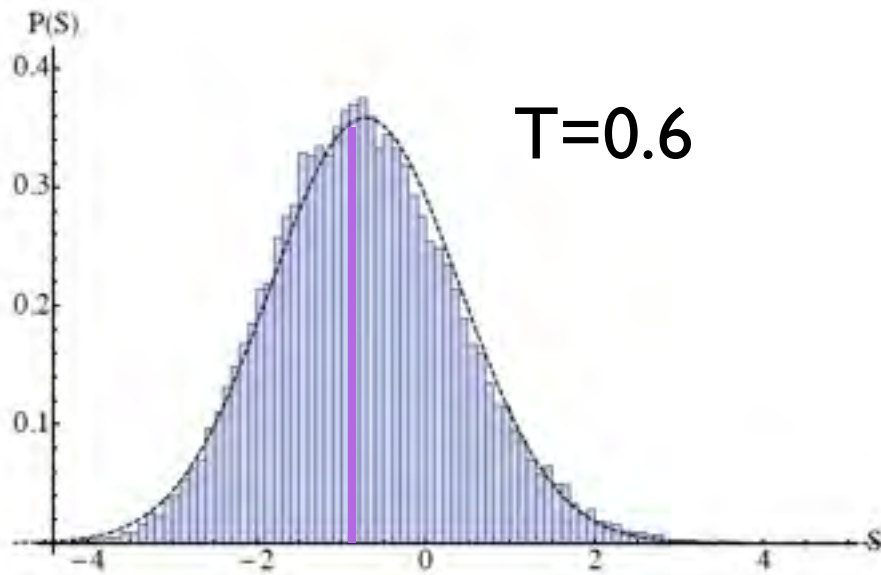
- Works well even at high temperatures ( $T_g=0.58$ )
  - At  $T=0.6$ , 24% of particles capture 72% of rearrangements (kinetic heterogeneities)
- average softness is high for rearranging particles
- Accuracy comparable in 2D and 3D

# Softness Lifetime I

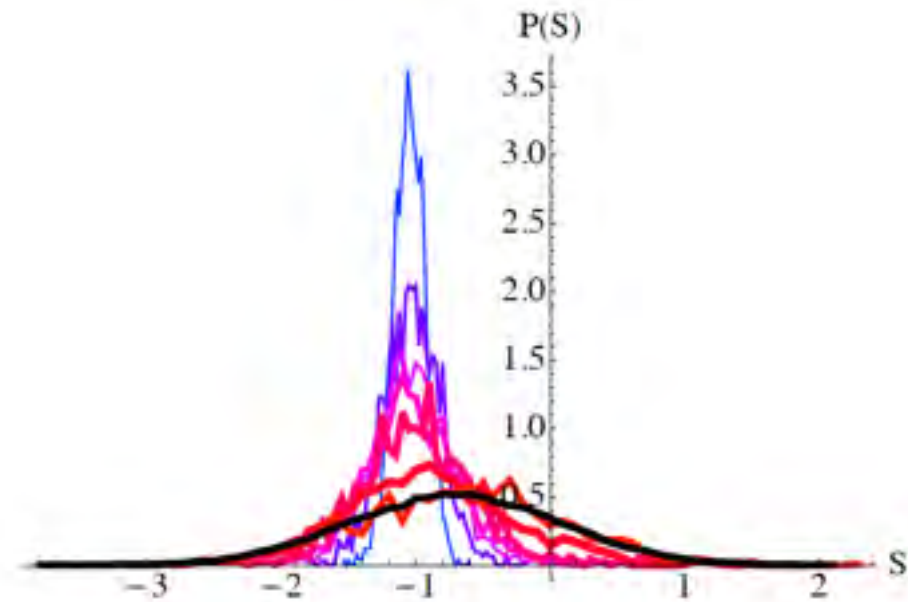


- Softness lifetime is comparable to relaxation time

# Softness Lifetime II

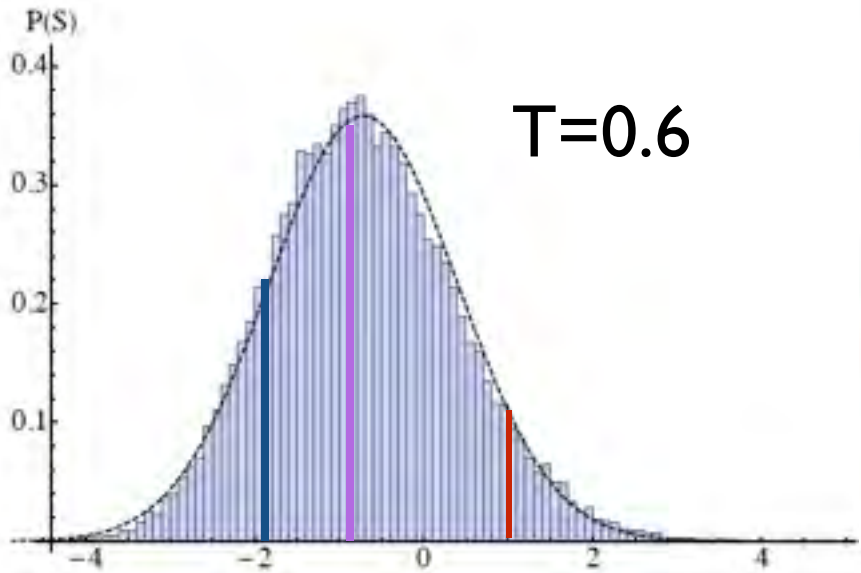


$T=0.6$

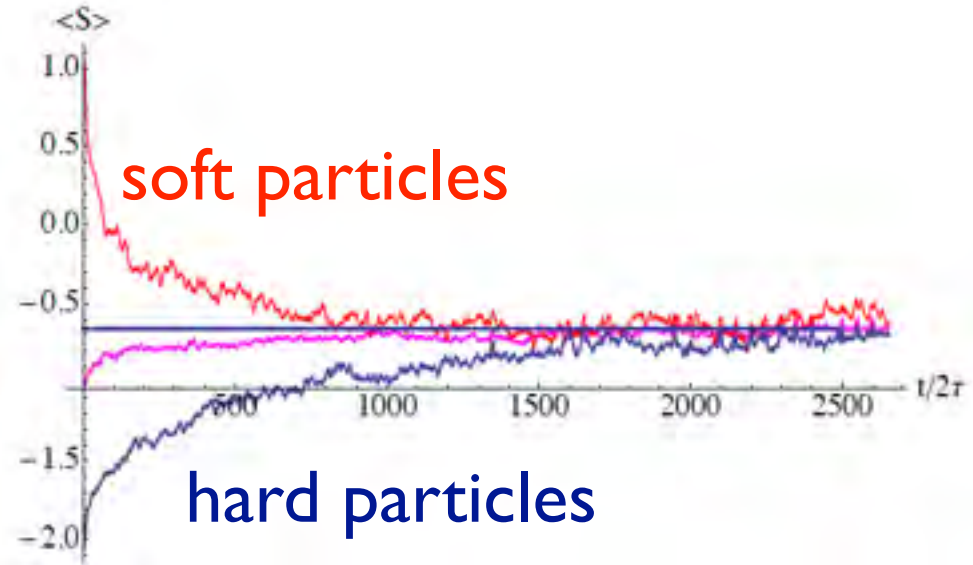


- Softness equilibrates over time to approach equilibrium softness distribution

# Softness Lifetime III

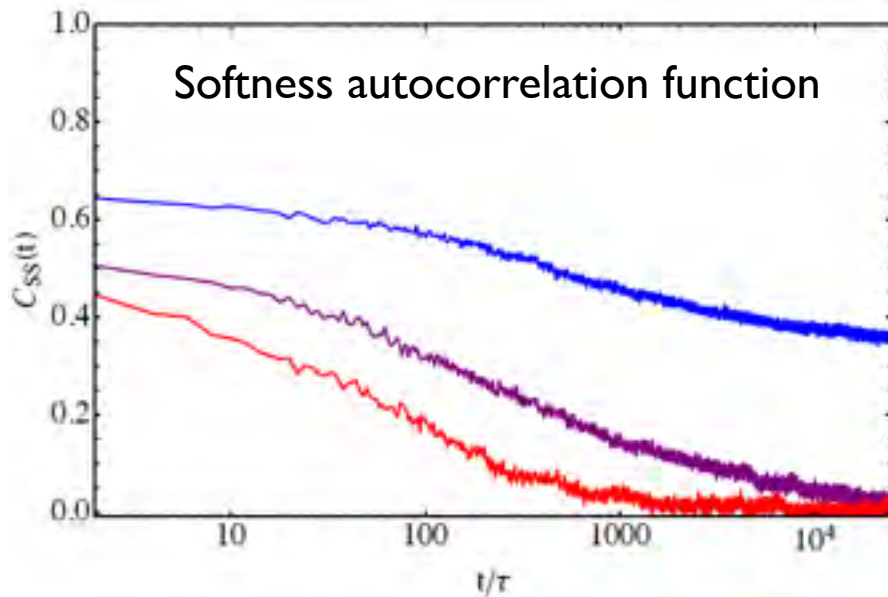


$T=0.6$



soft particles

hard particles



Softness autocorrelation function

- Soft particles equilibrate faster than hard particles
- Softness is promising as structural signature of kinetic heterogeneities

# Dividing Surface Contains Physics

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- Recall structural variables

$$S_Y^X(i; \mu) = \sum_j e^{-(R_{ij} - \mu)^2 / L^2}$$

$$Q_{YZ}^X(i; \xi, \lambda, \zeta) = \sum_j \sum_k e^{-(R_{ij}^2 + R_{ik}^2 + R_{jk}^2) / \xi^2} (1 + \lambda \cos \theta_{ijk})$$

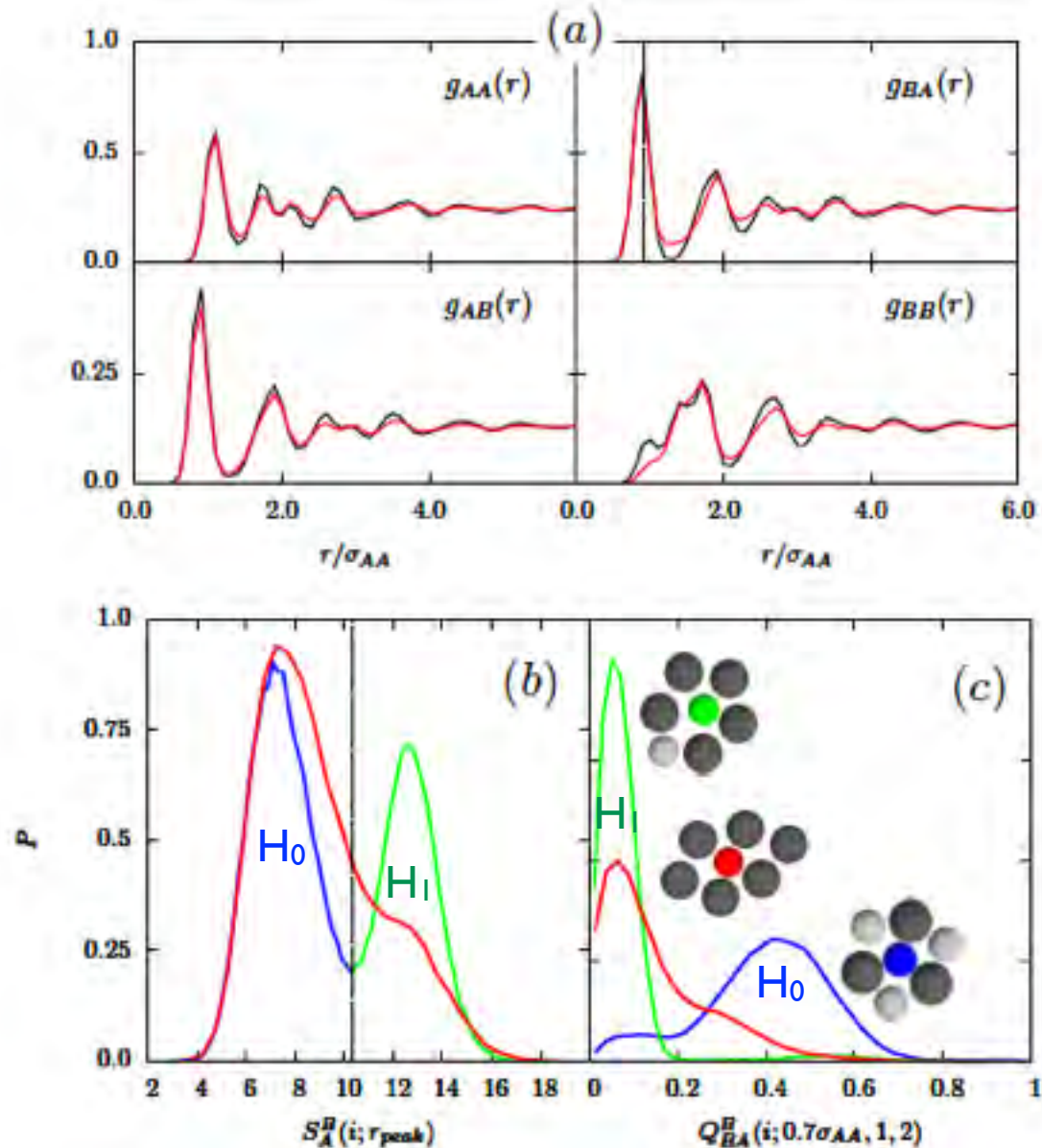
- S measures radial density
- Q measures relative bond orientation

- **Note**  $g_{XY}(r) = \lim_{L \rightarrow 0} \frac{1}{N} \sum_{i=1}^N S_Y^X(i; r) / 2\pi r$  so S corresponds to local  $g(r)$



# How are Soft Particles Different?

- Soft particles (red) are more “liquidlike” than system as whole
- S distinguishes between **soft** particles and  $H_I$  particles but not **soft** and  $H_0$
- Q distinguishes between soft and  $H_0$  but not **soft** and  $H_I$
- Need both S and Q (at least)



# Conclusions

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- Quasilocalized modes yields structural signatures of mobility near free surfaces and at flow defects
- Soft particles are structurally distinct, but in subtle ways that we can pick out using machine learning methods
- SVM method
  - Promising for identifying particles likely to rearrange under temperature or stress
  - Fast, and only requires positions of particles, not interactions
  - Works on experimental as well as simulation data!

# QLM: Thanks to

---



Lisa Manning



Daniel Sussman

DOE



Carl Goodrich



Sid Nagel

# SVM: Thanks to

---



Sam Schoenholz



Brad Malone



Dogus Cubuk



Tim Kaxiras

UPenn MRSEC



Jennifer Rieser

Joerg Rottler



Doug Durian