Patchy memories in low dimensions: Disorder, noise, and genesis; recovering after erasure

> Jie Yang, AAM, Syracuse University PRB 96, 214208 (2017) & to be submitted

Creighton Thomas, Olivia White; support from the NSF

KITP - Memory Formation in Matter - February 20, 2018

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Theory & computation: explore possibilities for particular class of memory

Many examples of memories in matter

Stone and chisel / Paper and pencil Photograph / Phonograph Computer: e.g., magnetic domains Associative memory in neural nets (Hopfield model)

Kaiser effect: remembers largest strain Kovacs effect: remembers waiting time Return-point memory in magnets: Nested hysteresis curves Pulse duration memory Multiple transient memories (charge density waves; non-Brownian suspensions) Multiple memories in jammed solids

Echoes: spin; (anharmonic) phonon Aging, rejuvenation and memory in glasses Dynamical systems - remembering initial conditions

Shape-memory alloys Designing in function: memory

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- Expt: bulk susceptibility



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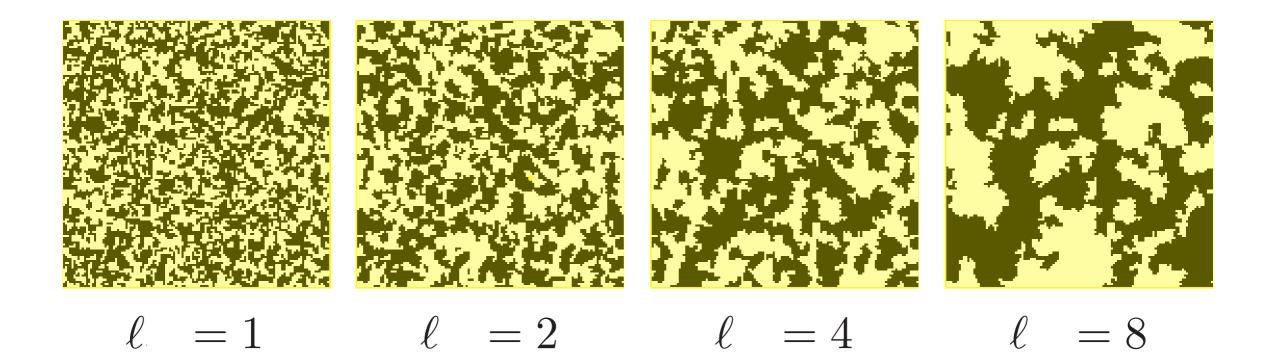
2D, ID Ising spin glass, T=0

 $\mathcal{H}(\vec{s}) = -\sum_{\langle ij \rangle} J_{ij} s_i s_j$ Square/linear/ladder lattice, index *i* for spins, Ising spins $s_i = \pm 1$, Gaussian distributed J_{ij} , mean 0

Two global ground states Two configurations denoted by A, \overline{A} Pictures: not s_i , instead whether s_i aligned with A or \overline{A}

Coarsening via patches

Coarsening via patches



Start with random spins, ℓ is patch size Dark = A phase, light = \overline{A} phase.

Inspired by, compare with, other spin glass work:

Ye, Gheissari, Machta, Newman, Stein (2016): Detailed study with of single spin flips Dependence of local T = 0 aging on dimensionality

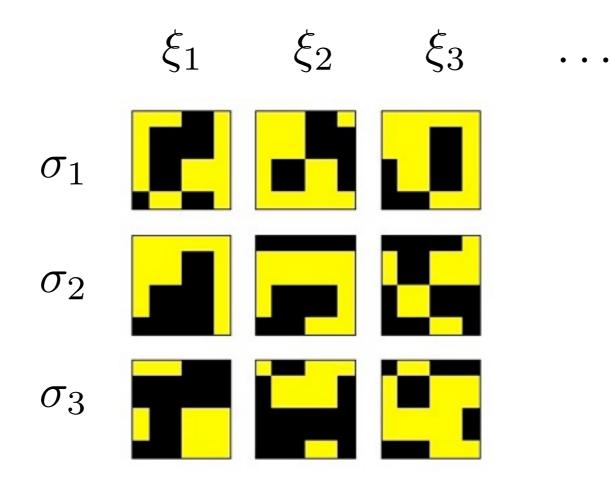
Chanal and Krauth (2010):

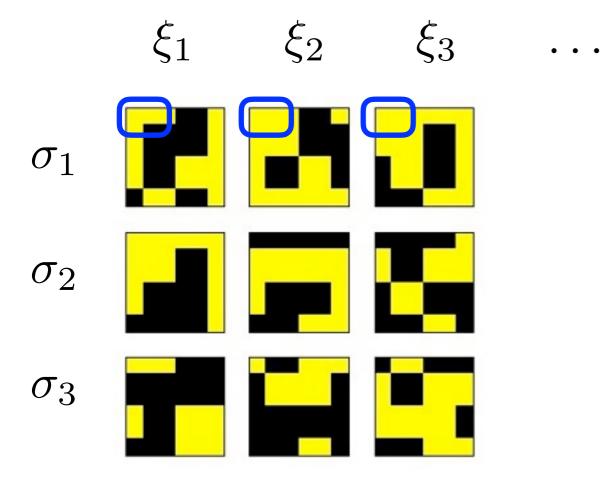
Multi-scale coupling from the past in 2DISG $T \neq 0$ Found: final configuration not dependent on initial configuration at high T Nature/environment, Genesis/initial conditions, Nurture/history of noise

Given

- environment = J_{ij}
- genesis = initial spins σ_i
- nurture = history of patch placements = random noise ξ

Find configurations $s_i(J_{ij}, \sigma, \xi)$

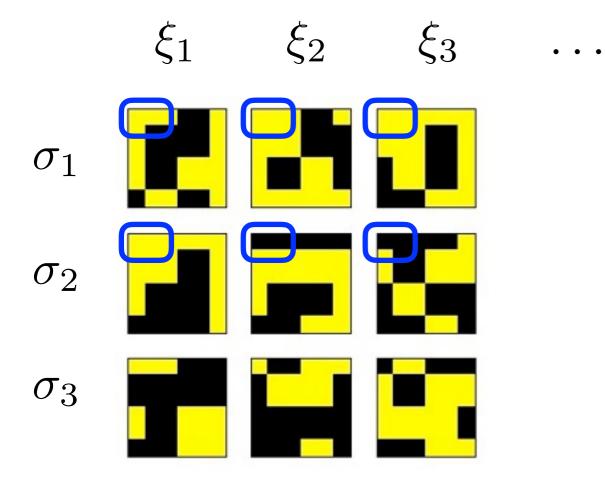




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Average over noise history ξ :

 $\langle s_0 s_1
angle (\sigma_1)$

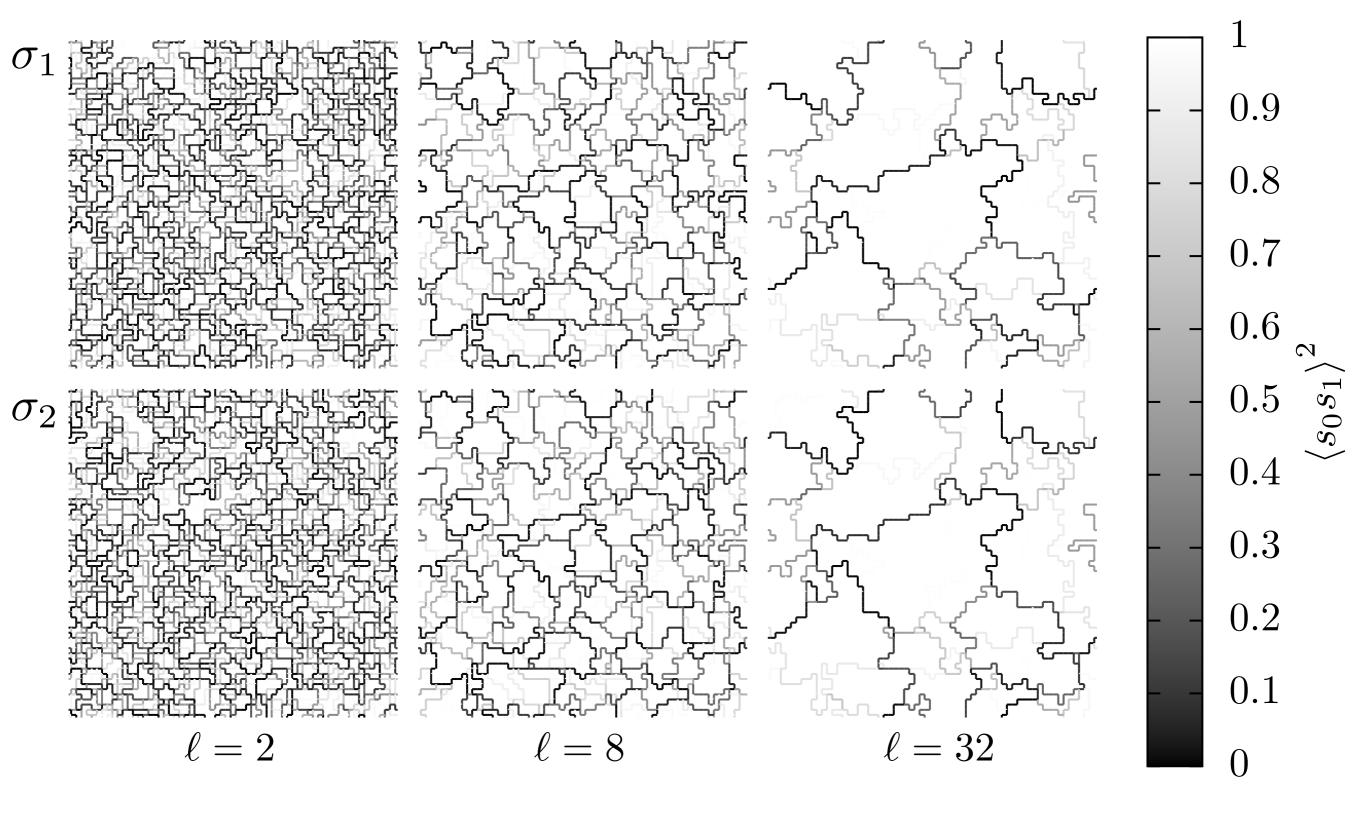


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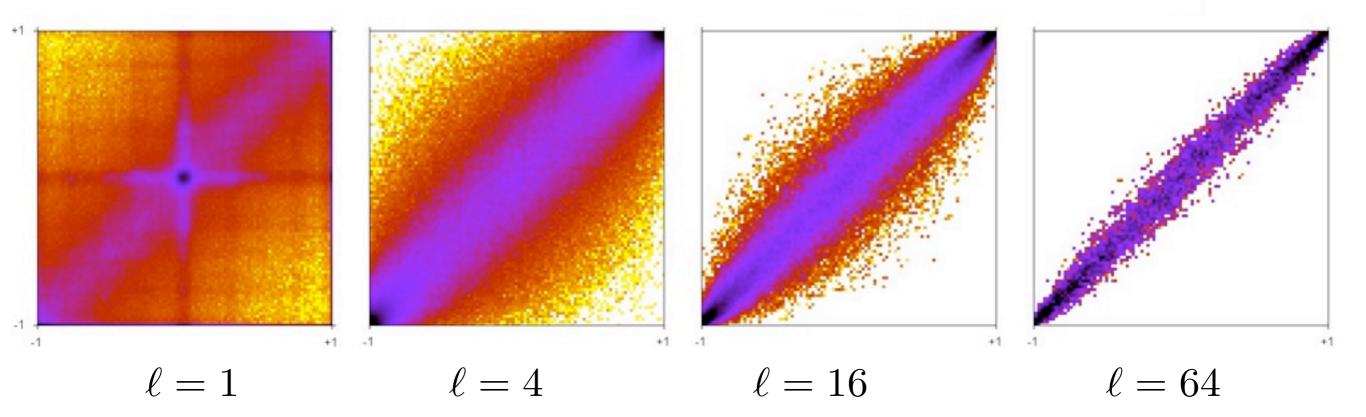
Average over noise history ξ : $\langle s_0 s_1 \rangle (\sigma_1)$

 $\langle s_0 s_1
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Two initial configs / Multiple noise histories dark (light) bonds = "floppy" (rigid) with respect to noise

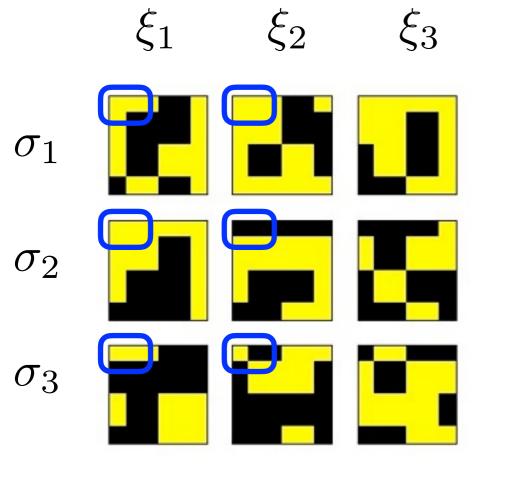


Noise-history-averaged bond correlations for two initial σ , $\langle s_i s_j \rangle (J_{ij}, \sigma_2)$ vs. $\langle s_i s_j \rangle (J_{ij}, \sigma_1)$



(all bonds $s_i s_j$, 50 samples J_{ij} with 256² spins, 2 initial conditions $\sigma_{1,2}$, 1000 noise histories ξ)

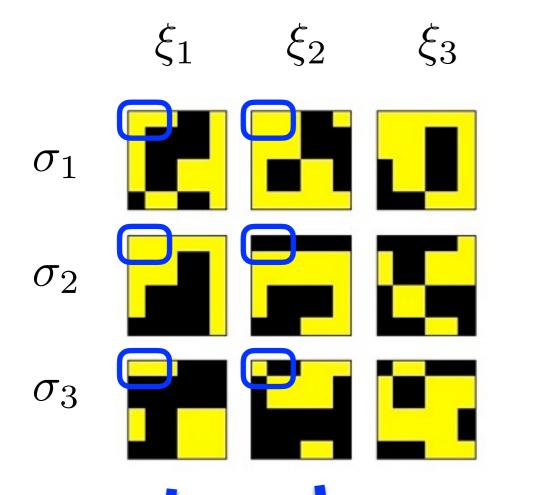
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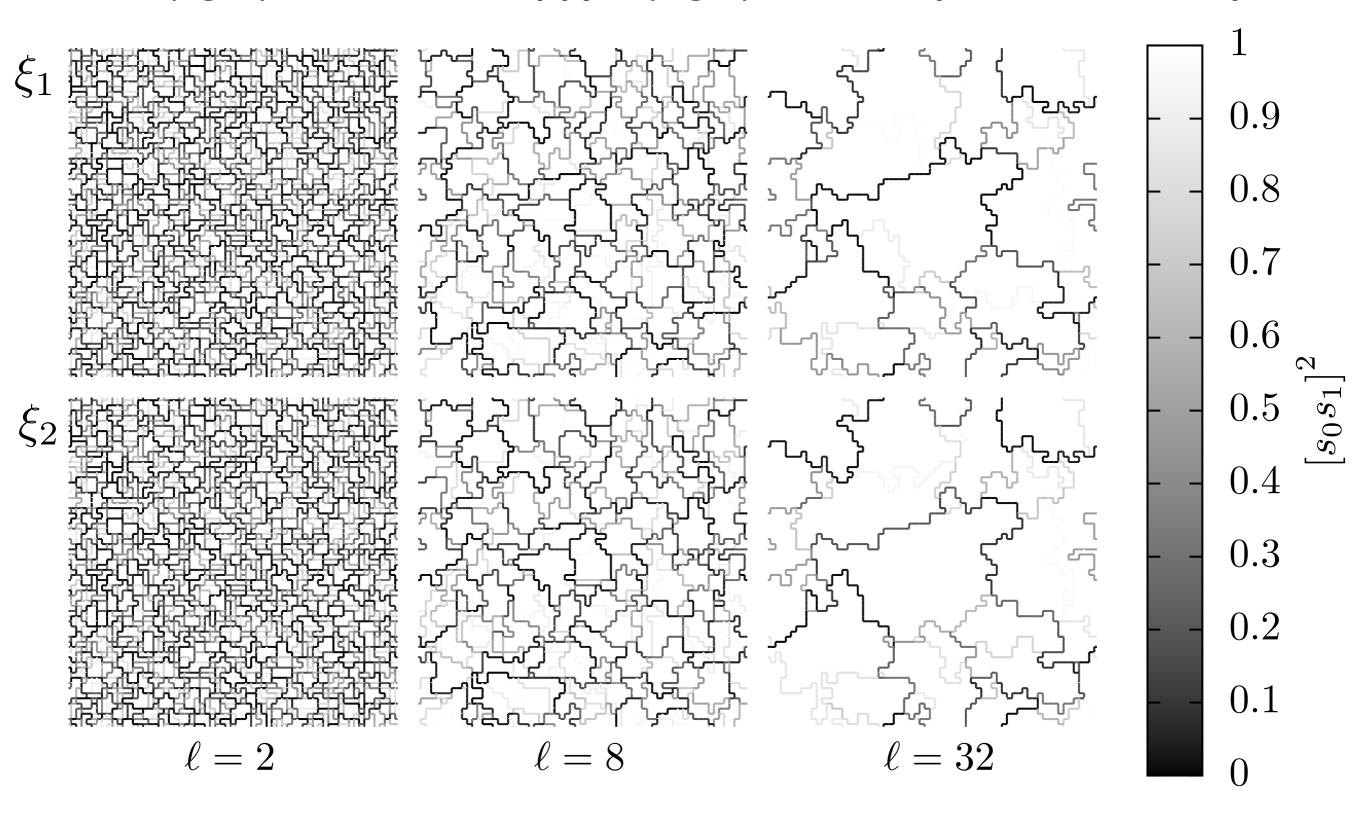
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 $[s_0s_1](J_{ij},\xi_1) \ [s_0s_1](J_{ij},\xi_2)$

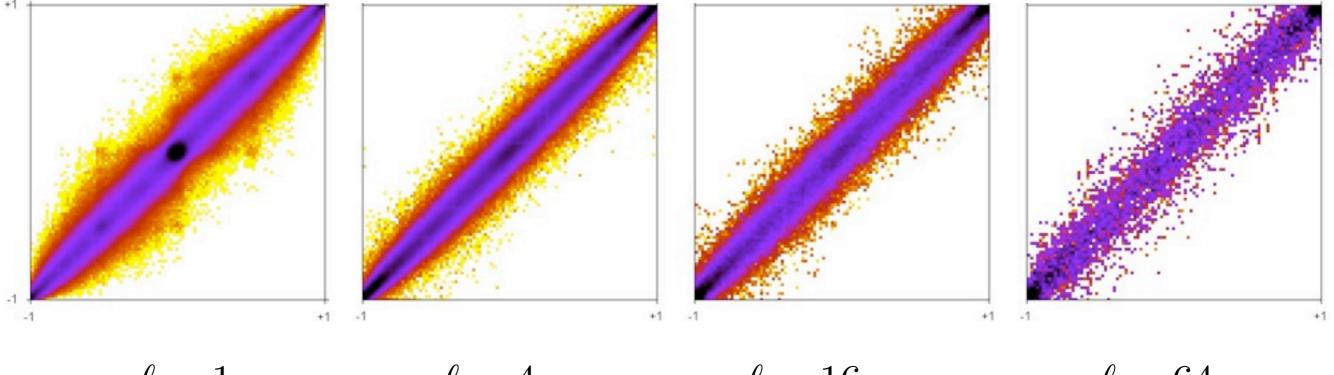
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Two noise histories / Multiple initial configs

dark (light) bonds = "floppy" (rigid) with respect to initial spins

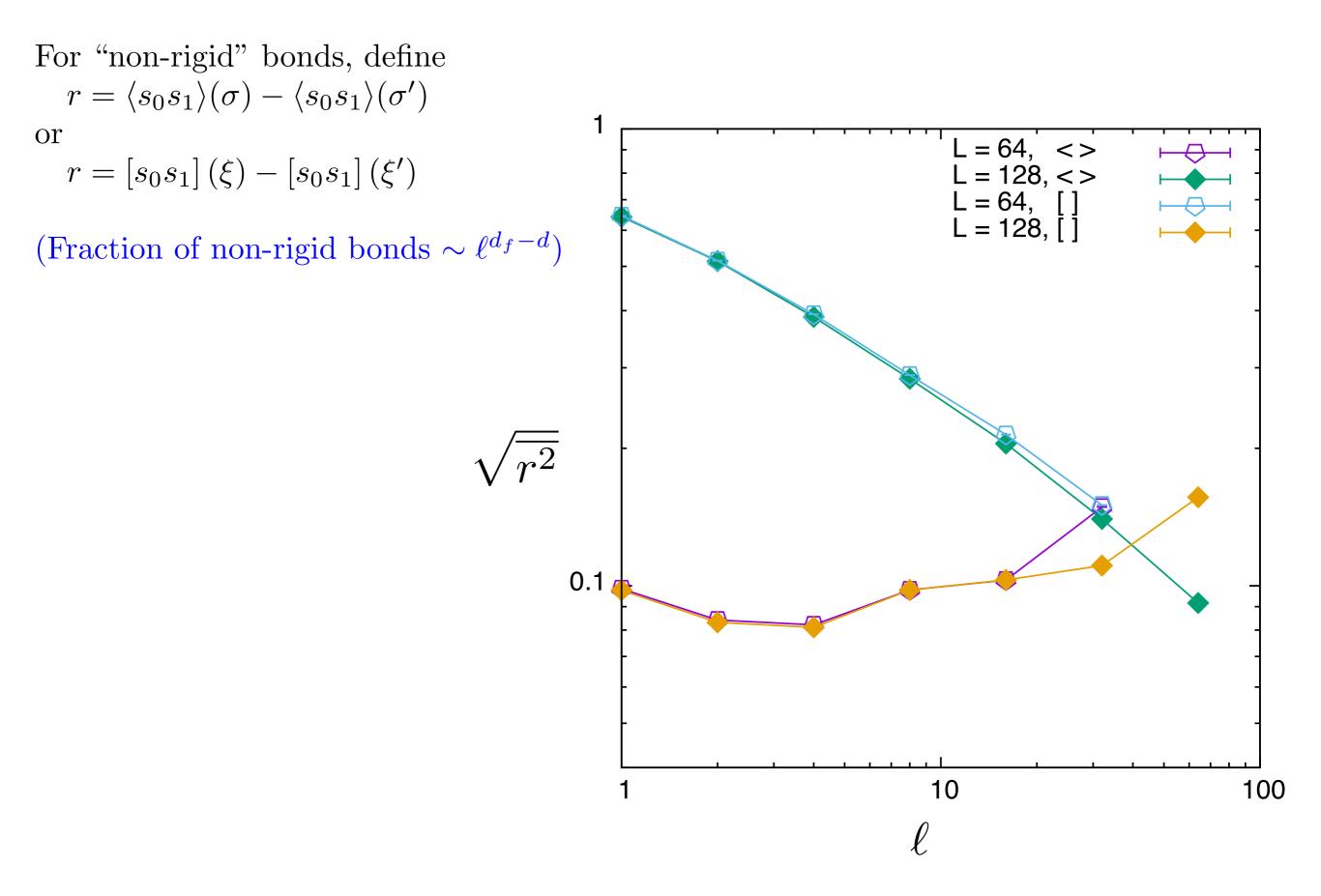


Initial-configuration-averaged bond correlations for two noise histories ξ , $\langle s_i s_j \rangle (J_{ij}, \xi_2)$ vs. $\langle s_i s_j \rangle (J_{ij}, \xi_1)$



 $\ell = 1 \qquad \qquad \ell = 4 \qquad \qquad \ell = 16 \qquad \qquad \ell = 64$

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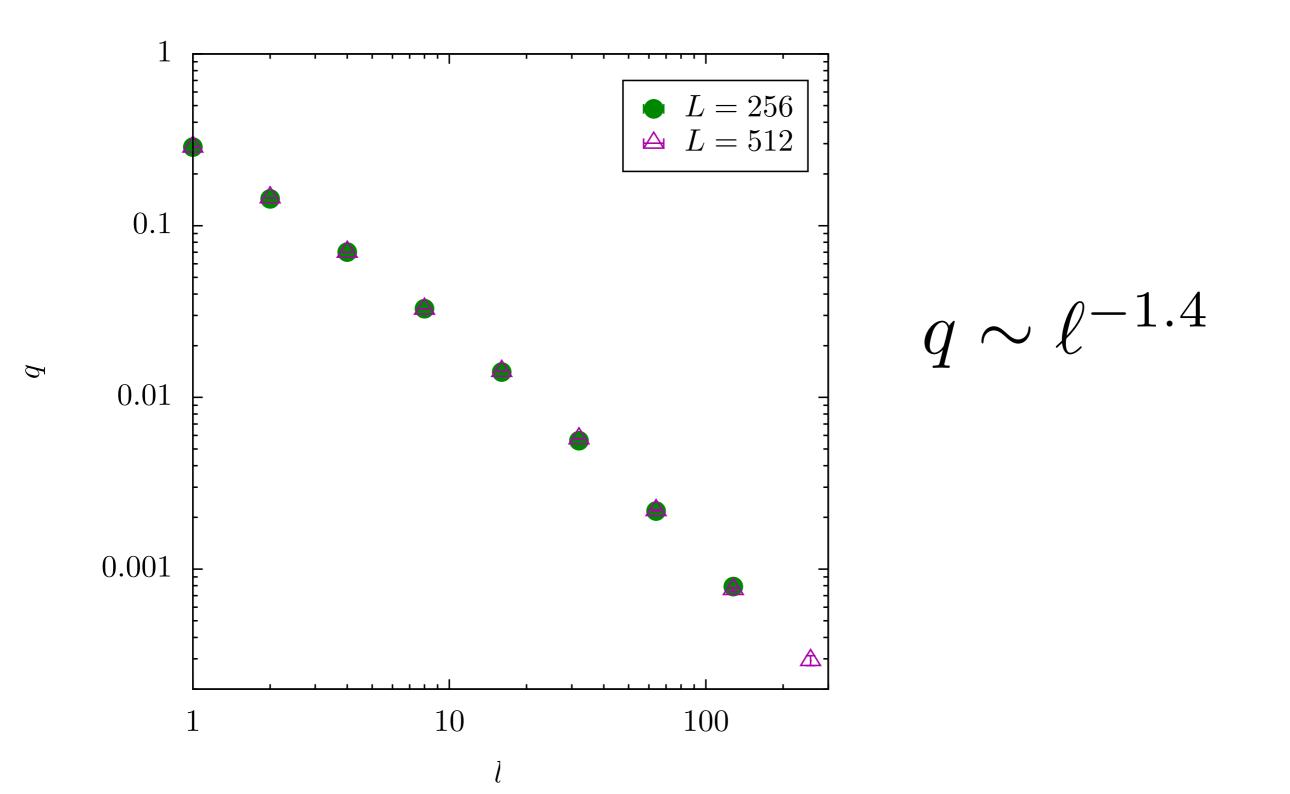
Noise, genesis, environment w/frustration

- Both $\langle \ldots \rangle$ and $[\ldots]$: same rigid domains, floppy walls, given J_{ij} .
- Average over histories ξ : floppiness independent of initial configuration σ .
- Average over starts σ : floppiness depends on history ξ .
- Even for [...], domain walls not fixed by history.
- ... scale-invariant uncertainty arising from initial conditions.
- Cf. partial-ordering/coupling-from-the-past: $\xi \rightarrow$ outcome.

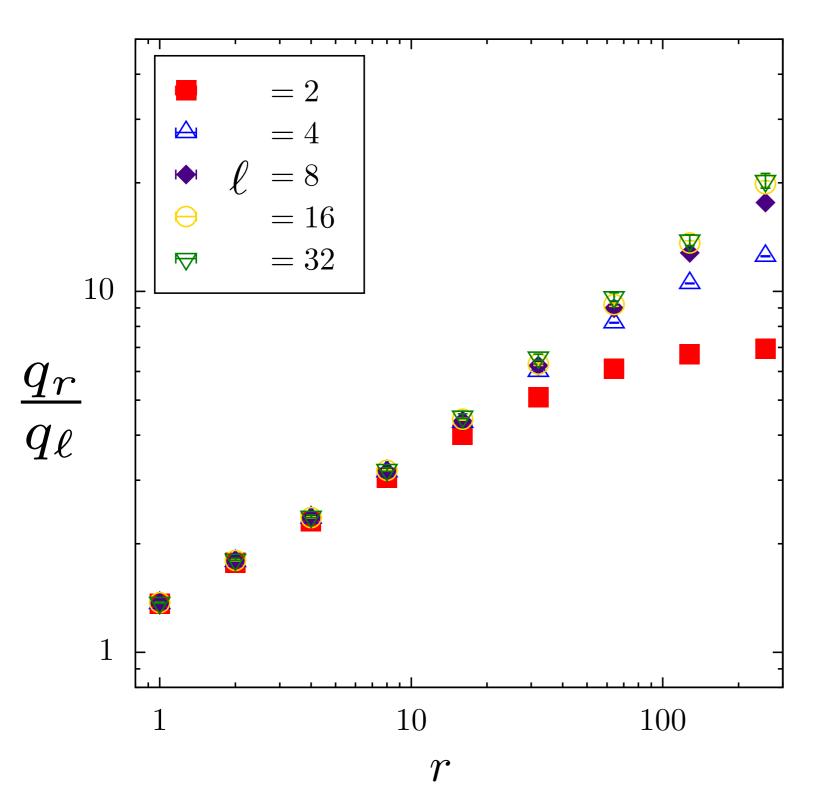
Memory test

- 1. Given J_{ij} , set spins to global ground state $s_i^A(\{J_{ij}\})$.
- 2. Scramble disorder independent J'_{ij} .
- 3. Grow patches to scale ℓ under new landscape.
- 4. Reset to J_{ij} .
- 5. Recover by growing patches to scale r.
- 6. Track overlap $q = N^{-1} \sum_i s_i^A s_i$ all along.

2D aging, couplings J'

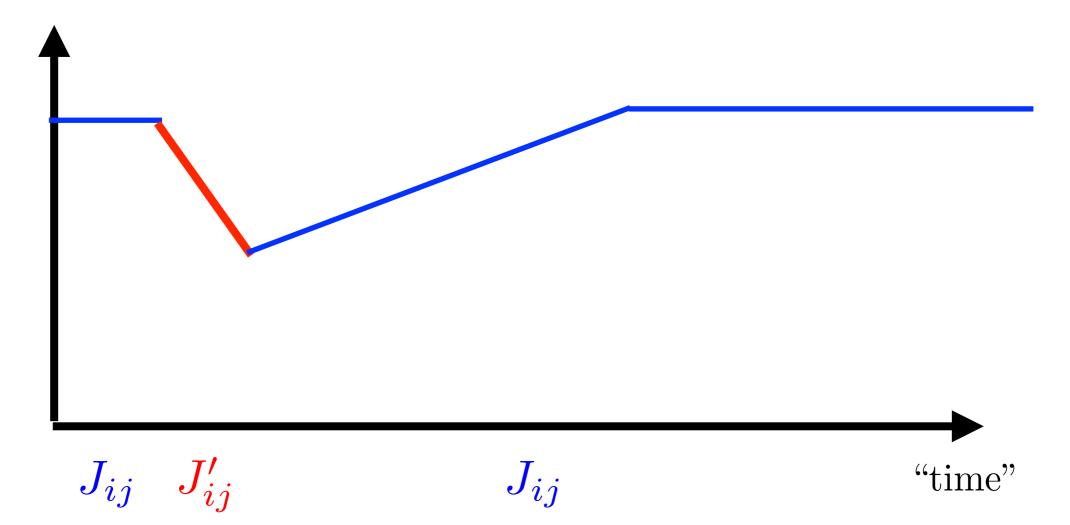


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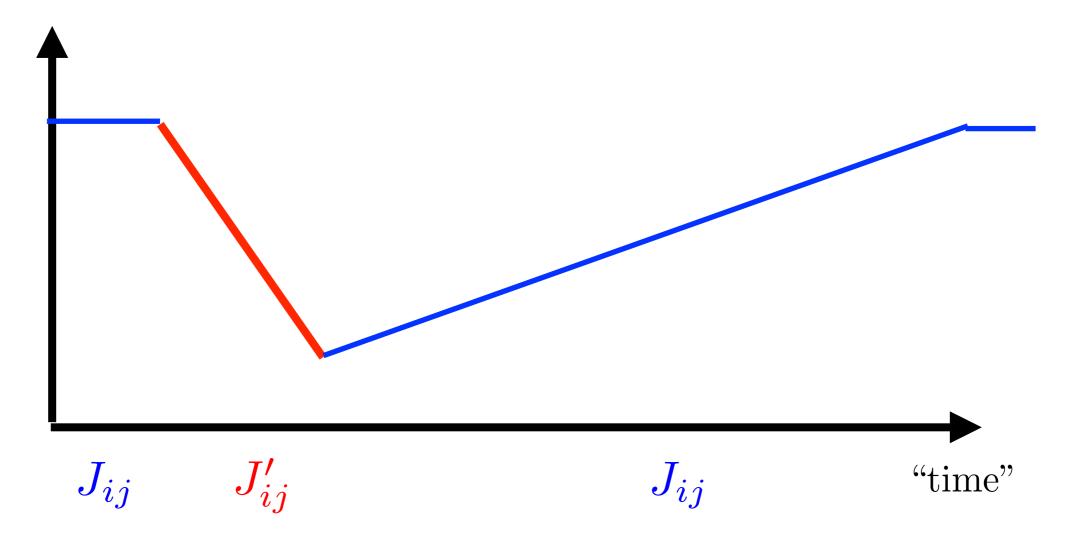


 $\frac{q_r}{q_\ell} \sim r^{0.5 \pm 0.05}$

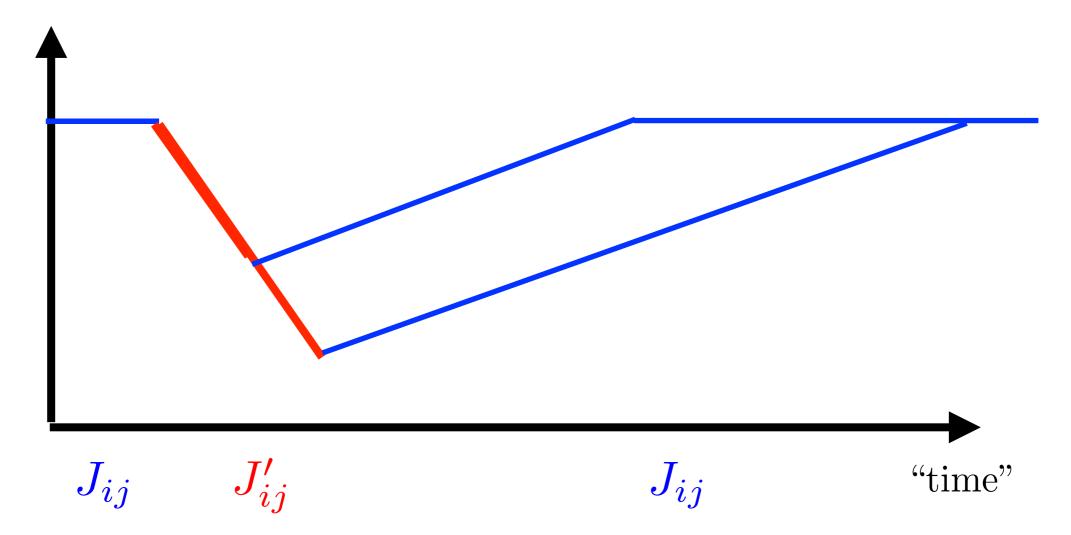
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Recovery length for 2D patches

Scaling argument:

Full recovery at

$$r_c \sim \ell^{1.4/0.5} = \ell^{2.8}$$

 \Rightarrow Recovery at scales $r_c \gg$ than aging scale ℓ

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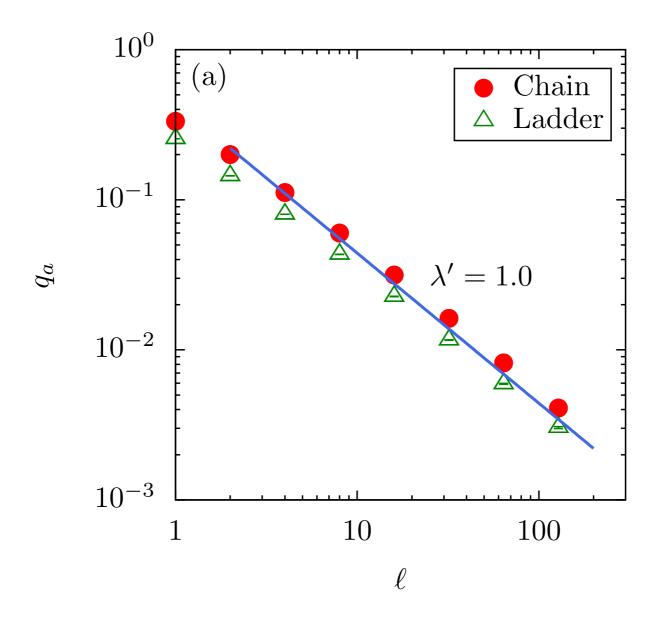
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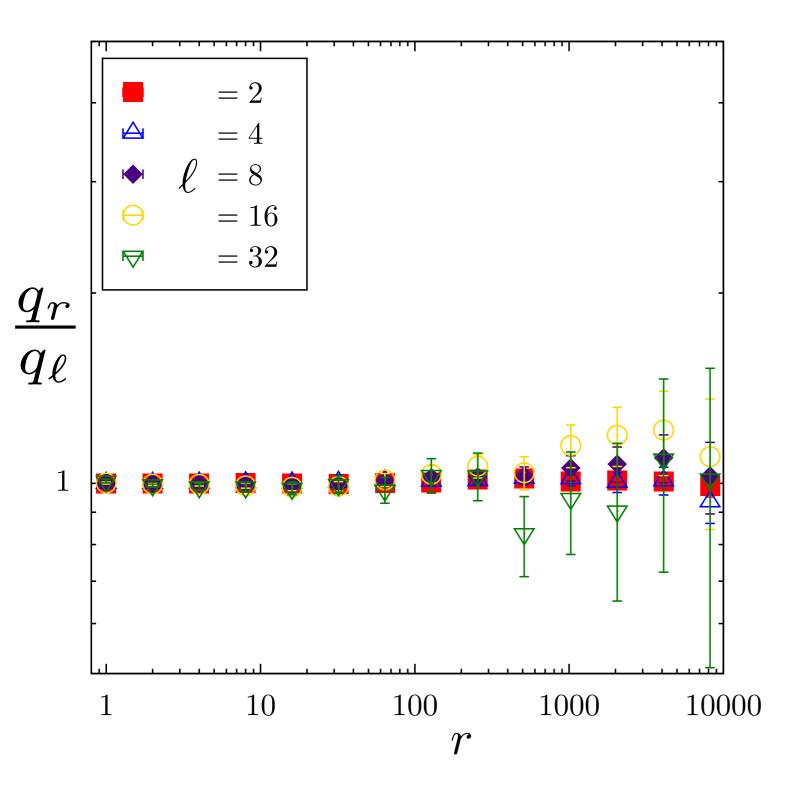
Why?

ID aging

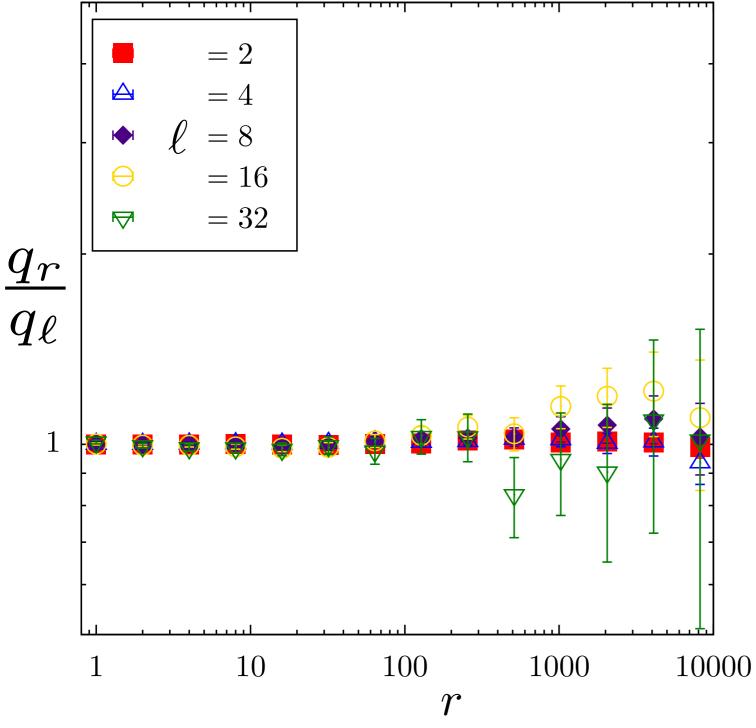




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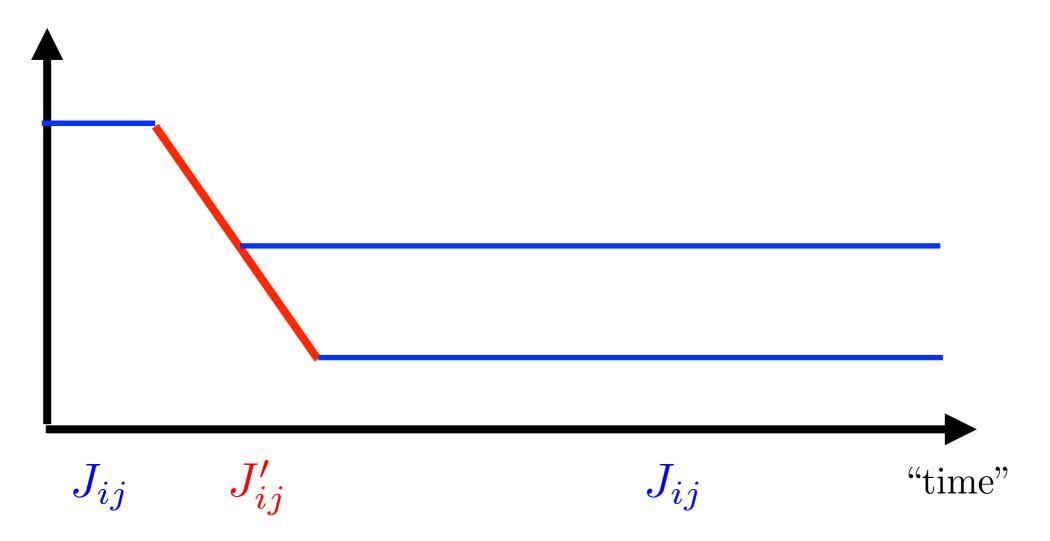


Can prove.

Coarsening while maintaining A to \overline{A} ratio.

Flips to A more frequently, but being "wrong" causes more damage.

$$d = 1$$



Sid Nagel, Feb. 15

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Scaling test

