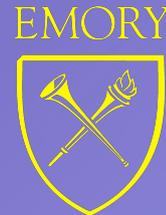


Exploring Spin Glass Ground States
with
Extremal Optimization

Stefan
Boettcher

www.physics.emory.edu/faculty/boettcher/



Find at: www.physics.emory.edu/faculty/boettcher





Collaborator:

- ▶ Allon Percus (Los Alamos/UCLA)

Funding:

- ▶ NSF-DMR, Los Alamos-LDRD, Emory-URC

Find at: www.physics.emory.edu/faculty/boettcher





Overview:





Overview:

- Extremal Optimization (EO) Heuristic
 - EO Algorithm
 - τ -EO, optimizing at the “ergodic edge”





Overview:

- Extremal Optimization (EO) Heuristic
 - EO Algorithm
 - τ -EO, optimizing at the “ergodic edge”
- EO Results for NP-hard Problems
 - Graph Partitioning
 - Coloring
 - Spin Glasses (MAX-CUT)





Overview:

- **Extremal Optimization (EO) Heuristic**
 - **EO** Algorithm
 - **τ -EO**, optimizing at the “ergodic edge”
- **EO Results for NP-hard Problems**
 - Graph Partitioning
 - Coloring
 - Spin Glasses (MAX-CUT)
- **Spin Glass Ground States with τ -EO**
 - Dilute **E**dwards-**A**nderson in $d=3, \dots, 7$
 - Mean-Field: **S**herrington-**K**irkpatrick & **B**ethe **L**attice
 - A Comprehensive View
 - **SK** with Power-Law Bonds $P(J) \sim 1/|J|^{1+\mu}$





Extremal Optimization (EO)

- Motivated by Self-Organized Criticality





Extremal Optimization (EO)

- Motivated by Self-Organized Criticality
 - Emergent Structure
 - ★ *without* tuning any Control Parameters
 - ★ despite (or because of) Large Fluctuations





Extremal Optimization (EO)

- Motivated by Self-Organized Criticality
 - Emergent Structure
 - ★ *without* tuning any Control Parameters
 - ★ despite (or because of) Large Fluctuations
- How can we use it to optimize?





Extremal Optimization (EO)

- Motivated by Self-Organized Criticality
 - Emergent Structure
 - ★ *without* tuning any Control Parameters
 - ★ despite (or because of) Large Fluctuations
- How can we use it to optimize?
 - Extremal Driving:
 - ★ Select and eliminate the “bad”,
 - ★ Replace it *at random*,
 - ★ Eventually, only the “good” is left!





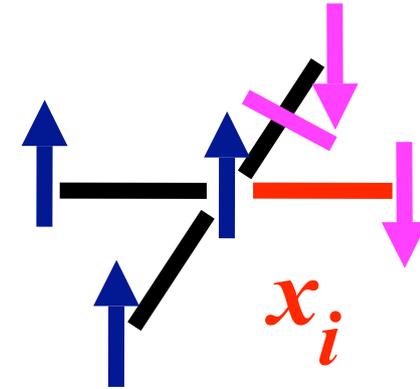
“Fitness” λ for various Problems:



“Fitness” λ for various Problems:

• Spin Glasses (eg. MAX-CUT):

$$\lambda_i = x_i \sum_j J_{i,j} x_j$$



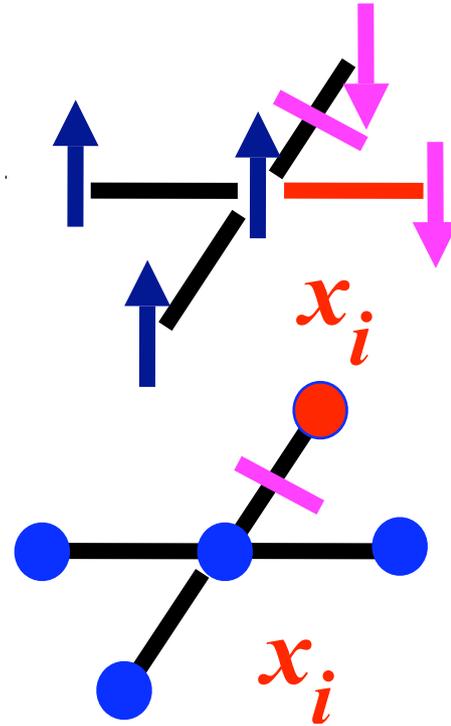
“Fitness” λ for various Problems:

- Spin Glasses (eg. MAX-CUT):

$$\lambda_i = x_i \sum_j J_{i,j} x_j$$

- Partitioning (eg. MIN-CUT):

$$\lambda_i = - (\# \text{-cut edges of } x_i)$$



“Fitness” λ for various Problems:

- Spin Glasses (eg. MAX-CUT):

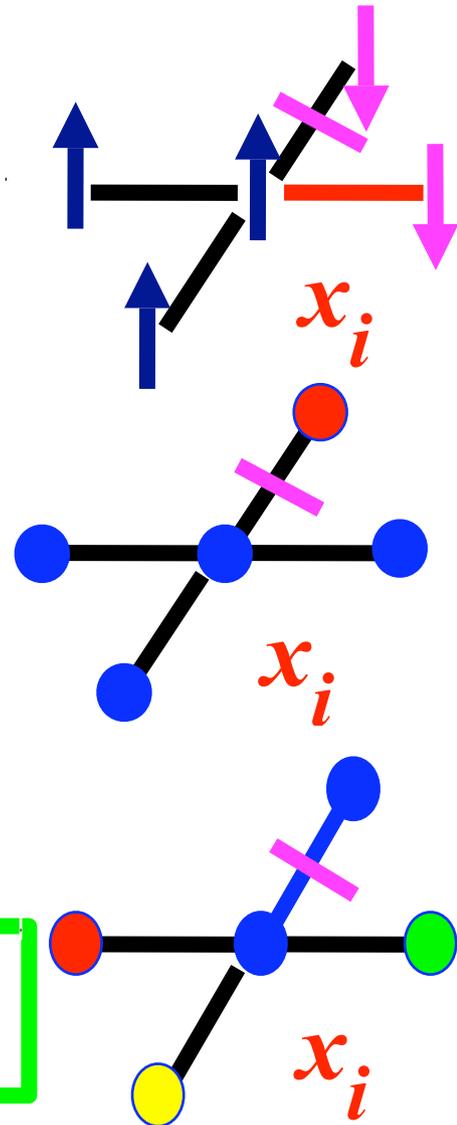
$$\lambda_i = x_i \sum_j J_{i,j} x_j$$

- Partitioning (eg. MIN-CUT):

$$\lambda_i = - (\# \text{-cut edges of } x_i)$$

- Coloring (eg. Potts Anti-ferro):

$$\lambda_i = - (\# \text{-monochrome edges of } x_i)$$



“Fitness” λ for various Problems:

- Spin Glasses (eg. MAX-CUT):

$$\lambda_i = x_i \sum_j J_{i,j} x_j$$

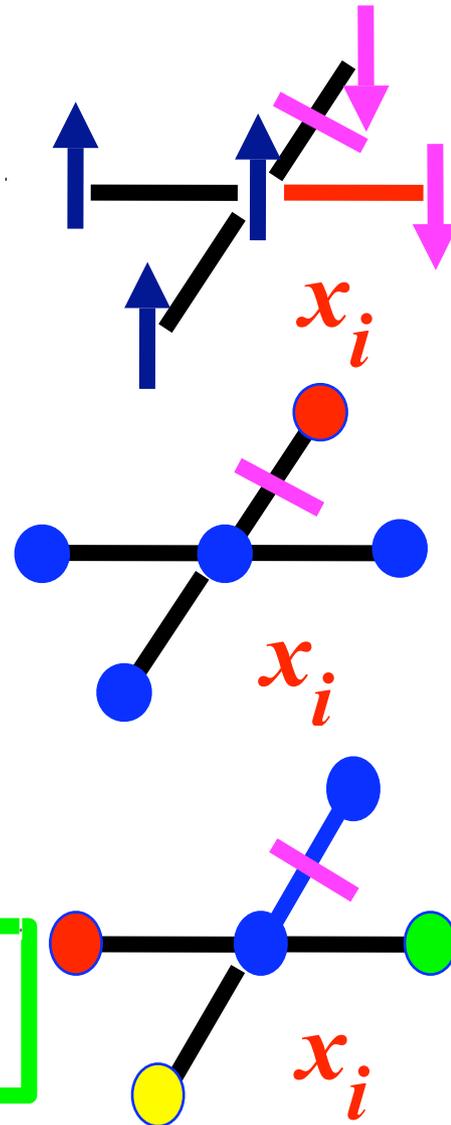
- Partitioning (eg. MIN-CUT):

$$\lambda_i = - (\# \text{-cut edges of } x_i)$$

- Coloring (eg. Potts Anti-ferro):

$$\lambda_i = - (\# \text{-monochrome edges of } x_i)$$

$$\text{Cost} \propto H = - \sum_i \lambda_i$$



>> “Extremal Optimization” (EO): <<



>> “Extremal Optimization” (EO): <<

(1) Provide initial Configuration $S=(x_1, \dots, x_n)$,



>> “Extremal Optimization” (EO): <<

- (1) Provide initial Configuration $S=(x_1, \dots, x_n)$,
- (2) Determine “Fitness” λ_i for each Variable x_i ,



>> “Extremal Optimization” (EO): <<

- (1) Provide initial Configuration $S=(x_1, \dots, x_n)$,
- (2) Determine “Fitness” λ_i for each Variable x_i ,
- (3) Rank all $i=\Pi(k)$ according to

$$\lambda_{\Pi(1)} \leq \lambda_{\Pi(2)} \leq \dots \leq \lambda_{\Pi(n)}$$



>> “Extremal Optimization” (EO): <<

- (1) Provide initial Configuration $S=(x_1, \dots, x_n)$,
- (2) Determine “Fitness” λ_i for each Variable x_i ,
- (3) Rank all $i=\Pi(k)$ according to

$$\lambda_{\Pi(1)} \leq \lambda_{\Pi(2)} \leq \dots \leq \lambda_{\Pi(n)}$$

- (4) Select x_w , $w=\Pi(1)$, i.e. x_w has worst Fitness!



>> “Extremal Optimization” (EO): <<

- (1) Provide initial Configuration $S=(x_1, \dots, x_n)$,
- (2) Determine “Fitness” λ_i for each Variable x_i ,
- (3) Rank all $i=\Pi(k)$ according to

$$\lambda_{\Pi(1)} \leq \lambda_{\Pi(2)} \leq \dots \leq \lambda_{\Pi(n)}$$

- (4) Select x_w , $w=\Pi(1)$, i.e. x_w has worst Fitness!
- (5) Update x_w unconditionally,



>> “Extremal Optimization” (EO): <<

(1) Provide initial Configuration $S=(x_1, \dots, x_n)$,

→ (2) Determine “Fitness” λ_i for each Variable x_i ,

(3) Rank all $i=\Pi(k)$ according to

$$\lambda_{\Pi(1)} \leq \lambda_{\Pi(2)} \leq \dots \leq \lambda_{\Pi(n)}$$

(4) Select x_w , $w=\Pi(1)$, i.e. x_w has worst Fitness!

(5) Update x_w unconditionally,

← (6) For t_{max} times, Repeat at (2),



>> “Extremal Optimization” (EO): <<

(1) Provide initial Configuration $S=(x_1, \dots, x_n)$,

(2) Determine “Fitness” λ_i for each Variable x_i ,

(3) Rank all $i=\Pi(k)$ according to

$$\lambda_{\Pi(1)} \leq \lambda_{\Pi(2)} \leq \dots \leq \lambda_{\Pi(n)}$$

(4) Select x_w , $w=\Pi(1)$, i.e. x_w has worst Fitness!

(5) Update x_w unconditionally,

(6) For t_{max} times, Repeat at (2),

(7) Return: Best $C(S)$ found along the way!





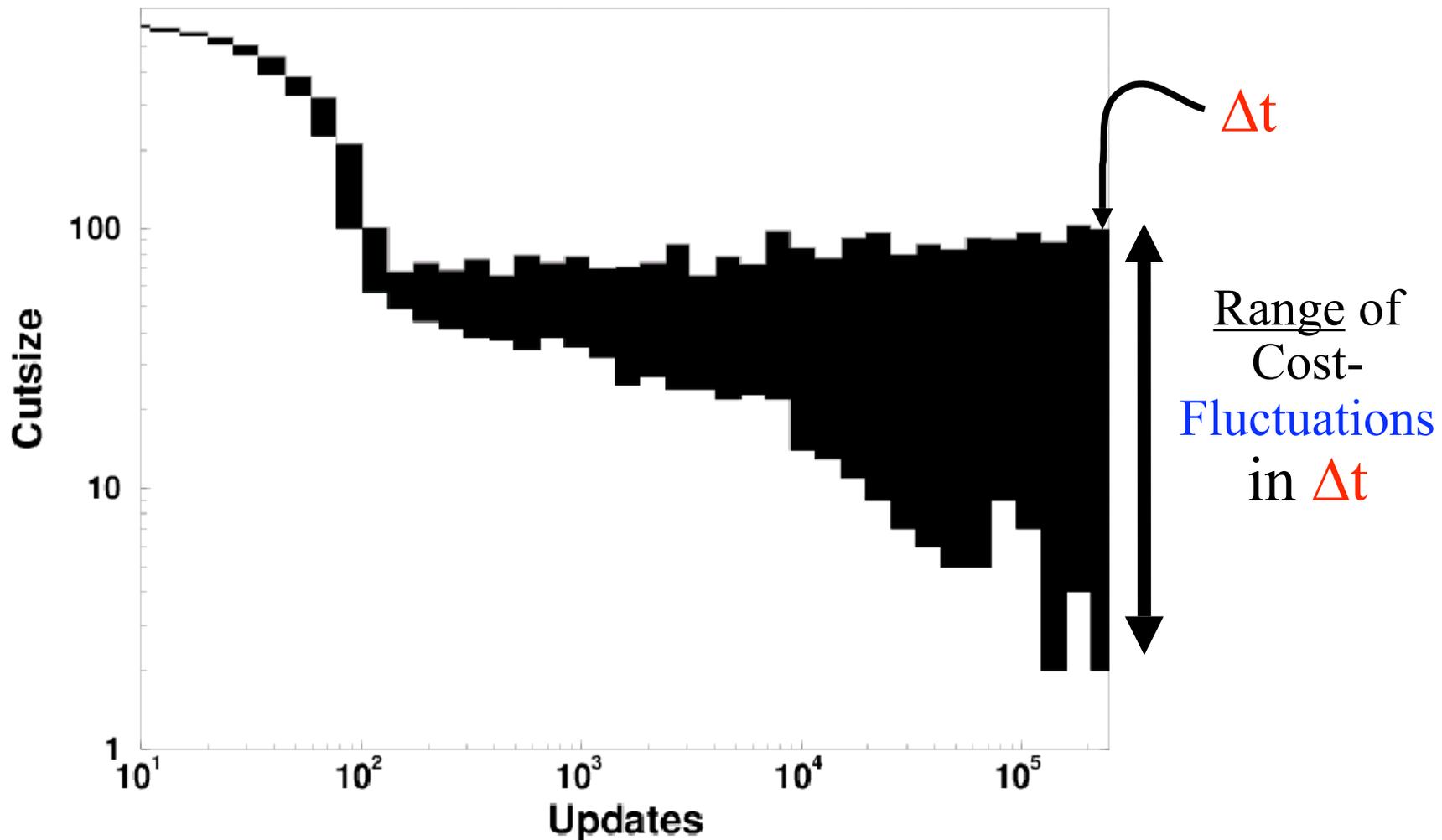
Typical Extremal Optimization Run:

EO-run for Partitioning ($n=500$):



Typical Extremal Optimization Run:

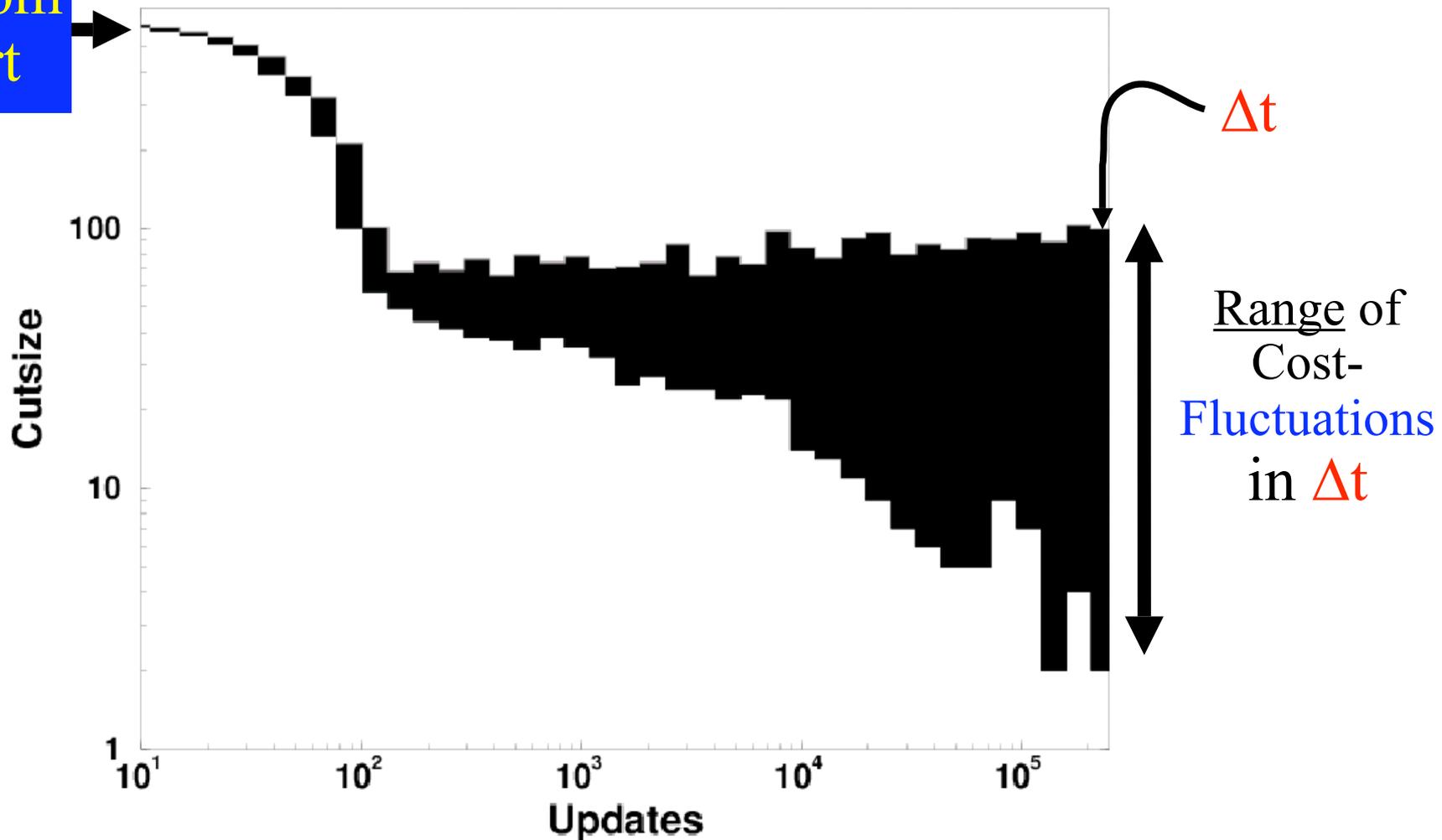
EO-run for Partitioning ($n=500$):



Typical Extremal Optimization Run:

EO-run for Partitioning ($n=500$):

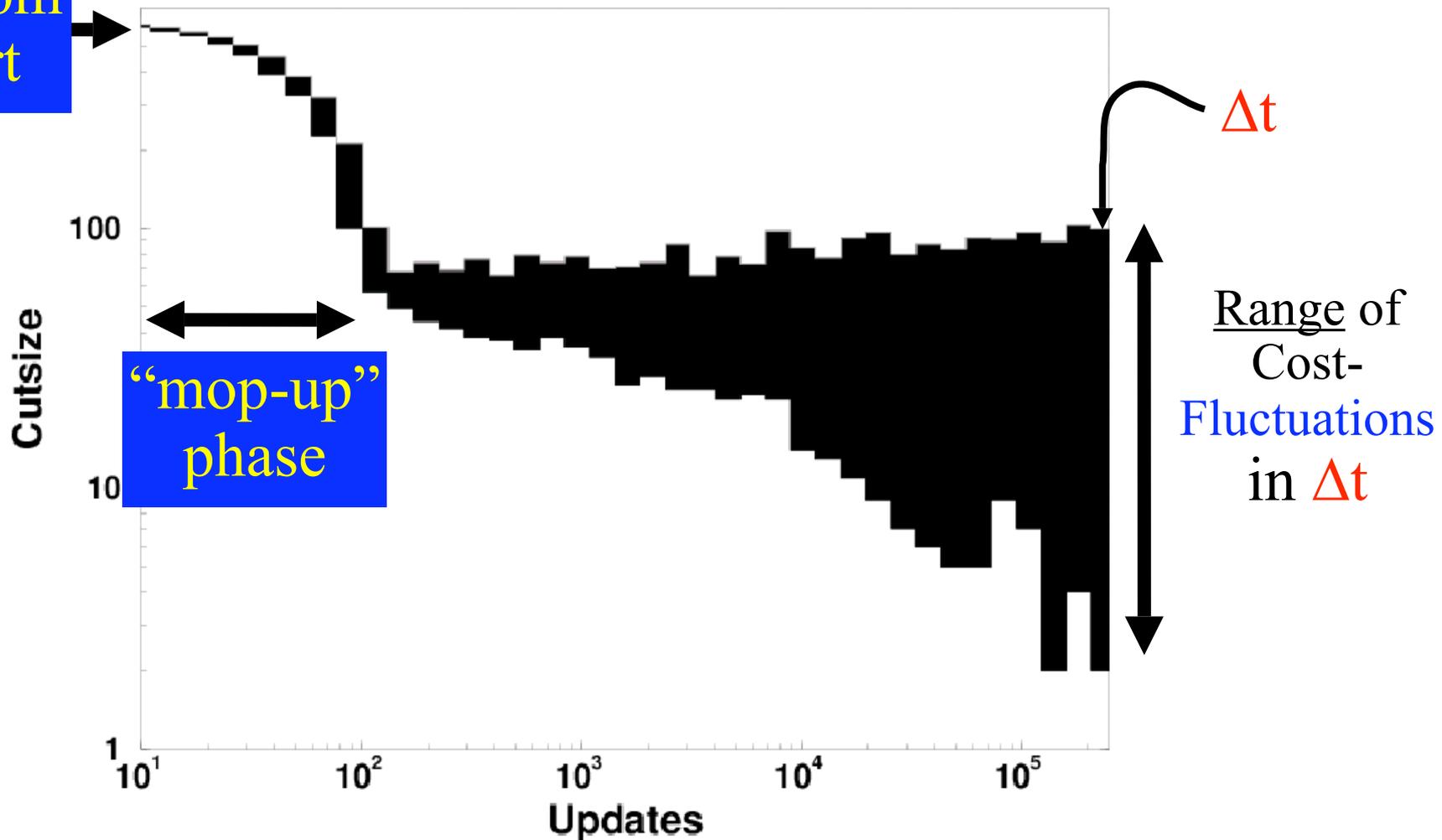
random
start



Typical Extremal Optimization Run:

EO-run for Partitioning ($n=500$):

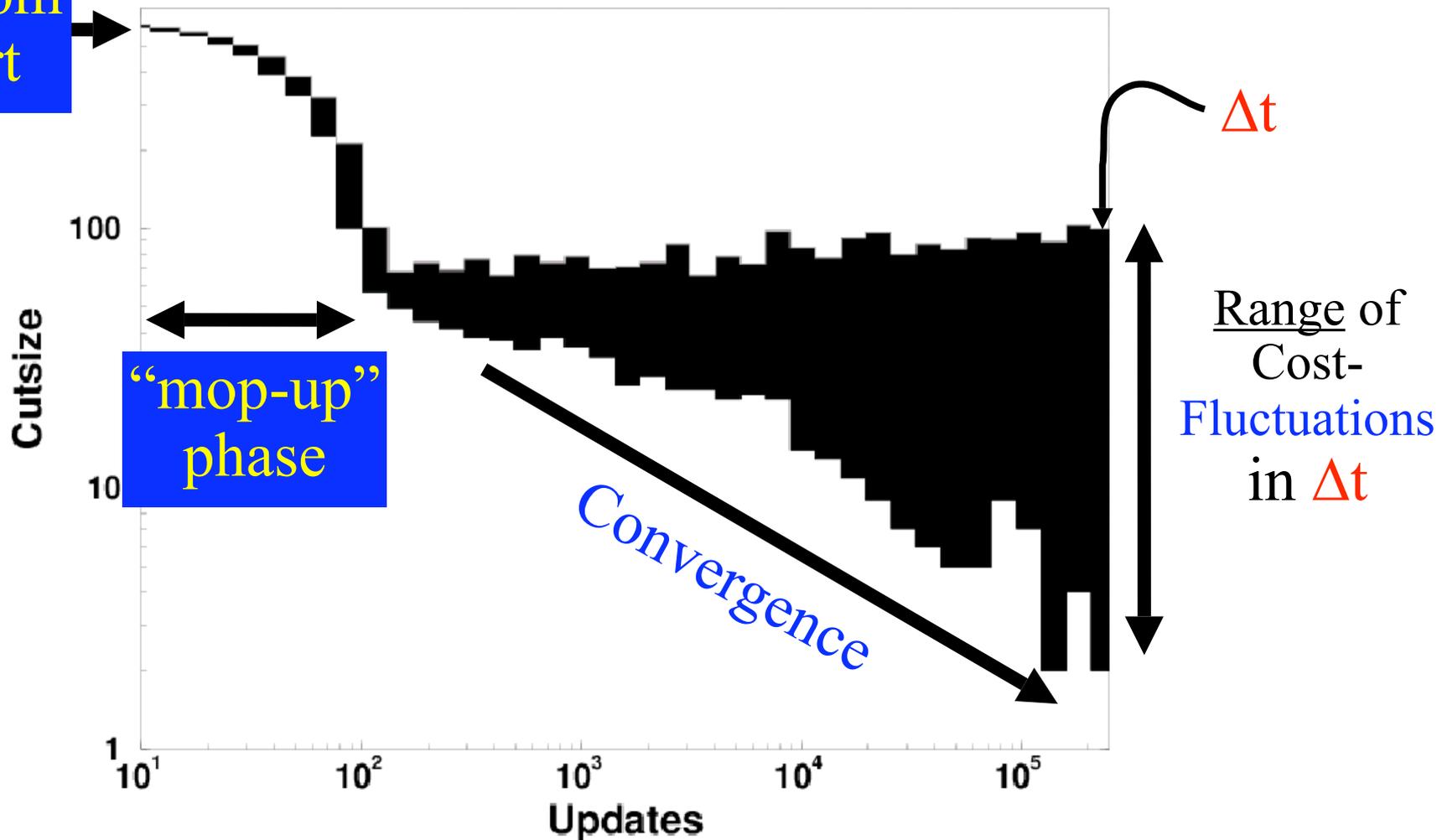
random
start



Typical Extremal Optimization Run:

EO-run for Partitioning ($n=500$):

random
start



τ -EO - Searching at the “Ergodic Edge”:



τ -EO - Searching at the “Ergodic Edge”:

For Ranks $\lambda_{\Pi(1)} \leq \dots \leq \lambda_{\Pi(n)}$, update $i = \Pi(k)$ with



τ -EO - Searching at the “Ergodic Edge”:

For Ranks $\lambda_{\Pi(1)} \leq \dots \leq \lambda_{\Pi(n)}$, update $i = \Pi(k)$ with scale-free, power-law distribution

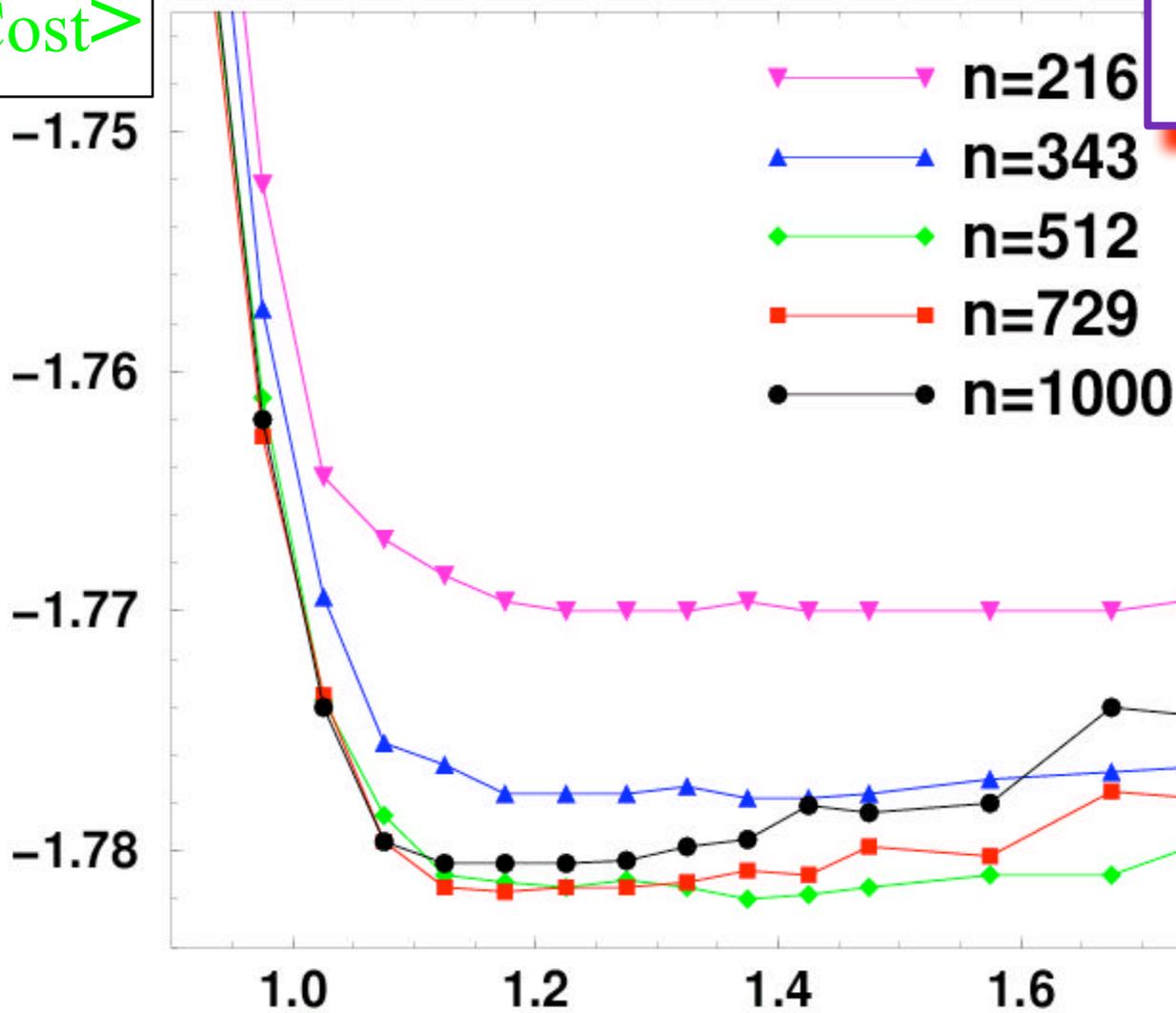
$$P(k) \propto k^{-\tau}$$



τ -EO - Searching at the “Ergodic Edge”:

For Ranks $\lambda_{\Pi(1)} \leq \dots \leq \lambda_{\Pi(n)}$, update $i = \Pi(k)$ with

$\langle \text{Cost} \rangle$



$$P(k) \propto k^{-\tau}$$

τ



τ -EO - Searching at the “Ergodic Edge”:

For Ranks $\lambda_{\Pi(1)} \leq \dots \leq \lambda_{\Pi(n)}$, update $i = \Pi(k)$ with

$\langle \text{Cost} \rangle$

-1.75

-1.76

- ∇ n=216
- \blacktriangle n=343
- \blacklozenge n=512
- \blacksquare n=729
- \bullet n=1000

$$P(k) \propto k^{-\tau}$$

$0 \leftarrow \tau$
random walk,
too ergodic

1.0

1.2

1.4

1.6

1.8

τ

τ -EO - Searching at the “Ergodic Edge”:

For Ranks $\lambda_{\Pi(1)} \leq \dots \leq \lambda_{\Pi(n)}$, update $i = \Pi(k)$ with

$\langle \text{Cost} \rangle$

-1.75

-1.76

- ∇ n=216
- \blacktriangle n=343
- \blacklozenge n=512
- \blacksquare n=729
- \bullet n=1000

$$P(k) \propto k^{-\tau}$$

$0 \leftarrow \mathcal{T}$
random walk,
too ergodic

$\mathcal{T} \rightarrow \infty$
greedy + frozen,
non-ergodic

1.0

1.2

1.4

1.6

1.8

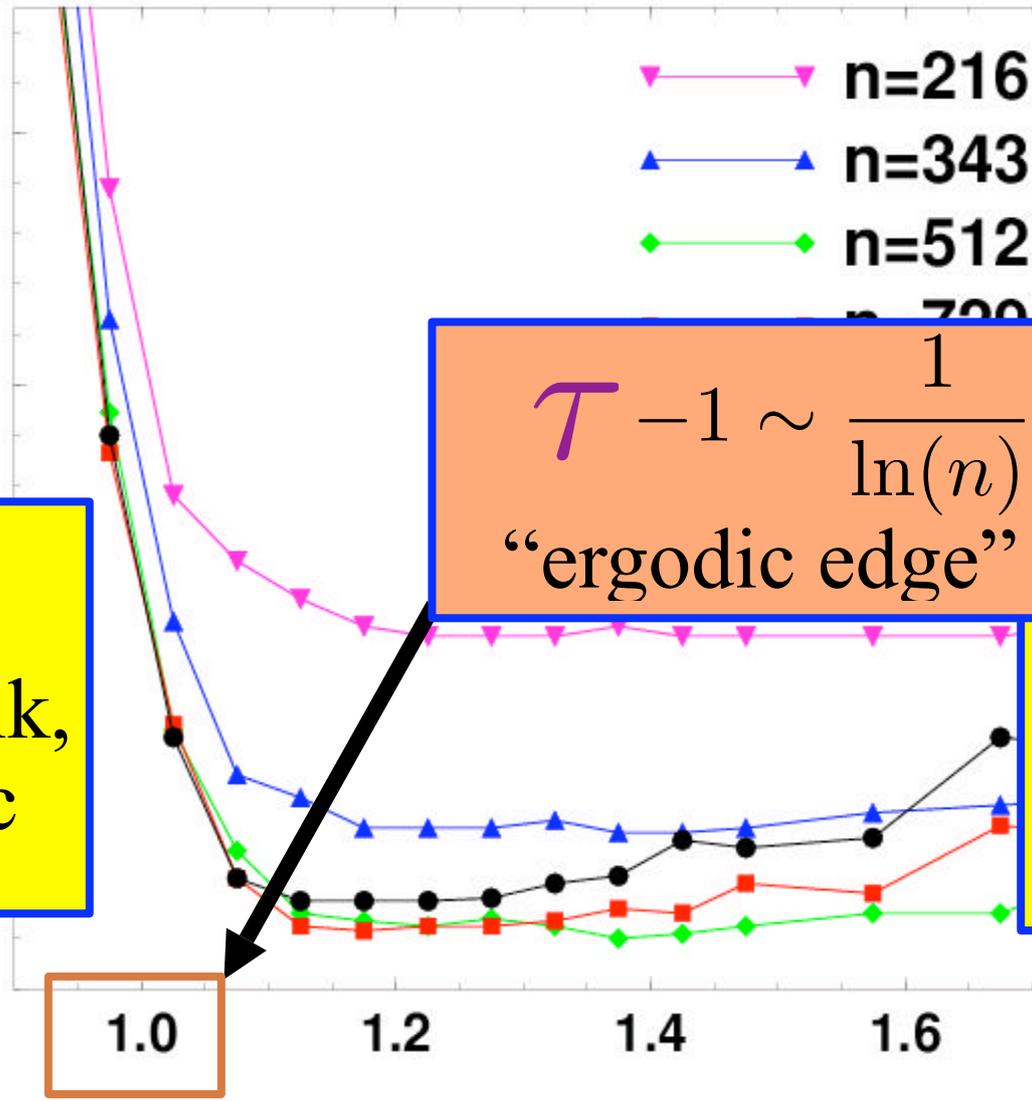
\mathcal{T}

τ -EO - Searching at the "Ergodic Edge":

For Ranks $\lambda_{\Pi(1)} \leq \dots \leq \lambda_{\Pi(n)}$, update $i = \Pi(k)$ with

$$P(k) \propto k^{-\tau}$$

$\langle \text{Cost} \rangle$



$$\tau^{-1} \sim \frac{1}{\ln(n)}$$

"ergodic edge"

$0 \leftarrow \tau$
random walk,
too ergodic

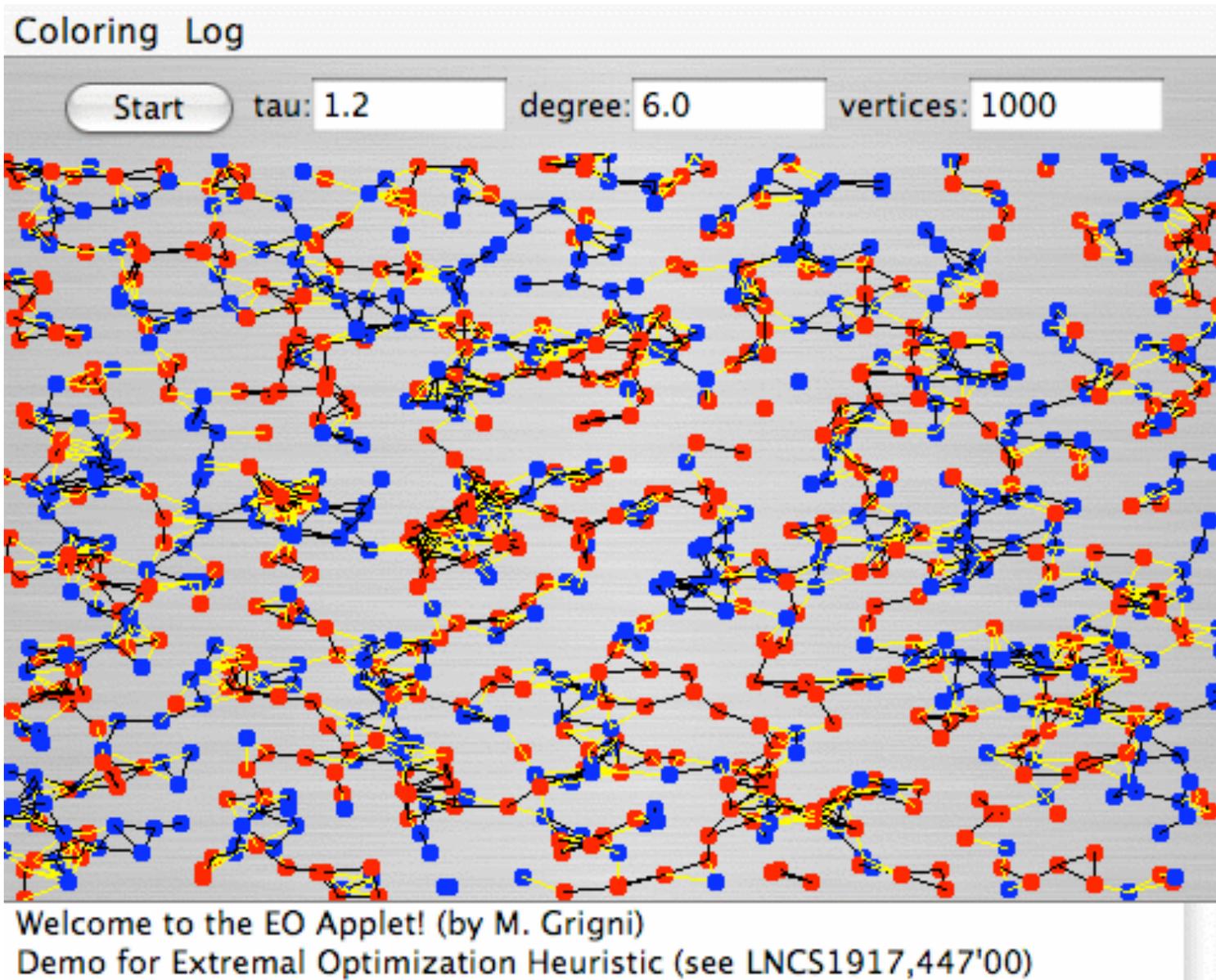
$\tau \rightarrow \infty$
greedy + frozen,
non-ergodic

1.0

τ



Animation of τ -EO for Graph-Partitioning



Results for T-EO:

◦ For Graph Bi-Partitioning:

Graph	Size n	EO	GA	heuristics
<i>Hammond</i>	4720	90 (42s)	90 (1s)	97 (8s)
<i>Barth5</i>	15606	139 (64s)	139 (44s)	146 (28s)
<i>Brack2</i>	62632	731 (12s)	731 (255s)	—
<i>Ocean</i>	143437	464 (200s)	464 (1200s)	499 (38s)

Comparison on Testbed of Graphs [[AI119\('00\)275](#)],

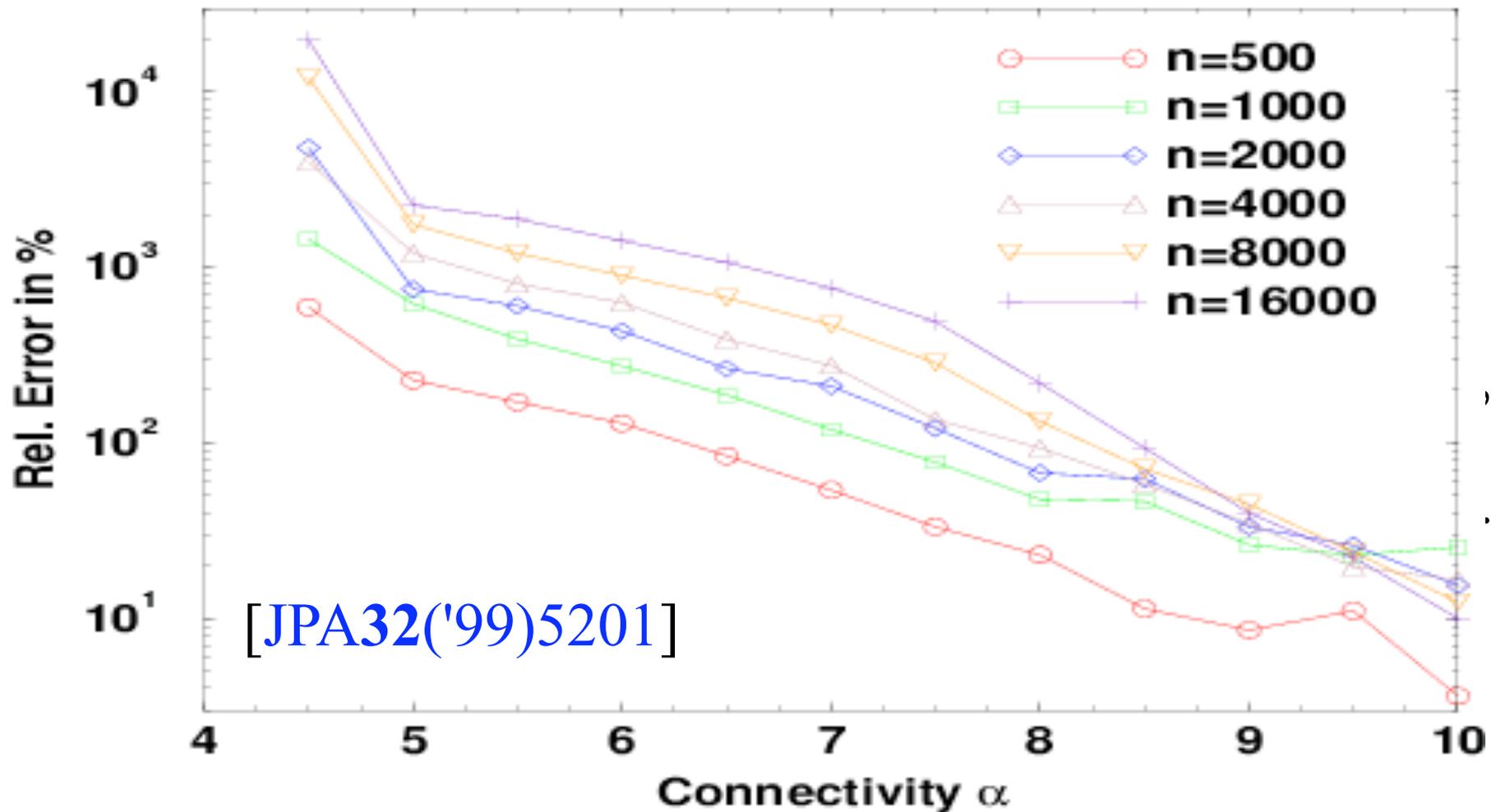
◦ GA by Merz et al. [[LNCS1498\('98\)765](#)],

◦ Spectral Heuristic by Hendrickson et al. [[Supercomputing '95](#)].

Results for τ -EO:

- For Graph Bi-Partitioning:

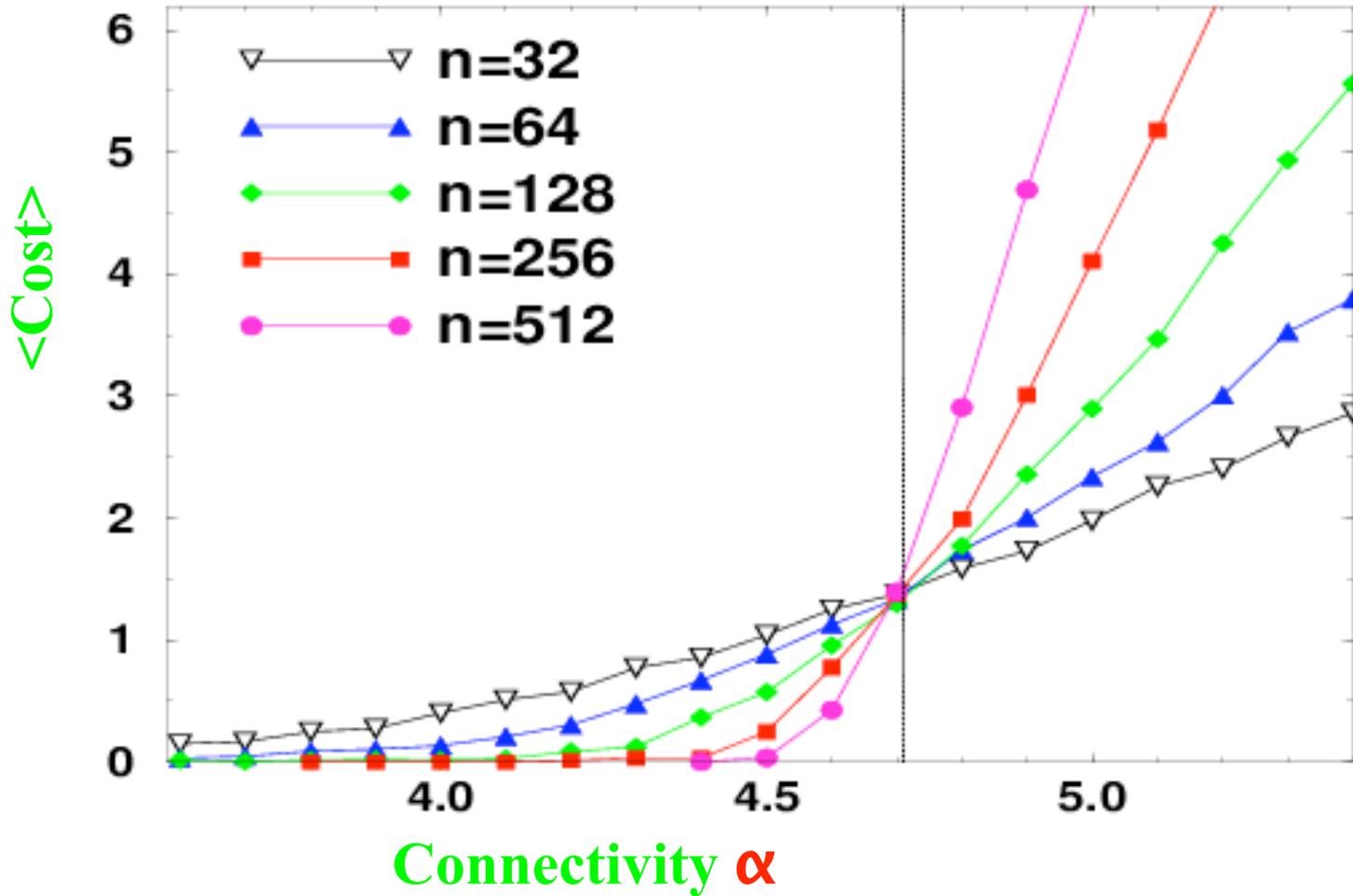
EO vs SA near Percolation for geom. Graphs:





Results for T-EO:

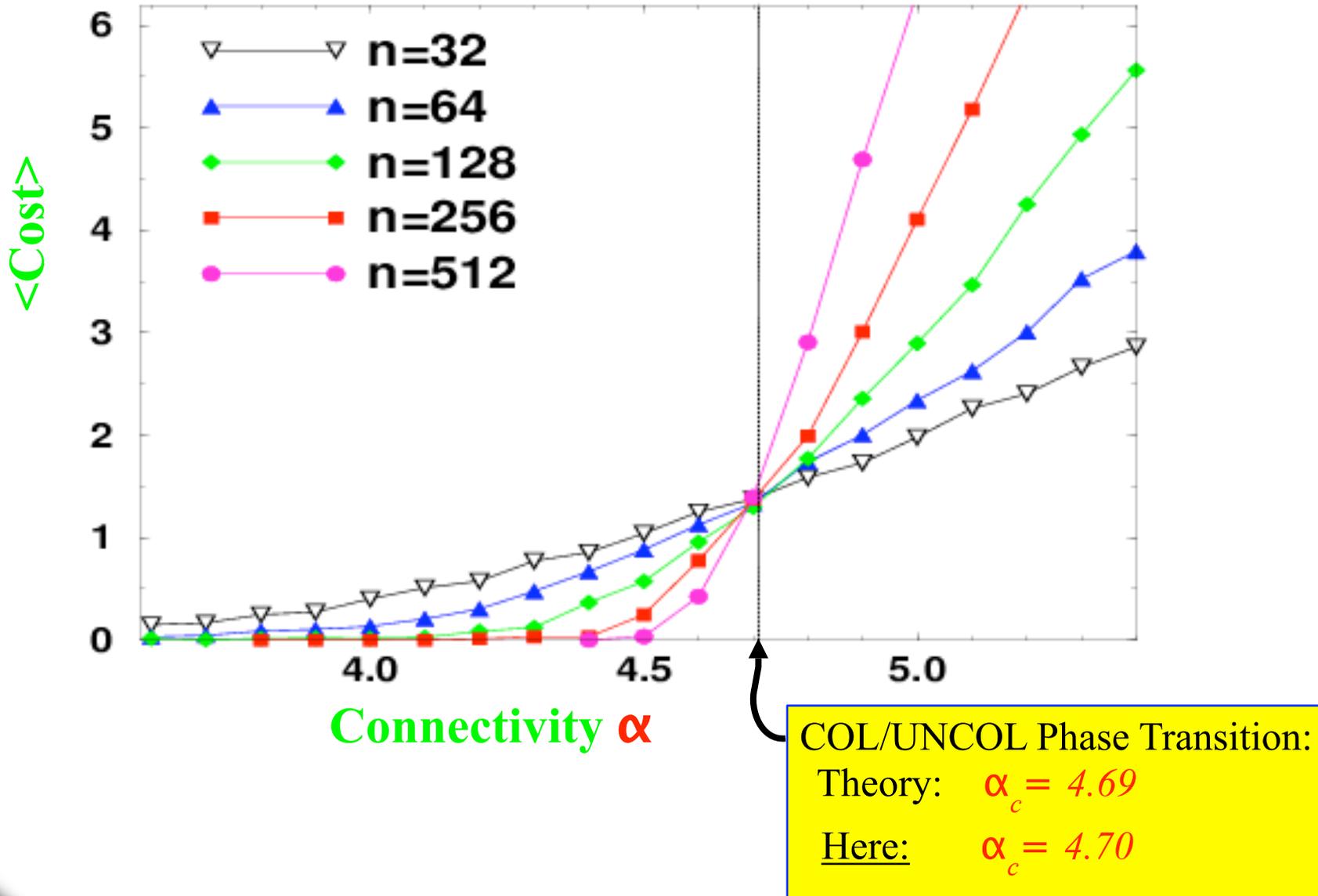
• For Graph-Coloring (MAX-3-COL):





Results for τ -EO:

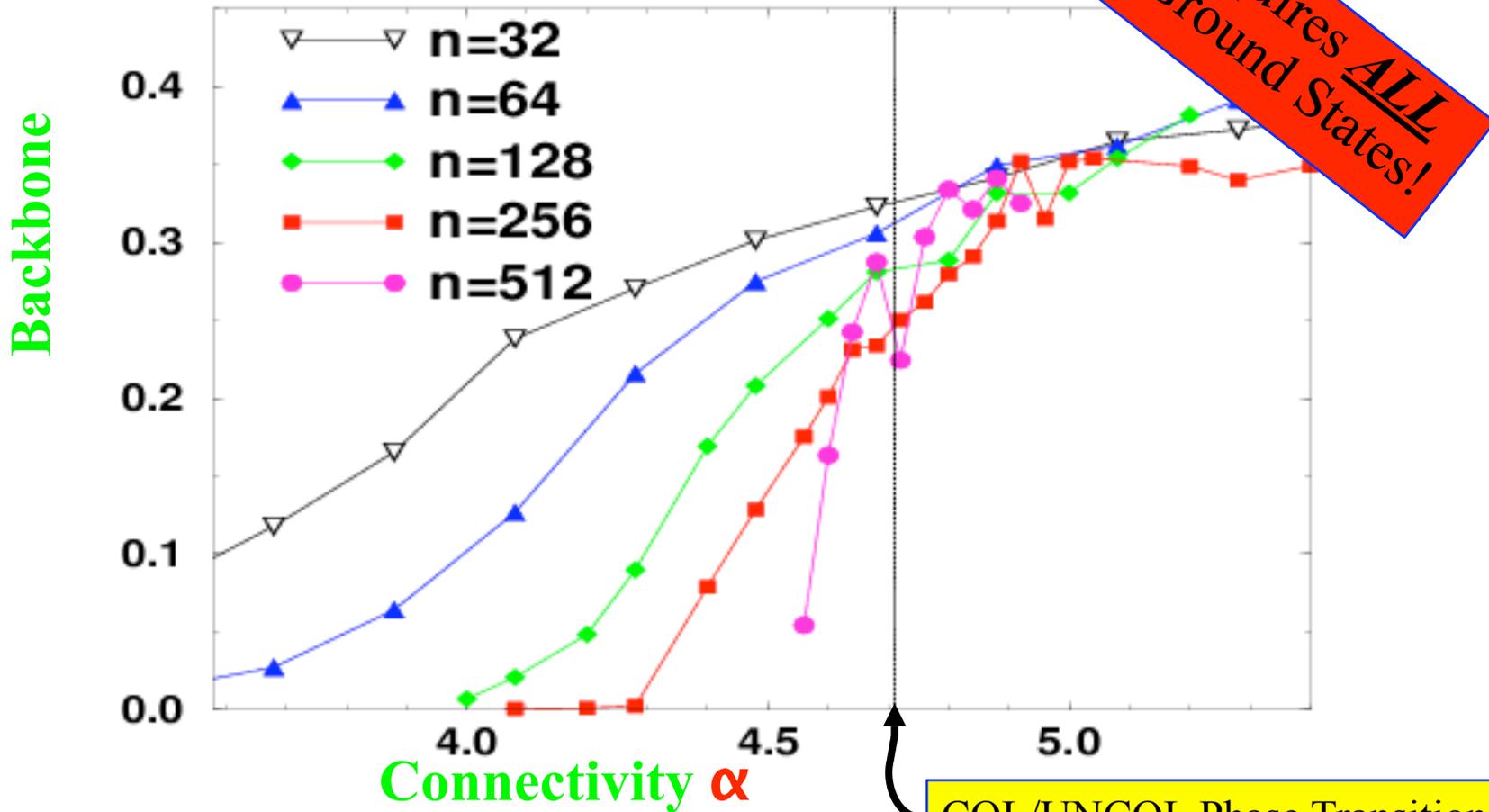
• For Graph-Coloring (MAX-3-COL):





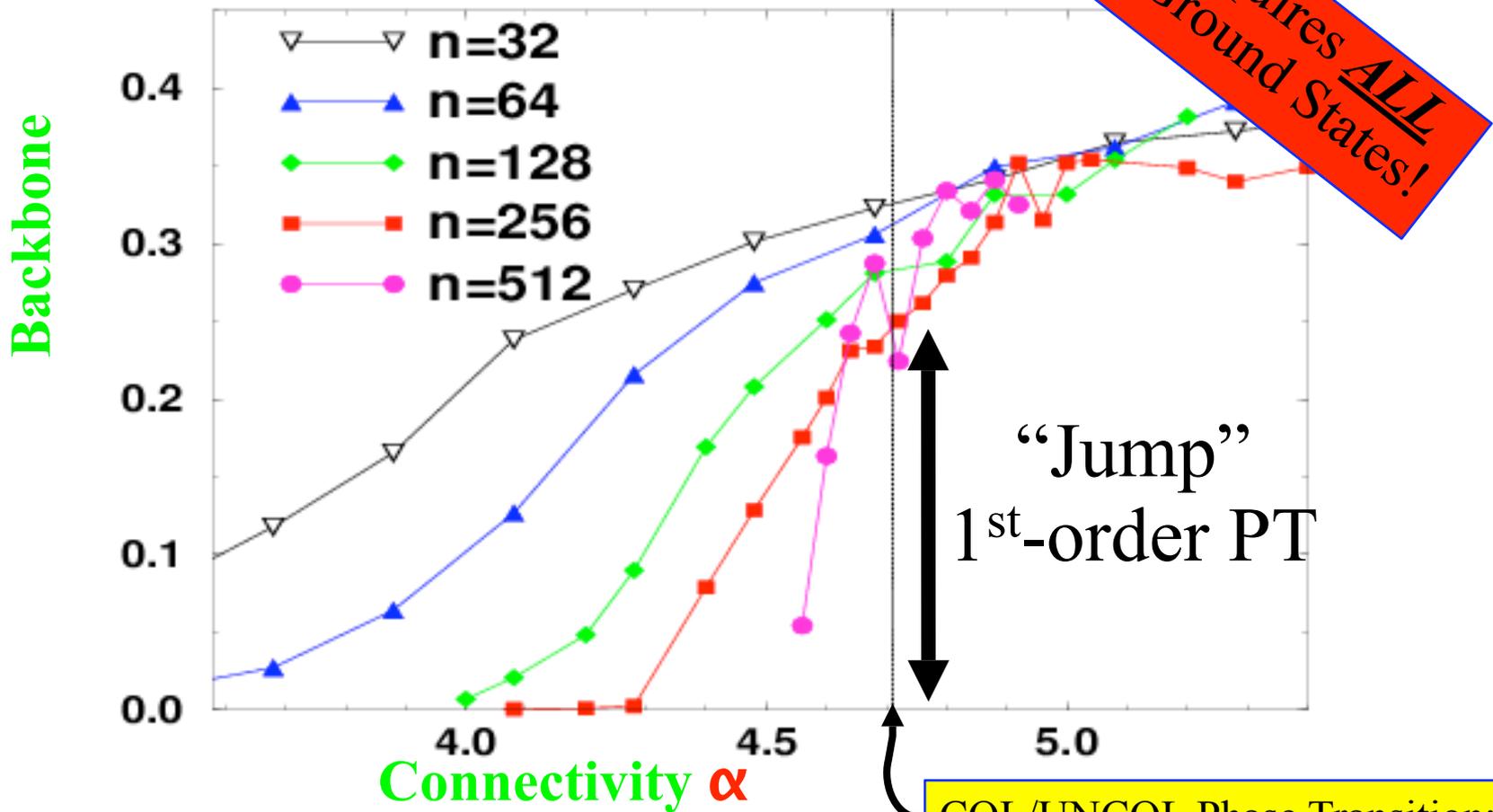
Results for τ -EO:

• For Graph-Coloring (MAX-3-COL):



Results for T-EO:

• For Graph-Coloring (MAX-3-COL):



COL/UNCOL Phase Transition:

Theory: $\alpha_c = 4.69$

Here: $\alpha_c = 4.70$





Results for T-EO:

• For Spin Glasses:

EO for *3d*-Lattice Spin Glasses [[PRL86\('01\)5211](#)]

L	t	E/L^3	Pal96	Hartmann97
3	0.0006	-1.6712(6)	-1.67171(9)	-1.6731(19)
4	0.0071	-1.7377(3)	-1.73749(8)	-1.7370(9)
5	0.0653	-1.7609(2)	-1.76090(12)	-1.7603(8)
6	0.524	-1.7712(2)	-1.77130(12)	-1.7723(7)
7	3.87	-1.7764(3)	-1.77706(17)	
8	22.1	-1.7796(5)	-1.77991(22)	-1.7802(5)
9	100	-1.7822(5)		
10	424.	-1.7832(5)	-1.78339(27)	-1.7840(4)
12	9720.	-1.7857(16)	-1.78407(121)	-1.7851(4)
∞	$O(n^4)$	-1.7865(3)	-1.7863(4)	-1.7868(3)

Genetic Algorithms by Pal [[PhysicaA223\('96\)283](#)]
and by Hartmann [[EPL40\('97\)492](#)]

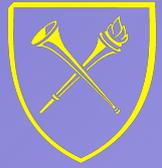




Lattice Spin Glasses (at $T=0$):

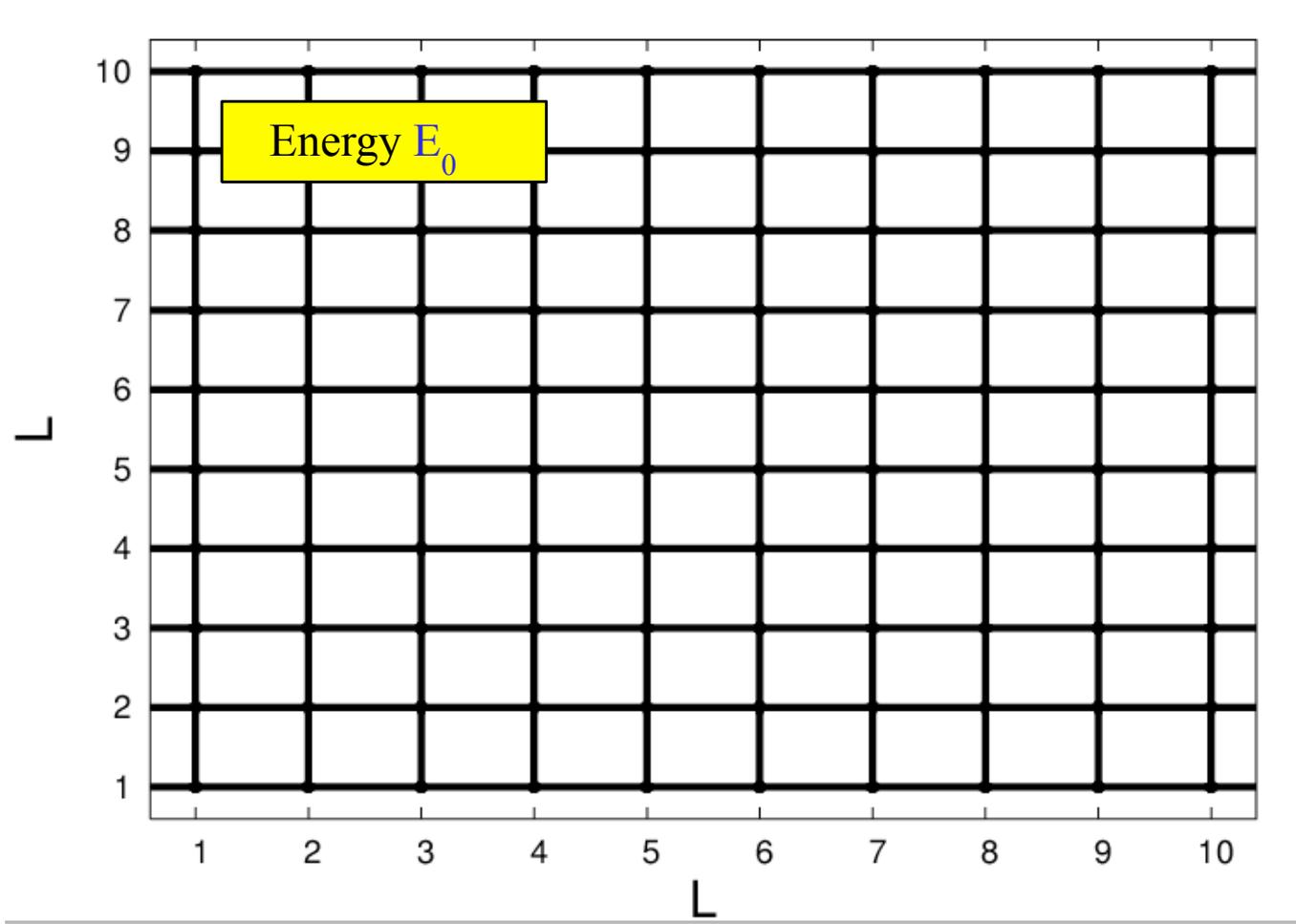
Defect-Energy:





Lattice Spin Glasses (at $T=0$):

Defect-Energy:

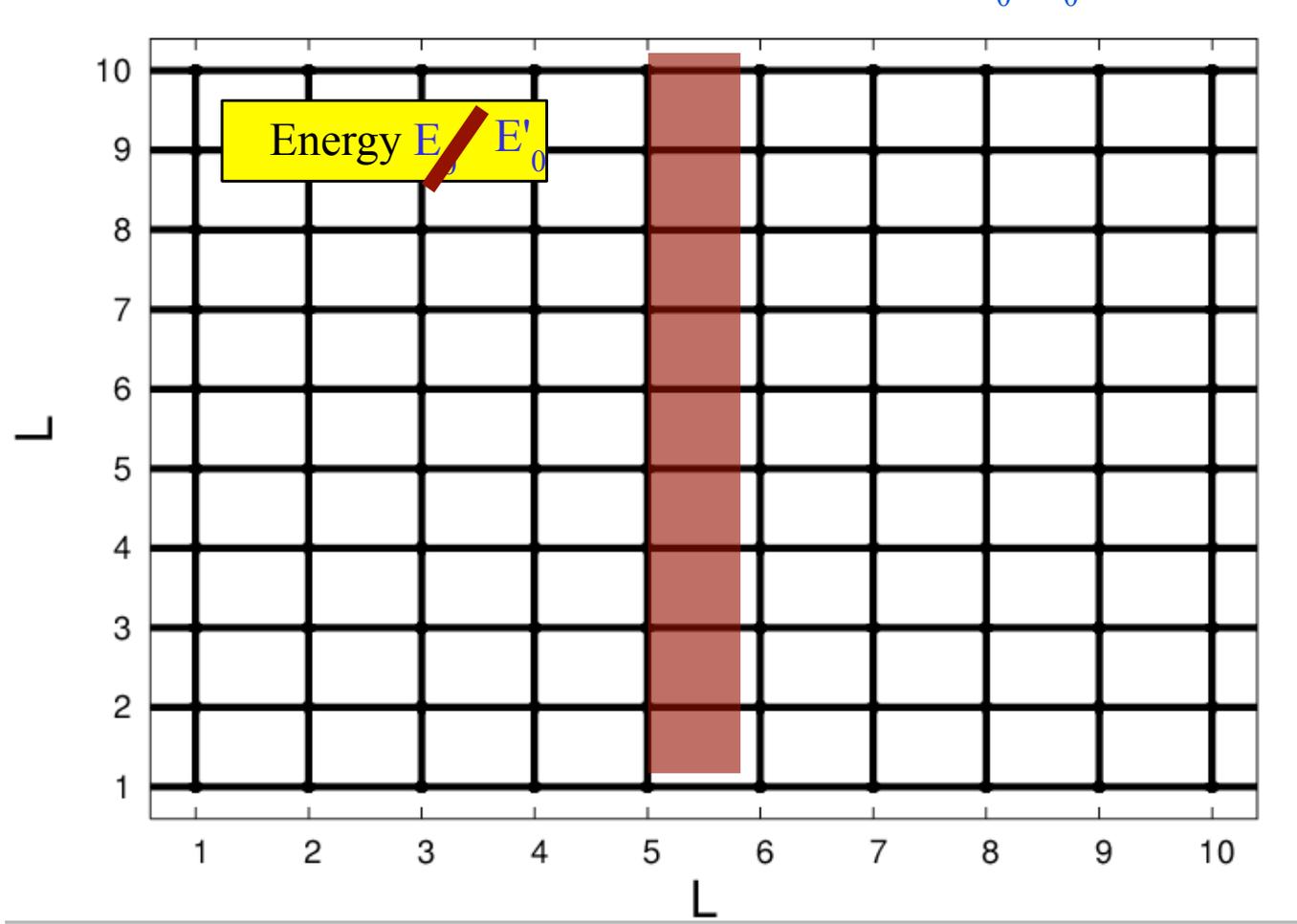




Lattice Spin Glasses (at $T=0$):

Defect-Energy:

Measure Defect Energy $\Delta E = E_0 - E'_0$

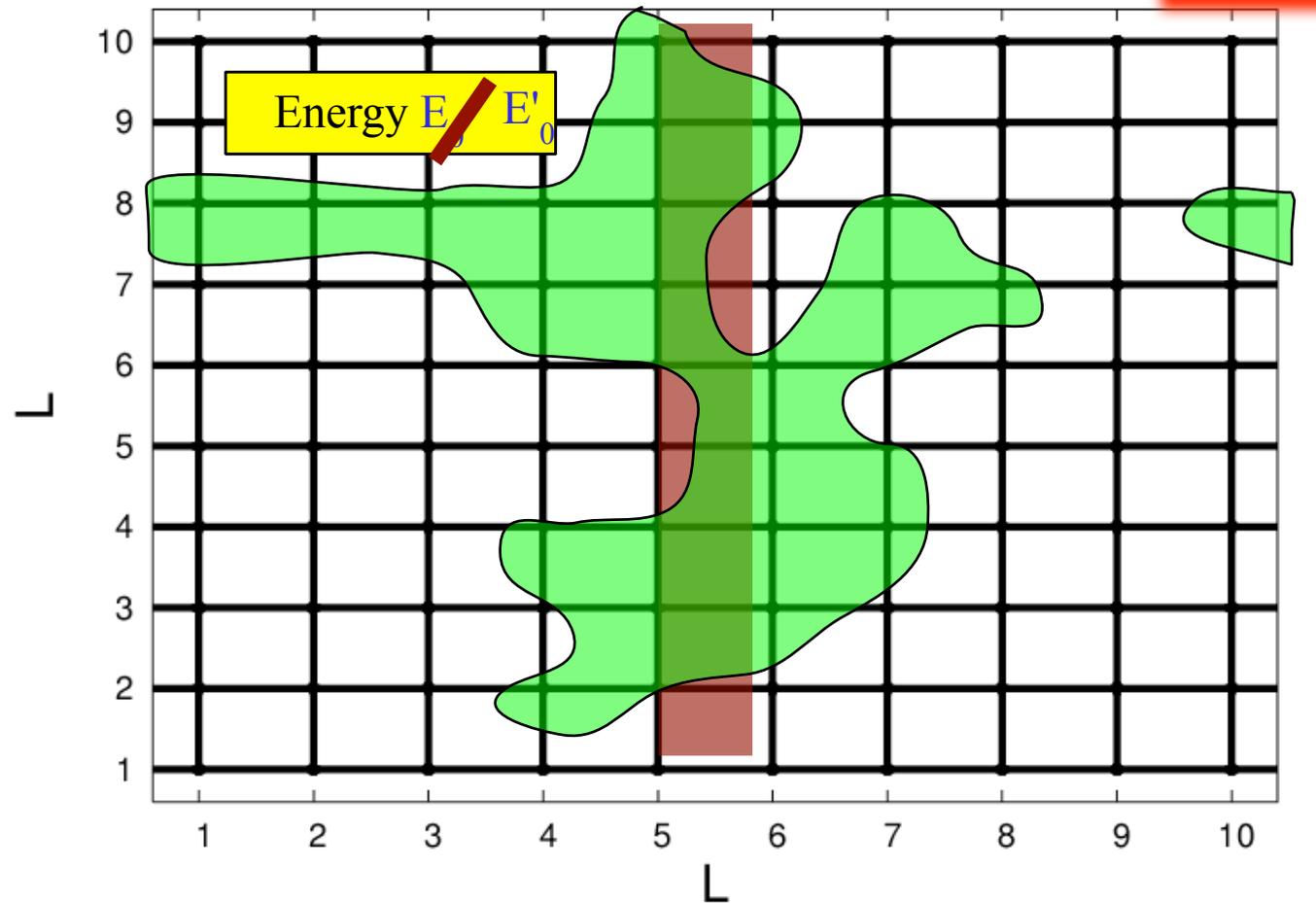


Lattice Spin Glasses (at $T=0$):

Defect-Energy:

Measure Defect Energy $\Delta E = E_0 - E'_0$

$$\Rightarrow \sigma(\Delta E) \sim L^y$$



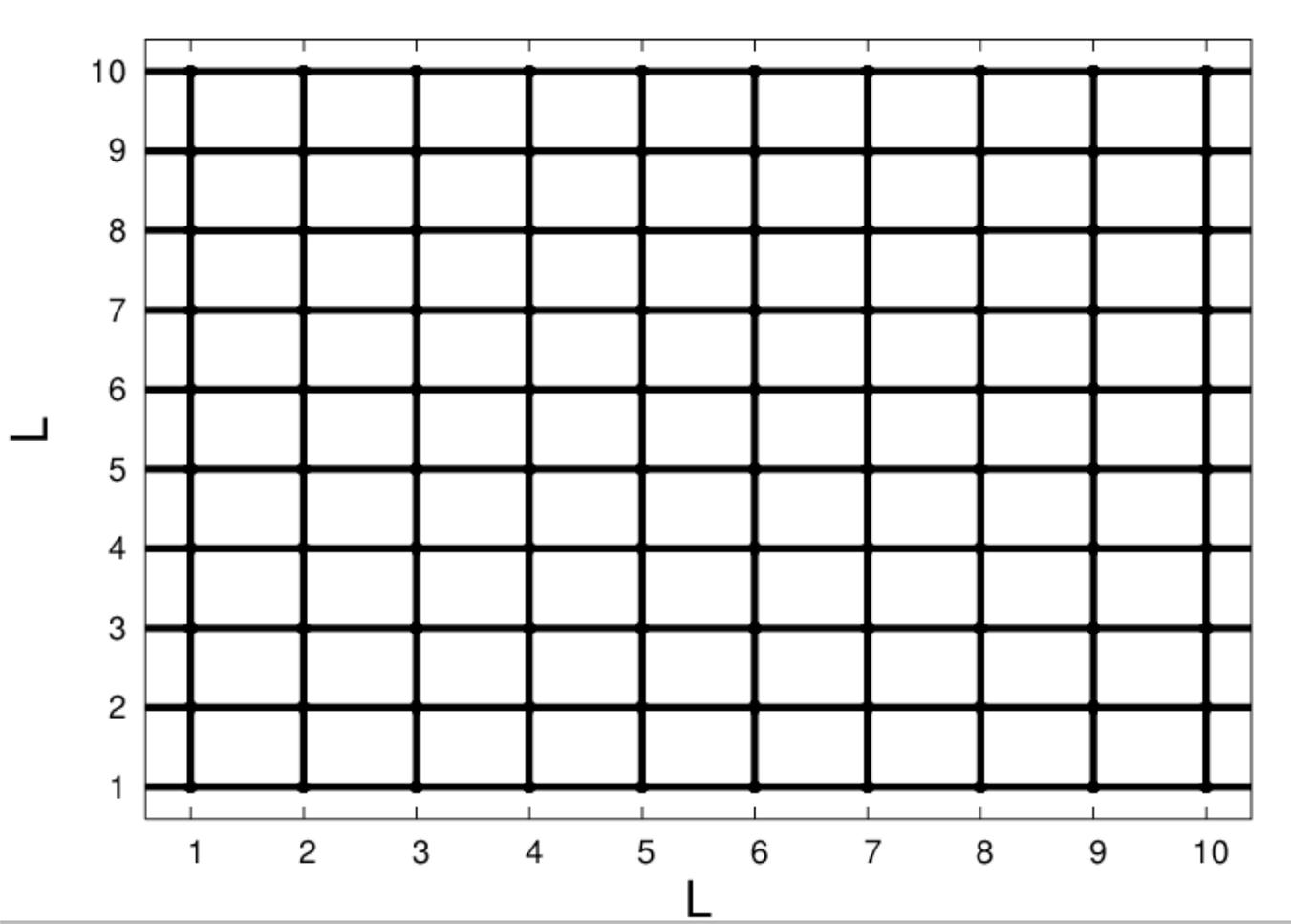
\Rightarrow Low Energy Excitations (like “small oscillations”)





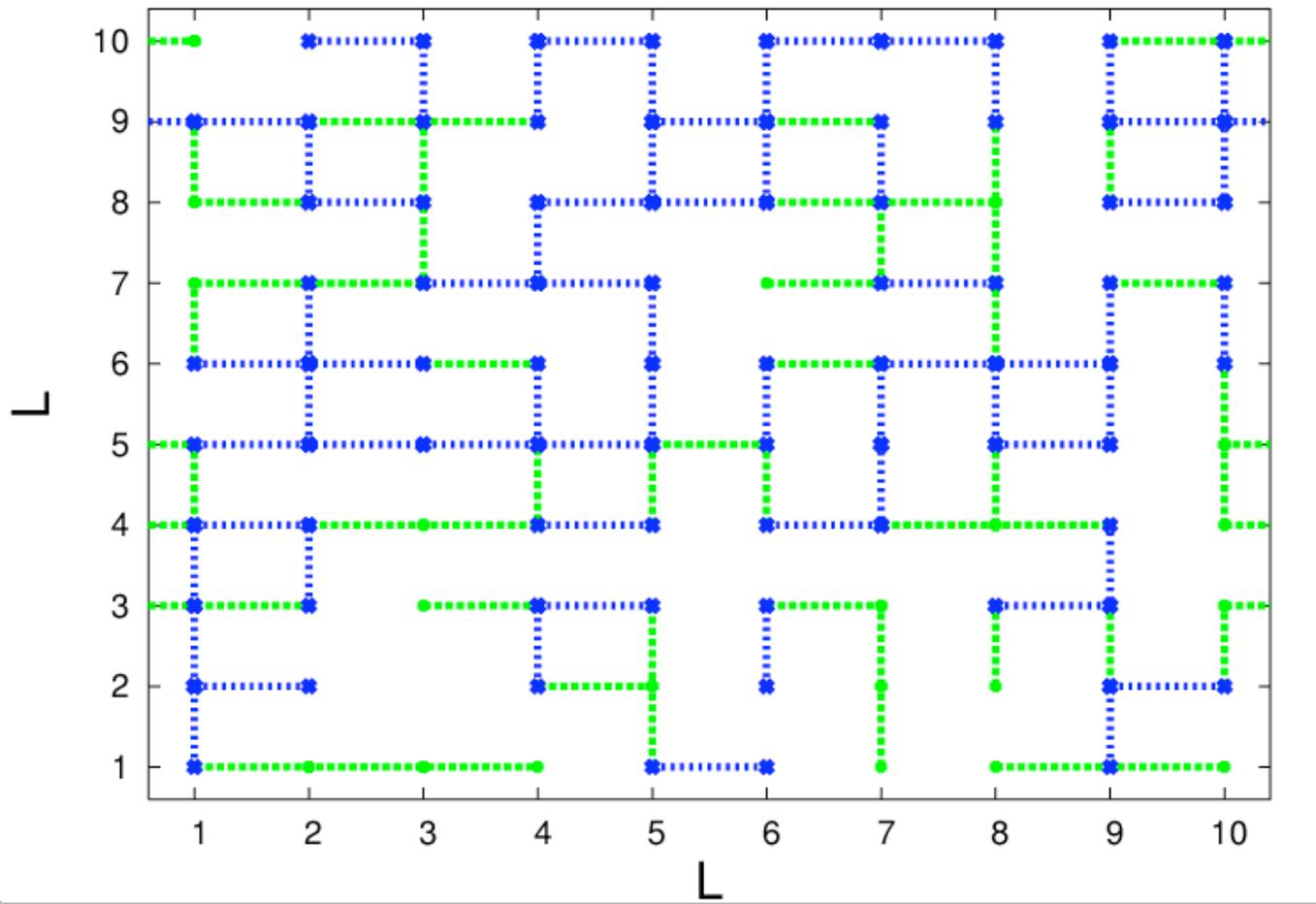
Lattice Spin Glasses (at $T=0$):

Defect-Energy:



Lattice Spin Glasses (at $T=0$):

Defect-Energy:



Before:
100 Spins

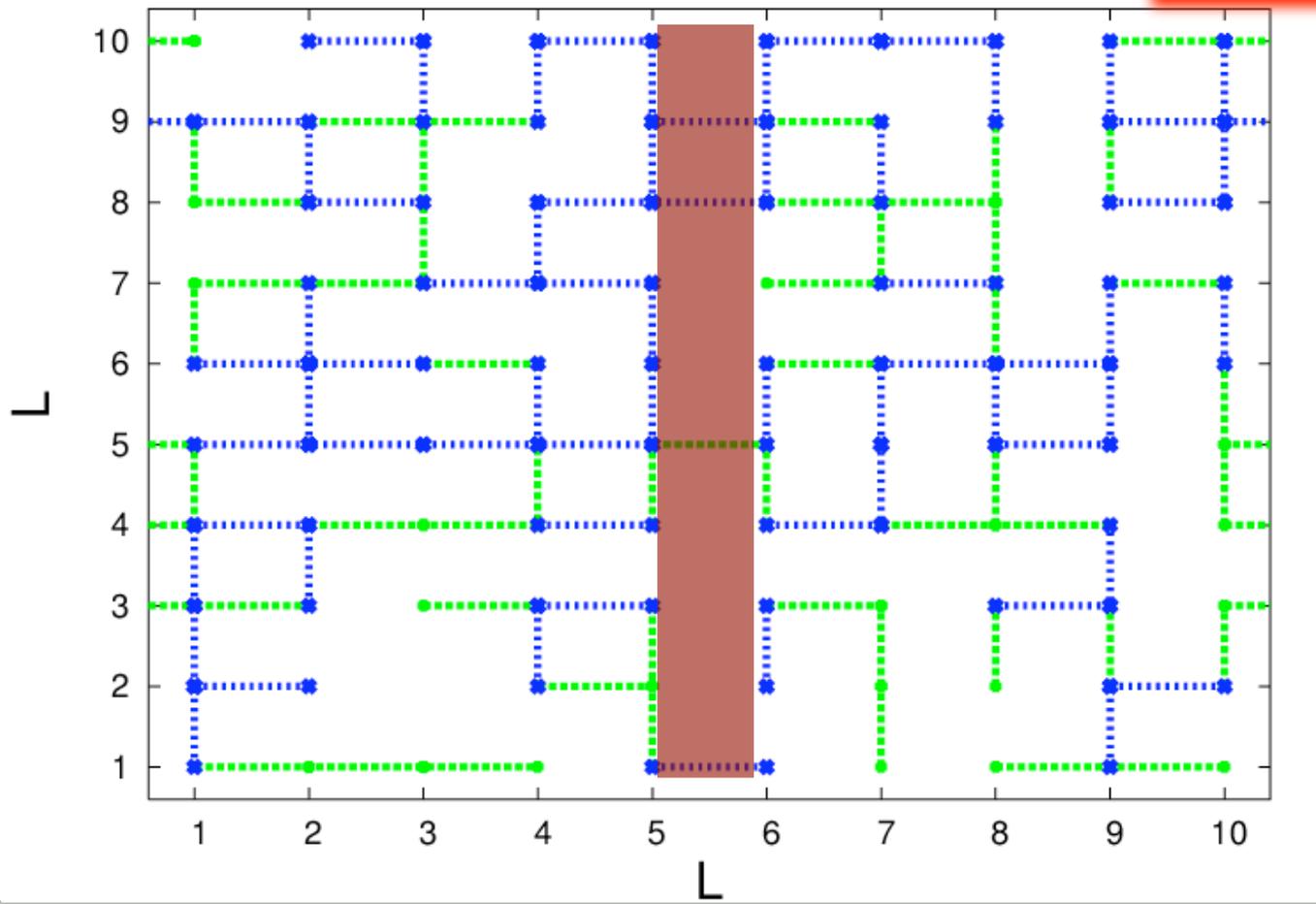


Lattice Spin Glasses (at $T=0$):

Defect-Energy:

Measure Defect Energy $\Delta E = E_0 - E'_0$

$$\Rightarrow \sigma(\Delta E) \sim L^y$$



Before:
100 Spins

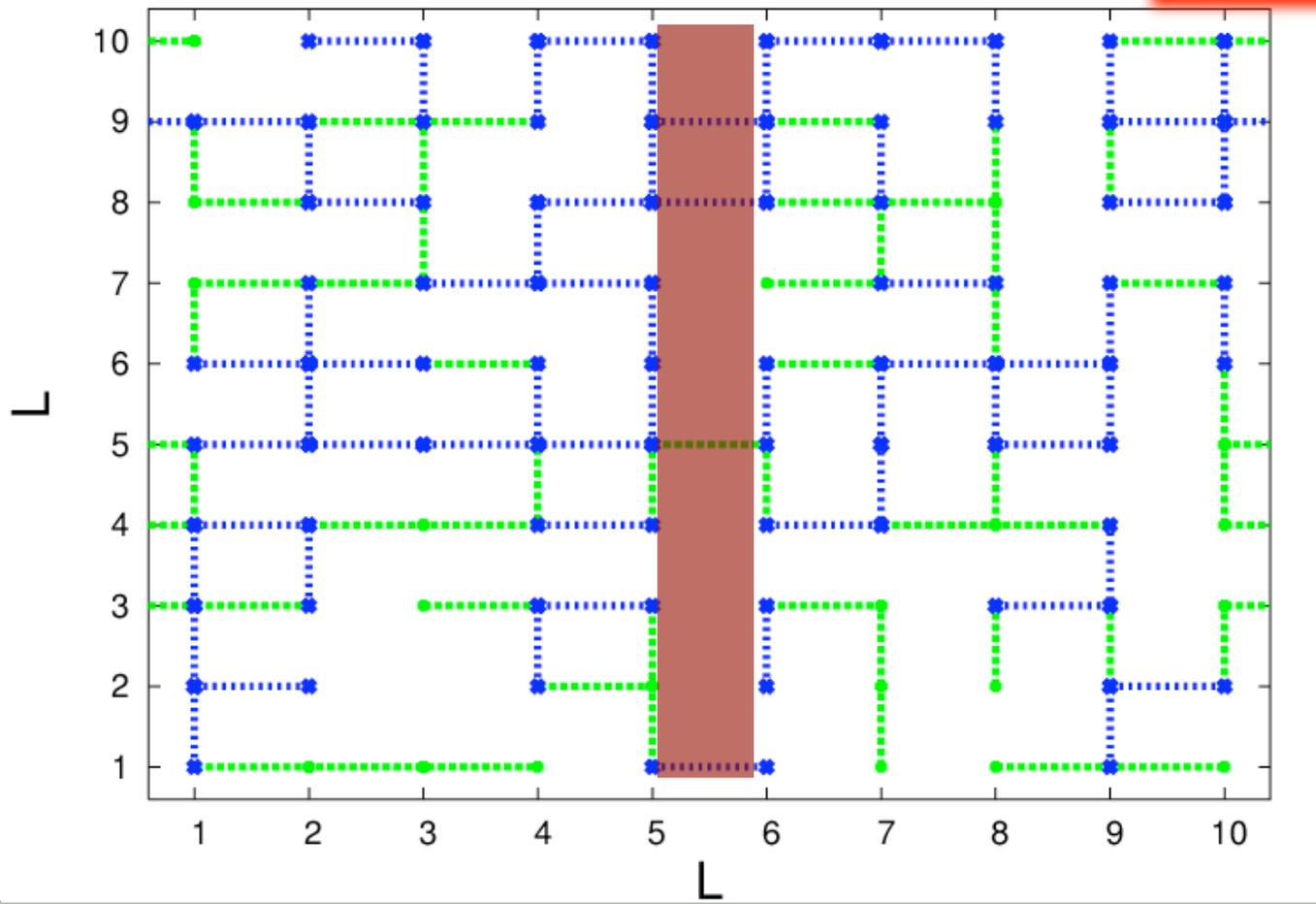


Lattice Spin Glasses (at $T=0$):

Defect-Energy:

Measure Defect Energy $\Delta E = E_0 - E'_0$

$$\Rightarrow \sigma(\Delta E) \sim L^y$$



Before:
100 Spins

\Rightarrow Low Energy Excitations of **bond-diluted Lattices**

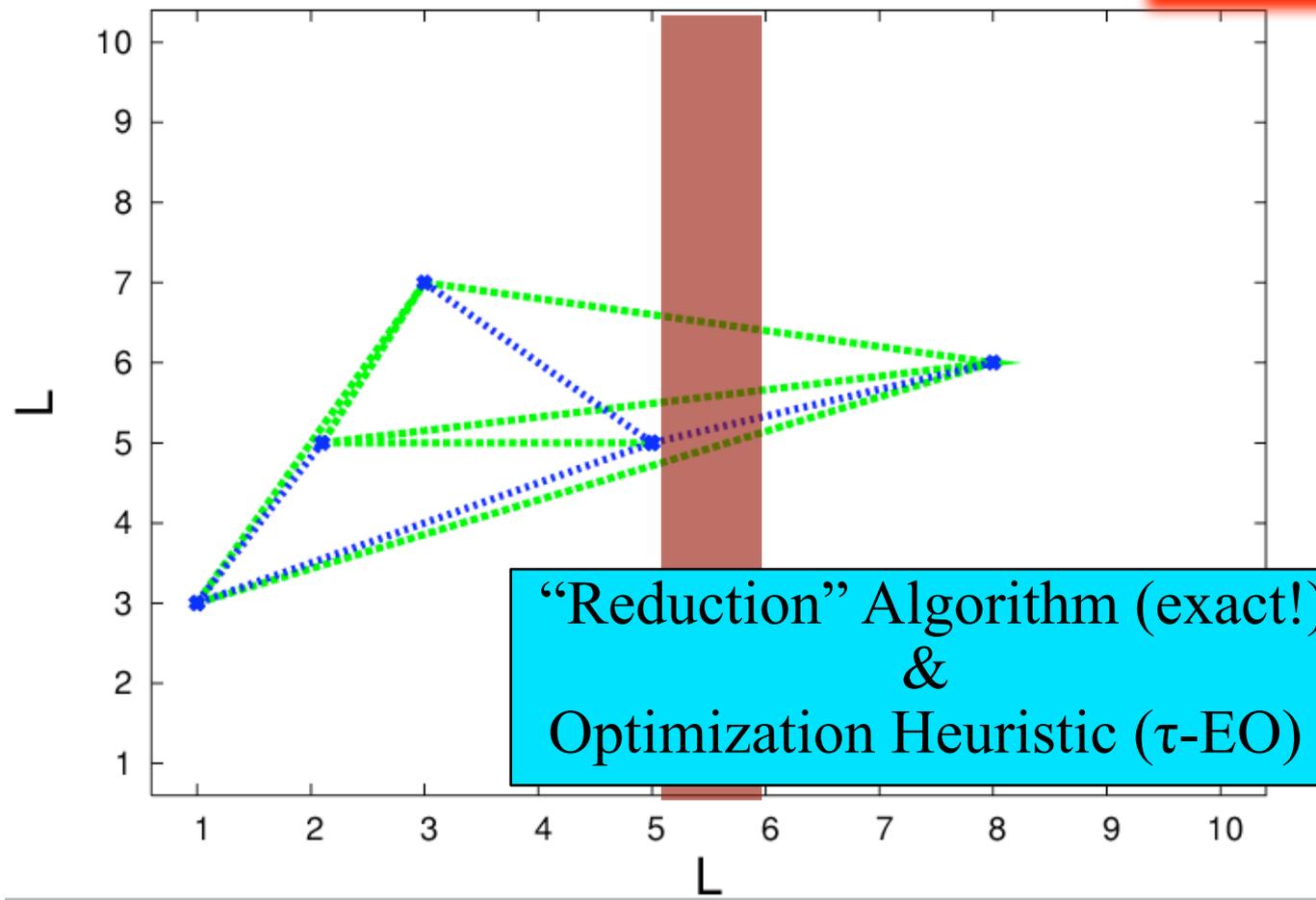


Lattice Spin Glasses (at $T=0$):

Defect-Energy:

Measure Defect Energy $\Delta E = E_0 - E'_0$

$$\Rightarrow \sigma(\Delta E) \sim L^y$$



Before:
100 Spins

After:
5 Spins

\Rightarrow Low Energy Excitations of **bond-diluted Lattices**





Lattice Spin Glasses (at $T=0$):

Defect-Energy: Measure “Stiffness”: $\sigma(\Delta E) \sim L^y$





Lattice Spin Glasses (at $T=0$):

Defect-Energy: Measure “Stiffness”: $\sigma(\Delta E) \sim L^y$

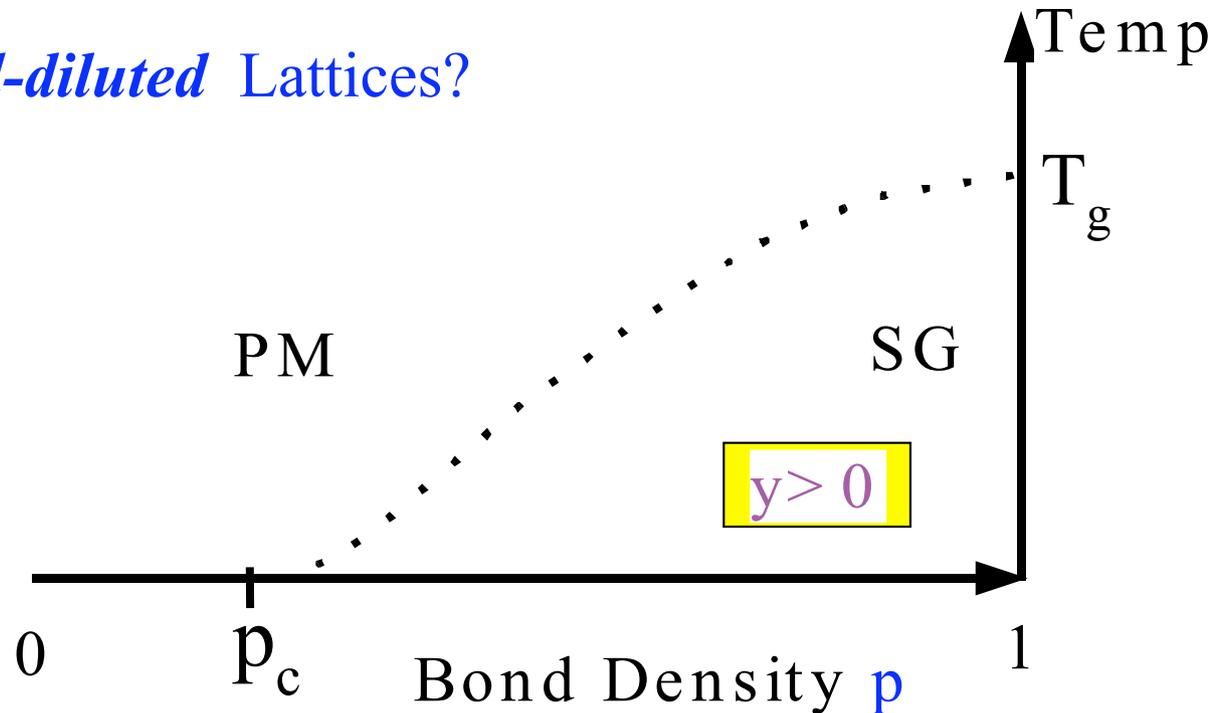
How *bond-diluted* Lattices?



Lattice Spin Glasses (at $T=0$):

Defect-Energy: Measure “Stiffness”: $\sigma(\Delta E) \sim L^y$

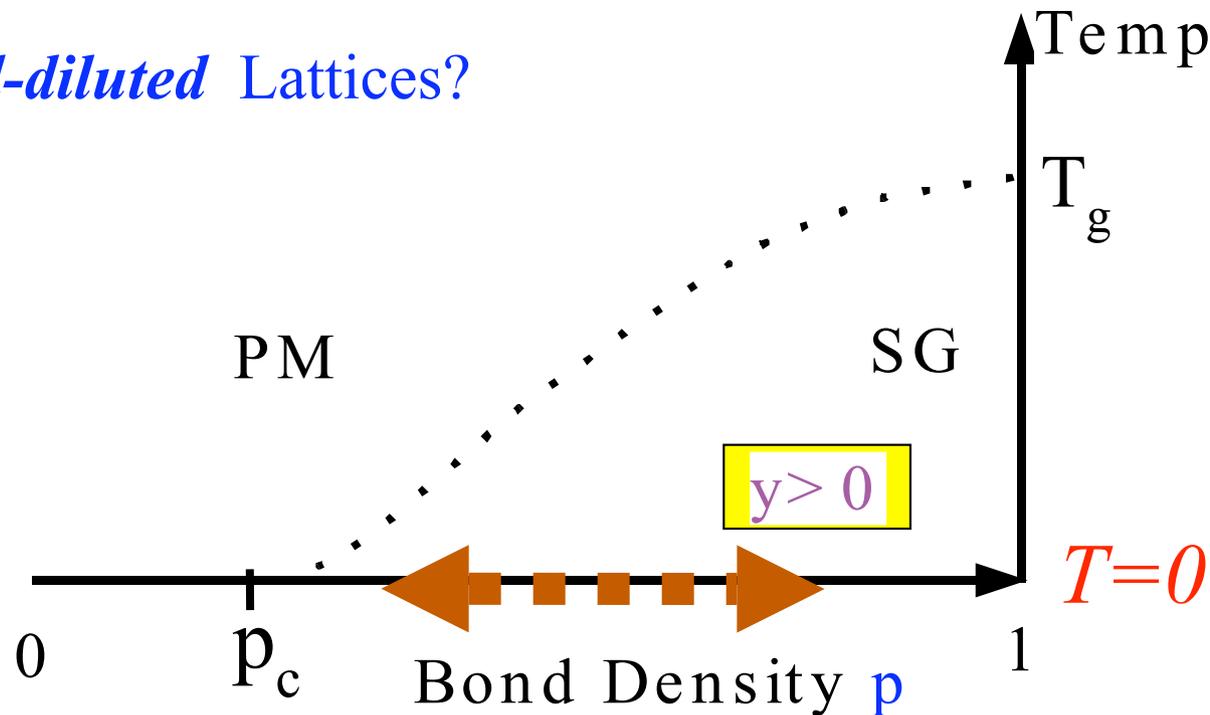
How *bond-diluted* Lattices?



Lattice Spin Glasses (at $T=0$):

Defect-Energy: Measure “Stiffness”: $\sigma(\Delta E) \sim L^y$

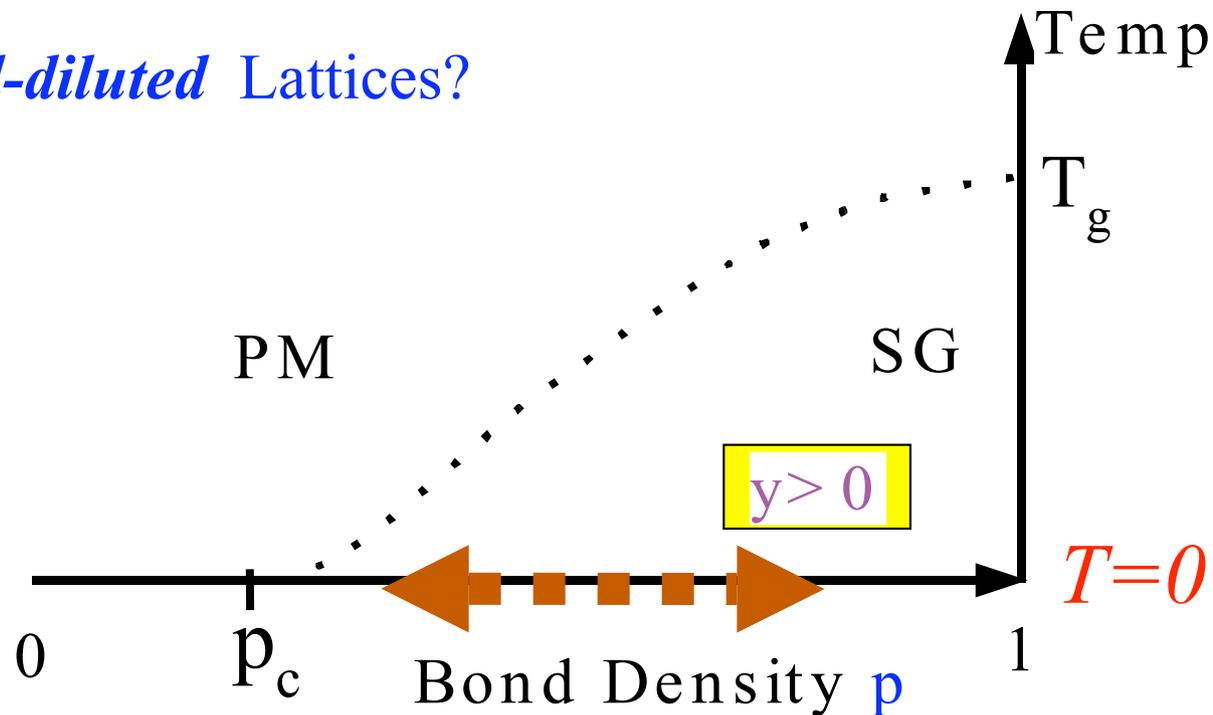
How *bond-diluted* Lattices?



Lattice Spin Glasses (at $T=0$):

Defect-Energy: Measure “Stiffness”: $\sigma(\Delta E) \sim L^y$

How bond-diluted Lattices?



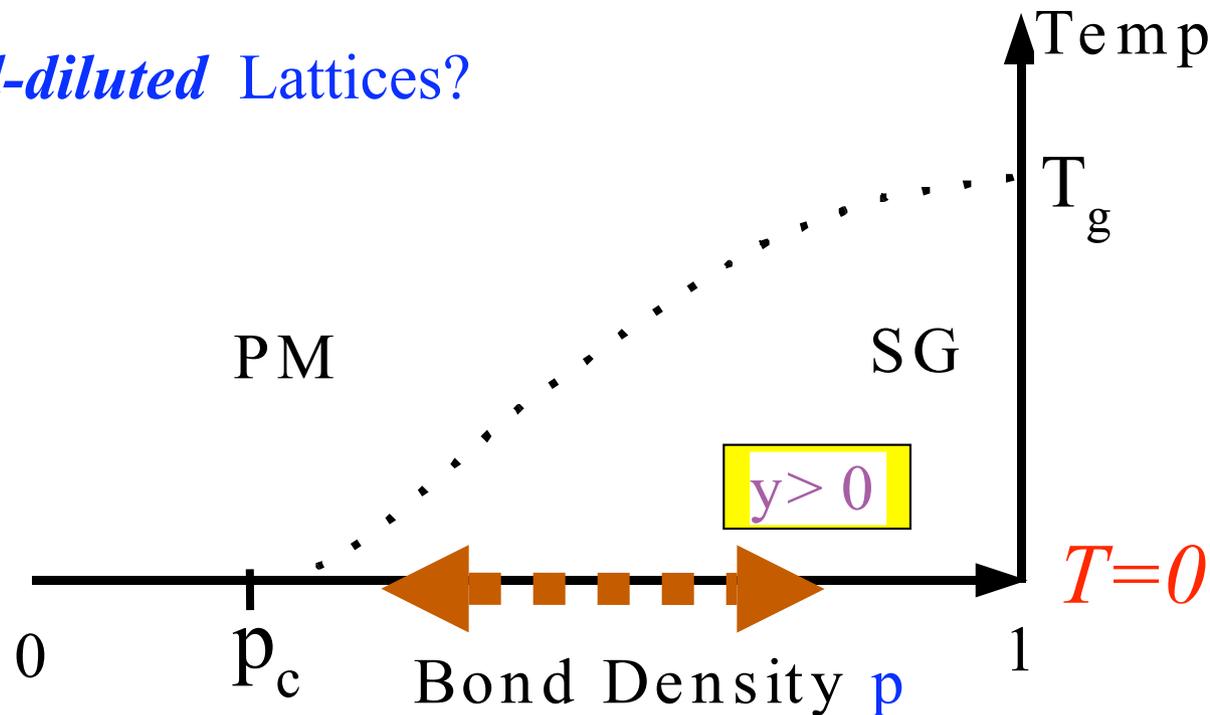
Why bond-diluted Lattices?



Lattice Spin Glasses (at $T=0$):

Defect-Energy: Measure “Stiffness”: $\sigma(\Delta E) \sim L^y$

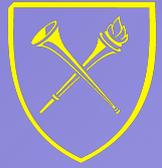
How bond-diluted Lattices?



Why bond-diluted Lattices?

- Simpler Problem
- Larger Sizes L
- Better Scaling





Defect-Energy of diluted Lattices:





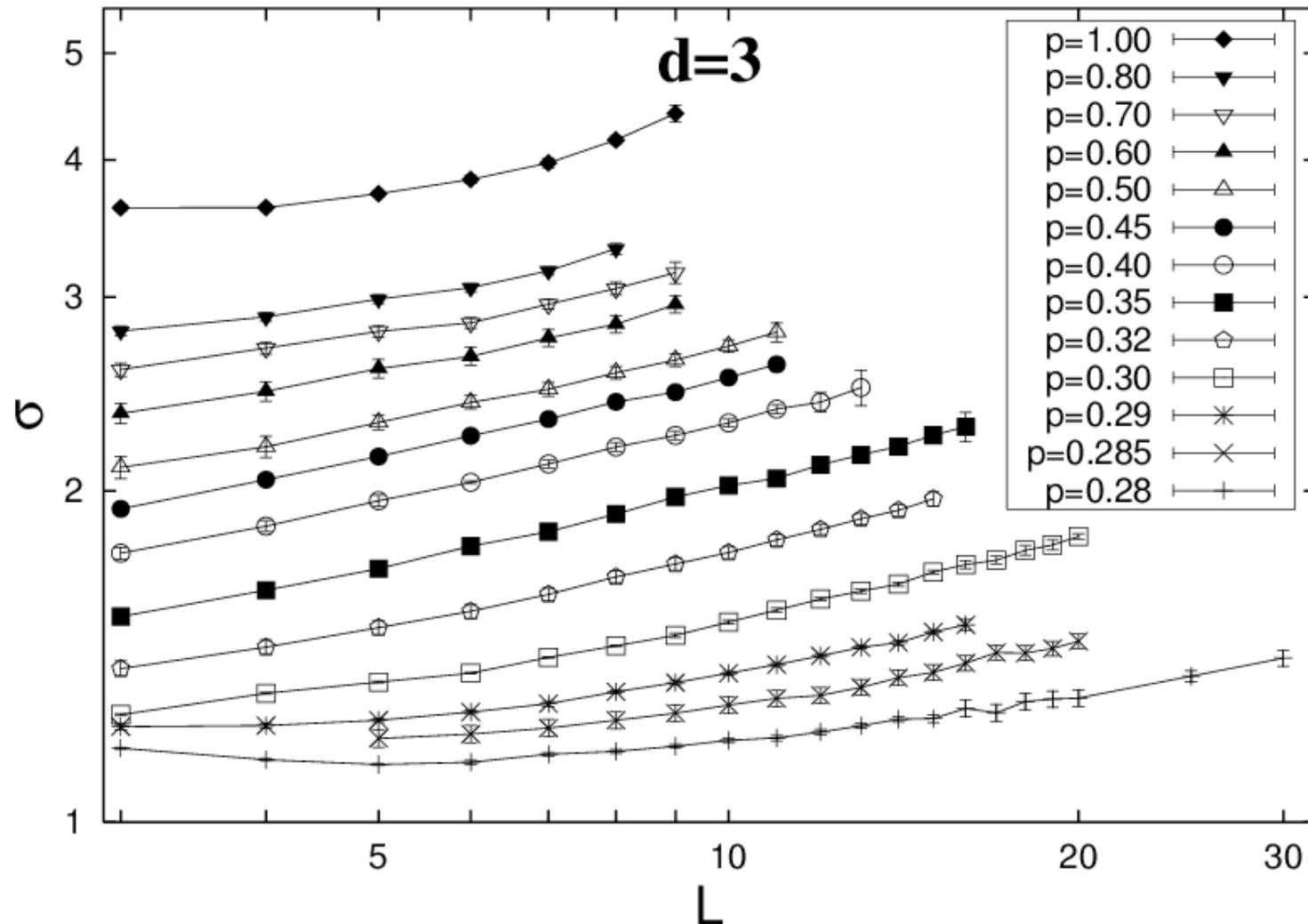
Defect-Energy of diluted Lattices:

- $\pm J$ -Glasses on Lattices of size L and density p .
- Defect-Energy $\sigma(\Delta E)$ with Reduction & Heuristic (τ -EO).



Defect-Energy of diluted Lattices:

- $\pm J$ -Glasses on Lattices of size L and density p .
- Defect-Energy $\sigma(\Delta E)$ with Reduction & Heuristic (τ -EO).





Stiffness Exponent γ for Lattice Glasses:

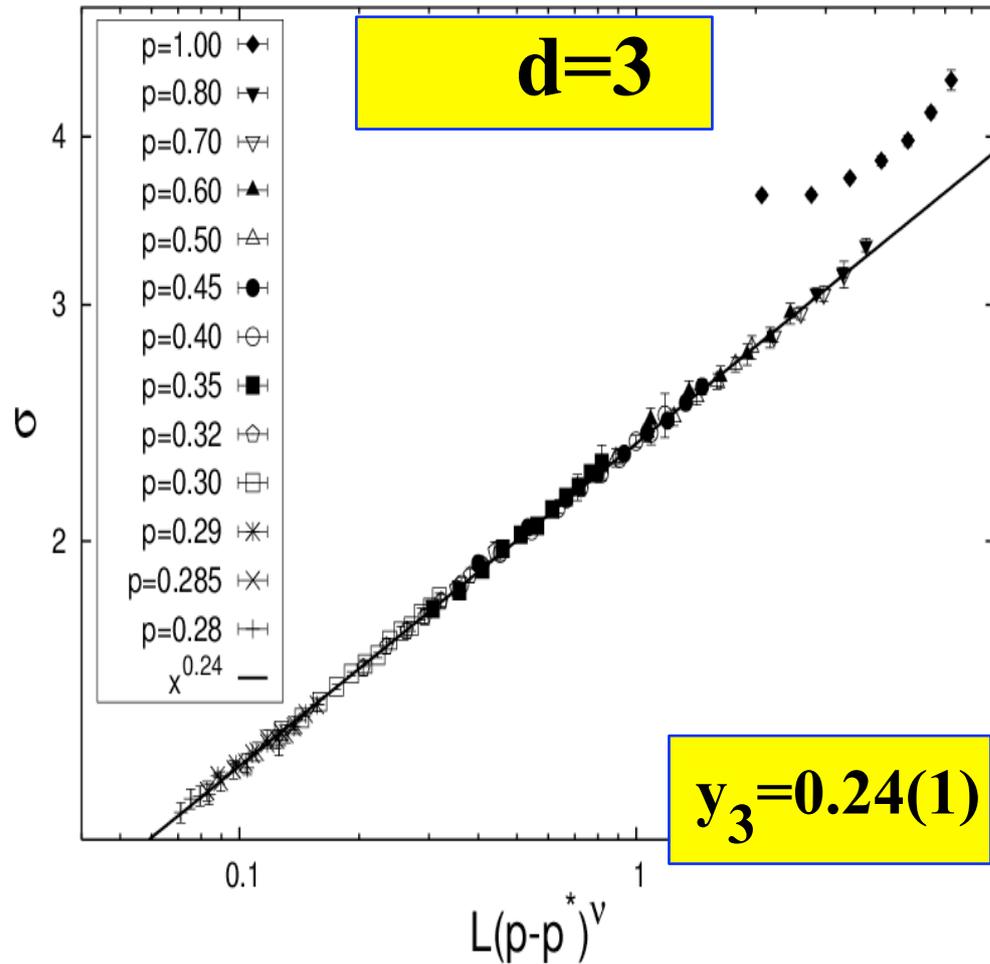
“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$





Stiffness Exponent γ for Lattice Glasses:

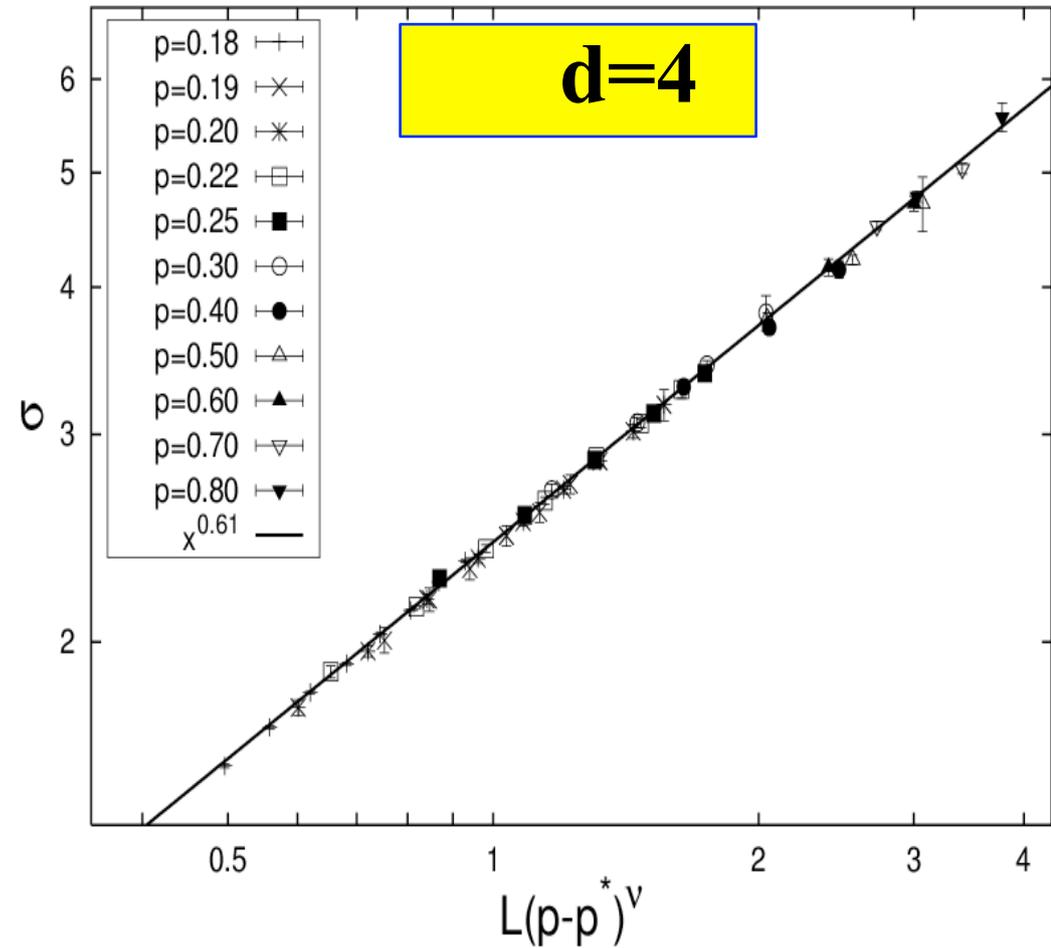
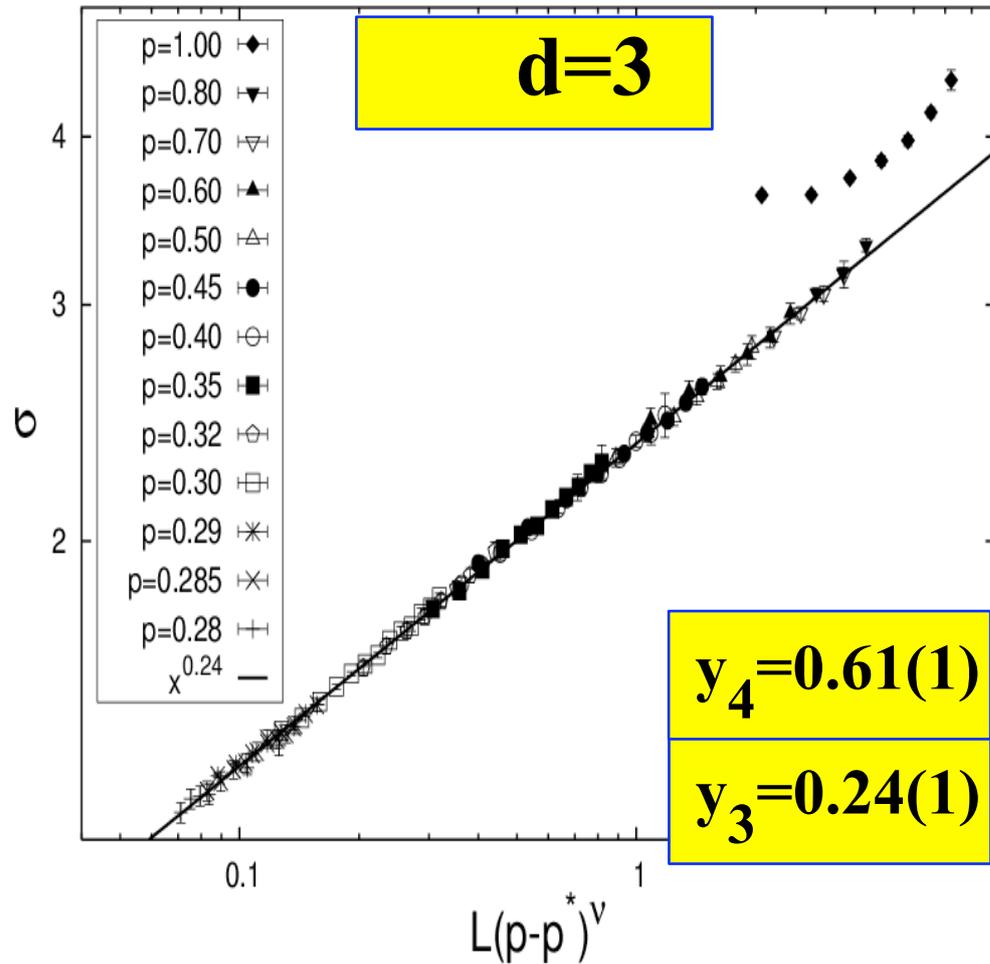
“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$





Stiffness Exponent γ for Lattice Glasses:

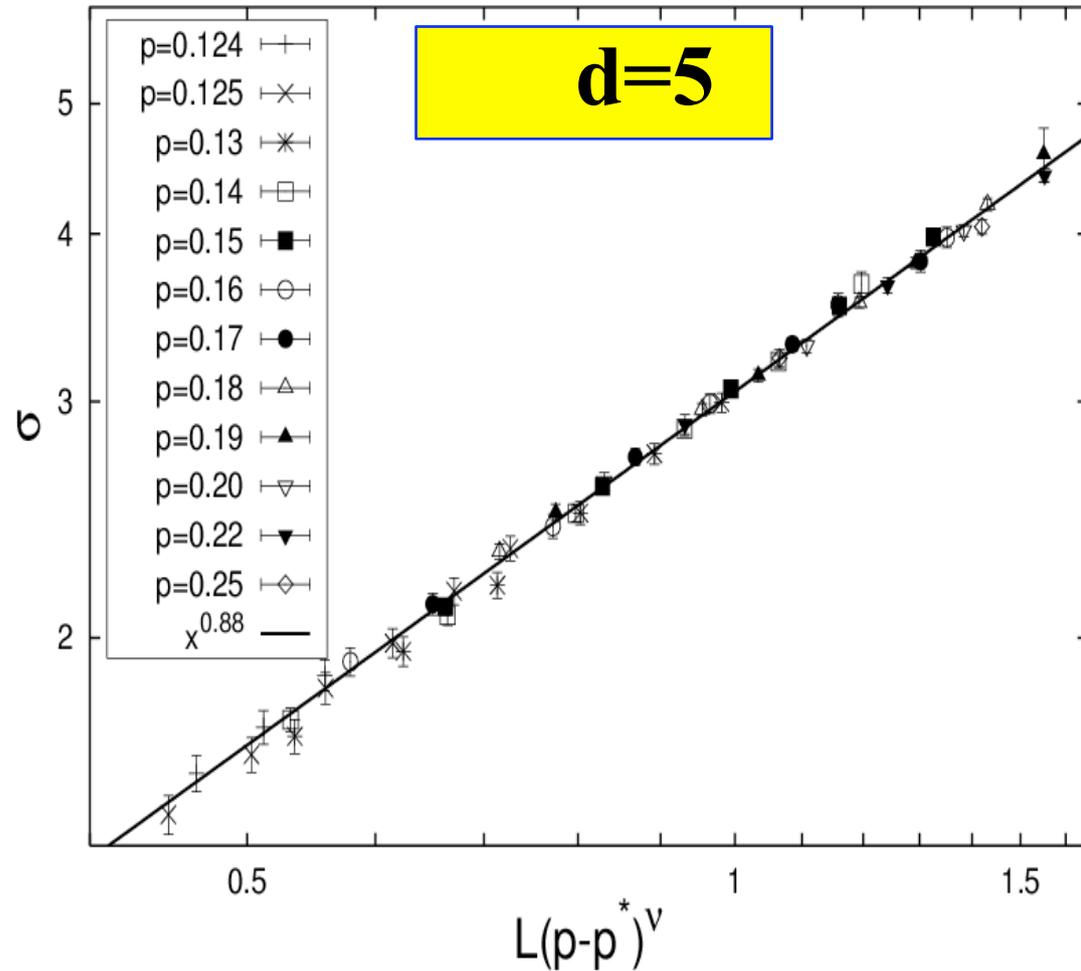
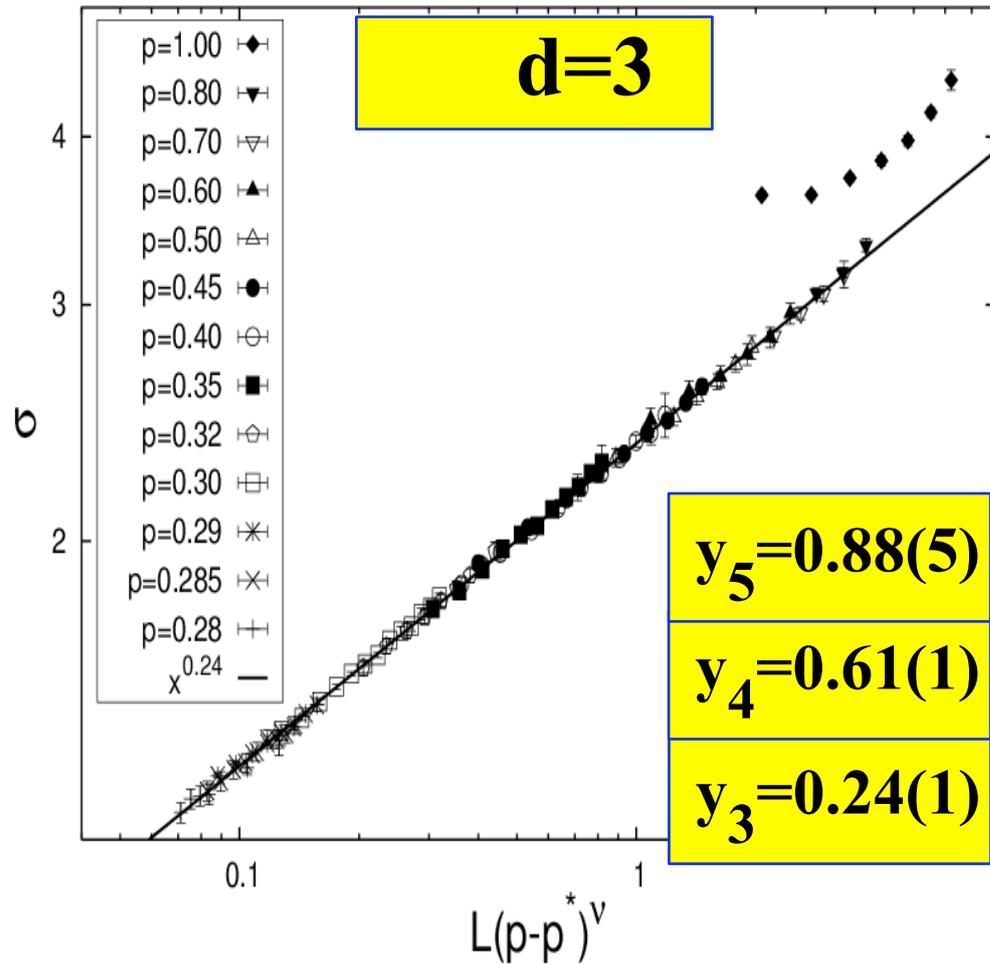
“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$

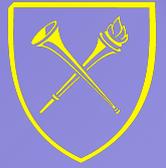




Stiffness Exponent γ for Lattice Glasses:

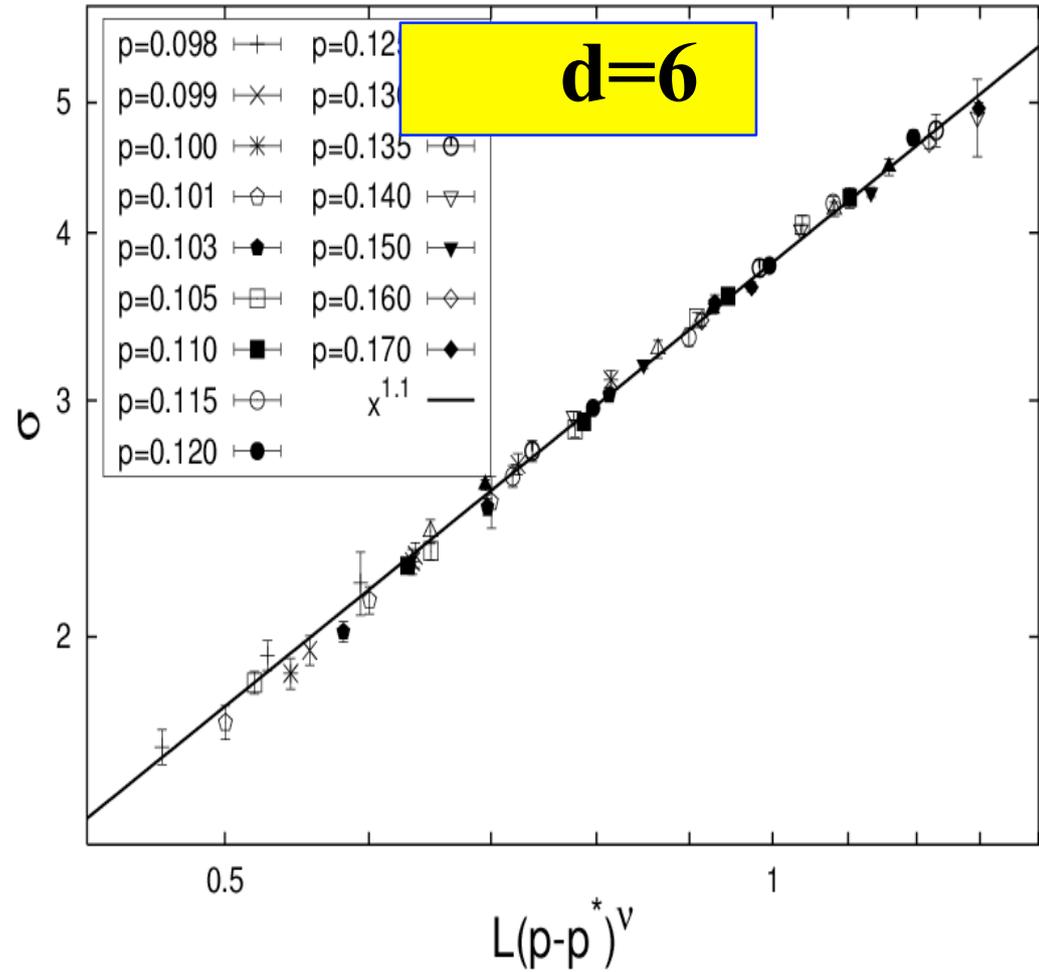
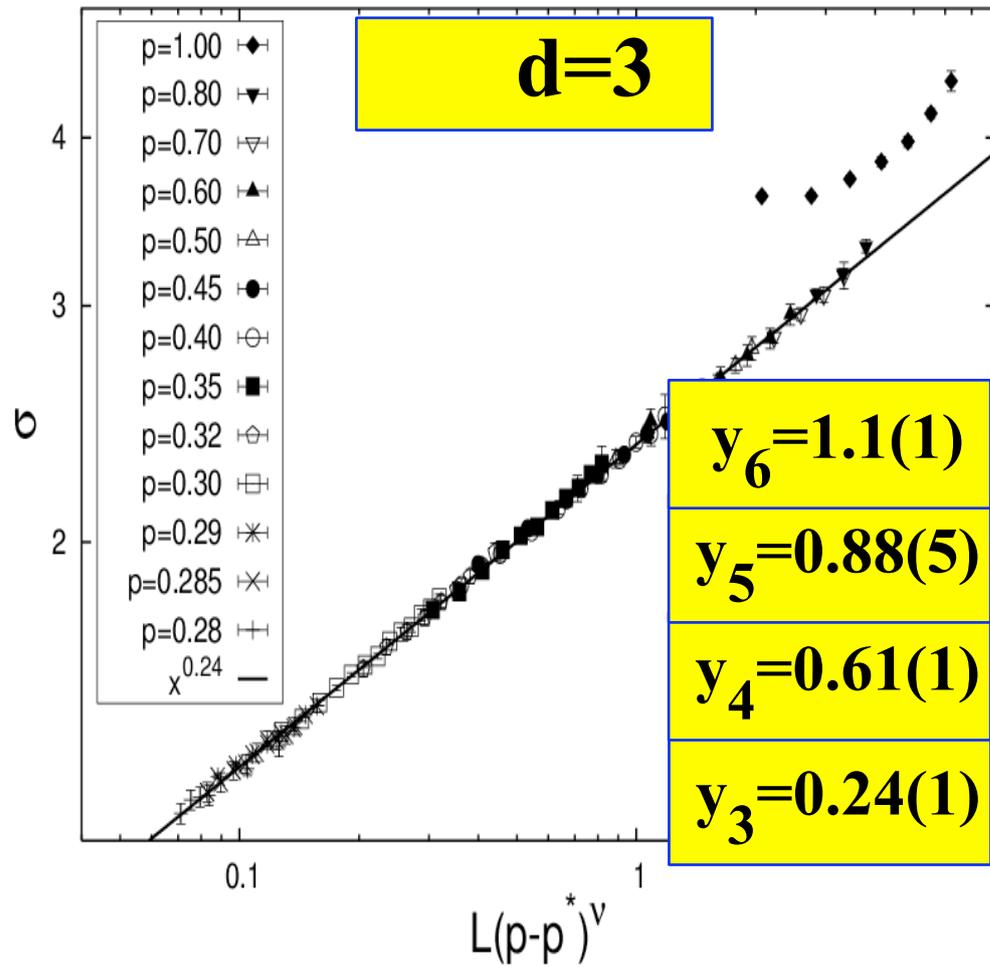
“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$

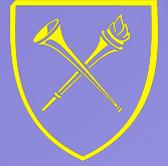




Stiffness Exponent γ for Lattice Glasses:

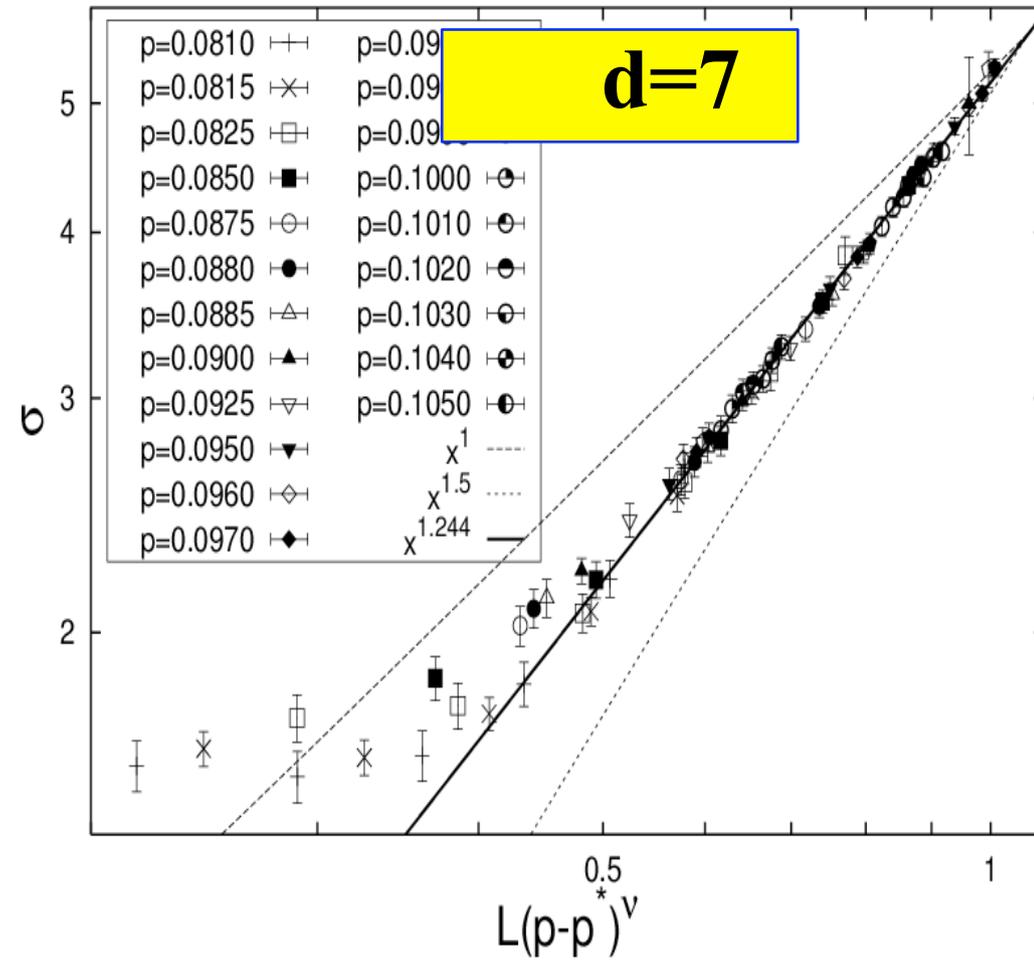
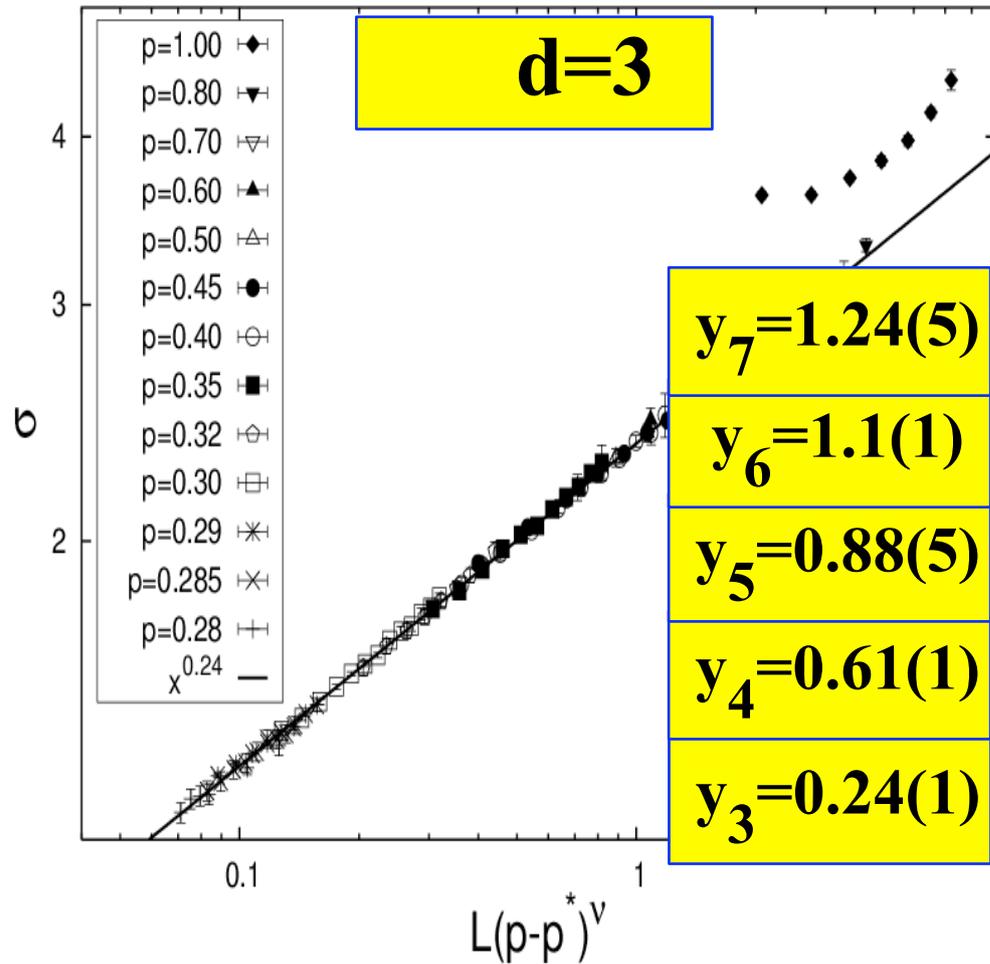
“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$

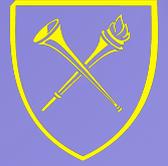




Stiffness Exponent γ for Lattice Glasses:

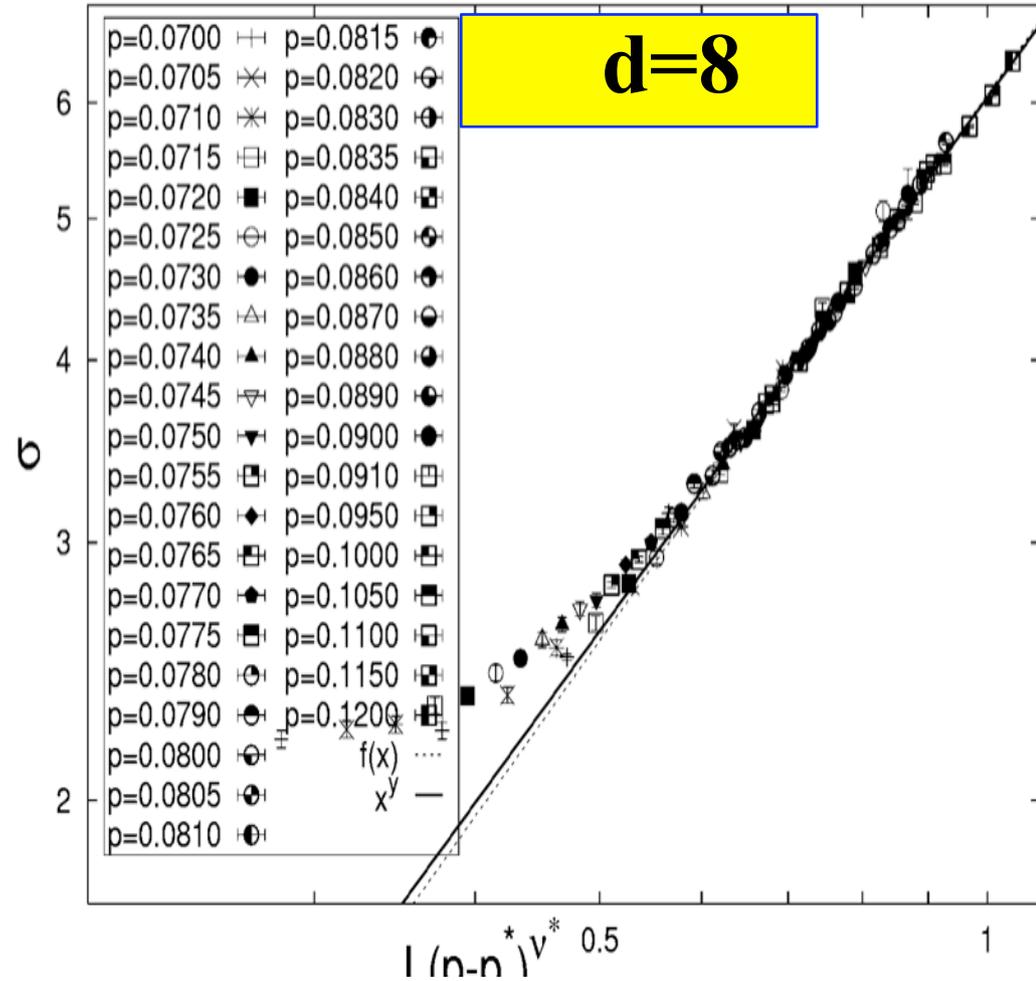
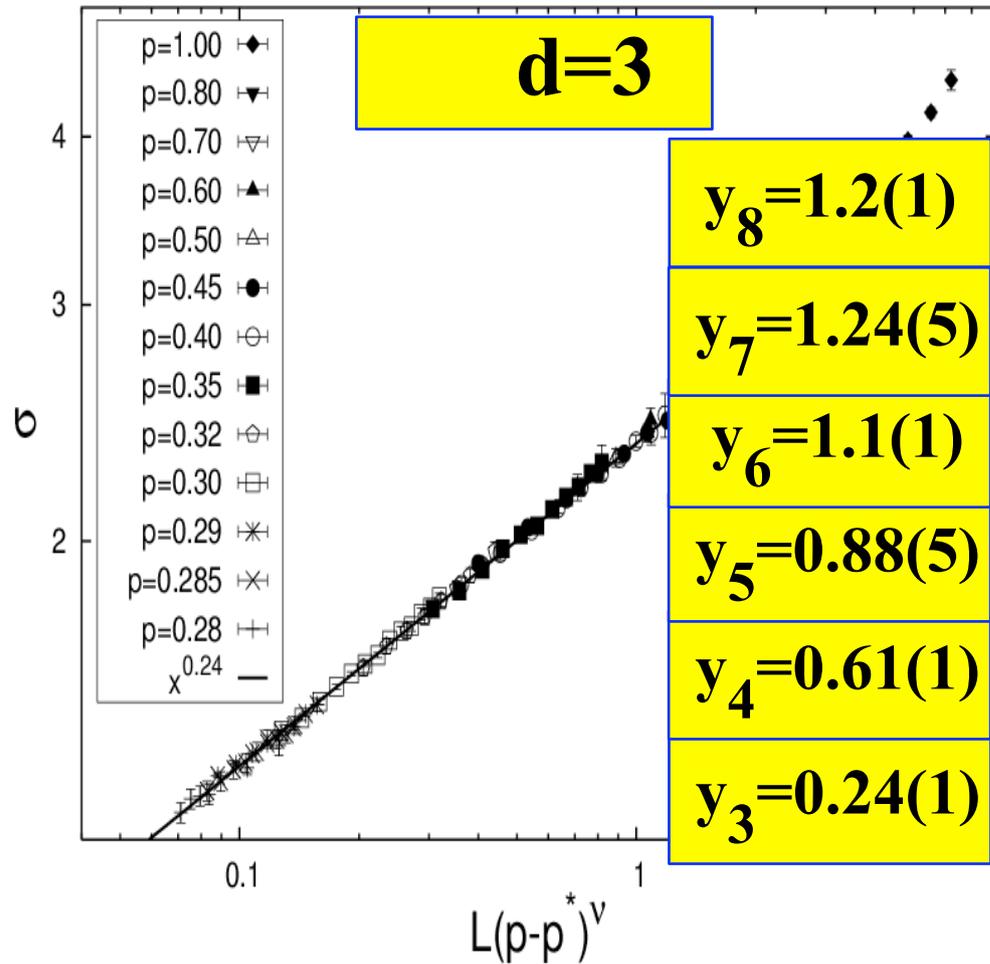
“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$





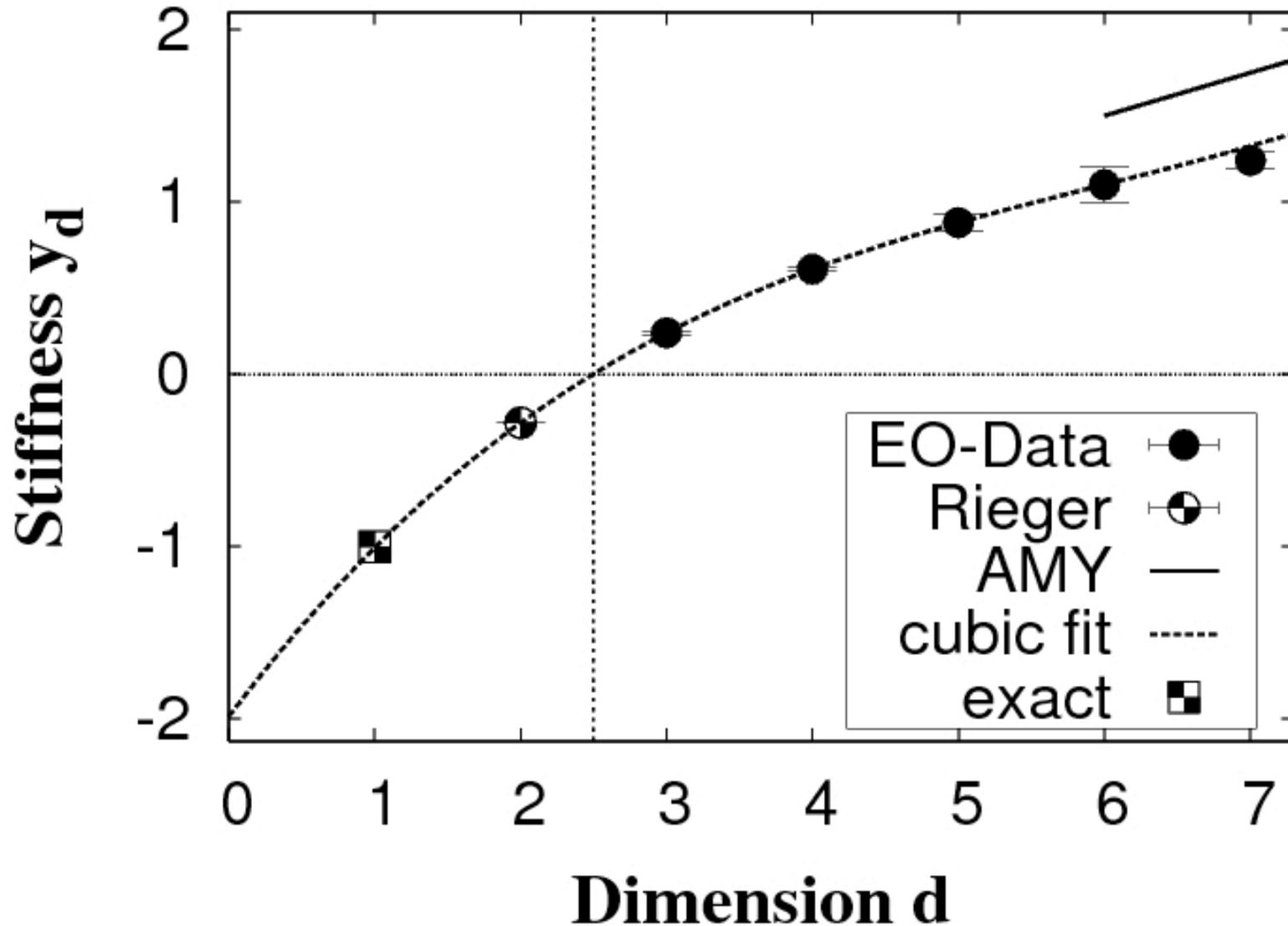
Stiffness Exponent γ for Lattice Glasses:

“Stiffness”: $\sigma(\Delta E) \sim L^\gamma$



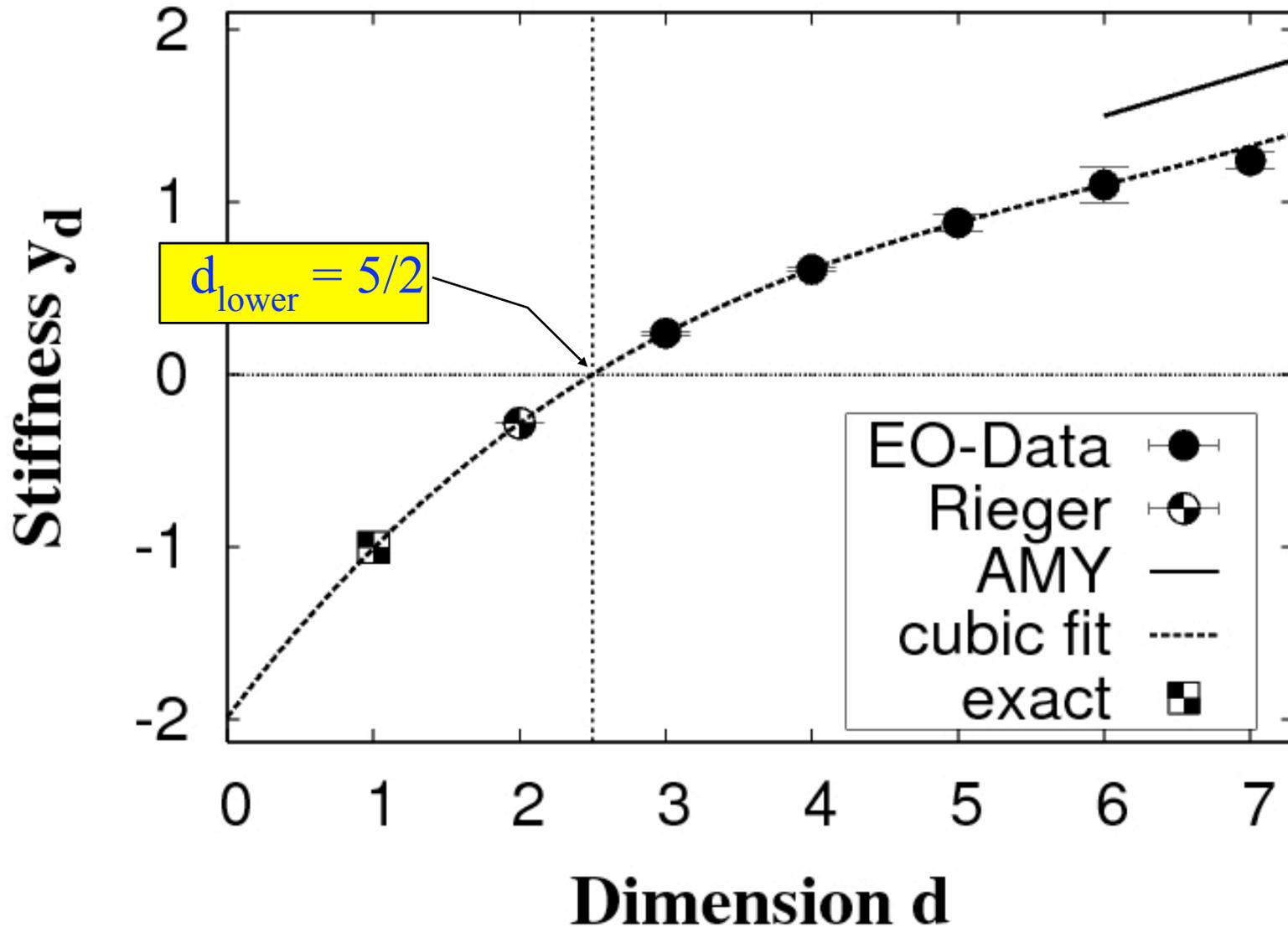
Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$



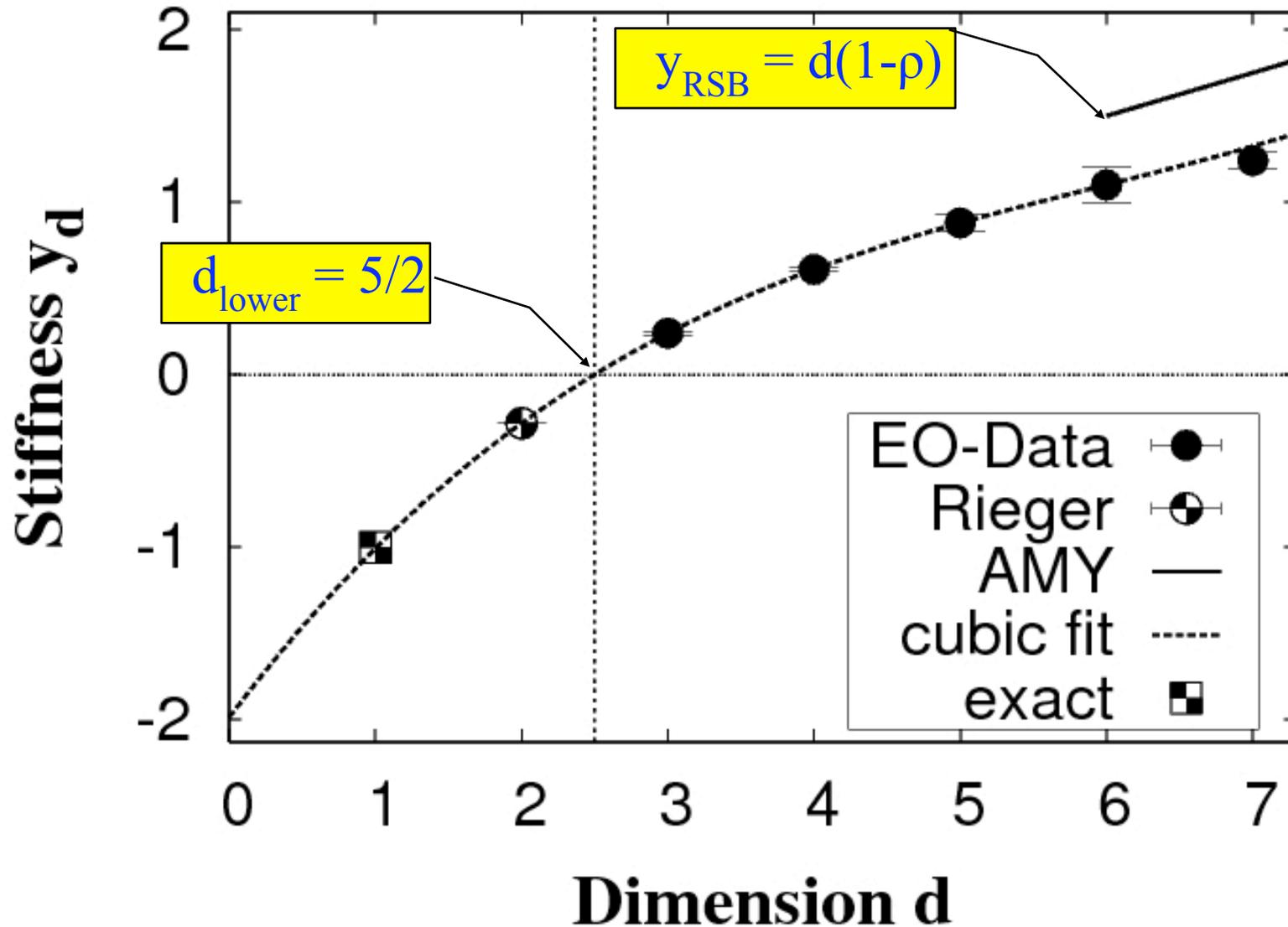
Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$



Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$





Other Evidence for $d_1=5/2$:





Other Evidence for $d_1=5/2$:

- From Theory: (Franz, Parisi & Virasoro, J. Phys. I **4**, 1657, '94)
Effective Mean Field calculation near T_g , where Replica Symmetry Breaking (RSB) disappears (ie. $T_g \rightarrow 0$) for $d_1=5/2$.





Other Evidence for $d_l=5/2$:

□ From Theory: (Franz, Parisi & Virasoro, J. Phys. I **4**, 1657, '94)
Effective Mean Field calculation near T_g , where Replica Symmetry Breaking (RSB) disappears (ie. $T_g \rightarrow 0$) for $d_l=5/2$.

□ From Numerics:

Know:

$$T_g \approx \sqrt{2d} \quad (d \rightarrow \infty)$$

$$T_g \approx \sqrt{2d - d_l} \quad (d \rightarrow d_l)$$



Other Evidence for $d_l=5/2$:

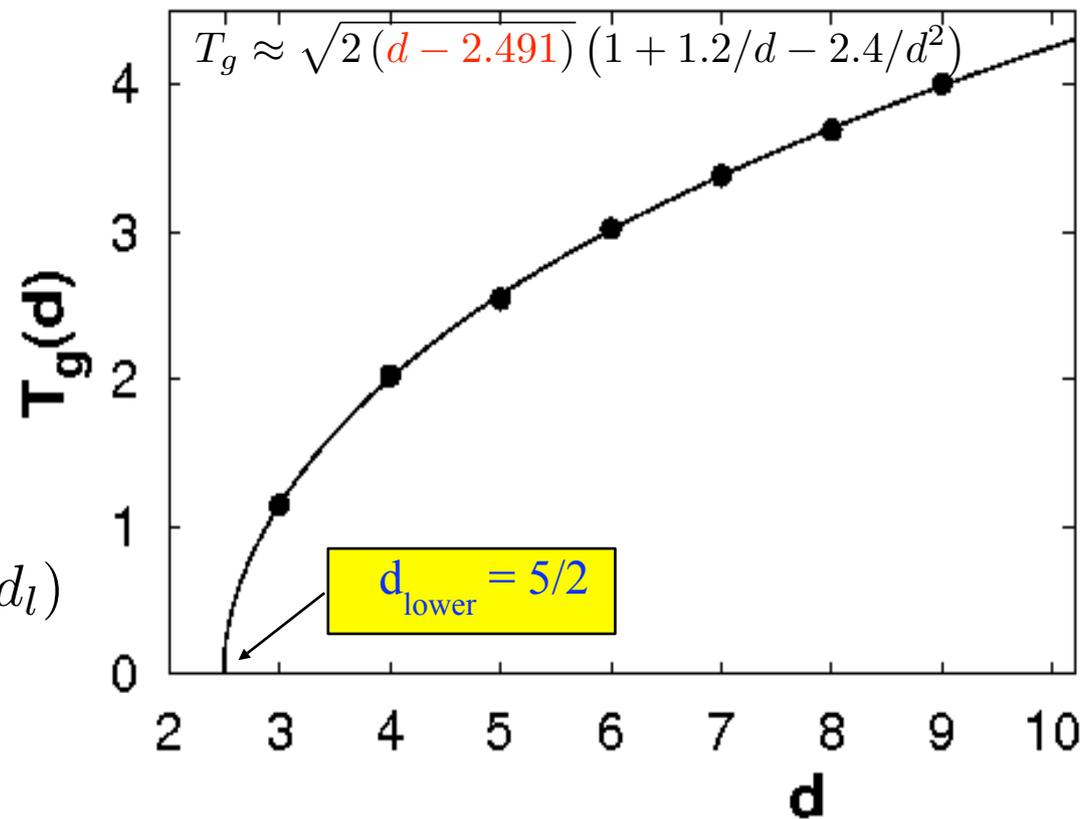
□ From Theory: (Franz, Parisi & Virasoro, J. Phys. I **4**, 1657, '94)
Effective Mean Field calculation near T_g , where Replica Symmetry Breaking (RSB) disappears (ie. $T_g \rightarrow 0$) for $d_l=5/2$.

□ From Numerics:

Know:

$$T_g \approx \sqrt{2d} \quad (d \rightarrow \infty)$$

$$T_g \approx \sqrt{2d - d_l} \quad (d \rightarrow d_l)$$



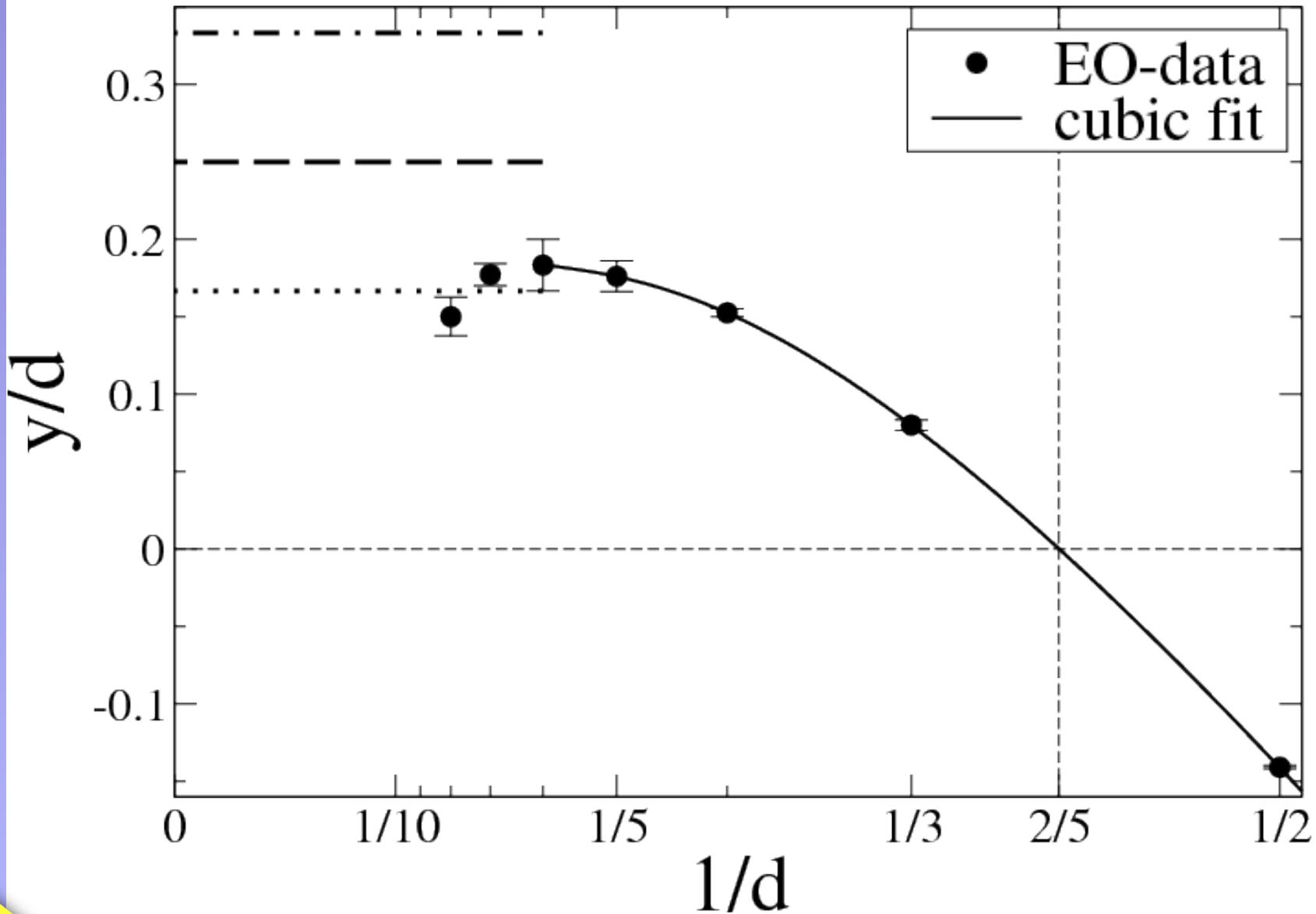
Data from:

- MC (Ballesteros et al) for $d=3,4$
- High-T Series (Klein et al) for $d \geq 5$



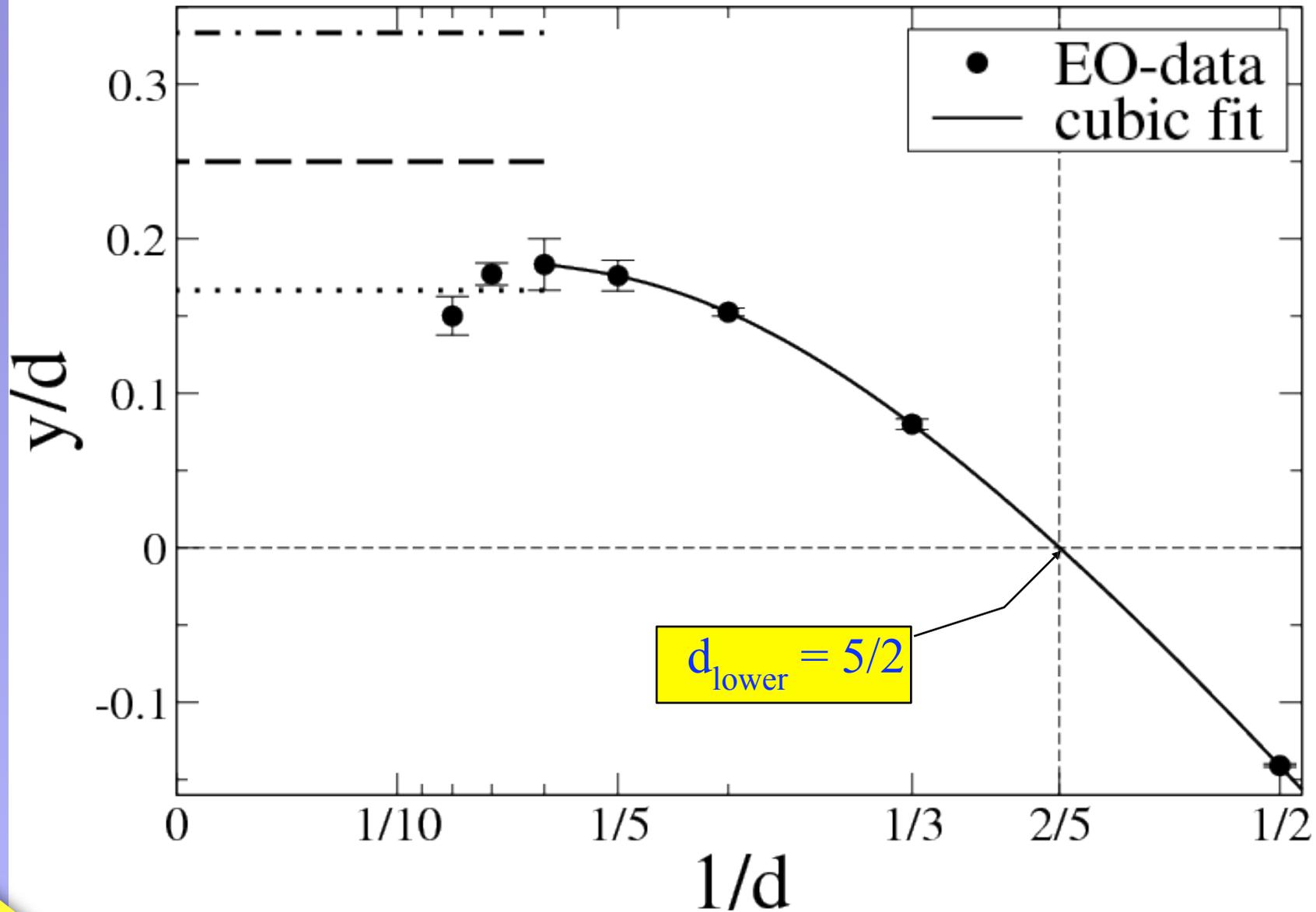
Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$



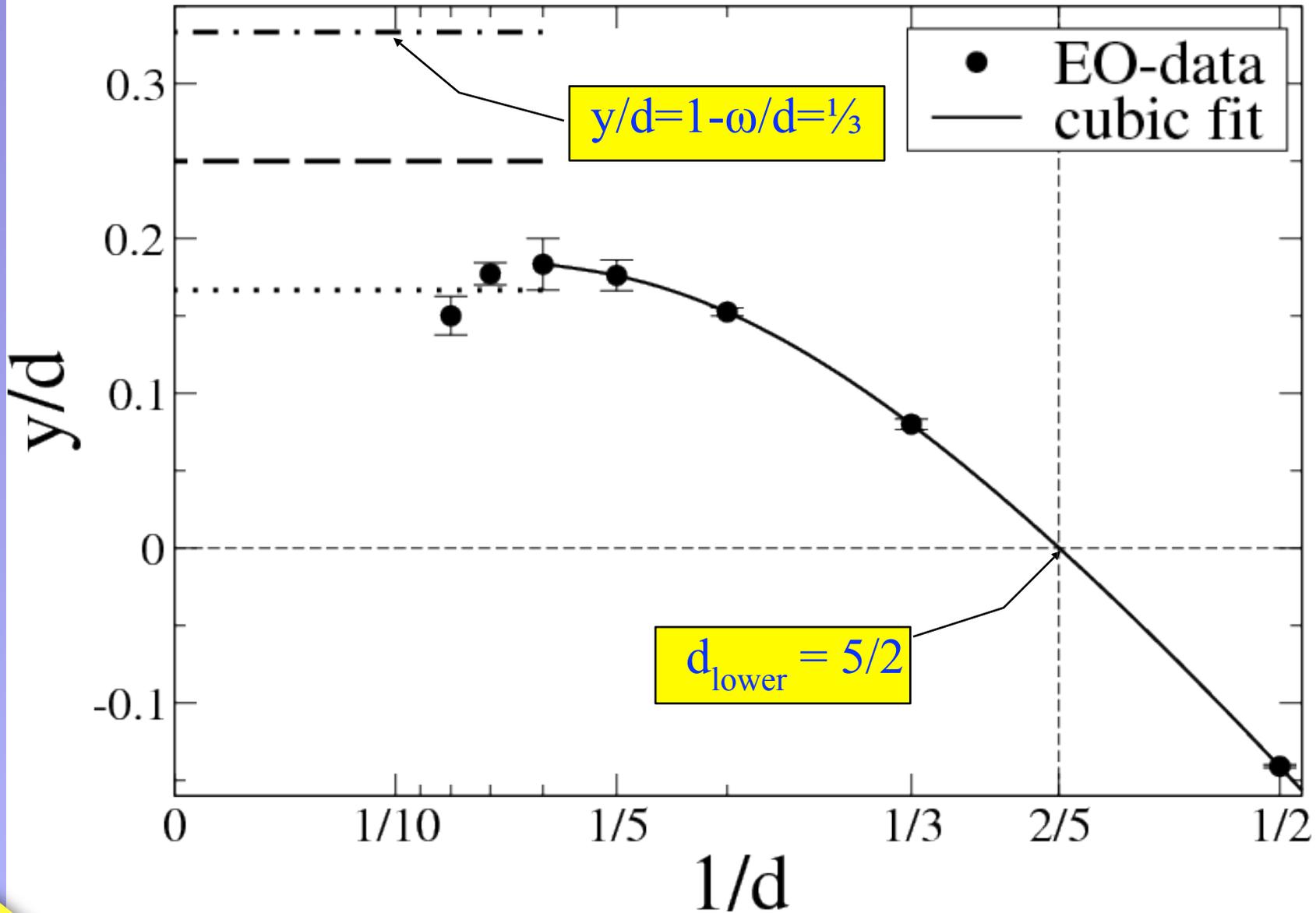
Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$



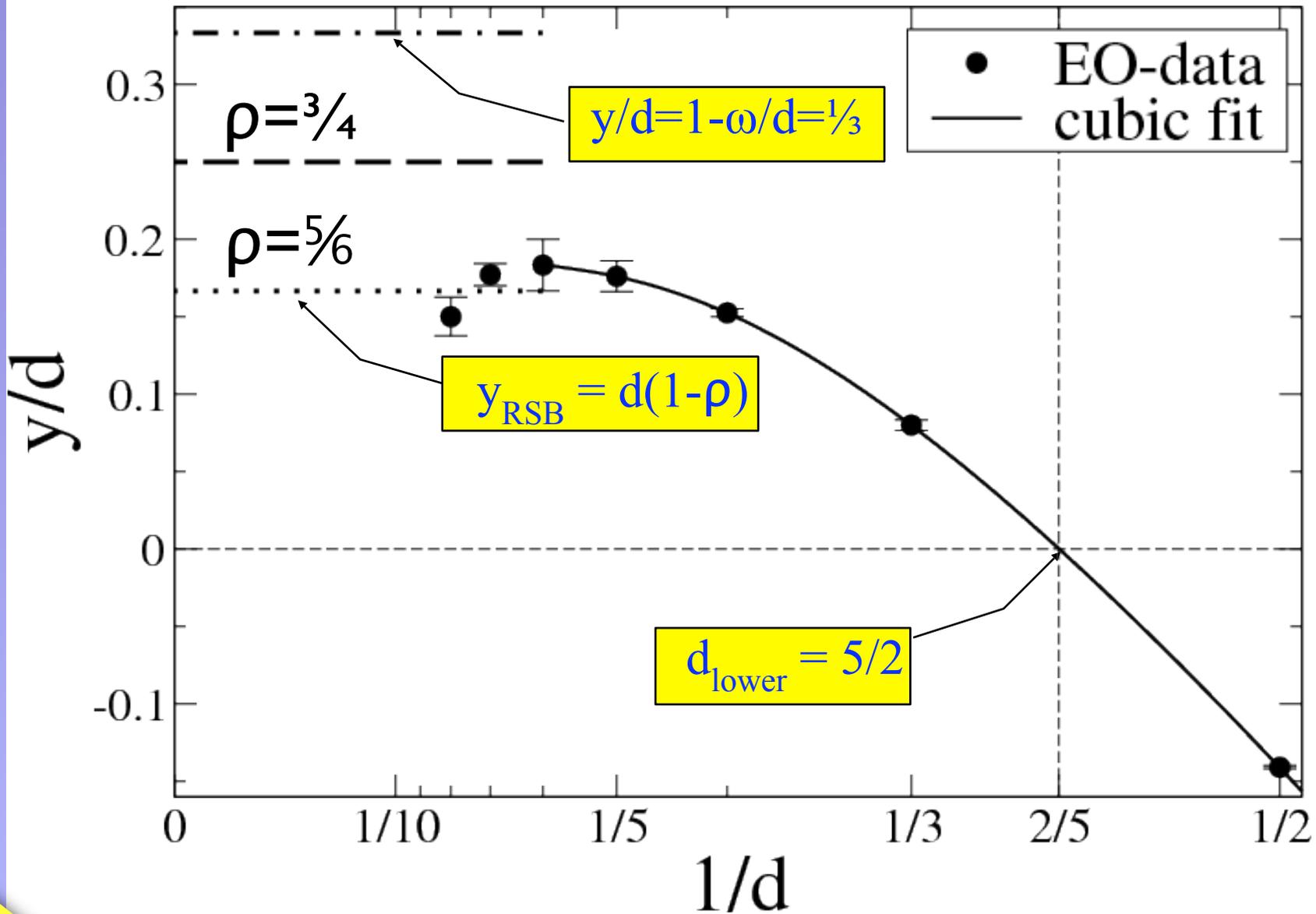
Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$



Comparing with Theory:

“Stiffness”: $\sigma(\Delta E) \sim L^y$





Corrections-to-Scaling in EA:

Ground State Energy: $E(L) \sim e_0 L^d + AL^y \quad (L \rightarrow \infty)$





Corrections-to-Scaling in EA:

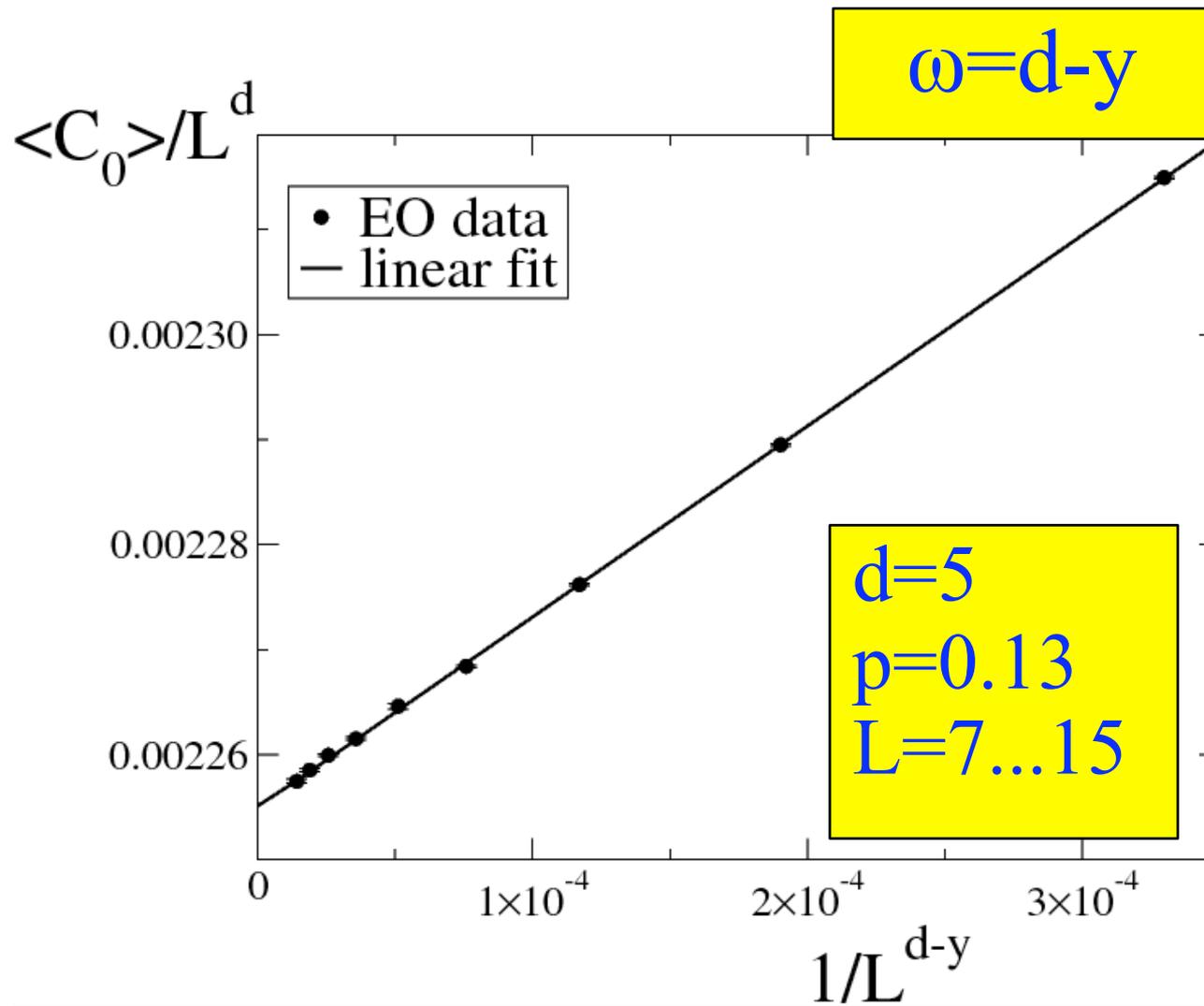
Ground State Energy: $E(L)/L^d \sim e_0 + A/L^{d-y} \quad (L \rightarrow \infty)$

$$\omega = d - y$$



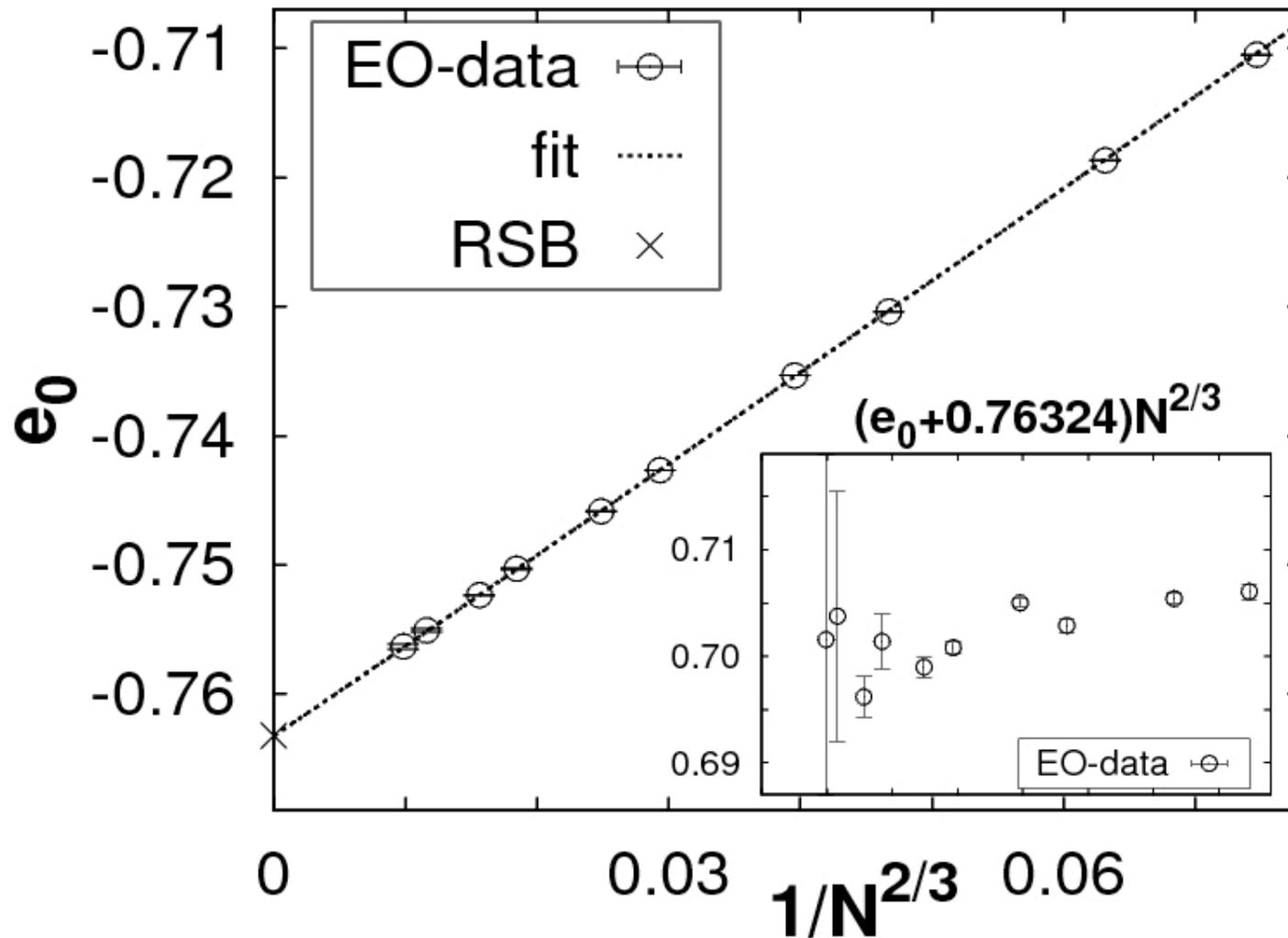
Corrections-to-Scaling in EA:

Ground State Energy: $E(L)/L^d \sim e_0 + A/L^{d-y} \quad (L \rightarrow \infty)$



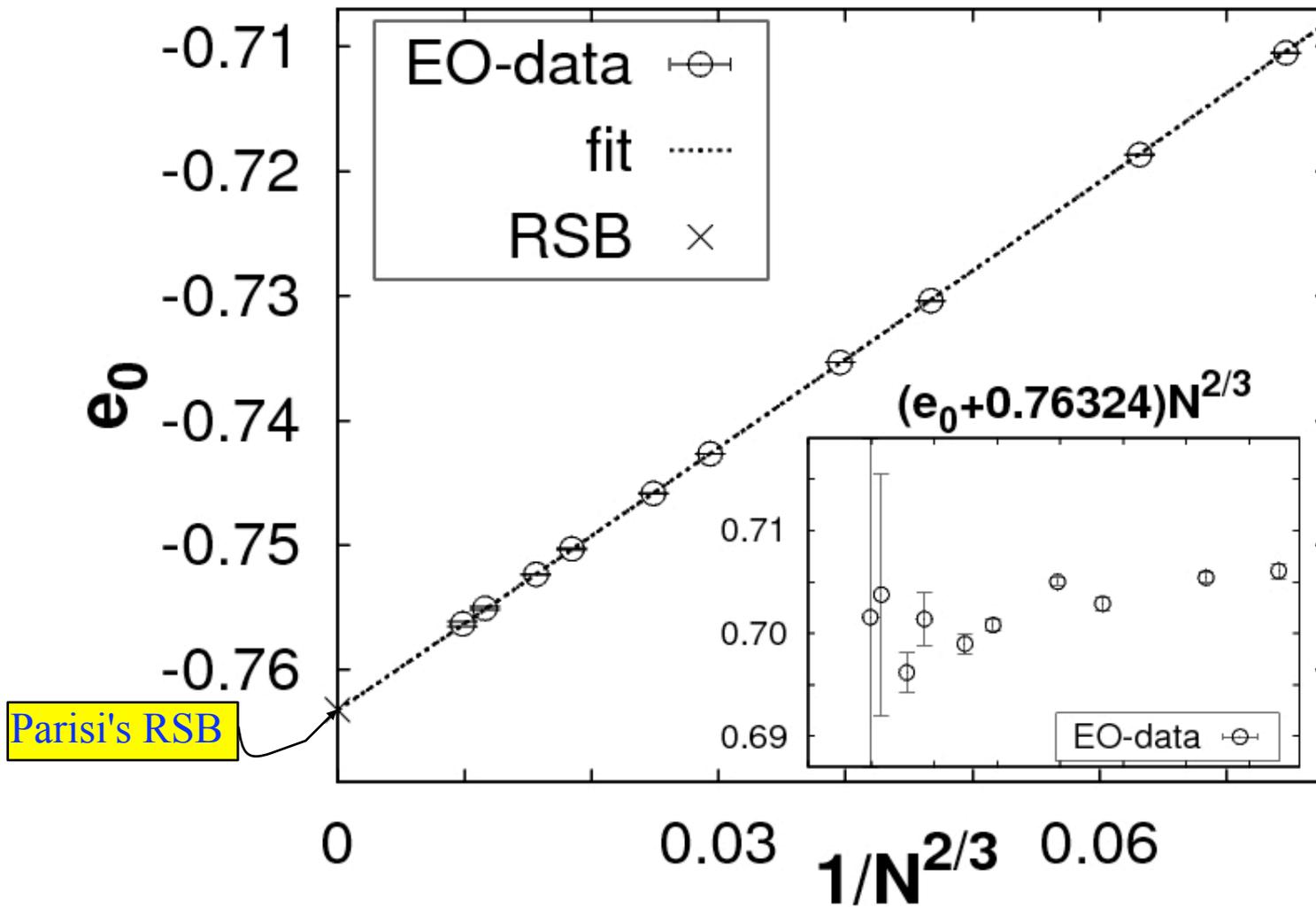
τ -EO for Sherrington-Kirkpatrick

- Mean-Field ($d \rightarrow \infty$) Spin Glasses:



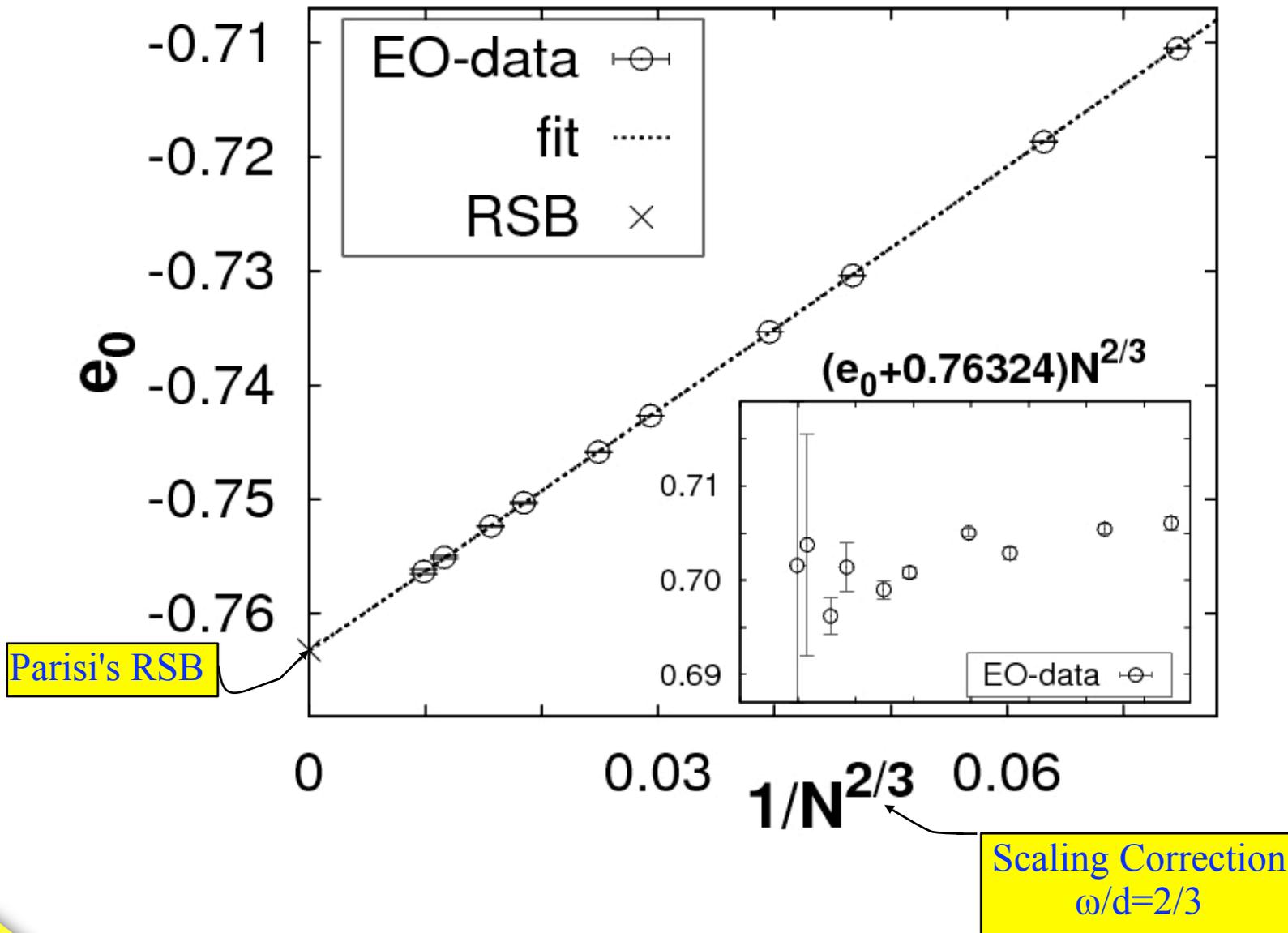
τ -EO for Sherrington-Kirkpatrick

- Mean-Field ($d \rightarrow \infty$) Spin Glasses:



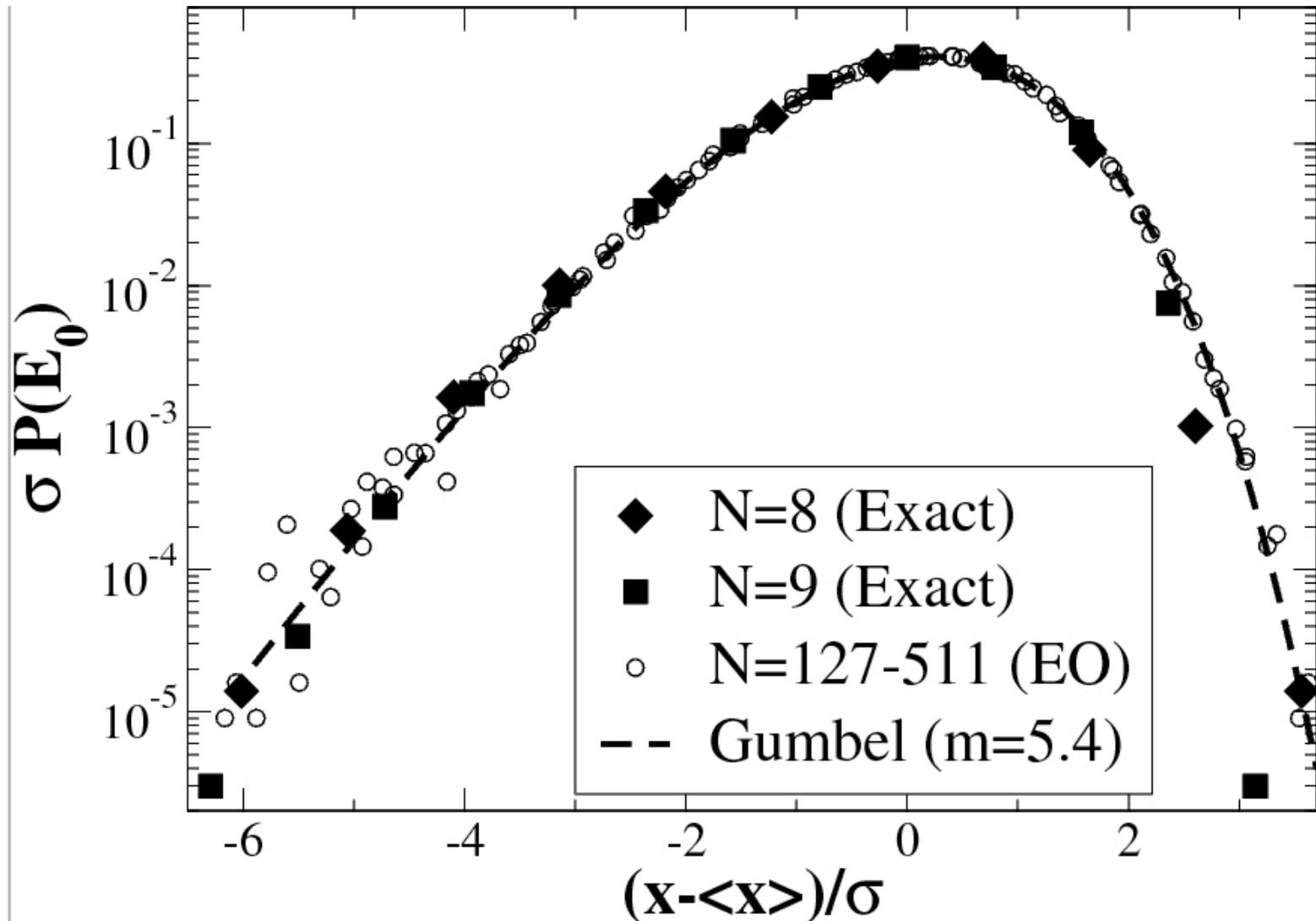
τ -EO for Sherrington-Kirkpatrick

- Mean-Field ($d \rightarrow \infty$) Spin Glasses:



τ -EO for Sherrington-Kirkpatrick

- Mean-Field ($d \rightarrow \infty$) Spin Glasses:





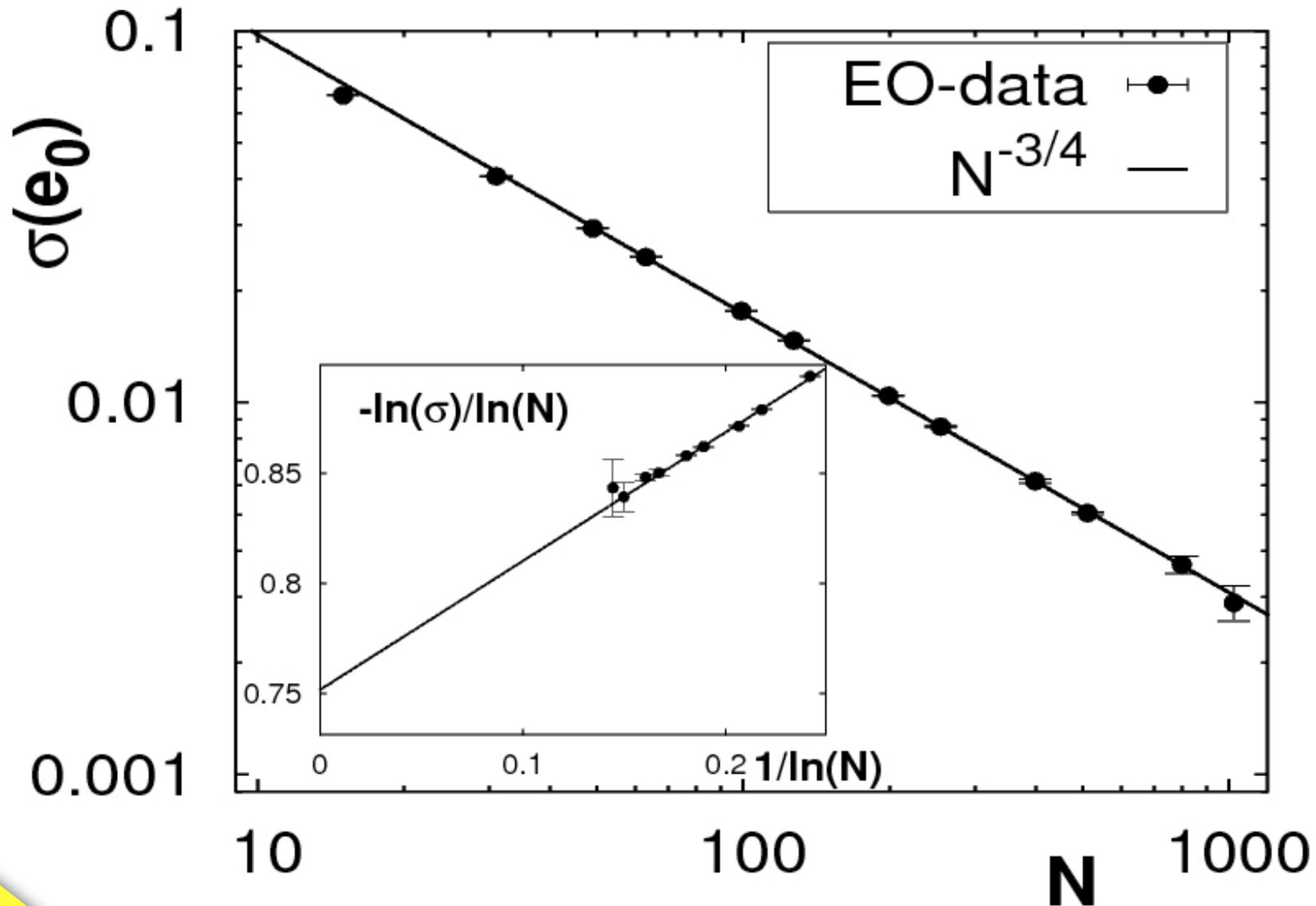
τ -EO for Sherrington-Kirkpatrick

- Mean-Field ($d \rightarrow \infty$) Spin Glasses:



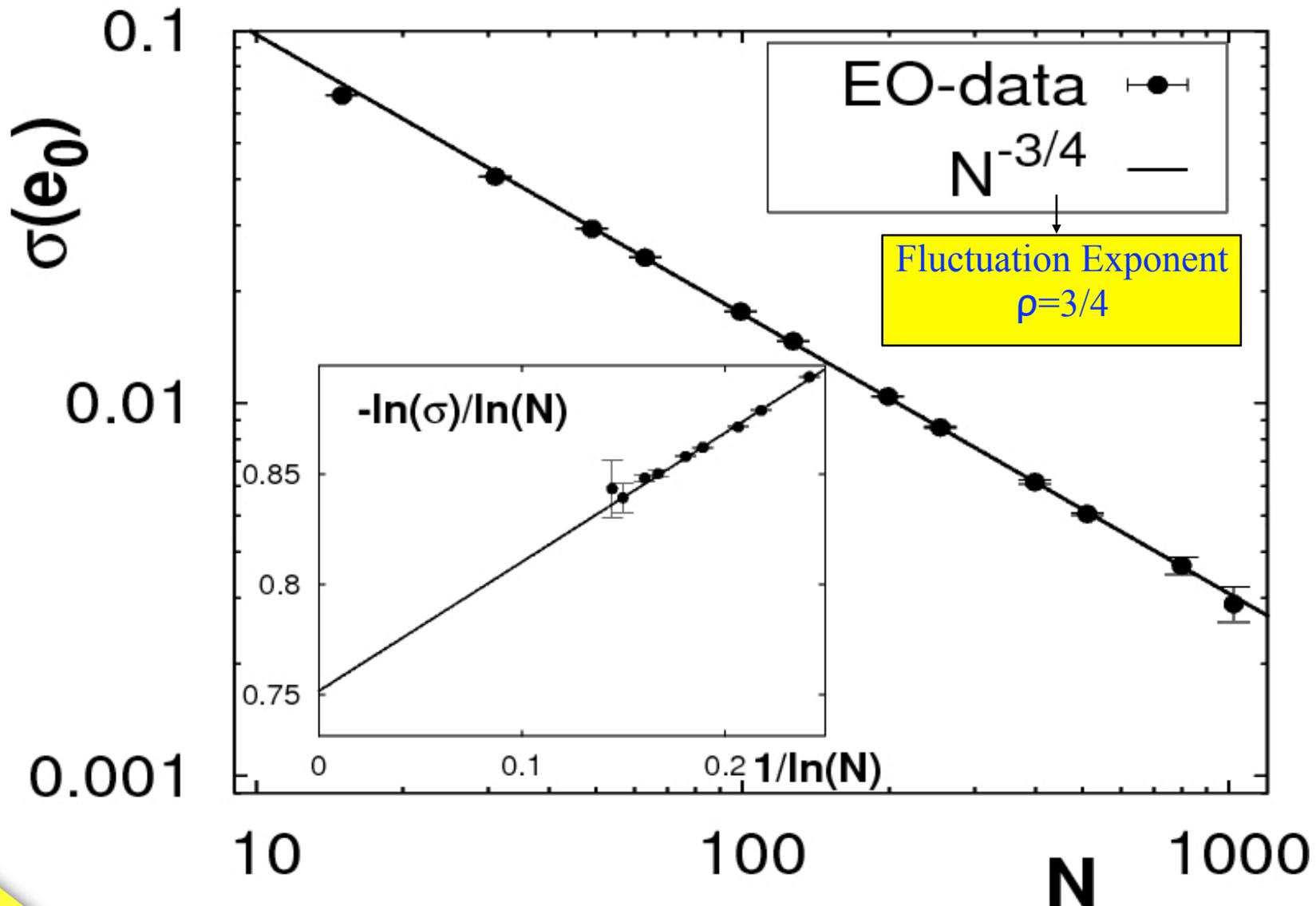
τ -EO for Sherrington-Kirkpatrick

- Mean-Field ($d \rightarrow \infty$) Spin Glasses:



τ -EO for Sherrington-Kirkpatrick

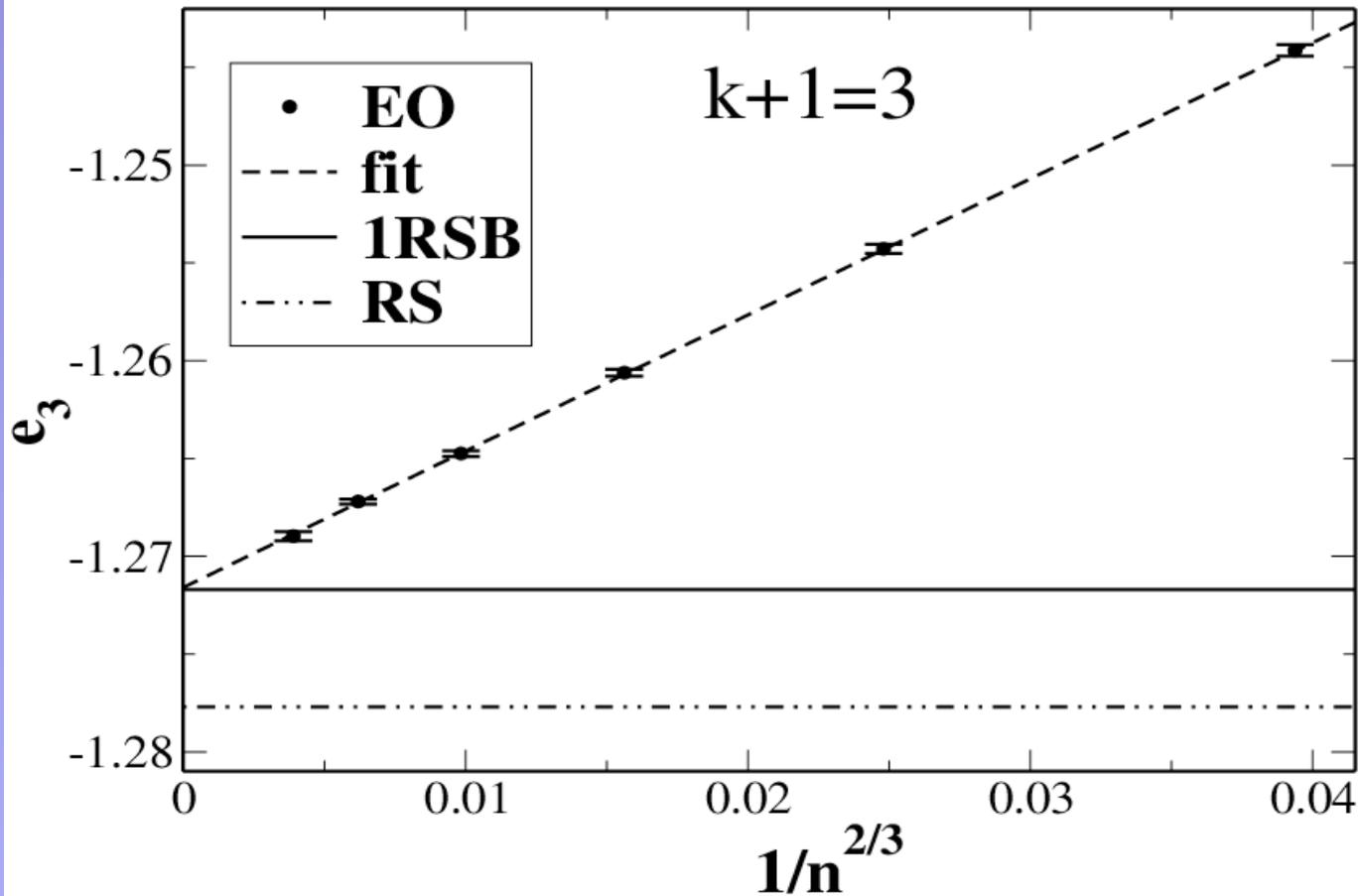
- Mean-Field ($d \rightarrow \infty$) Spin Glasses:

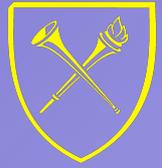




T-EO for Bethe Lattices:

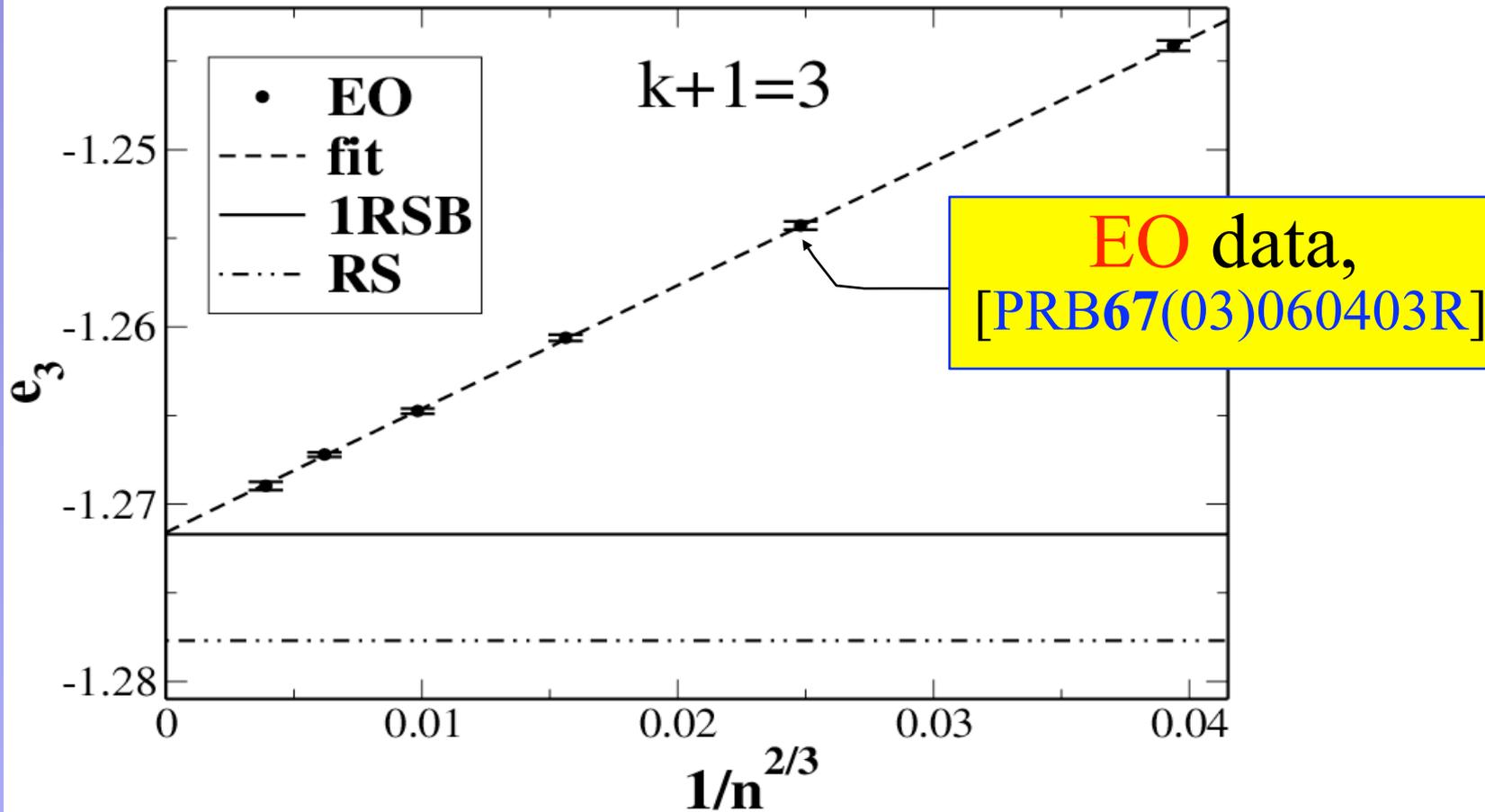
EO for 3-connected Bethe Lattice Glass w/ Replica Sym. Breaking:





T-EO for Bethe Lattices:

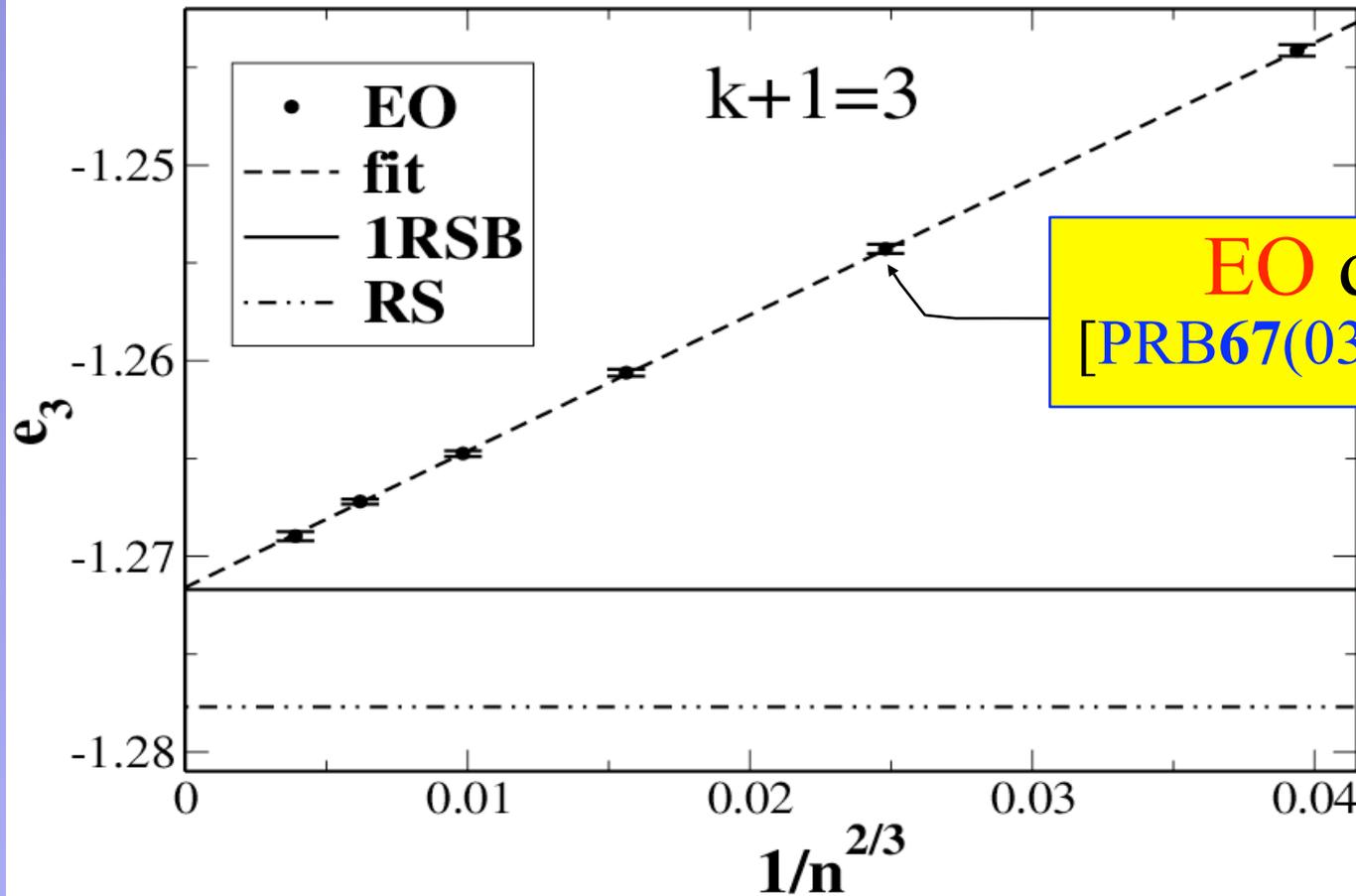
EO for 3-connected Bethe Lattice Glass w/ Replica Sym. Breaking:





T-EO for Bethe Lattices:

EO for 3-connected Bethe Lattice Glass w/ Replica Sym. Breaking:



EO data,
[PRB67(03)060403R]

Replica Theory:

⇐ 1RSB,

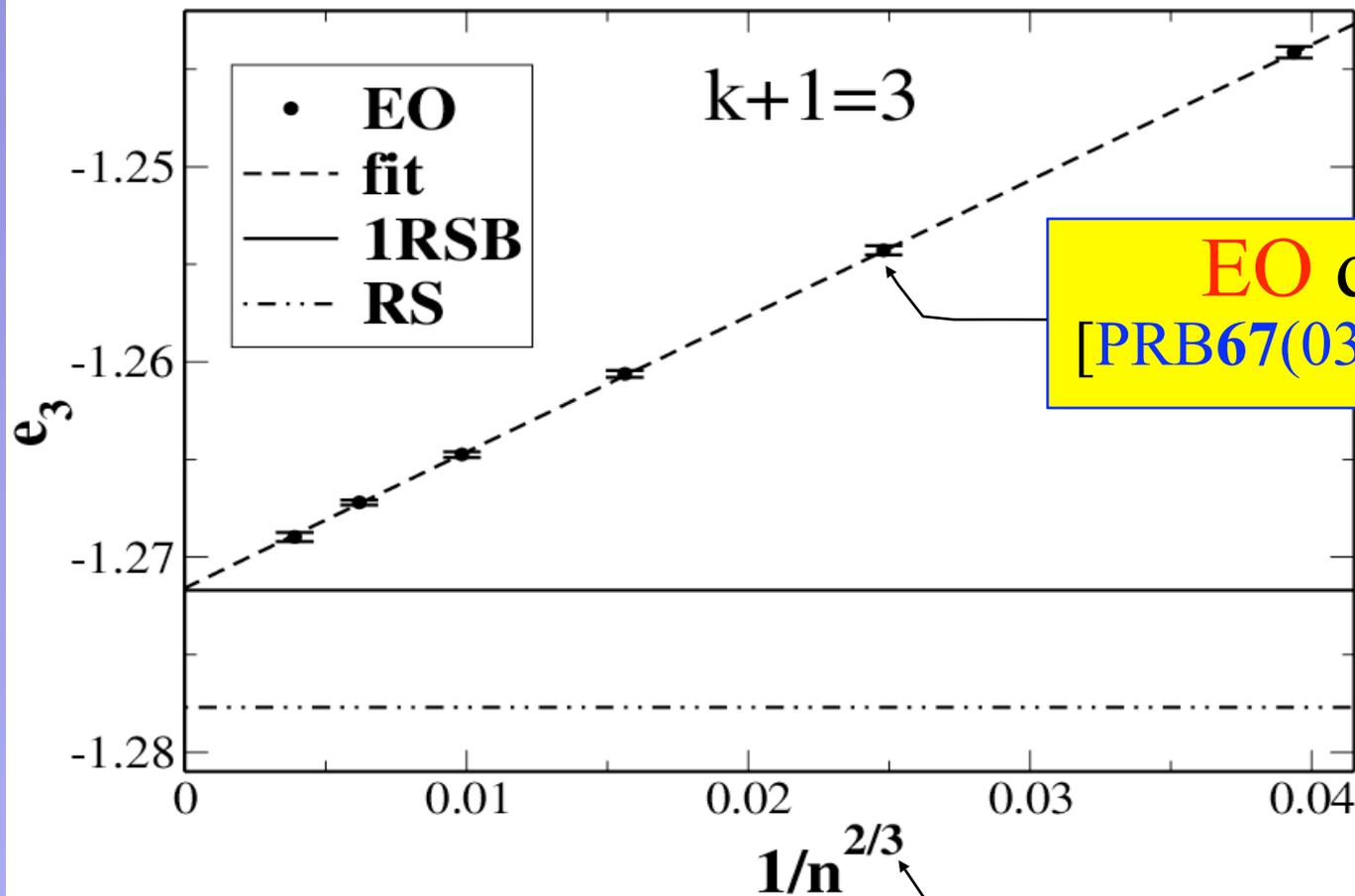
⇐ no RSB,

[J.Stat.Phys.111(03)1]



T-EO for Bethe Lattices:

EO for 3-connected Bethe Lattice Glass w/ Replica Sym. Breaking:



EO data,
[PRB67(03)060403R]

Replica Theory:

⇐ 1RSB,

⇐ no RSB,

[J.Stat.Phys.111(03)1]

Scaling corrections

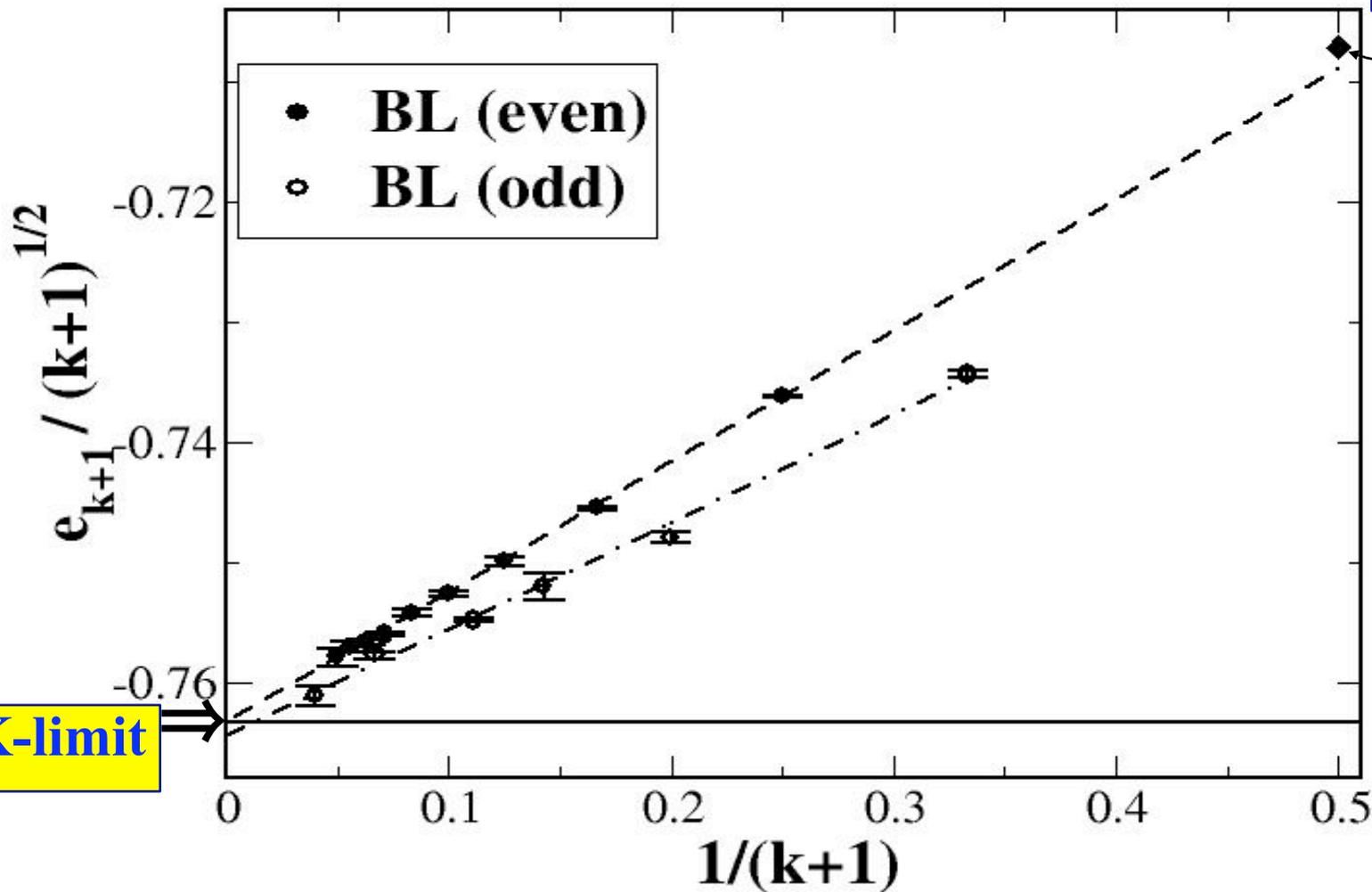


T-EO for Bethe Lattices:

EO for $(k+1)$ -connected Bethe Lattice Glasses for $(k+1) \rightarrow \infty$:

Energies

$k=1$, exact



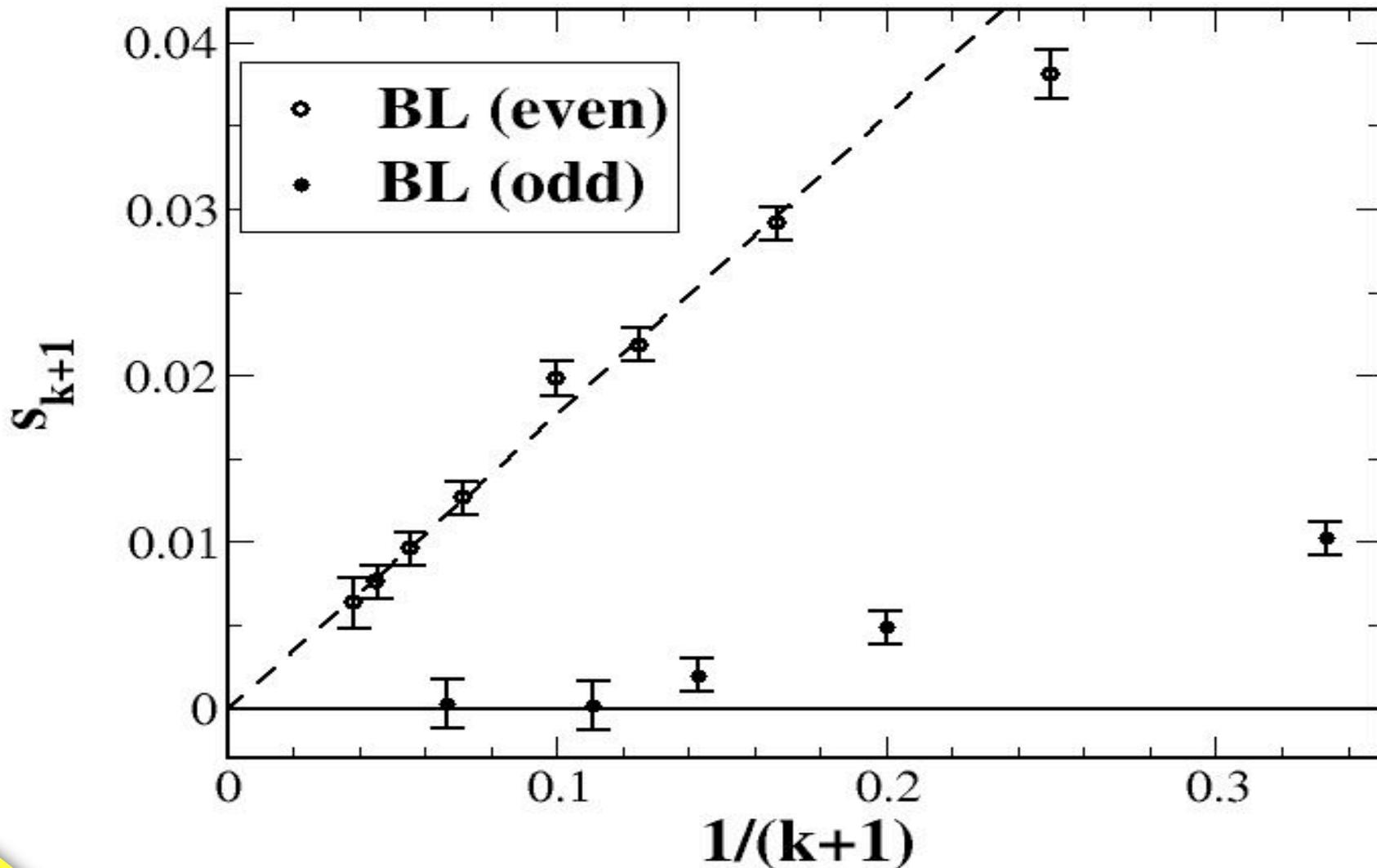
SK-limit



T-EO for Bethe Lattices:

EO for $(k+1)$ -connected Bethe Lattice Glasses for $(k+1) \rightarrow \infty$:

Entropies





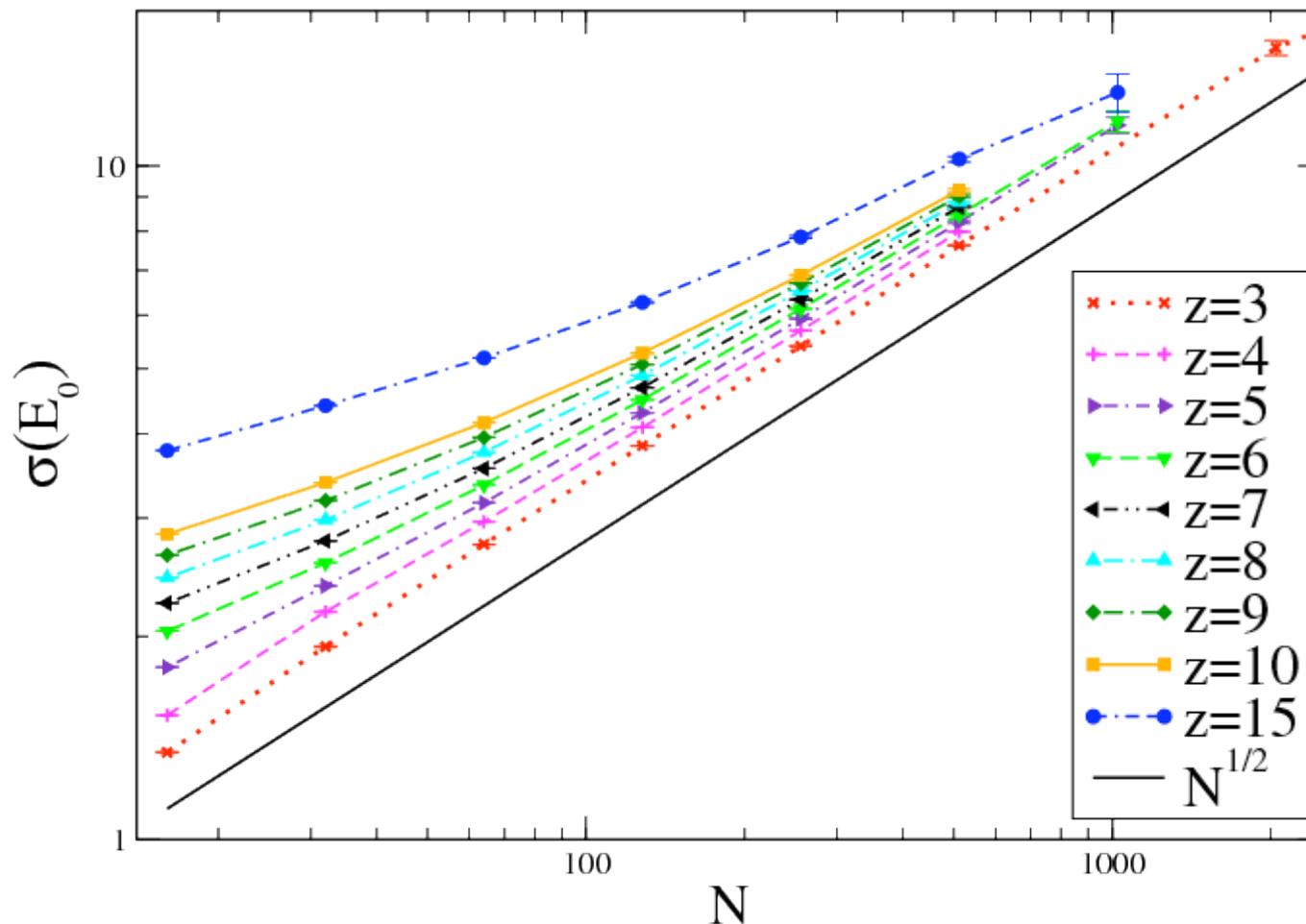
Distribution of Ground State Energies:

Deviation $\sigma(e_0)$ of PDF for **Bethe Lattices** of Degree $z(=k+1)$:



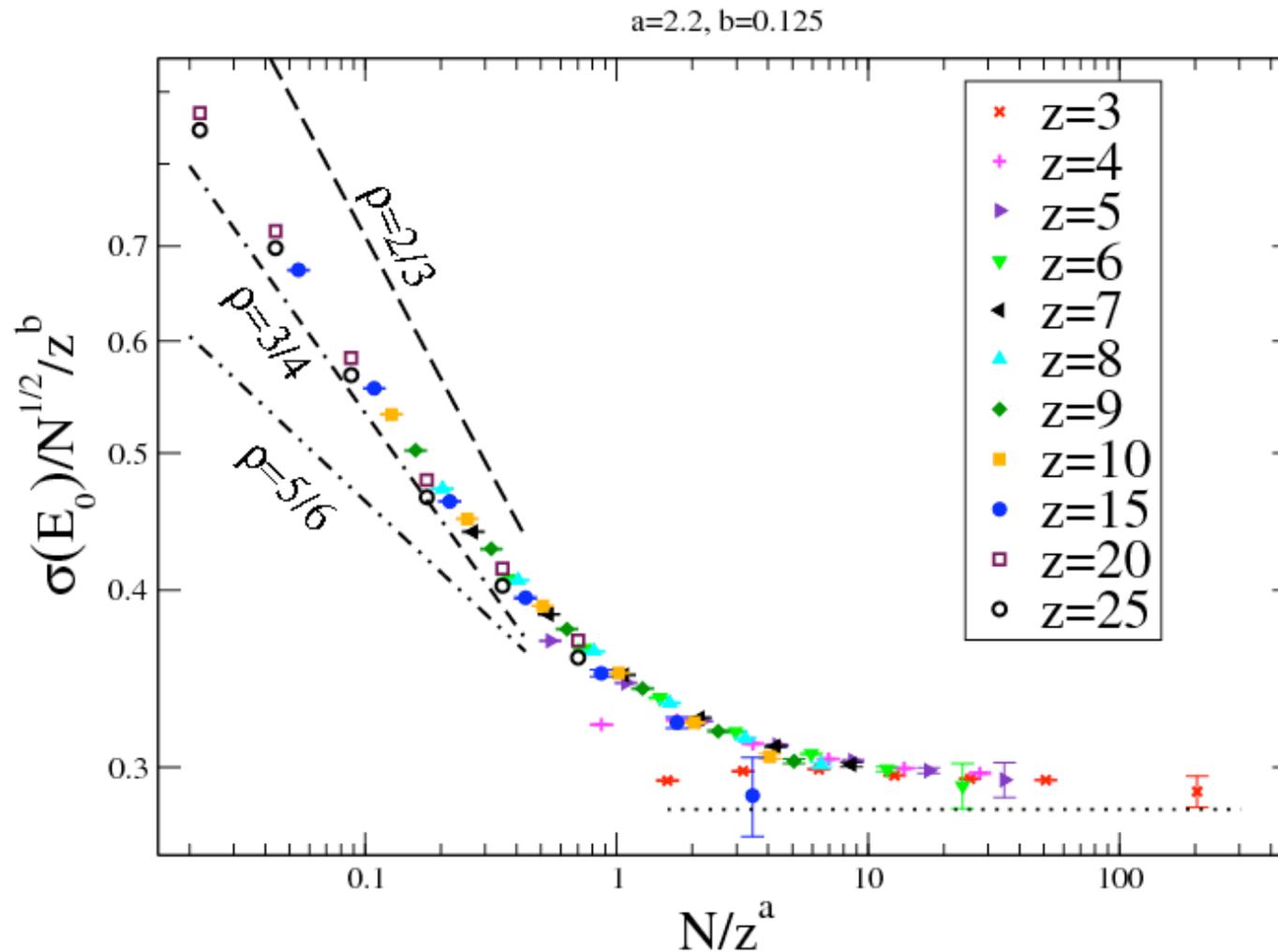
Distribution of Ground State Energies:

Deviation $\sigma(e_0)$ of PDF for Bethe Lattices of Degree $z(=k+1)$:



Distribution of Ground State Energies:

Deviation $\sigma(e_0)$ of PDF for Bethe Lattices of Degree $z(=k+1)$:





Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree

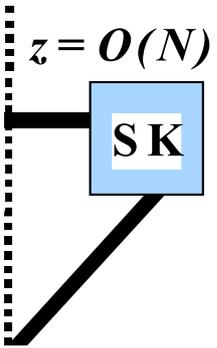


Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree

- Sherrington-Kirkpatrick (**SK**)
↳ dense Graph

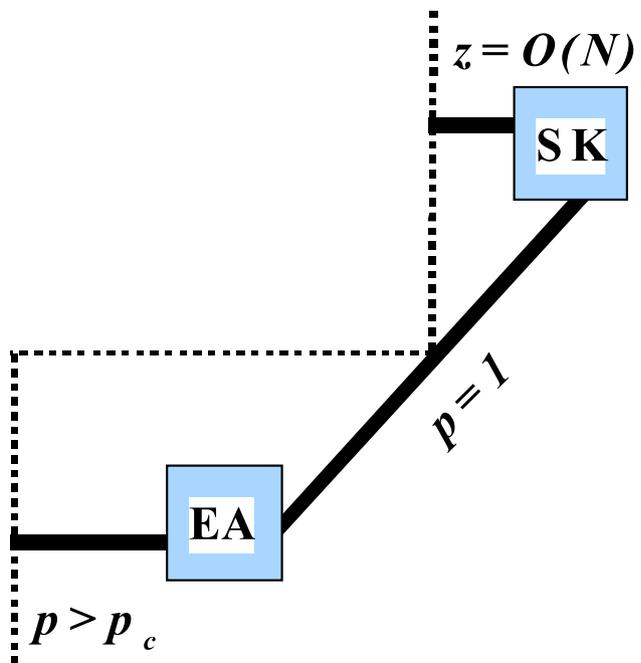


Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree

- Sherrington-Kirkpatrick (**SK**)
↳ dense Graph
- Edwards-Anderson Model (**EA**)
↳ hyper-cubic Lattice, dilute

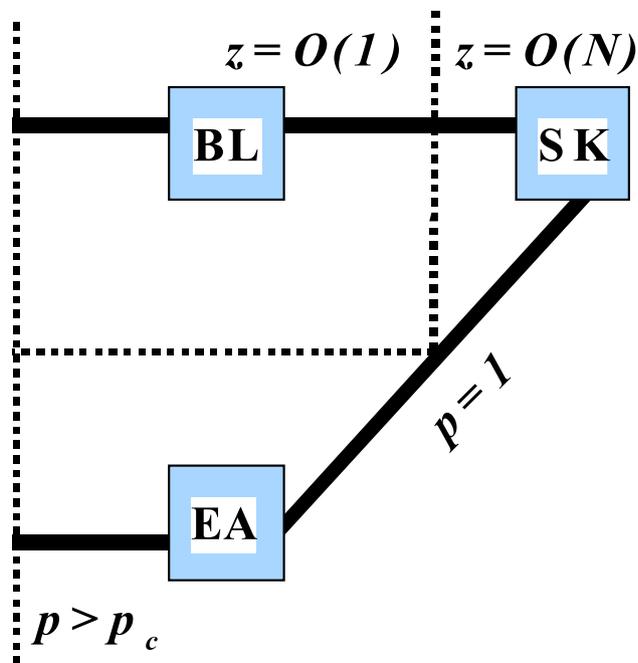


Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree

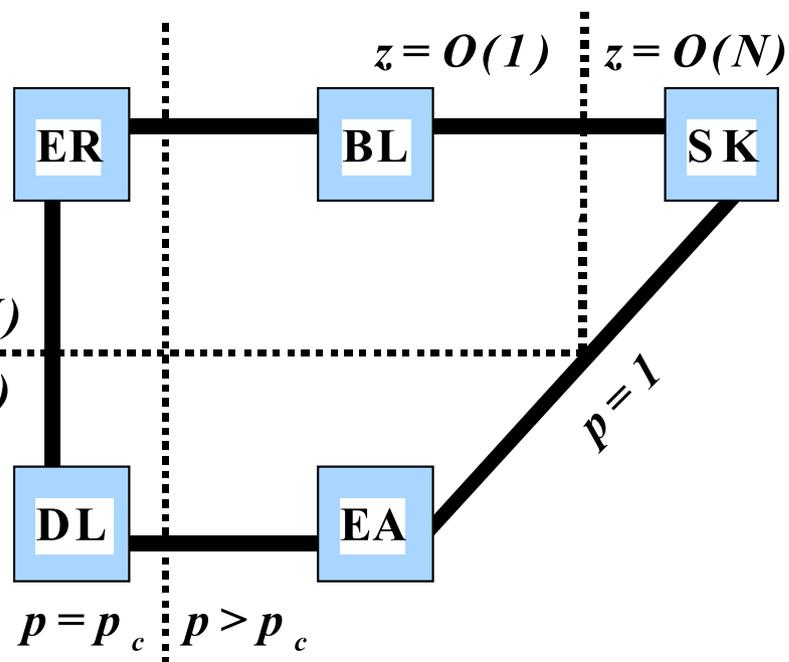
- Sherrington-Kirkpatrick (**SK**)
↳ dense Graph
- Edwards-Anderson Model (**EA**)
↳ hyper-cubic Lattice, dilute
- Bethe “Lattice” (**BL**)
↳ randomly diluted Graph



Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree



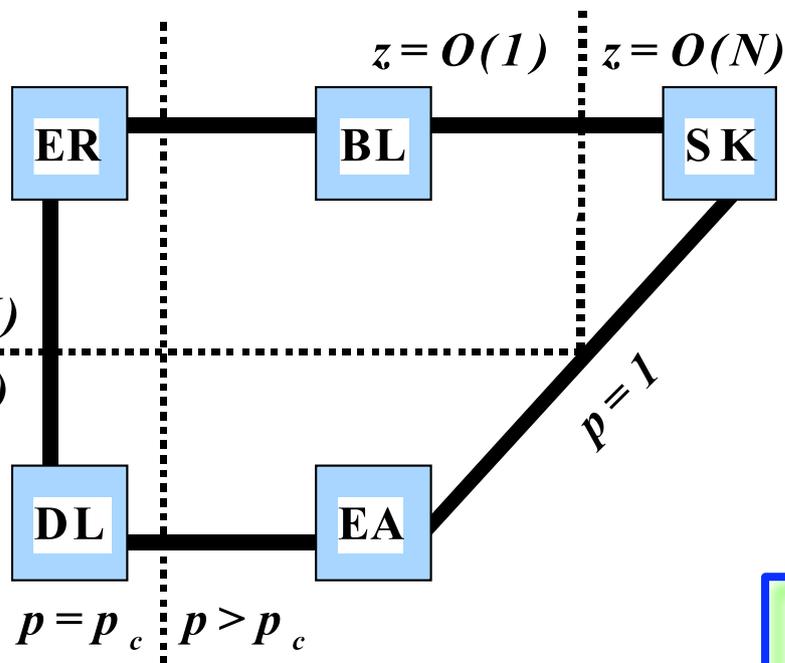
- Sherrington-Kirkpatrick (**SK**)
 - ↳ dense Graph
- Edwards-Anderson Model (**EA**)
 - ↳ hyper-cubic Lattice, dilute
- Bethe "Lattice" (**BL**)
 - ↳ randomly diluted Graph
- Dilute Lattice (**DL**)
 - ↳ **EA** at p_c
- Erdős-Rényi Graph (**ER**)
 - ↳ Random Graph at p_c



Comprehensive View on Spin Glasses:

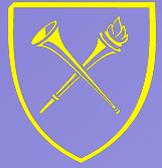
A Set of Models:

d: dimension,
p: bond density,
z: bond degree



- Sherrington-Kirkpatrick (**SK**)
↳ dense Graph
- Edwards-Anderson Model (**EA**)
↳ hyper-cubic Lattice, dilute
- Bethe “Lattice” (**BL**)
↳ randomly diluted Graph
- Dilute Lattice (**DL**)
↳ **EA** at p_c
- Erdős-Renyi Graph (**ER**)
↳ Random Graph at p_c

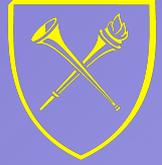
$$H = \frac{1}{\sqrt{2dp}} \sum_{\langle i,j \rangle} J_{i,j} x_i x_j$$



Comprehensive View on Spin Glasses:

A Set of Exponents:





Comprehensive View on Spin Glasses:

A Set of Exponents:

1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$

▫ In **EA**: $\rho = 1/2$ (Wehr&Aizenman: ρ exact!)



Comprehensive View on Spin Glasses:

A Set of Exponents:

1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$

▫ In **EA**: $\rho = 1/2$ (Wehr&Aizenman: ρ exact!)

2) Distribution $P(\Delta E_0)$, width $\sigma(\Delta E_0) \sim N^{y/d} = L^y$

▫ In **EA**: $y \approx 0.24, \dots, 1.2$ for $d=3, \dots, 7$



Comprehensive View on Spin Glasses:

A Set of Exponents:

- 1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$
 - In **EA**: $\rho = 1/2$ (Wehr&Aizenman: ρ exact!)

- 2) Distribution $P(\Delta E_0)$, width $\sigma(\Delta E_0) \sim N^{y/d} = L^y$
 - In **EA**: $y \approx 0.24, \dots, 1.2$ for $d=3, \dots, 7$

- 3) Corrections-to-Scaling : $e_0(N) - e_0(\infty) \sim N^{-\omega/d} = L^{-\omega}$
 - In **EA**: $\omega/d = 1 - y/d$



Comprehensive View on Spin Glasses:

A Set of Exponents:

- 1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$
 - In **EA**: $\rho = 1/2$ (Wehr&Aizenman ρ exact!)

- 2) Distribution $P(\Delta E_0)$, width $\sigma(\Delta E_0) \sim N^{y/d} = L^y$
 - In **EA**: $y \approx 0.24, \dots, 1.2$ for $d=3, \dots, 7$

- 3) Corrections-to-Scaling : $e_0(N) - e_0(\infty) \sim N^{-\omega/d} = L^{-\omega}$
 - In **EA**: $\omega/d = 1 - y/d$



Comprehensive View on Spin Glasses:

A Set of Exponents:

- 1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$
 - In **EA**: $\rho = 1/2$ (Wehr&Aizenman ρ exact!)
 - In **SK**: $\rho \approx 3/4$ (Highly Skewed)

- 2) Distribution $P(\Delta E_0)$, width $\sigma(\Delta E_0) \sim N^{y/d} = L^y$
 - In **EA**: $y \approx 0.24, \dots, 1.2$ for $d=3, \dots, 7$

- 3) Corrections-to-Scaling : $e_0(N) - e_0(\infty) \sim N^{-\omega/d} = L^{-\omega}$
 - In **EA**: $\omega/d = 1 - y/d$



Comprehensive View on Spin Glasses:

A Set of Exponents:

1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$

▫ In **EA**: $\rho = 1/2$ (Wehr&Aizenman ρ exact!)

▫ In **SK**: $\rho \approx 3/4$ (Highly Skewed)

2) Distribution $P(\Delta E_0)$, width $\sigma(\Delta E_0) \sim N^{y/d} = L^y$

▫ In **EA**: $y \approx 0.24, \dots, 1.2$ for $d=3, \dots, 7$

▫ In **SK**: $y/d = 1 - \rho \rightarrow 1/4$, too high for **EA** at $d \geq 6$

3) Corrections-to-Scaling : $e_0(N) - e_0(\infty) \sim N^{-\omega/d} = L^{-\omega}$

▫ In **EA**: $\omega/d = 1 - y/d$



Comprehensive View on Spin Glasses:

A Set of Exponents:

1) Distribution $P(e_0)$, width $\sigma(e_0) \sim N^{-\rho} = L^{-d\rho}$

▫ In **EA**: $\rho = 1/2$ (Wehr&Aizenman ρ exact!)

▫ In **SK**: $\rho \approx 3/4$ (Highly Skewed)

2) Distribution $P(\Delta E_0)$, width $\sigma(\Delta E_0) \sim N^{y/d} = L^y$

▫ In **EA**: $y \approx 0.24, \dots, 1.2$ for $d=3, \dots, 7$

▫ In **SK**: $y/d = 1 - \rho \rightarrow 1/4$, too high for **EA** at $d \geq 6$

3) Corrections-to-Scaling : $e_0(N) - e_0(\infty) \sim N^{-\omega/d} = L^{-\omega}$

▫ In **EA**: $\omega/d = 1 - y/d$

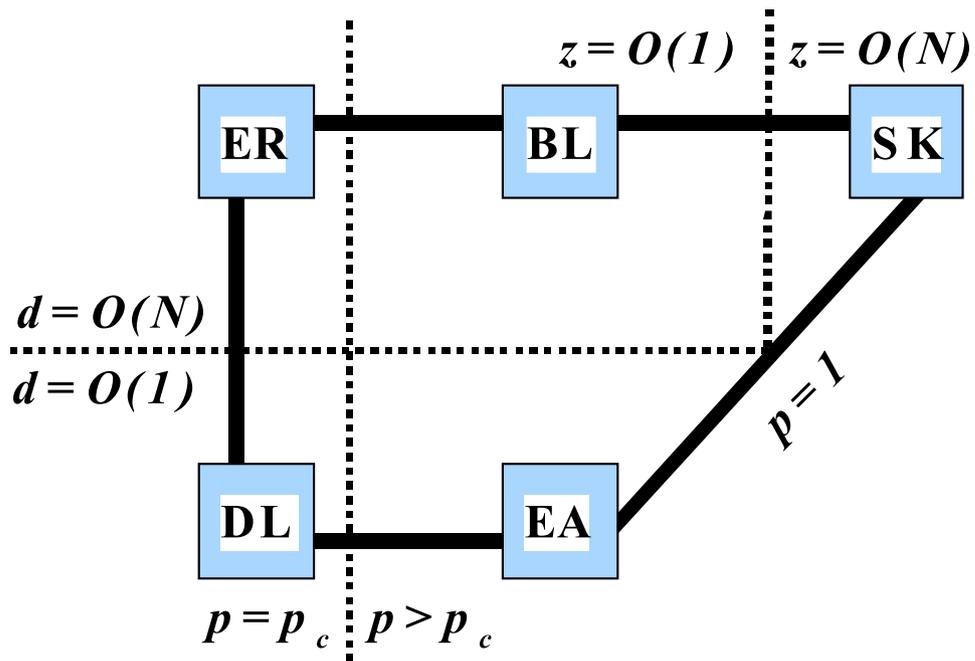
▫ In **SK**: $\omega/d \approx 2/3 \neq 1 - y/d$



Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree

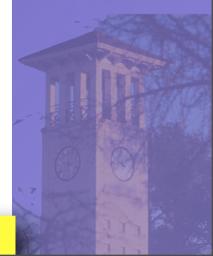
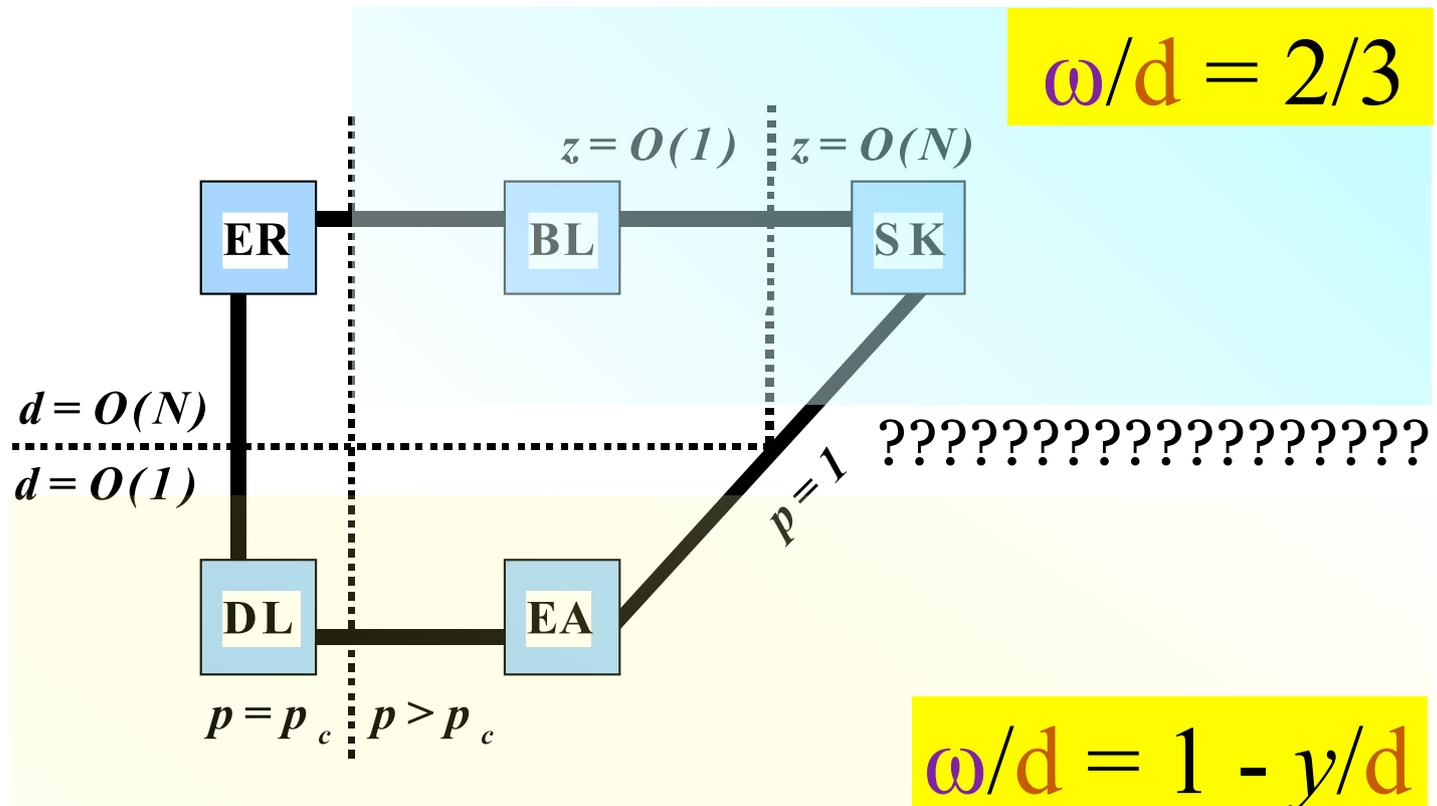


Comprehensive View on Spin Glasses:

A Set of Models:

d: dimension,
p: bond density,
z: bond degree

- Corrections-to-Scaling: ω



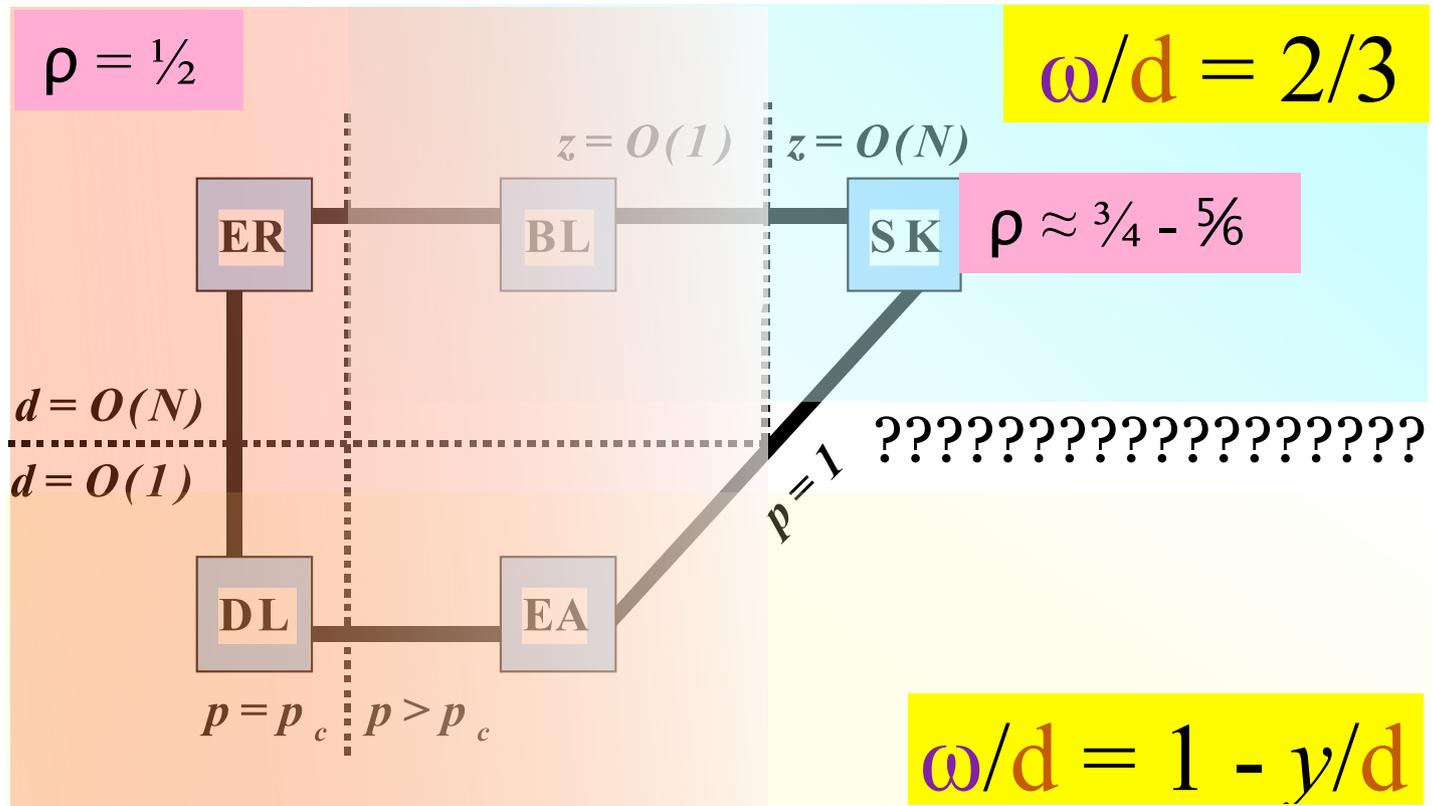
Comprehensive View on Spin Glasses:



A Set of Models:

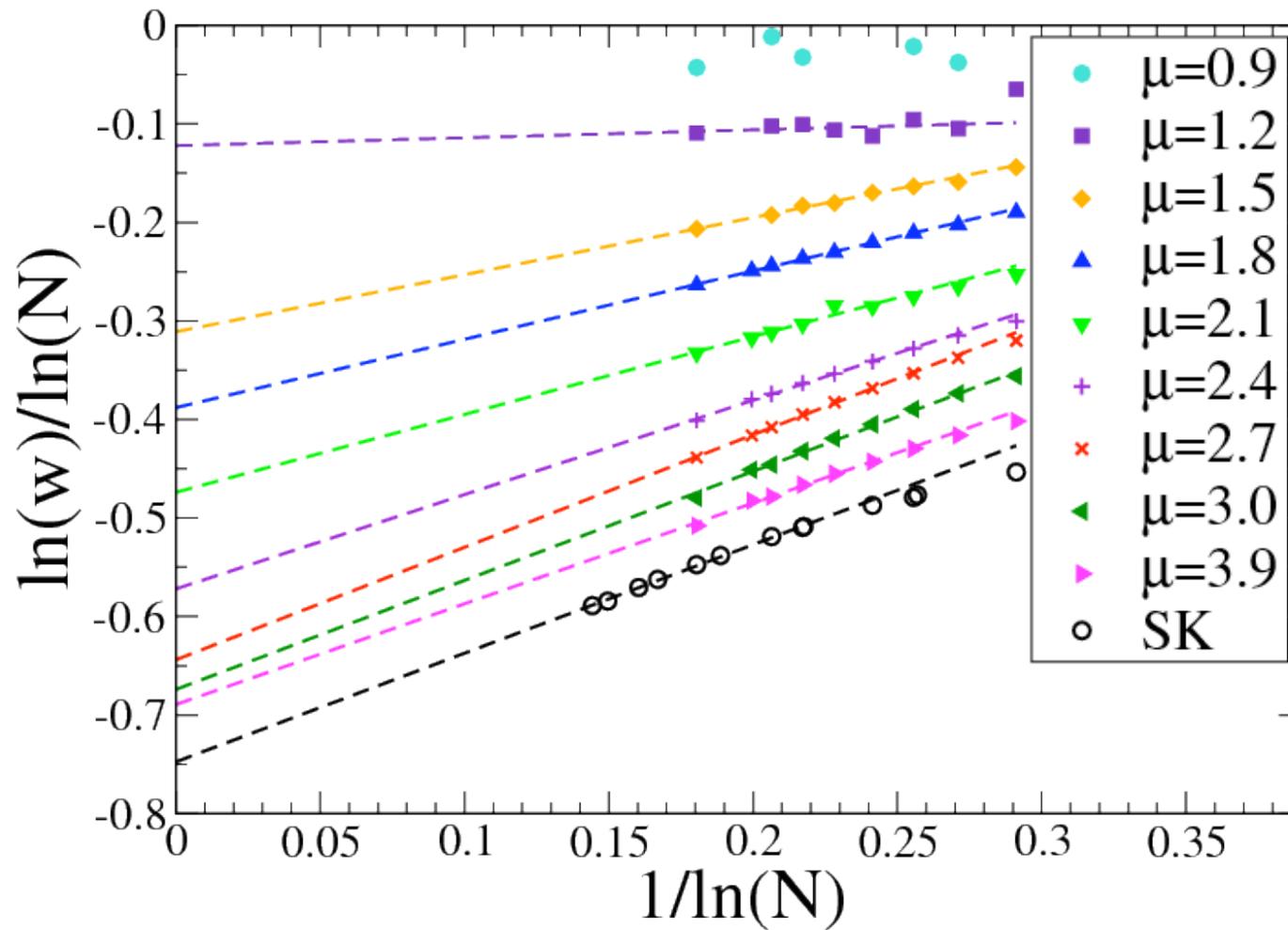
d: dimension,
p: bond density,
z: bond degree

- Corrections-to-Scaling: ω
- Energy Fluctuations: ρ



SK with Power-Law Bonds:

Power-Law Bonds: $P(J) \sim 1/|J|^{1+\mu}$ ($|J| > 1$)



SK with Power-Law Bonds:

Power-Law Bonds: $P(J) \sim 1/|J|^{1+\mu}$ ($|J| > 1$)

