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FORSCHUNGSZENTRUM JÜLICH

26 JUNE 2023

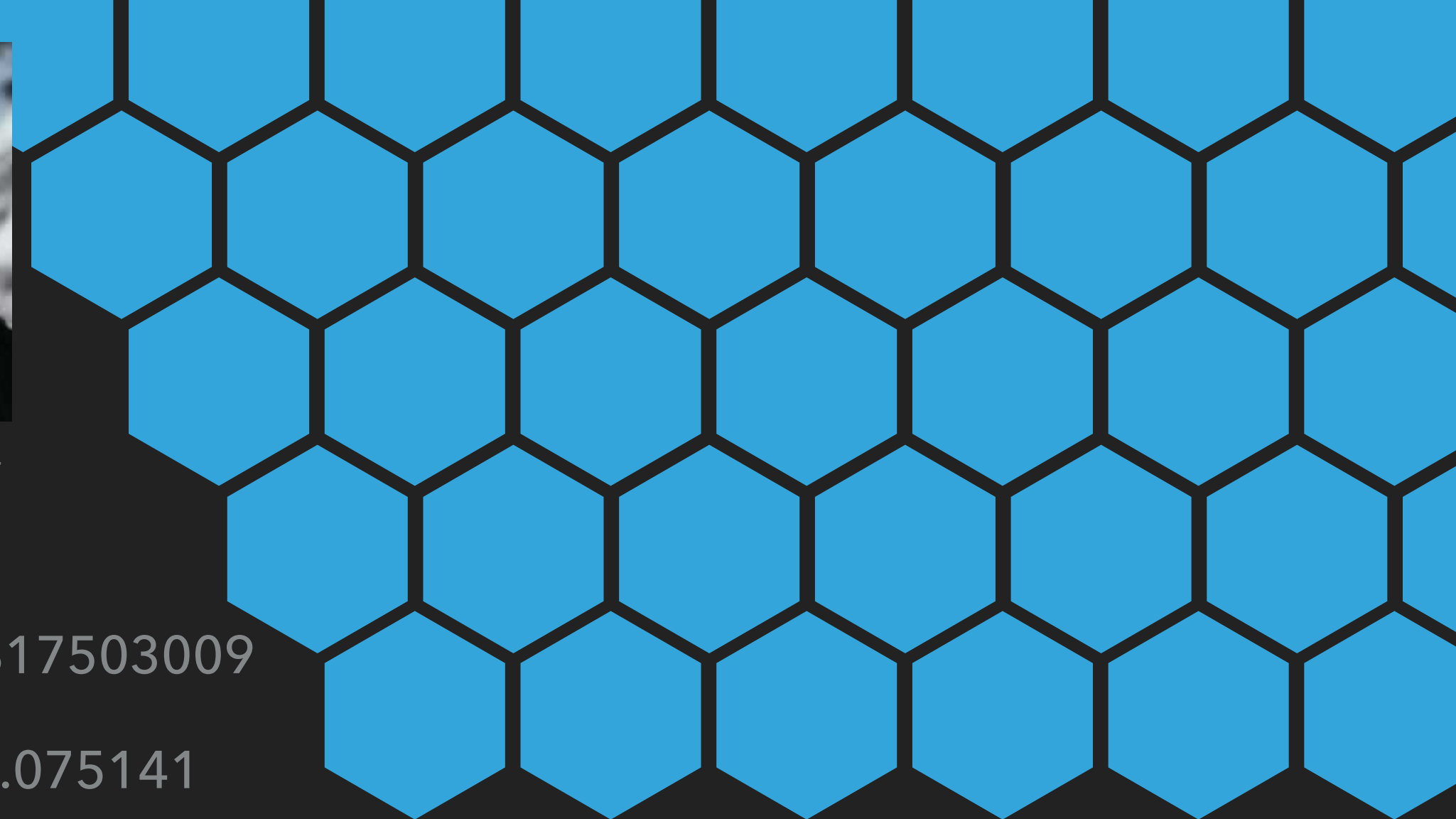
ML FOR LATTICE FIELD THEORY AND BEYOND

LEARNING ABOUT THE HUBBARD MODEL



The best way to stop the Batsignal
turned out to be Batnoise.

Saturday Morning Breakfast Cereal



Christopher Körber

Jan-Lukas Wynen

Stefan Krieg

Peter Labus

Timo Lähde

Tom Luu

Johann Ostmeyer

1710.06213 EB, Körber, Krieg, Labus, Lähde, Luu, Lattice 2017 10.1051/epjconf/201817503009

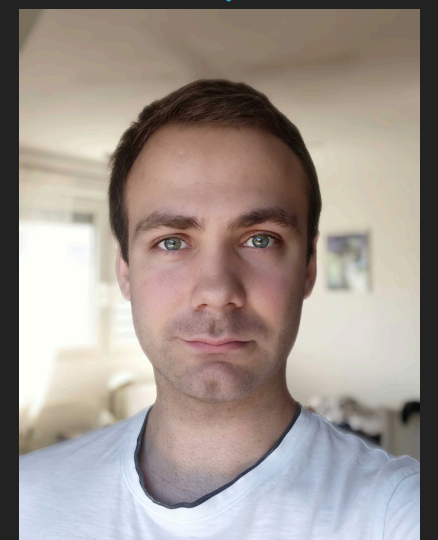
1812.09268 Wynen, EB, Körber, Lähde, Luu, PRB 100:075141 10.1103/PhysRevB.100.075141

2005.11112 Ostmeyer, EB, Krieg, Lähde, Luu, Urbach, PRB 102:245105 10.1103/PhysRevB.102.245105

2006.11221 Wynen, EB, Krieg, Luu, Ostmeyer, PRB 103:125153 10.1103/PhysRevB.103.125153

2105.06936 Ostmeyer, EB, Krieg, Lähde, Luu, Urbach, PRB 104:155142 10.1103/PhysRevB.104.155142

2203.00390 Rodekamp, EB, Gäntgen, Krieg, Luu, Ostmeyer; PRB 106(12):125139 10.1103/PhysRevB.106.125139



Marcel Rodekamp

Christoph Gäntgen

LEARNING ABOUT THE DOPED HUBBARD MODEL



L. Ron Hubbard House
Washington DC

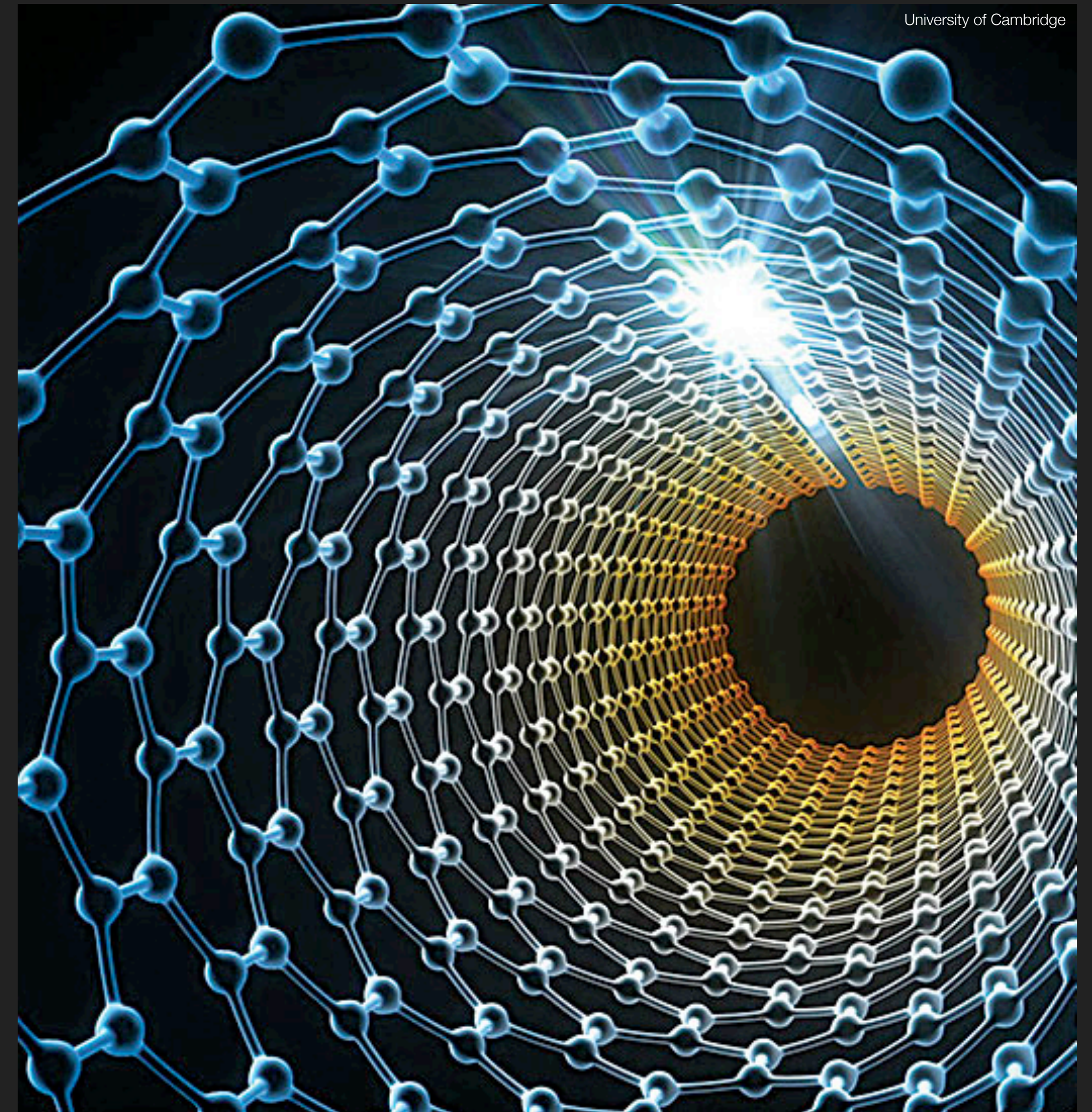
THE HUBBARD MODEL

INTRODUCTION

OVERSIMPLIFIED BUT RICH

WHY THE HUBBARD MODEL?

- ▶ simplest possible Hamiltonian with electronic movement and interactions
- ▶ captures the Mott transition
- ▶ strong interactions render the model intractable
- ▶ easy to decorate / enrich to model realistic materials → all sorts of industrial applications
- ▶ much simpler than QCD

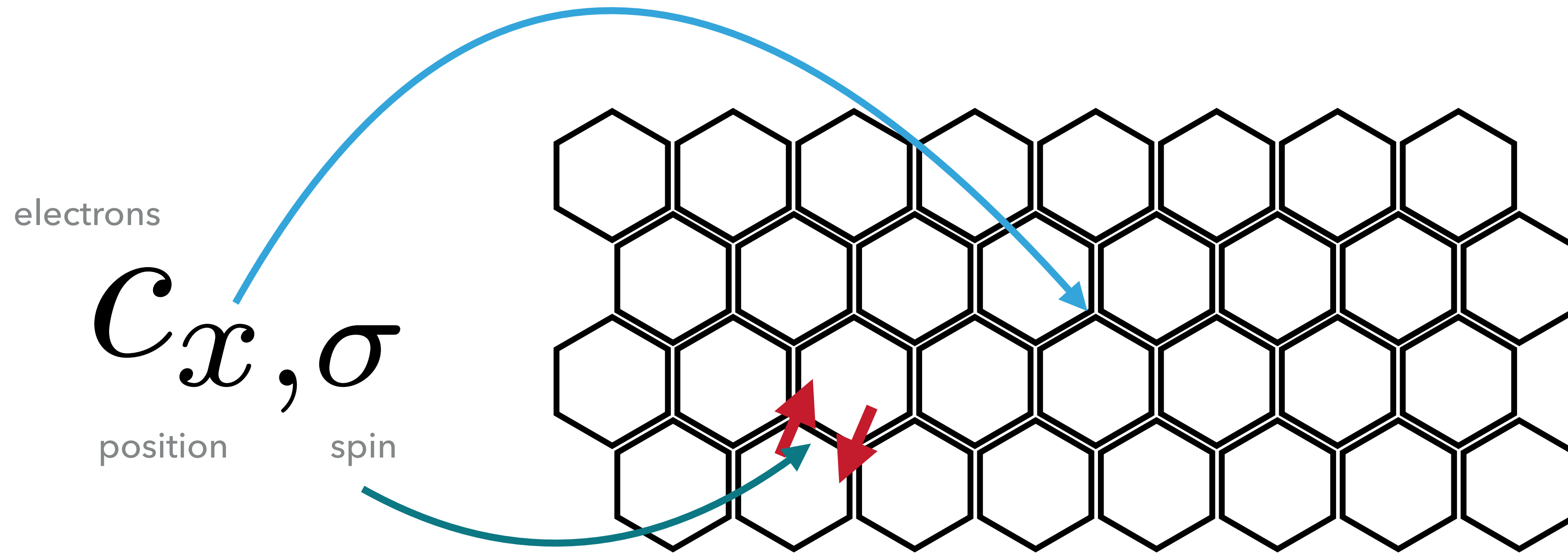




THE HUBBARD MODEL

BACKGROUND

Hubbard Gold Medal of Anne Morrow Lindbergh
Robert Lawton CC BY-SA 2.5



Tight Binding

$$H_0 = - \sum_{x, y, \sigma} c_{x, \sigma}^\dagger h_{xy} c_{y, \sigma}$$

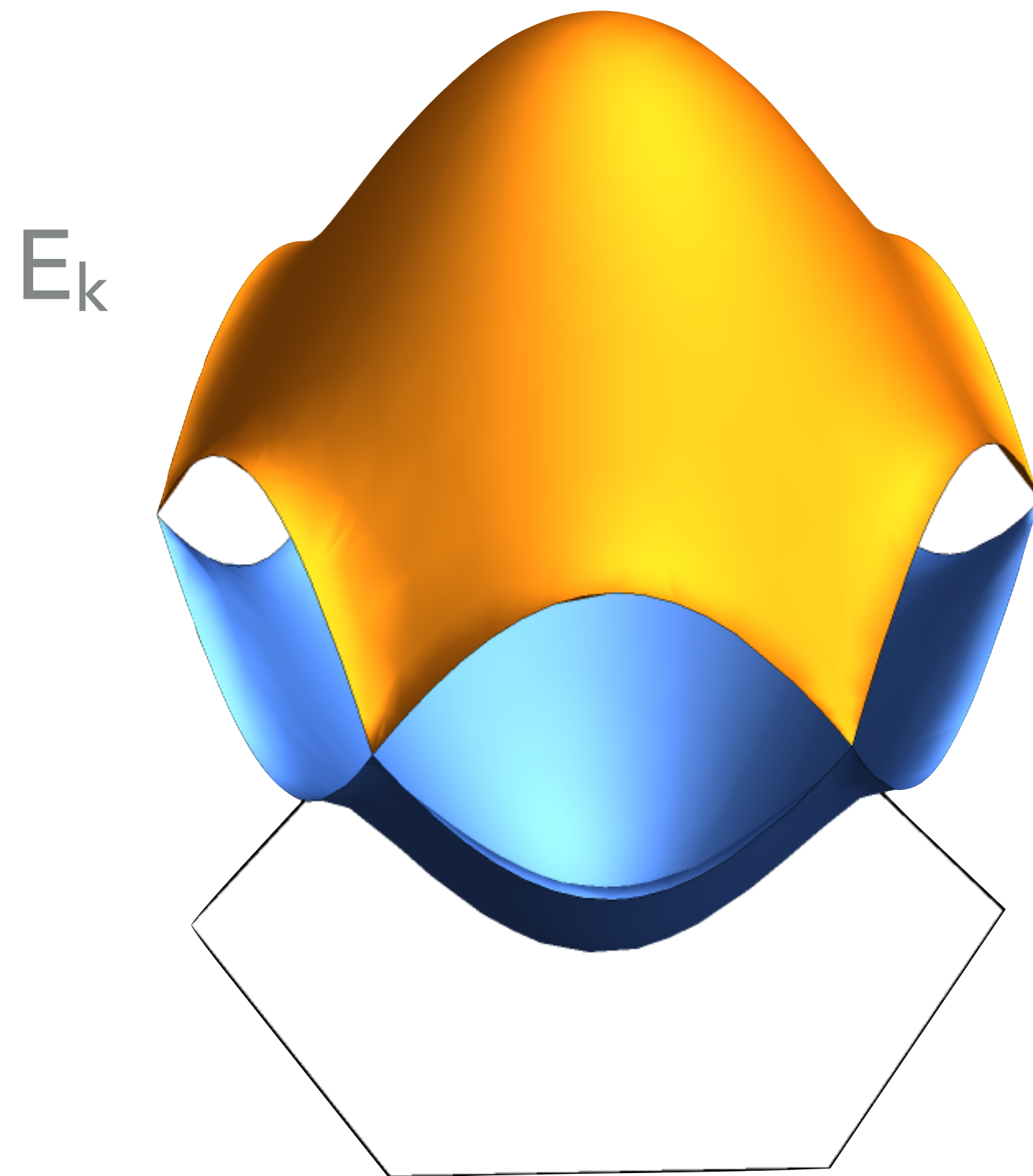
$$h_{xy} = \kappa \delta_{\langle xy \rangle}$$

Nearest-neighbor hopping

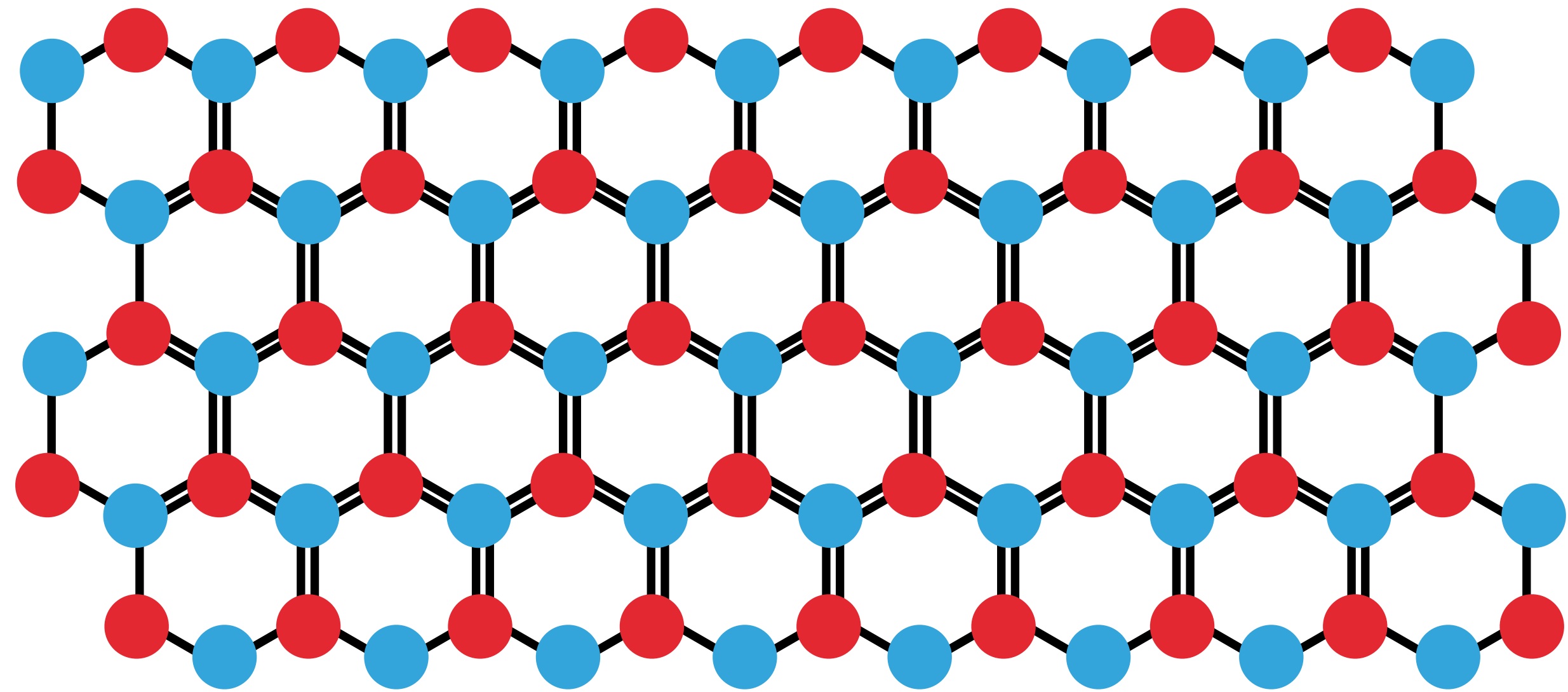
- ▶ Honeycomb:
 - ▶ triangular Bravais lattice
 - ▶ 2 sites per cell

THE TIGHT-BINDING MODEL

Wallace Phys. Rev. 71 622-634 (9 1947)



Brillouin Zone (BZ)

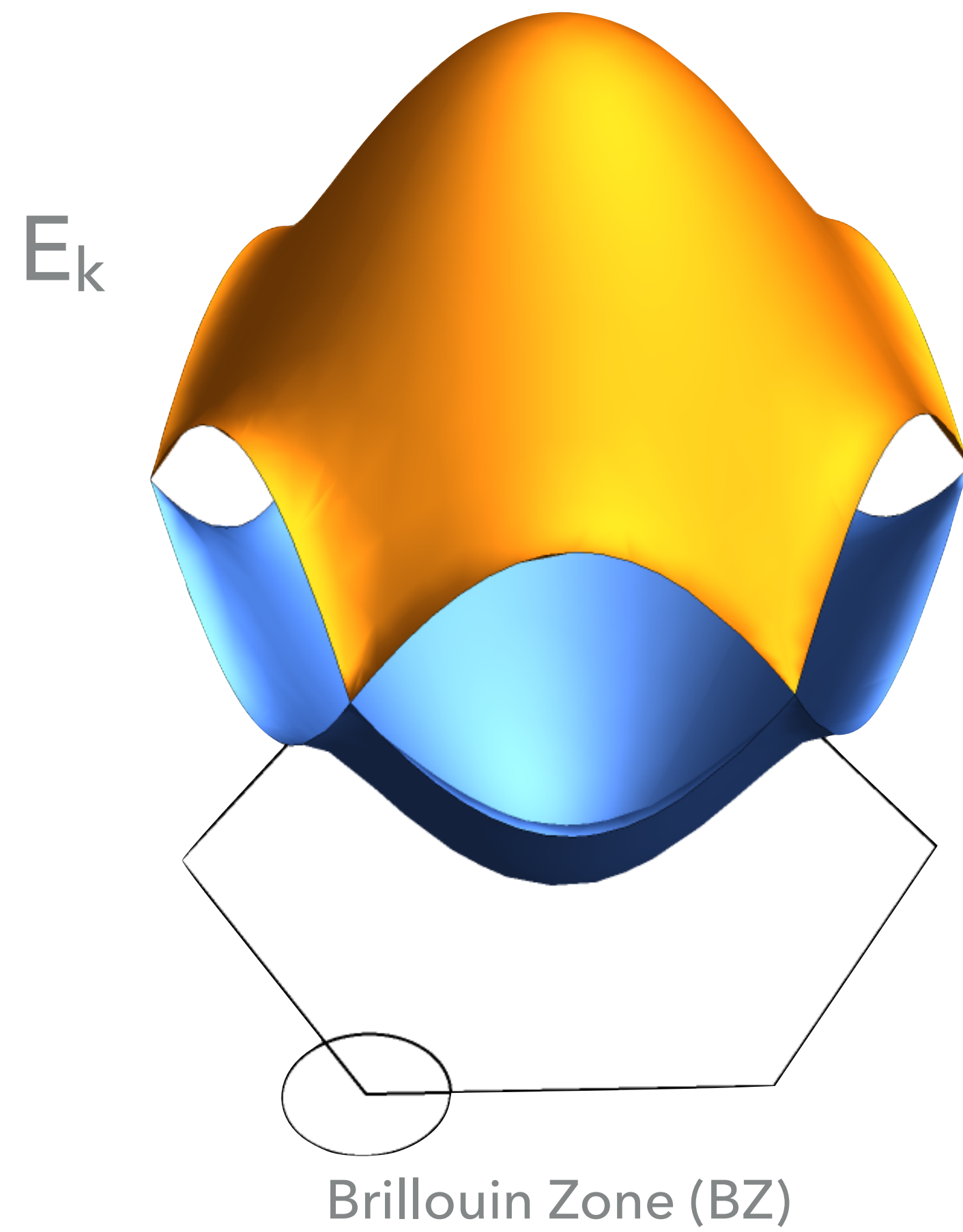


$$H_0 = - \sum_{x,y,\sigma} c_{x,\sigma}^\dagger h_{xy} c_{y,\sigma}$$

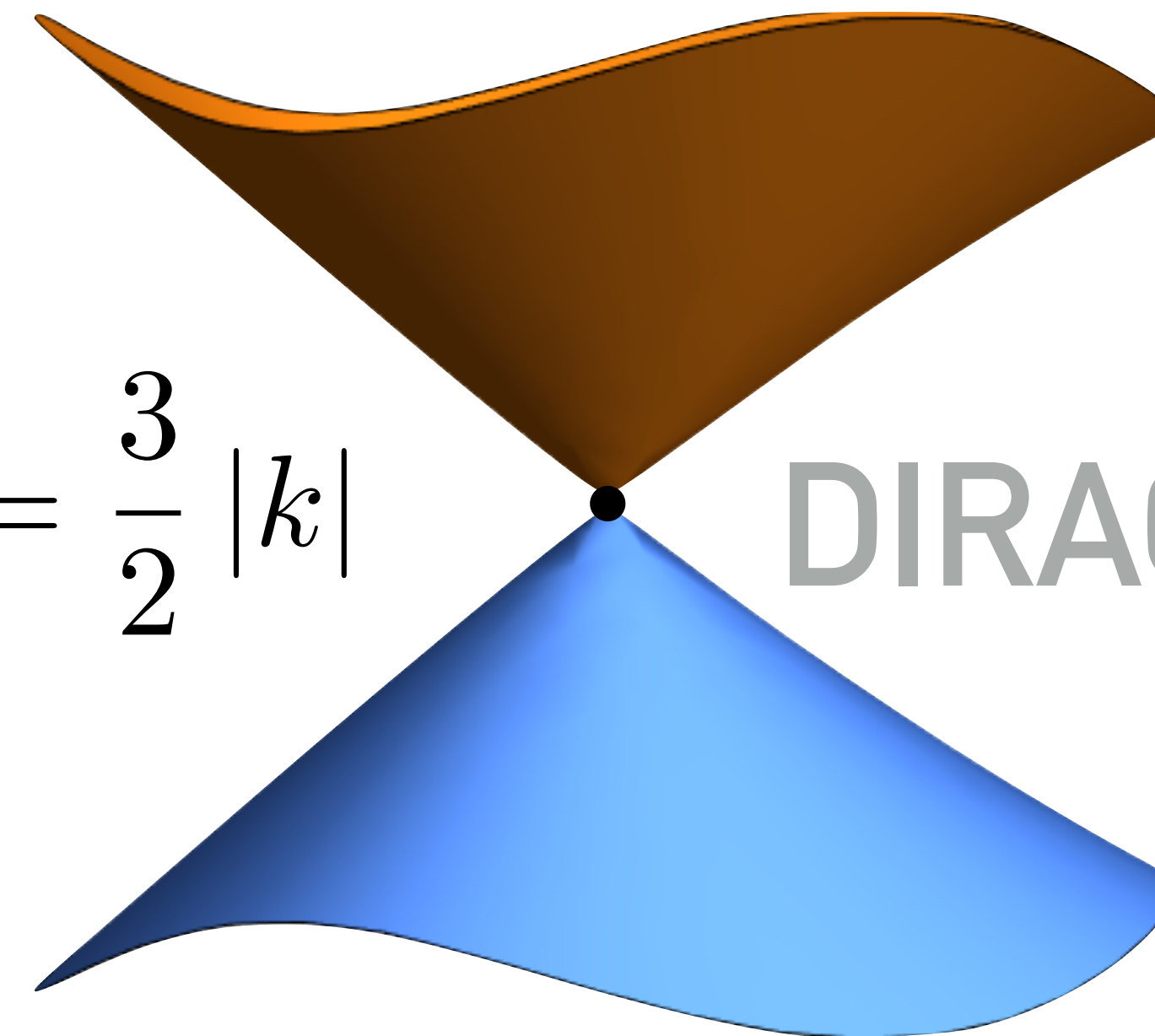
$$E_k/\kappa = \pm \sqrt{3 + 2 \left(\cos \left(\frac{3}{2}k_x + \frac{\sqrt{3}}{2}k_y \right) + \cos \left(\frac{3}{2}k_x - \frac{\sqrt{3}}{2}k_y \right) + \cos \left(\sqrt{3}k_y \right) \right)}$$

THE TIGHT-BINDING MODEL

Wallace Phys. Rev. 71 622-634 (9 1947)



$$E_k/\kappa = \frac{3}{2} |k|$$



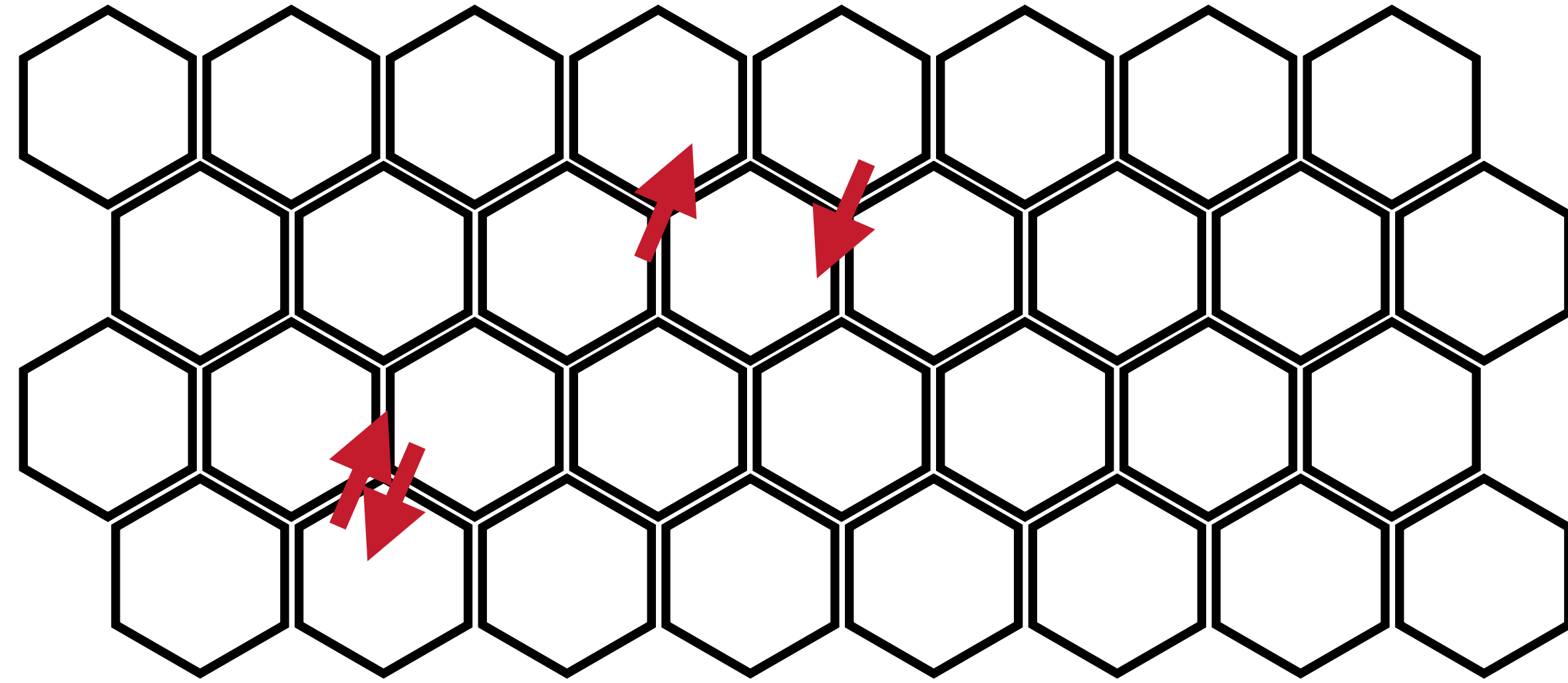
DIRAC POINT!

$$E_k/\kappa = \pm \sqrt{3 + 2 \left(\cos \left(\frac{3}{2}k_x + \frac{\sqrt{3}}{2}k_y \right) + \cos \left(\frac{3}{2}k_x - \frac{\sqrt{3}}{2}k_y \right) + \cos \left(\sqrt{3}k_y \right) \right)}$$

$$n_{x\sigma} = c_{x,\sigma}^\dagger c_{x,\sigma}$$

$$H_I = -\frac{U}{2} \sum_x (n_{x\uparrow} - n_{x\downarrow})^2$$

$U > 0$ is repulsive



Spin

$$S_x^i = \frac{1}{2} \sum_{ss'} c_{xs} \sigma_{ss'}^i c_{xs'}^\dagger$$

Charge

$$\rho_x = 1 - 2S_x^0$$

Half-filling: $\mu_\rho = 0$

$$[S_x^i, S_y^j] = i\delta_{xy} \epsilon^{ijk} S_x^k$$

THE HUBBARD MODEL

Néel Ann. de Phys. 12 (3): 137-198 (1948) 10.1051/anphys/194812030137

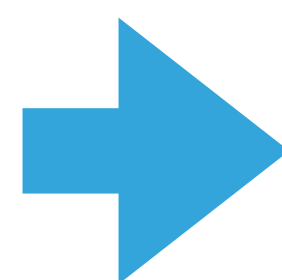
$$n_{x\sigma} = c_{x,\sigma}^\dagger c_{x,\sigma}$$

$$H_I = -\frac{U}{2} \sum_x (n_{x\uparrow} - n_{x\downarrow})^2$$

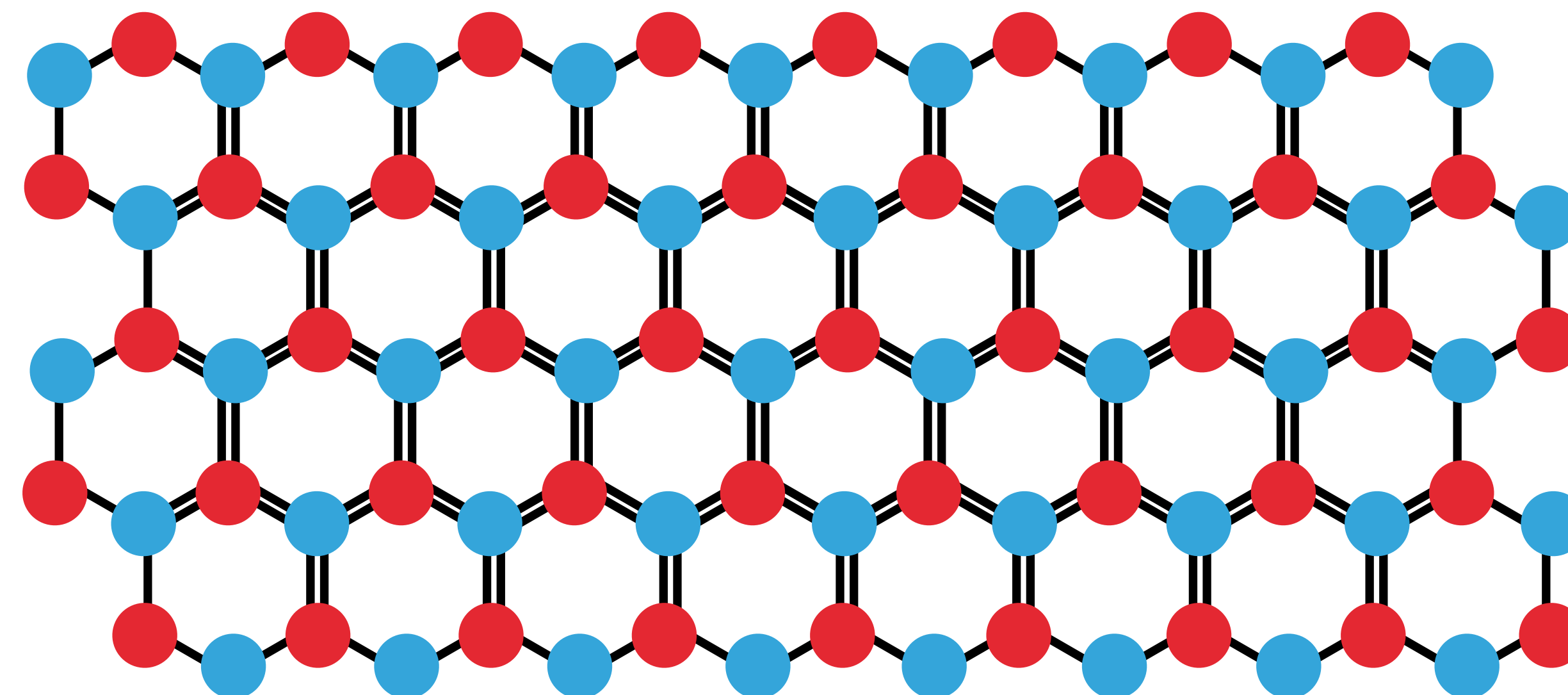
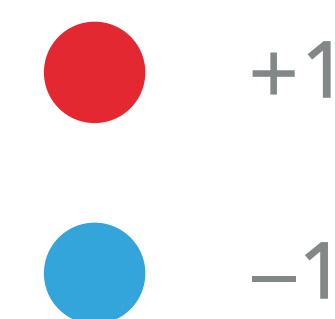


U/κ very large

$$H_{\text{eff}} \sim +\frac{\kappa^2}{U} \sum_{\langle x,y \rangle} \vec{S}_x \cdot \vec{S}_y$$



alternating, checkerboard,
or *Néel* order



antiferromagnetic (Mott) insulator! (AFMI)

THE HUBBARD MODEL

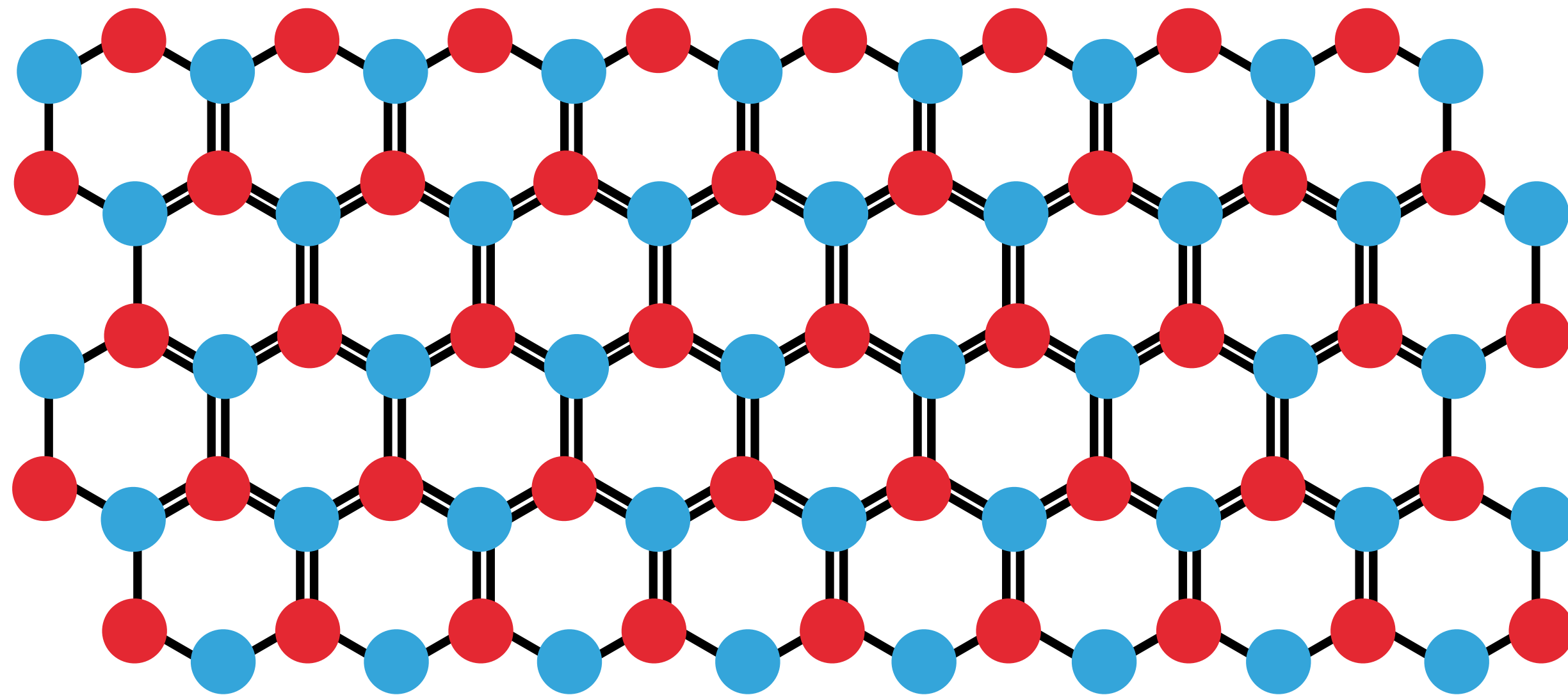
Hubbard, Proc. Roy. Soc. A 1963 10.1098/rspa.1963.0204

Bipartite lattice!

$$a_x = c_{x\uparrow} \quad a_x^\dagger = c_{x\uparrow}^\dagger$$

$$b_x = (-1)^x c_{x\downarrow} \quad b_x^\dagger = (-1)^x c_{x\downarrow}^\dagger$$

$$H = \sum_{xy} (a_x^\dagger h_{xy} a_y + b_x^\dagger h_{xy} b_y) \\ + \frac{U}{2} \sum_x (n_x^a - n_x^b)^2$$



Bipartite lattice!

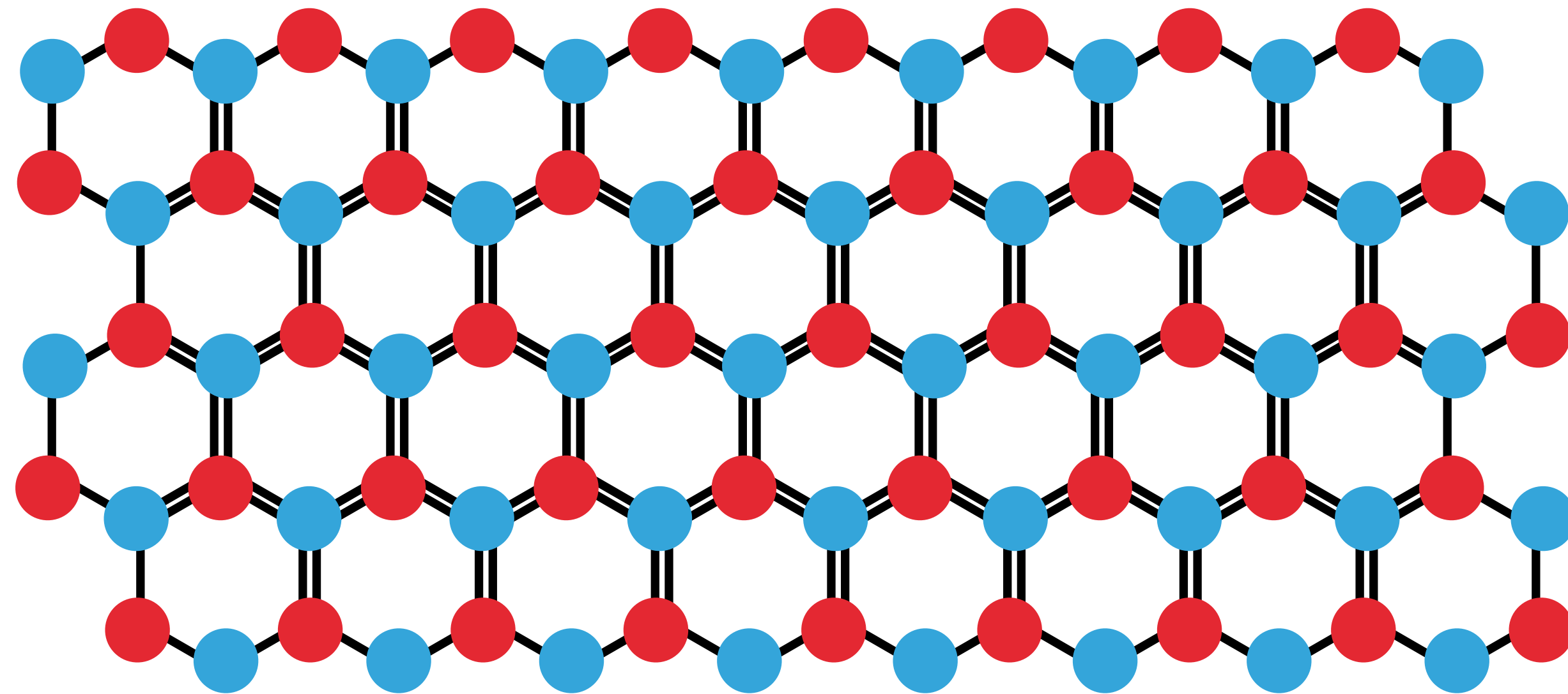
$$a_x = c_{x\uparrow} \quad a_x^\dagger = c_{x\uparrow}^\dagger$$

$$b_x = (-1)^x c_{x\downarrow} \quad b_x^\dagger = (-1)^x c_{x\downarrow}^\dagger$$

$$H = \sum_{xy} (a_x^\dagger h_{xy} a_y + b_x^\dagger h_{xy} b_y)$$

$$+ \frac{U}{2} \sum_x (n_x^a - n_x^b)^2$$

$$\left(\frac{1}{2} \sum_{xy} \rho_x V_{xy} \rho_y \right)$$



Half filling: $\langle \rho \rangle = 0$ $\rho_x = n_x^a - n_x^b$

$V =$ screened Coulomb

Ulybyshev, Buividovich, Katsnelson, Polikarpov PRL 111 056801

(2013) 2010.1103/PhysRevLett.111.056801 1403.3620

$V =$ on-site + nearest neighbor

Buividovich, Smith, Ulybyshev, von Smekal LATTICE2016 244

10.22323/1.256.0244 1610.09855

- ▶ + staggered mass
- ▶ + chemical potential
- ▶ + ...



**IT HAS YET TO RECEIVE ADEQUATE
MATHEMATICAL TREATMENT, AND ONE HAS
TO RESORT TO THE INDIGNITY OF
NUMERICAL SIMULATIONS TO SETTLE EVEN
THE SIMPLEST QUESTIONS ABOUT IT.**

Philip W. Anderson

Nobel Lecture

8 December 1977

The
COMIC ADVENTURES
of
OLD MOTHER HUBBARD
and
HER DOG.



Sarah Catherine Martin CC BY-SA 4.0

Published June 1-1805, by J Harris, Successor to E Newbery, Corner of S^t. Pauls Church Yard.

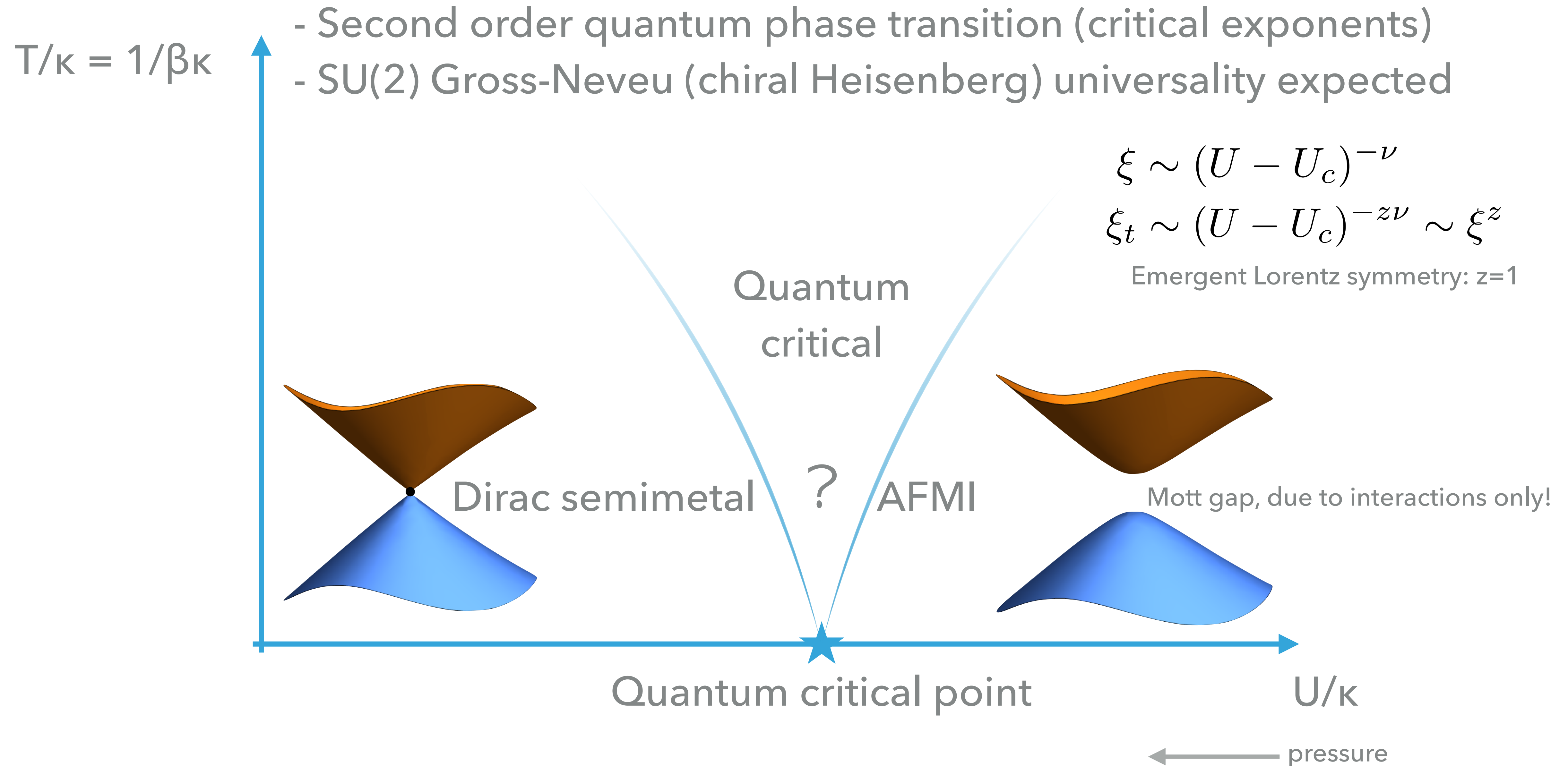
THE HUBBARD MODEL

**SIGN-PROBLEM-FREE
RESULTS**

PHASE DIAGRAM

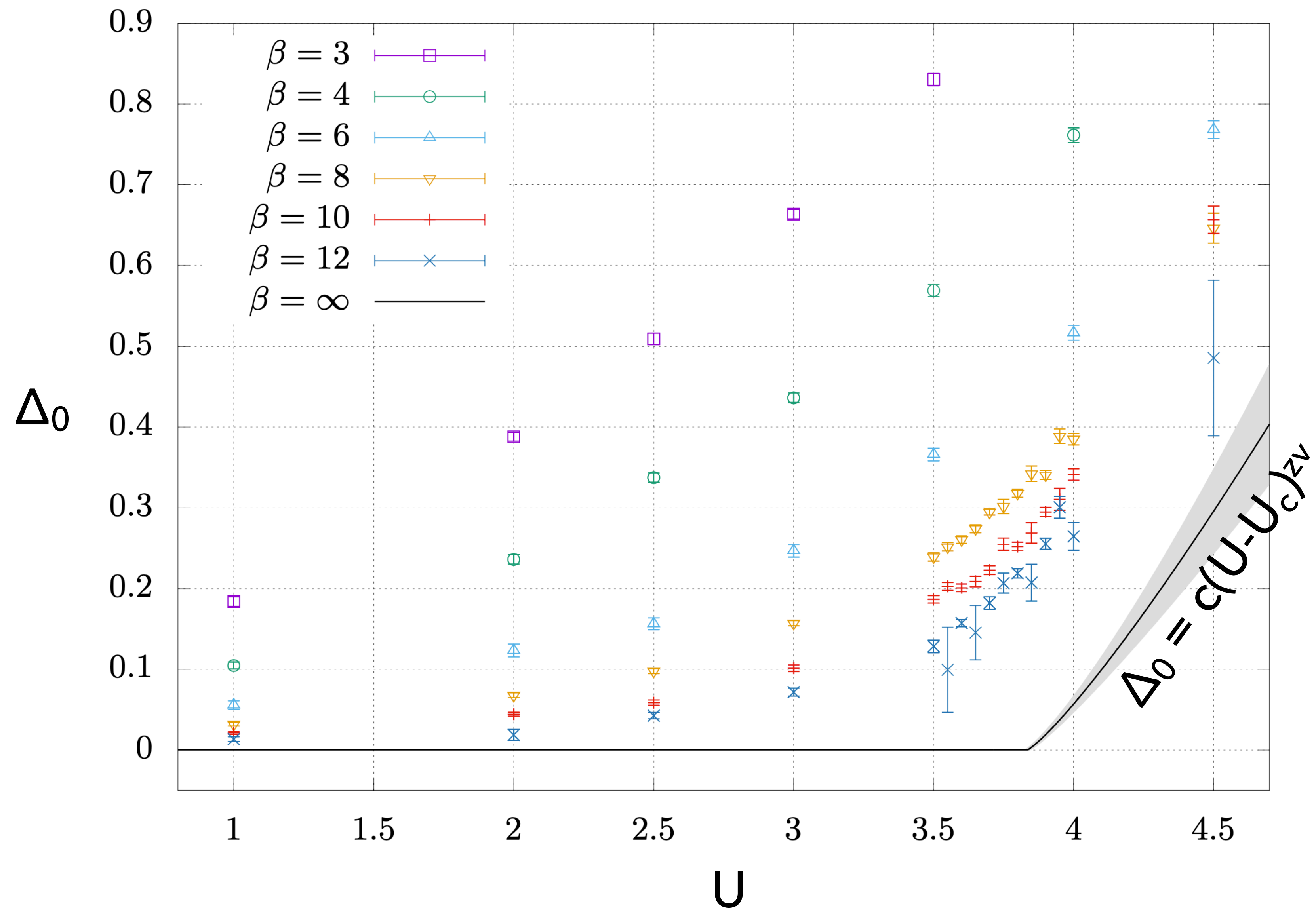
Herbut, Juričić, and Roy, PRB 79 085116 (2009)

0811.0610 10.1103/PhysRevB.79.085116



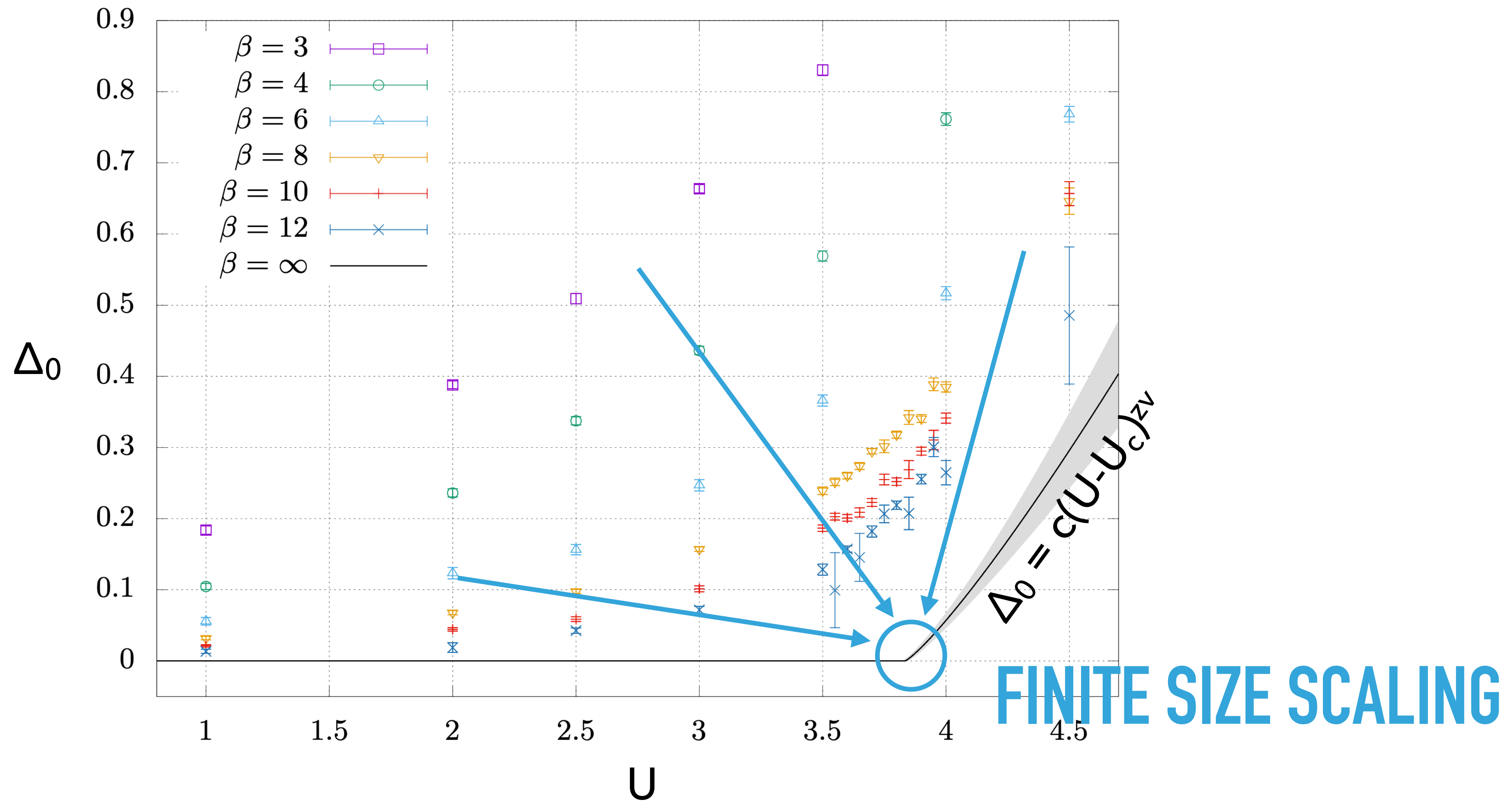
SEMIMETAL / AFMI TRANSITION

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112



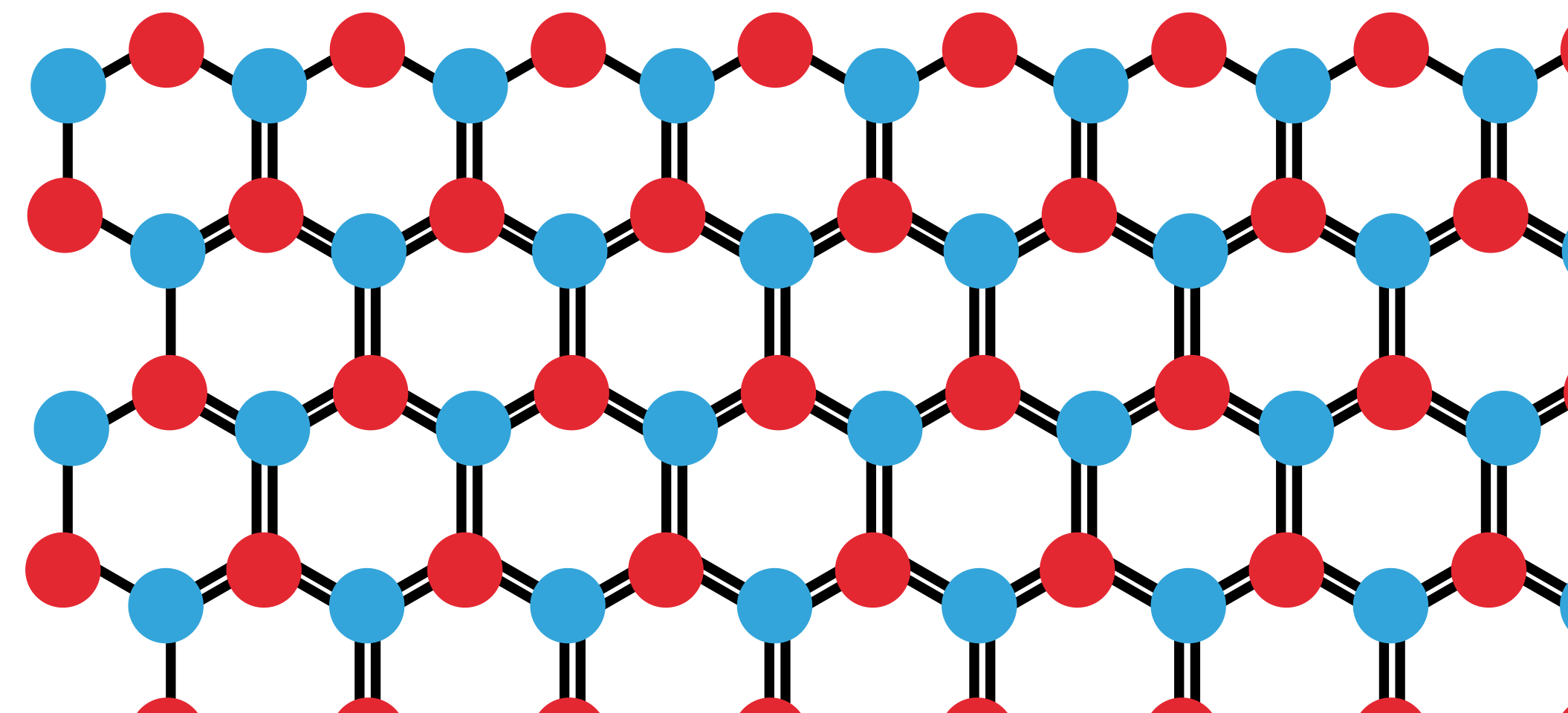
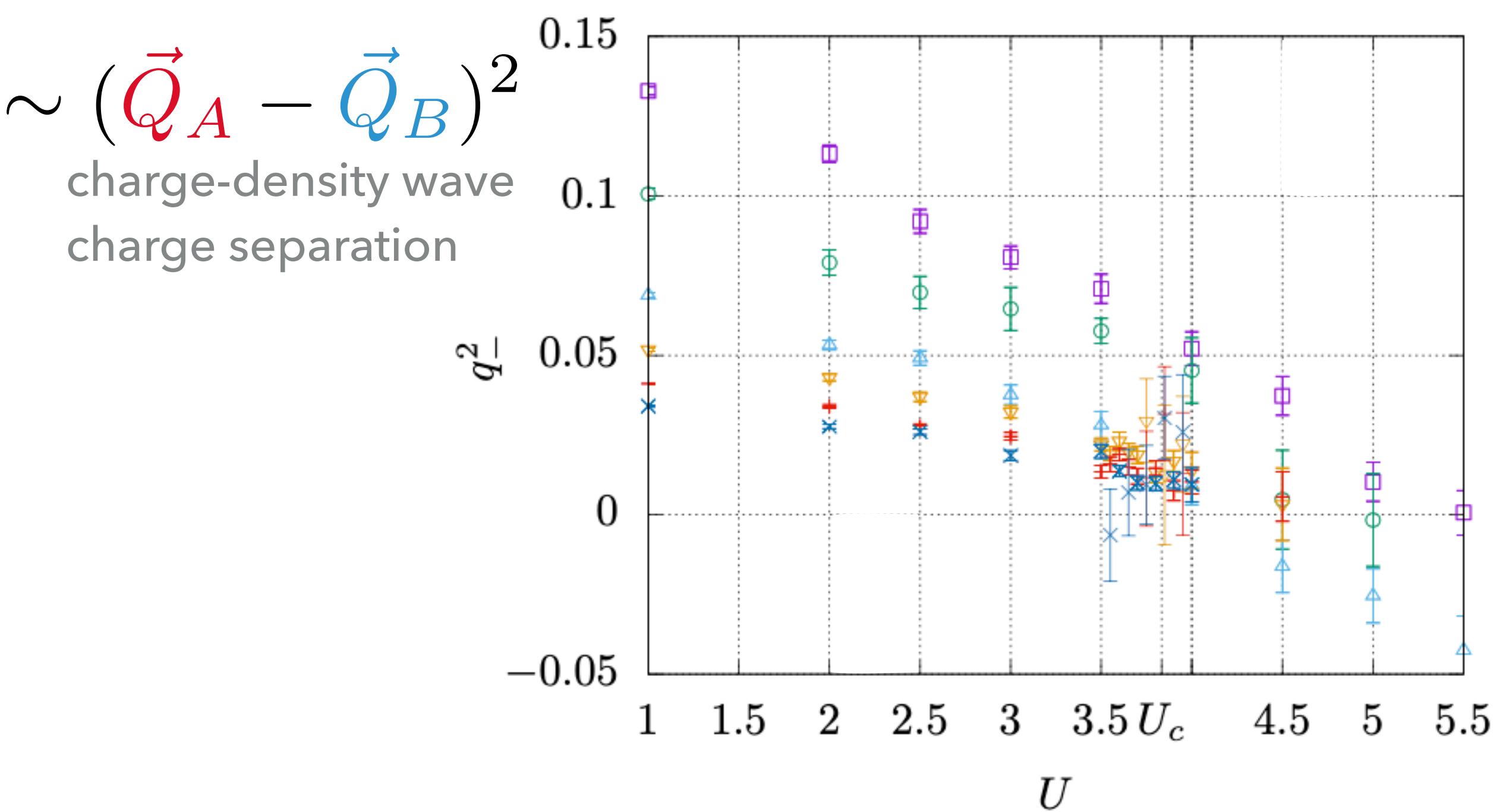
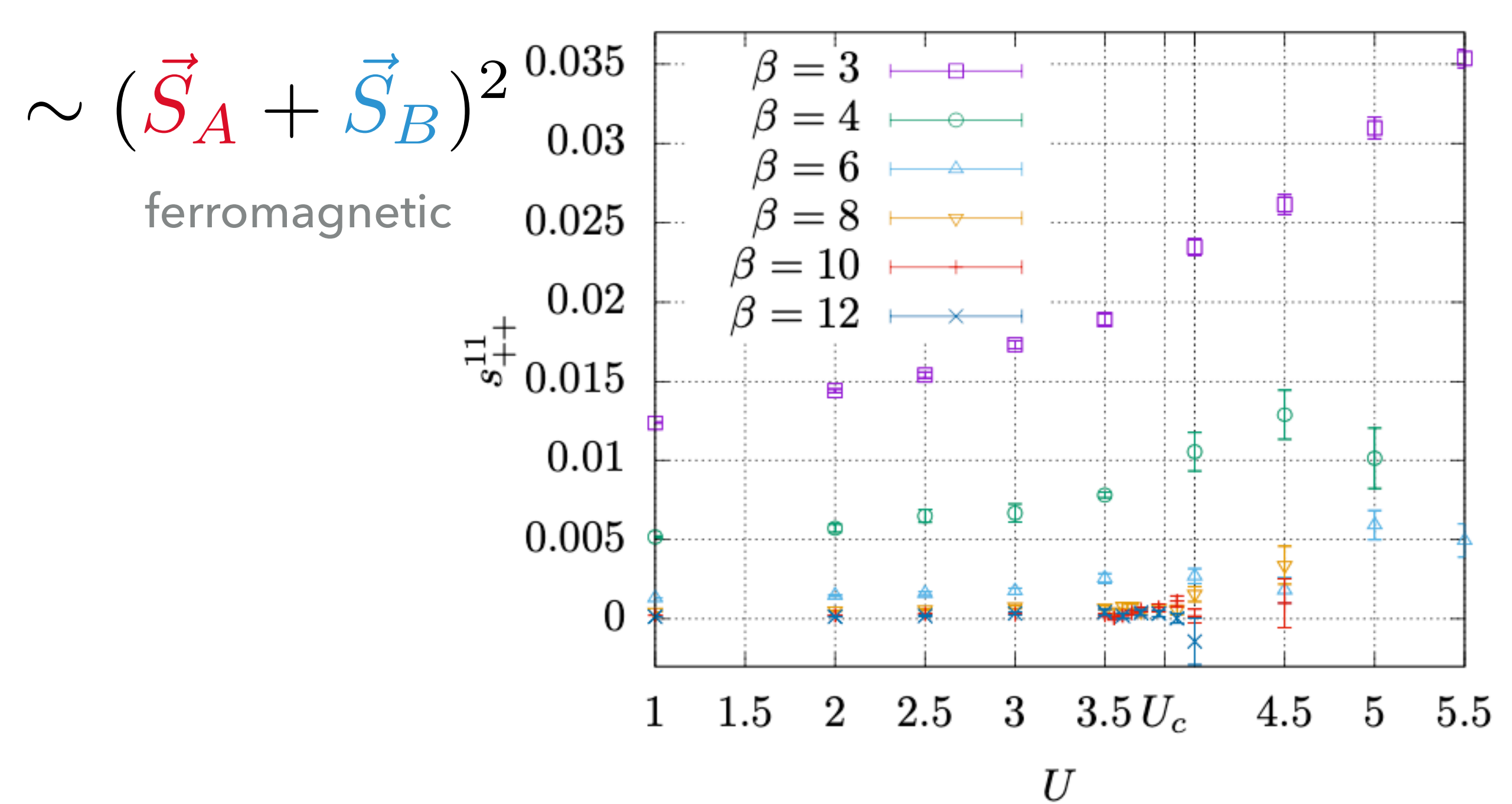
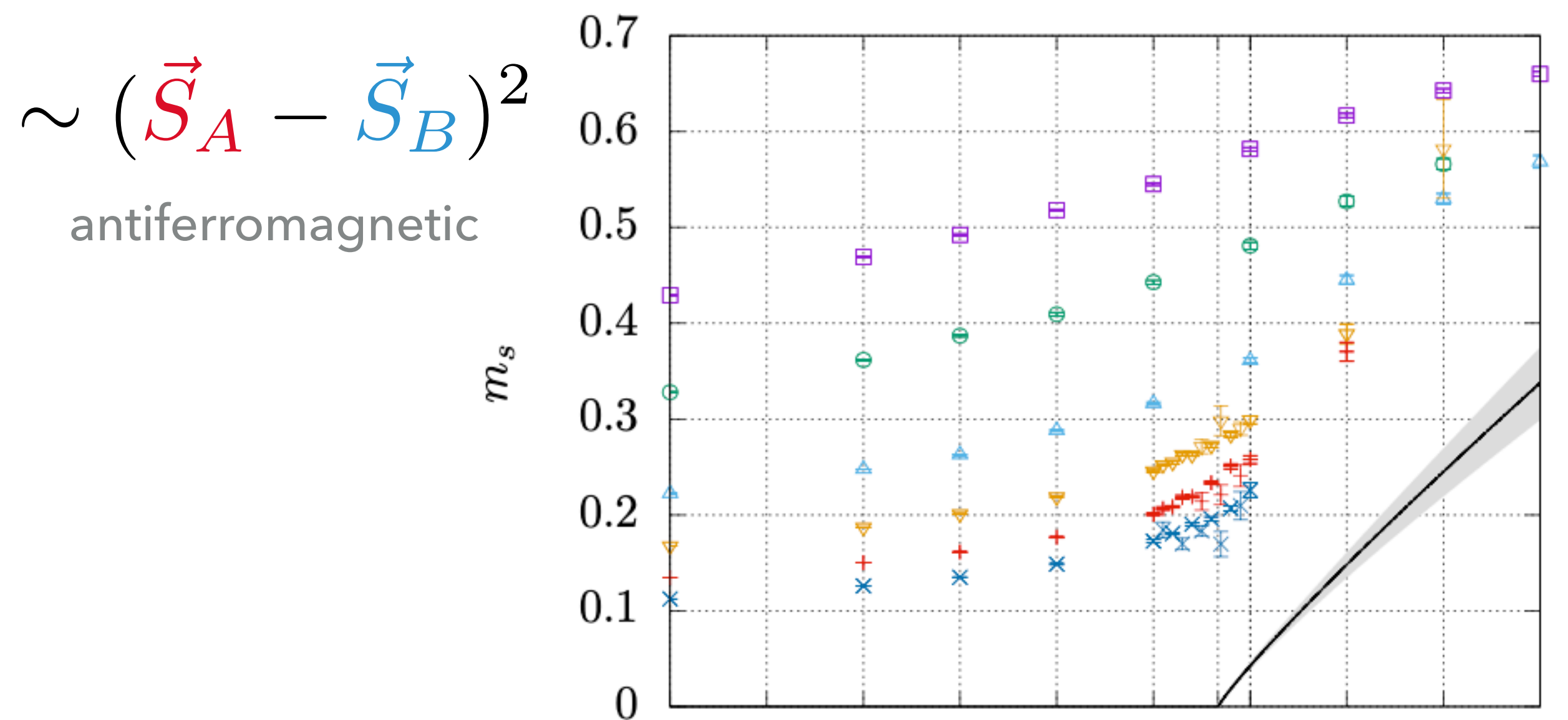
SEMIMETAL / AFMI TRANSITION

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112

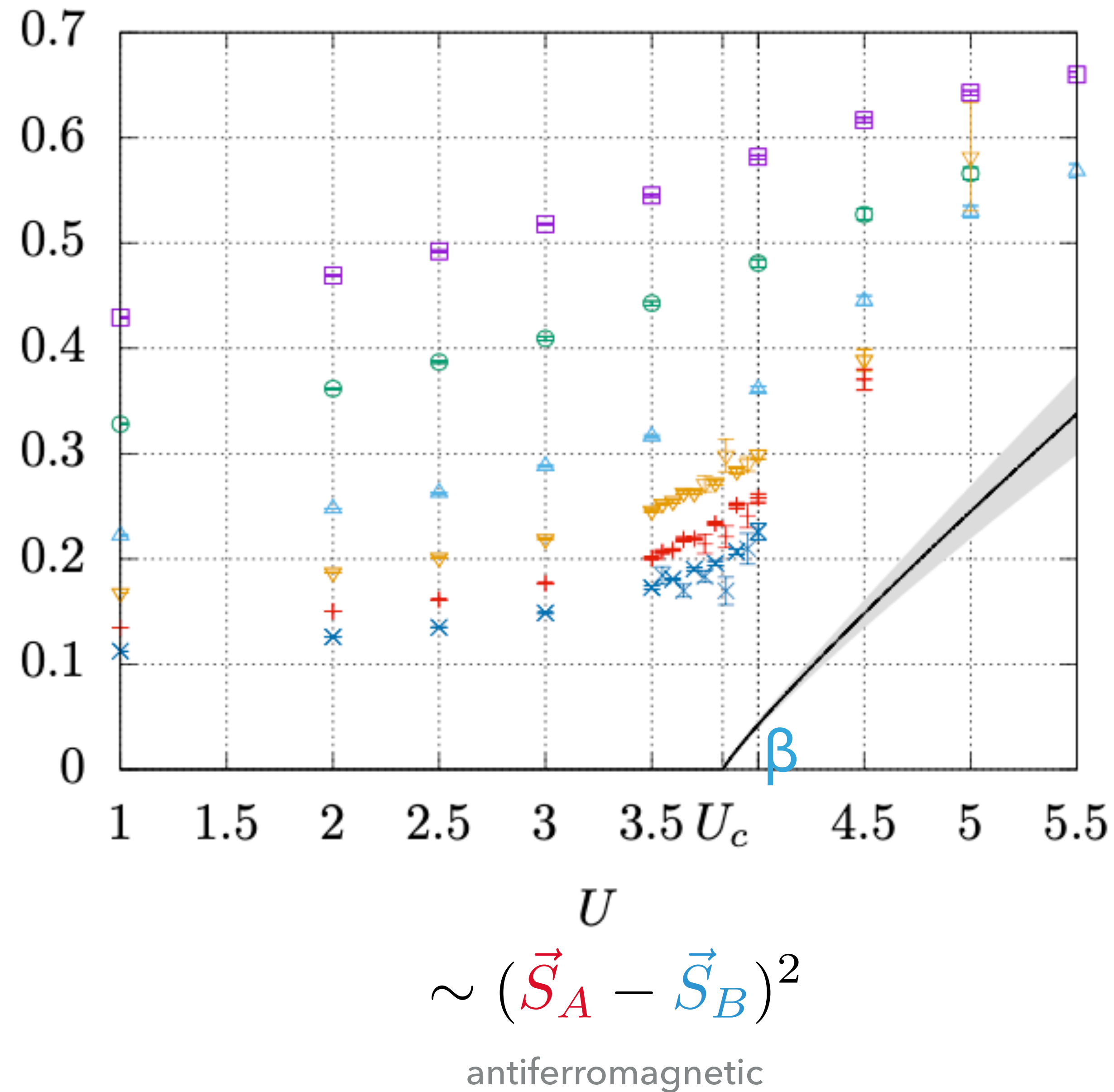
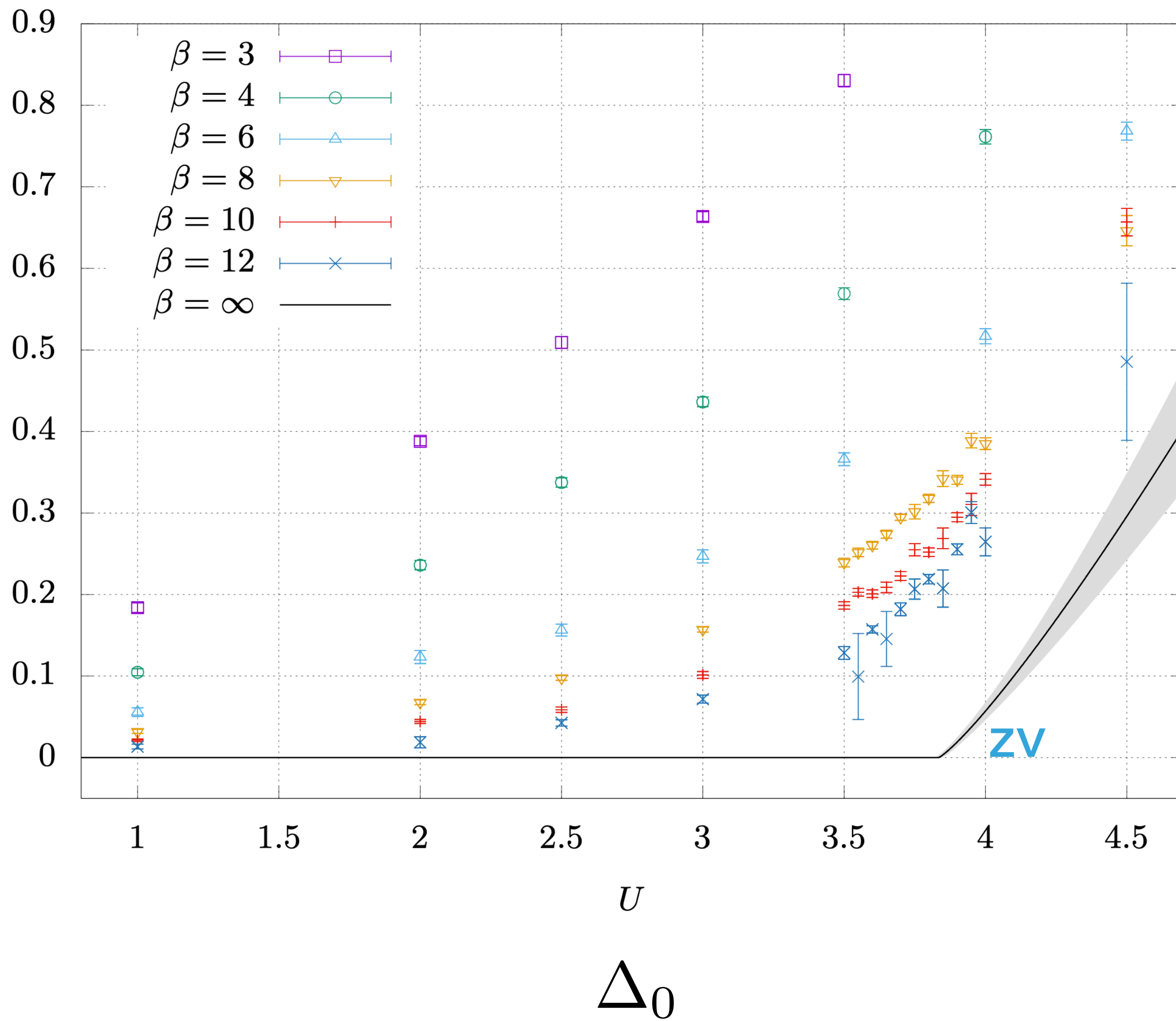


ZERO-TEMPERATURE LIMIT

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2105.06936

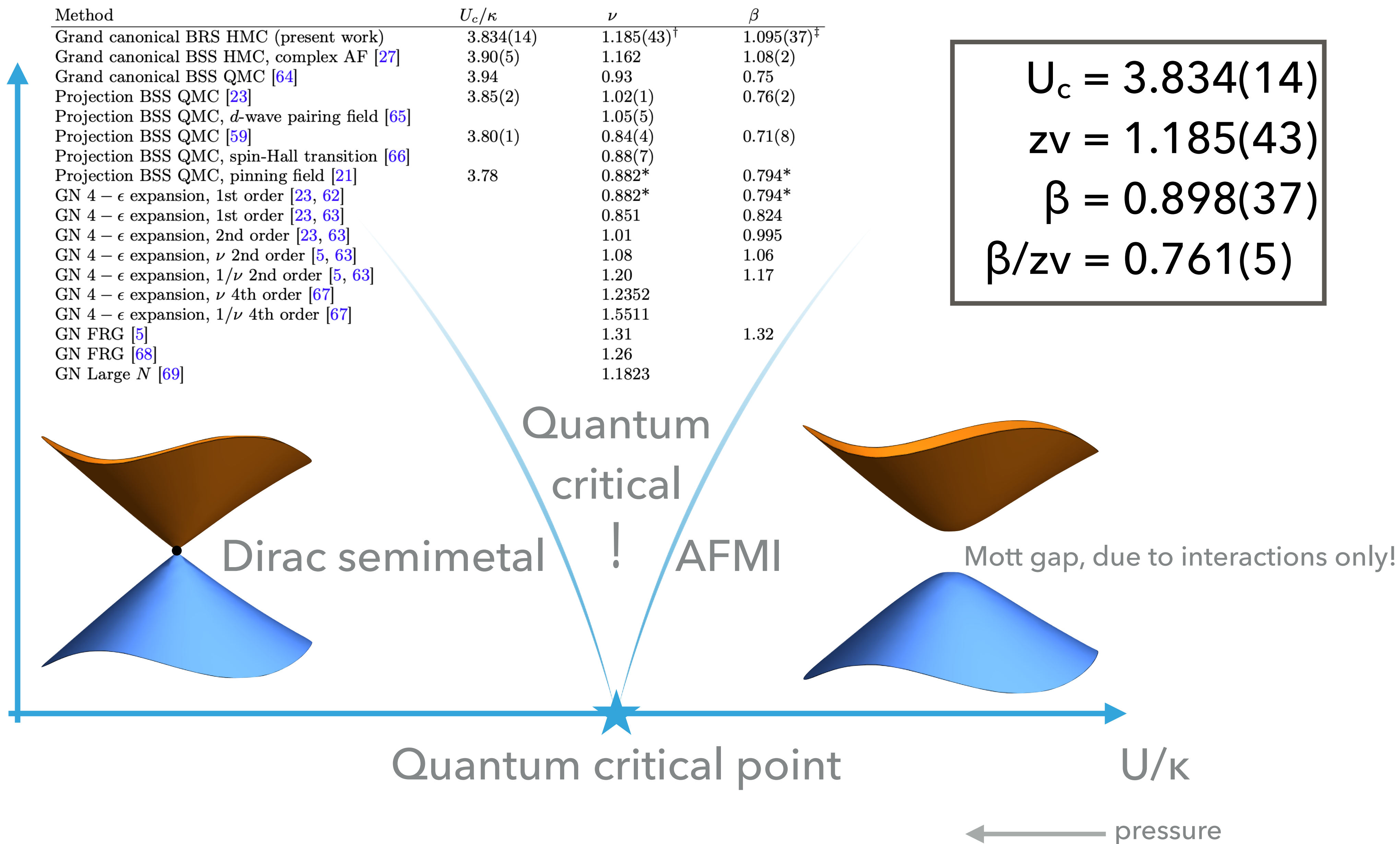


THE PHASE TRANSITION

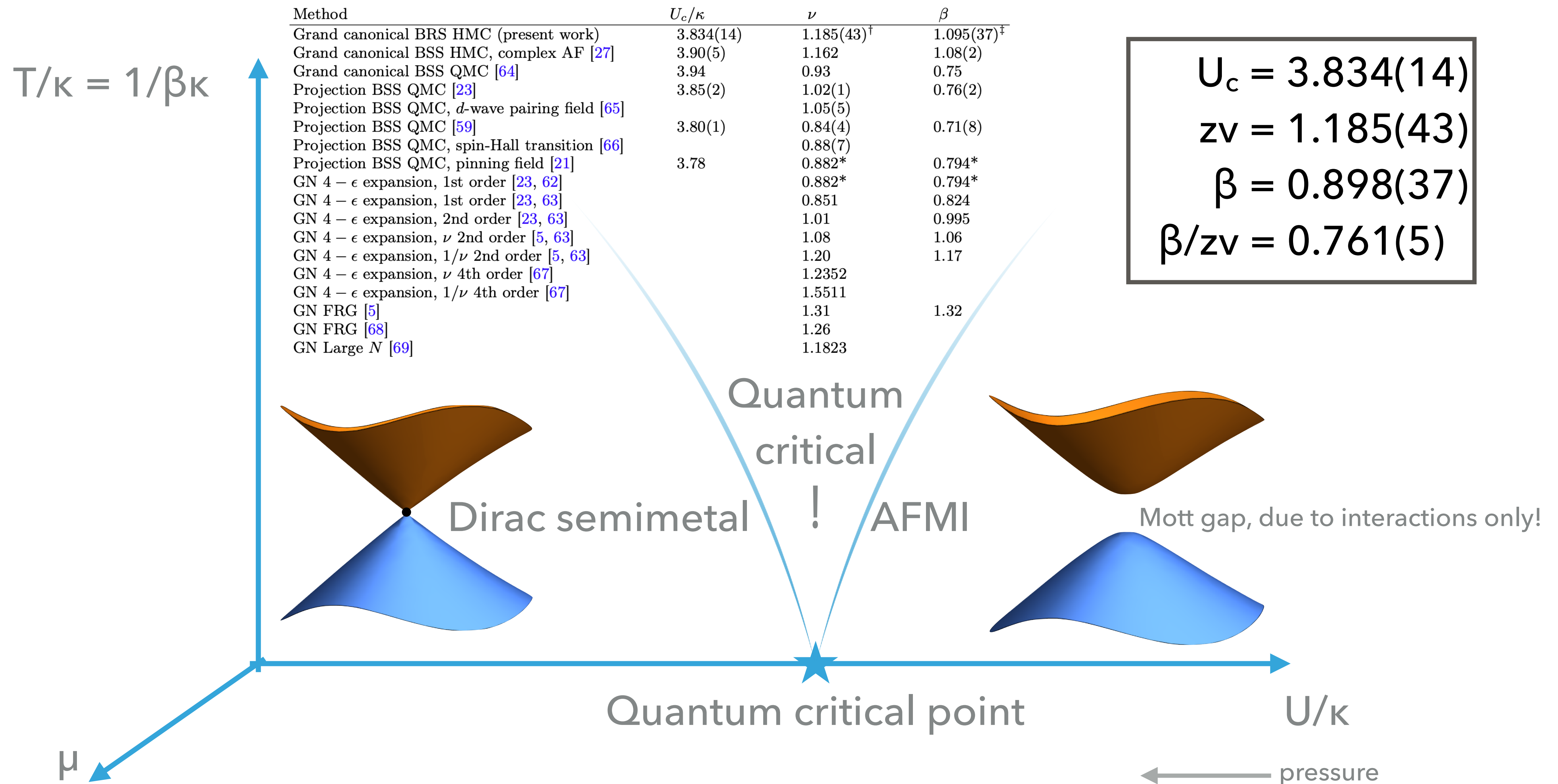


PHASE DIAGRAM

$$T/\kappa = 1/\beta\kappa$$



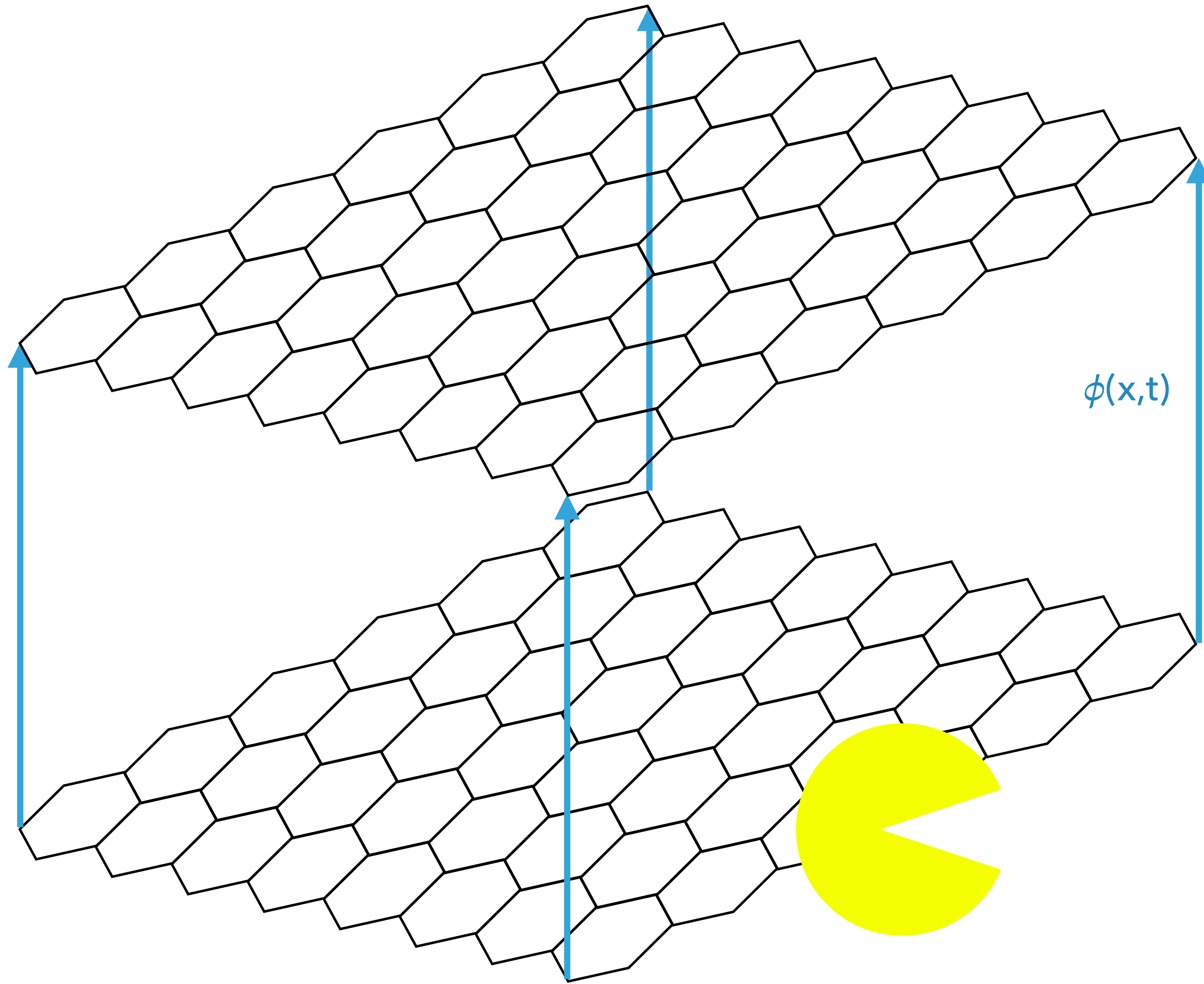
BEYOND HALF FILLING?





SIGN PROBLEMS

IMPORTANCE SAMPLING THE PATH INTEGRAL



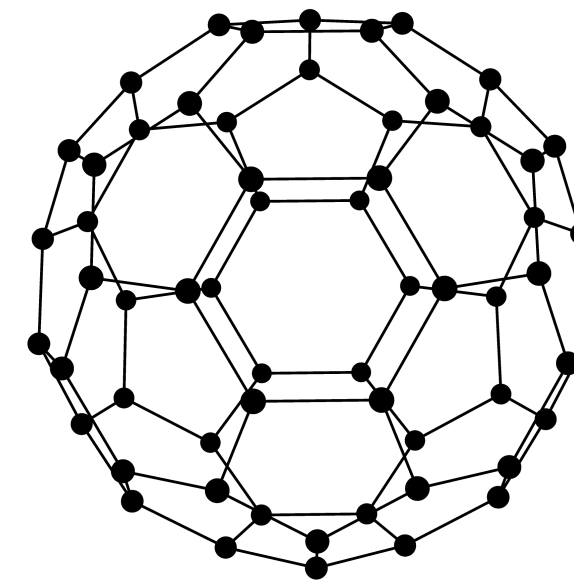
$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability?

$$\mathcal{E} \propto \int \mathcal{D}\phi \det (M[+\phi, +h, +\mu] M[-\phi, -h, -\mu]) e^{-\frac{1}{2} \phi V^{-1} \phi}$$

Trade for $M^\dagger[+\phi, \dots]$

Flip on a bipartite lattice



No problem in the bipartite $\mu=0$ case.

$$\langle \mathcal{O} \rangle = \frac{\langle \mathcal{O} e^{-iS^I} \rangle_R}{\langle e^{-iS^I} \rangle_R}$$

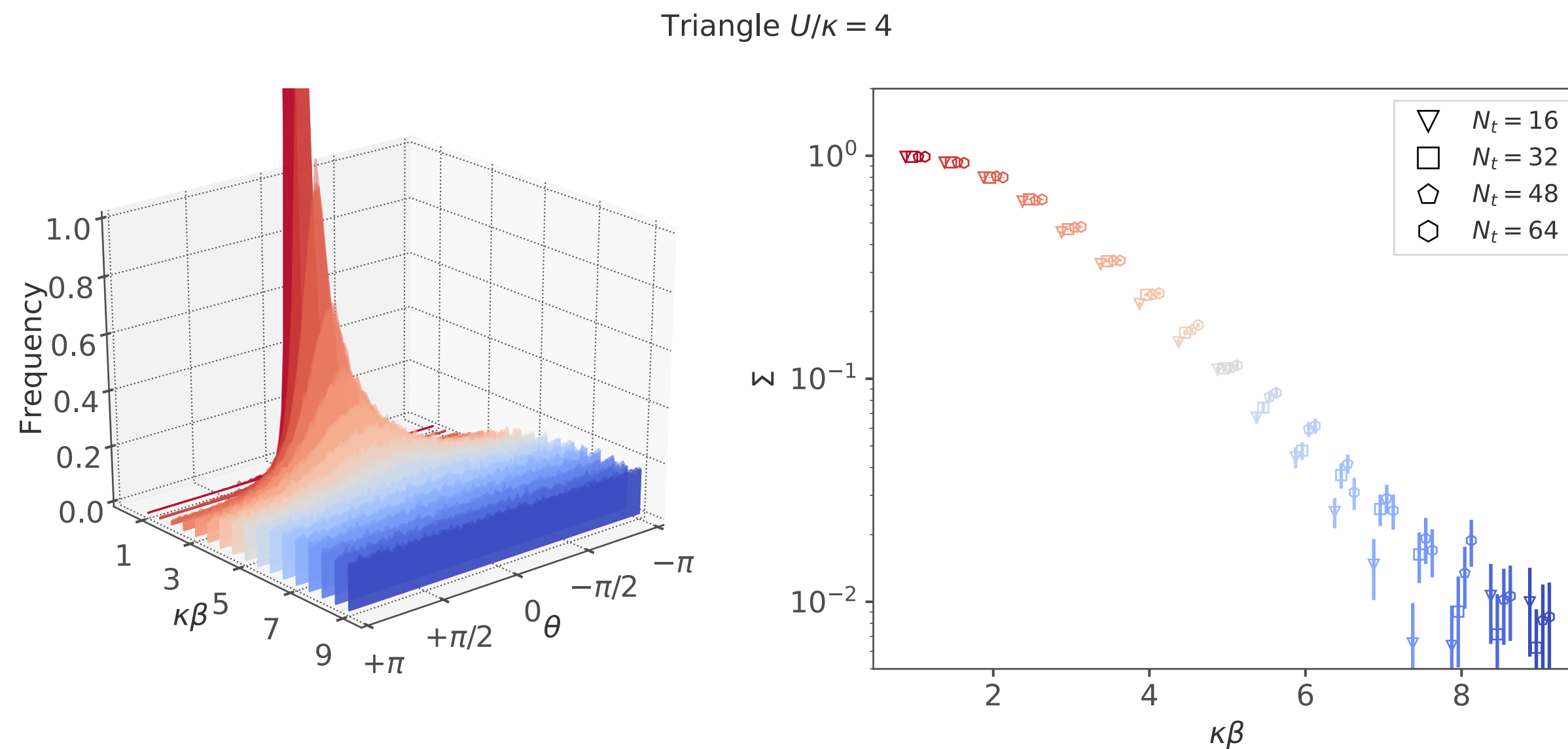
The average sign, or *statistical power* $\Sigma = \left| \langle e^{-iS^I} \rangle_R \right|$

COMPLEX WEIGHTS LEAD TO SIGN PROBLEMS

The average sign, or *statistical power* Σ

$$\Sigma = \left| \langle e^{-iS^I} \rangle_R \right|$$

is the ratio of two partition functions \sim differences of free energies
so is extensive in spacetime volume \rightarrow exponentially decays



$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability?

Generate an *ensemble* for fixed U, β, N_t, \dots

$$\{\phi_1, \phi_2, \dots, \phi_N\}$$

Hybrid Monte Carlo

Estimate any observable

$$\langle \mathcal{O} \rangle = \frac{\langle \mathcal{O} e^{-iS^I} \rangle_R}{\langle e^{-iS^I} \rangle_R}$$

Importance sample according to the real part
of the weight

SIGN OPTIMIZATION

Pick a simple contour deformation with few parameters λ .

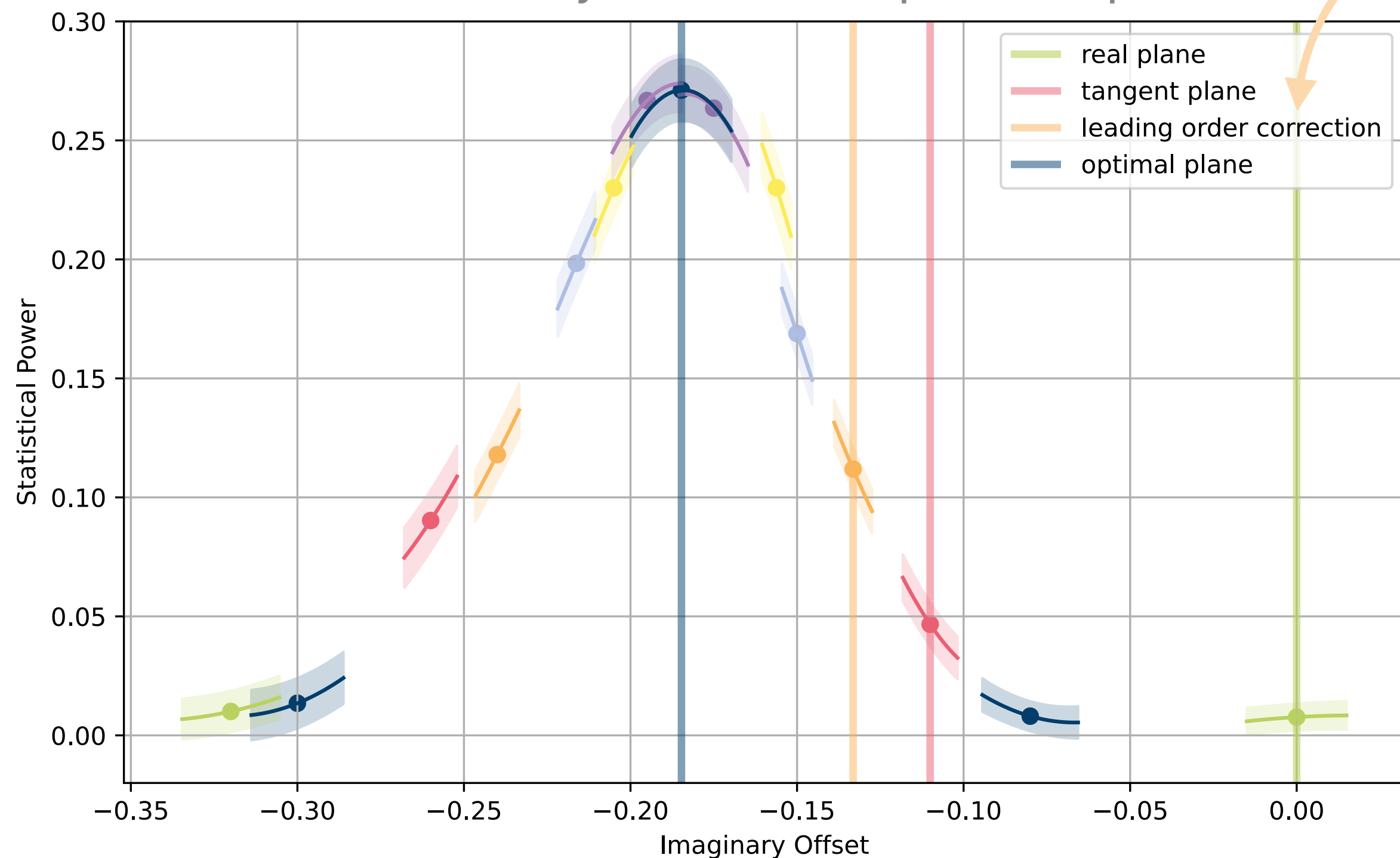
Example: the simplest imaginable deformation; integrate on a plane parallel to \mathbb{R}^N offset by a constant imaginary piece

Gradient of Σ is sign-problem free!

$$\nabla_{\lambda} \log \Sigma(\lambda) = \left\langle \nabla_{\lambda} \mathcal{S}_R - \text{ReTr} [J^{-1} \nabla_{\lambda} J] \right\rangle_R$$

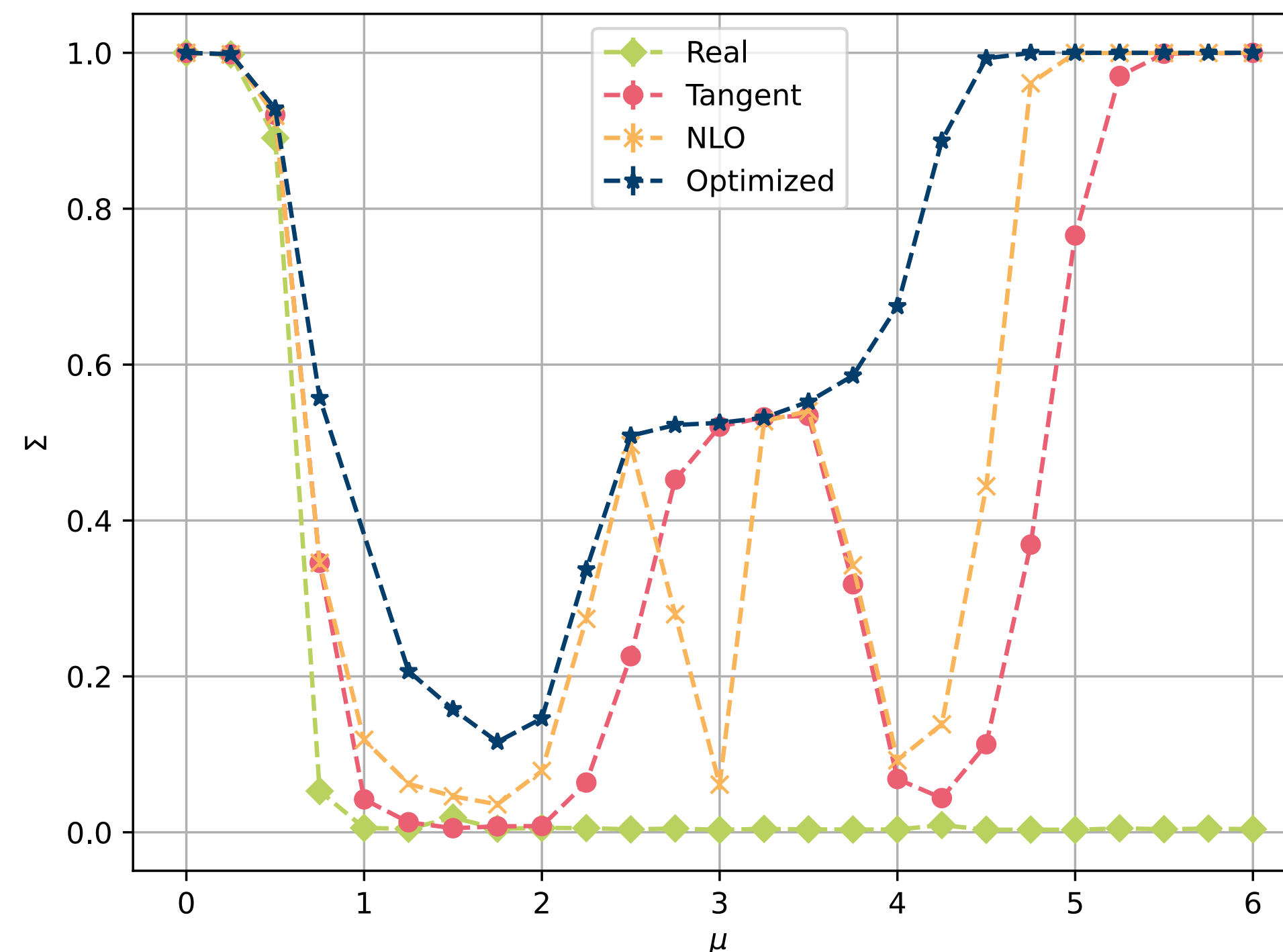
Alexandru, Bedaque, Lamm + Lawrence, PRD 97 094510 (2018)
1804.00697 10.1103/PhysRevD.97.094510

8-site honeycomb $N_t=16$ $\beta=8$ $U=2$ $\mu=1$



Gäntgen et al. 2208.XXXX

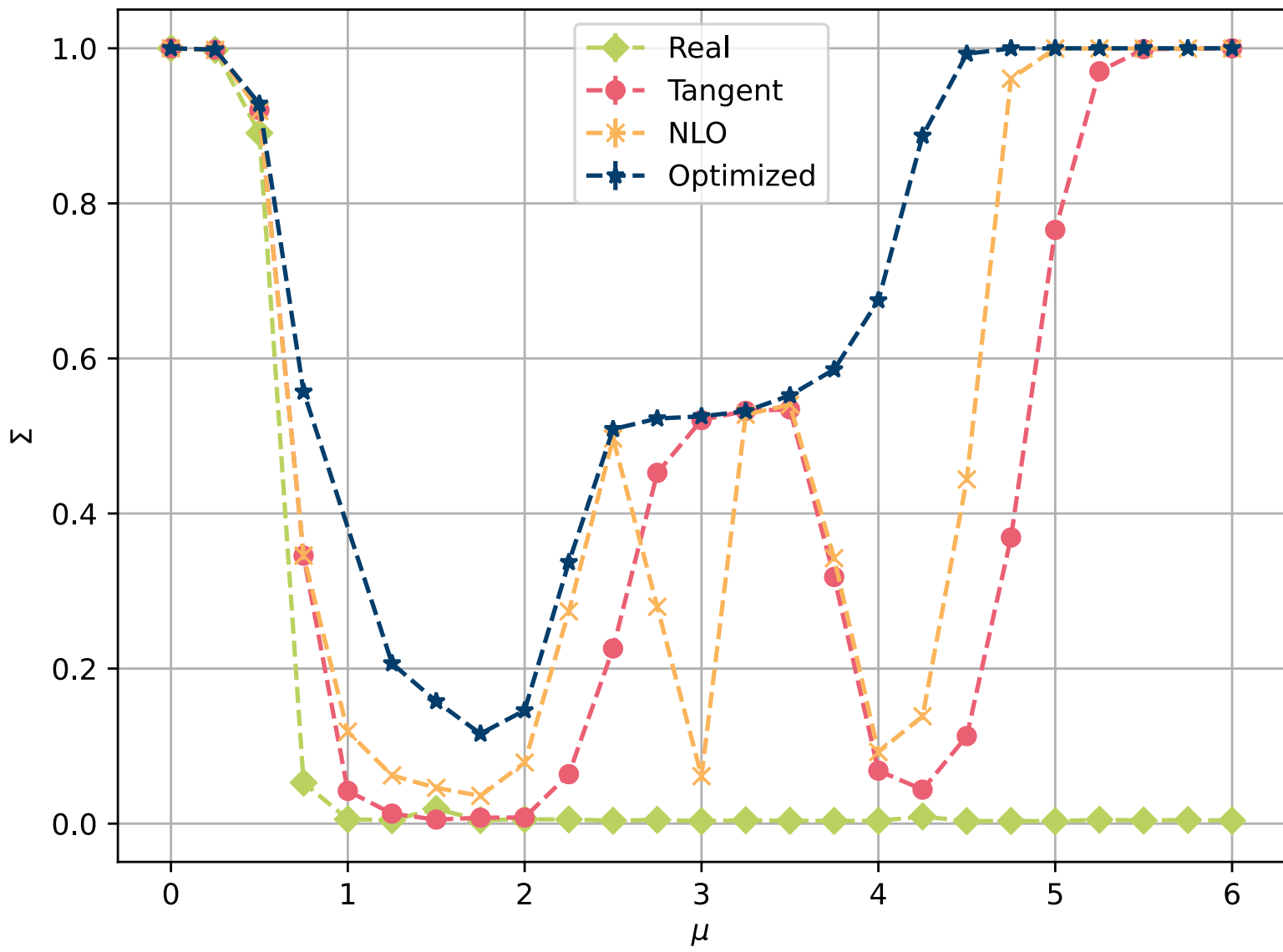
Sort of like the one-loop effective potential



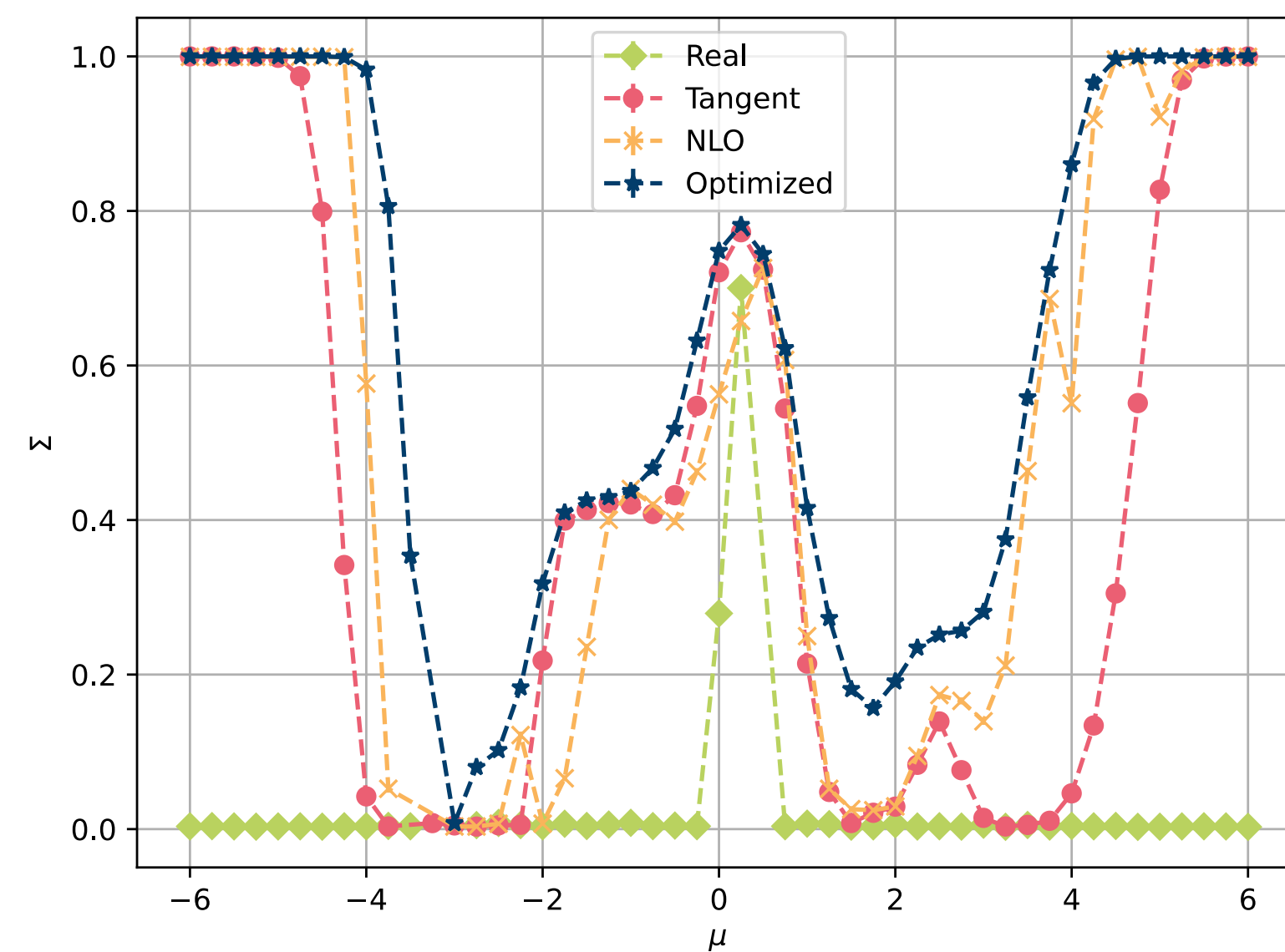
Unsupervised learning?

RESULTS

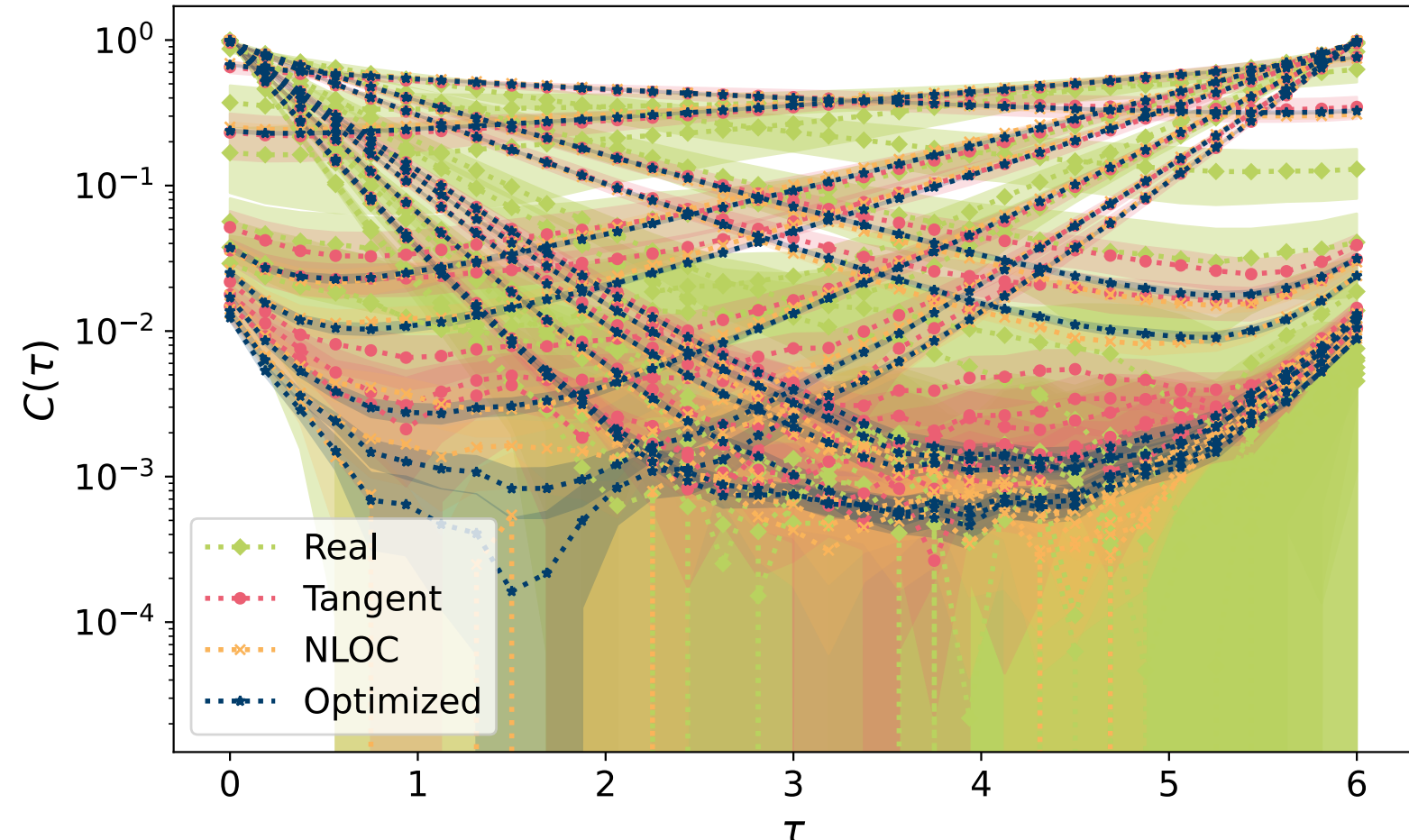
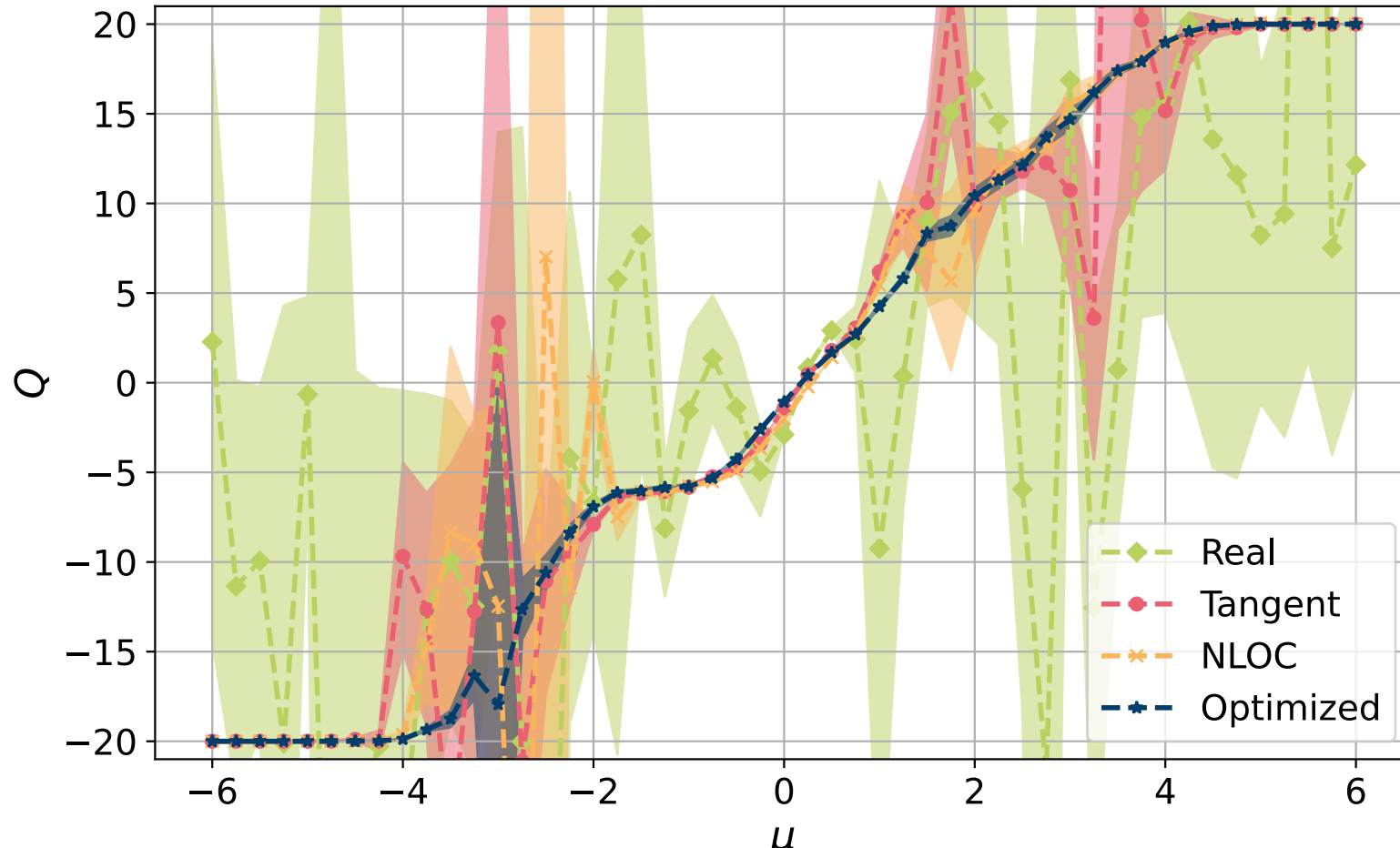
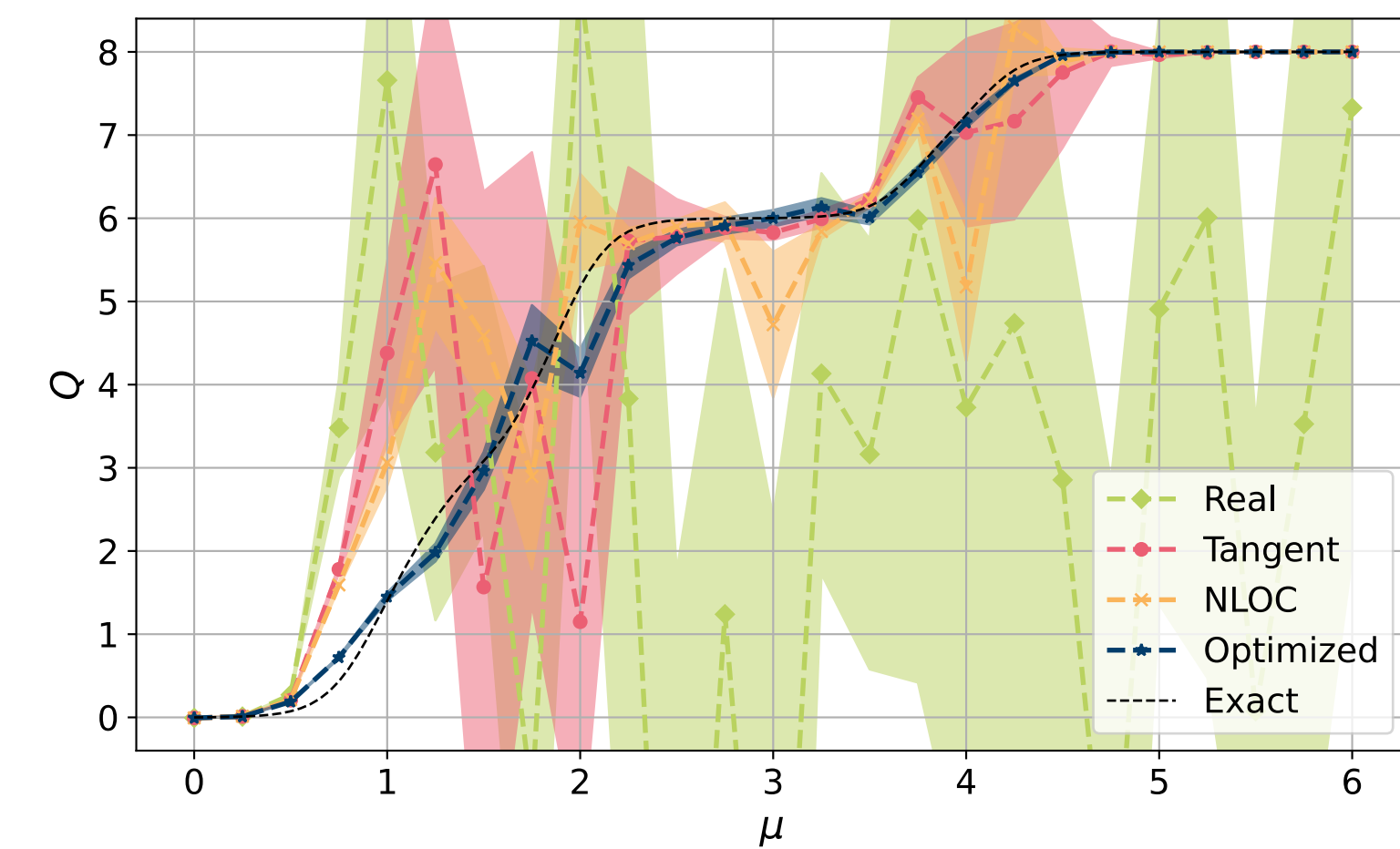
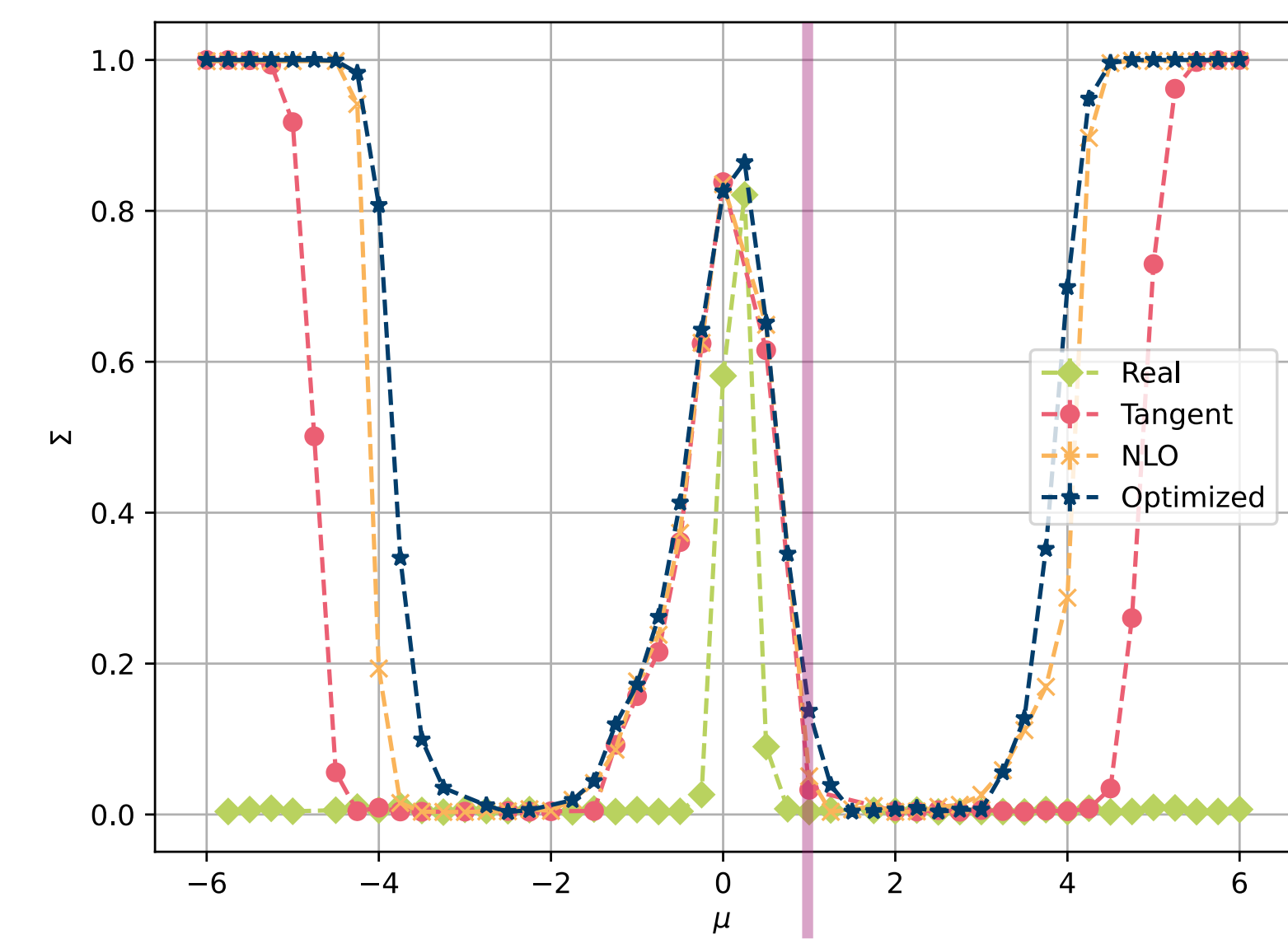
8-site honeycomb $N_t=16$ $\beta=8$ $U=2$

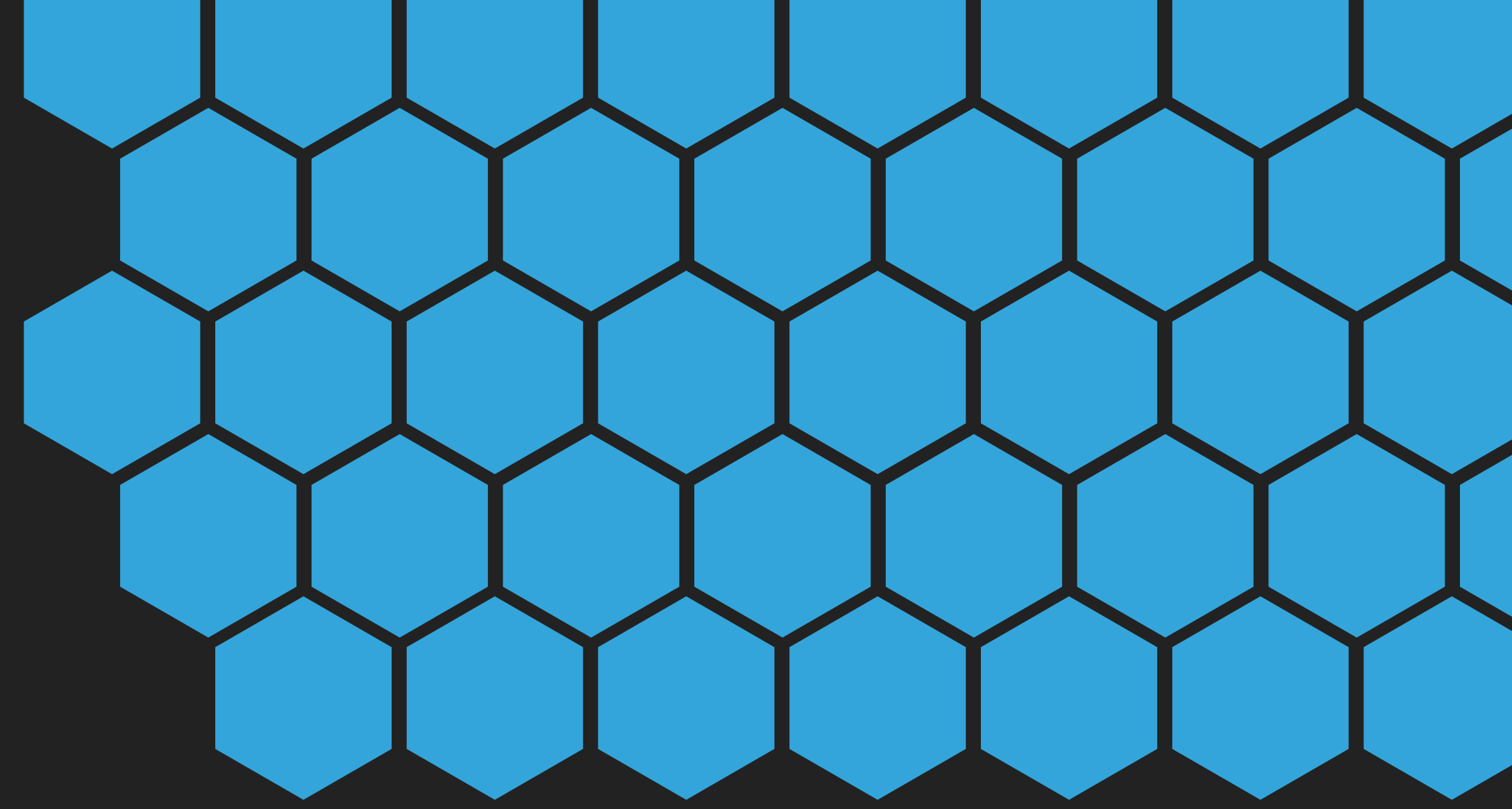


C_{20} $N_t=32$ $\beta=6$ $U=2$ $\mu=1$



C_{60} $N_t=32$ $\beta=6$ $U=2$ $\mu=1$

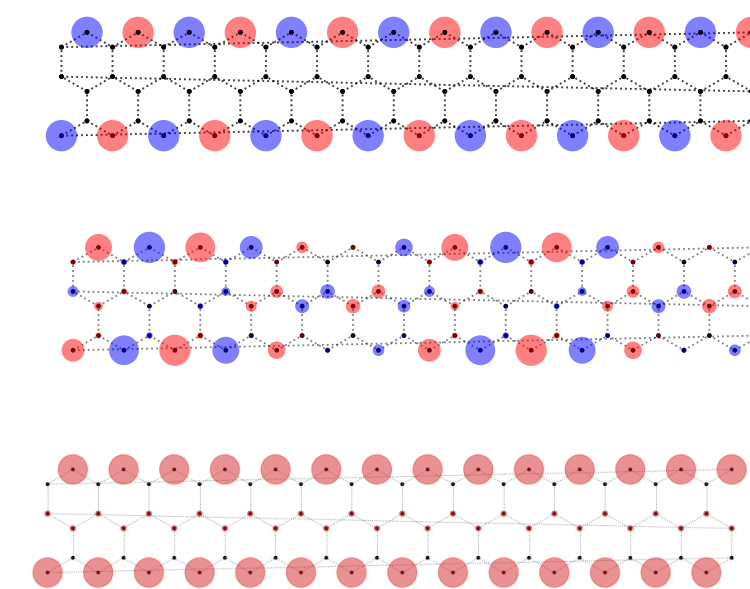
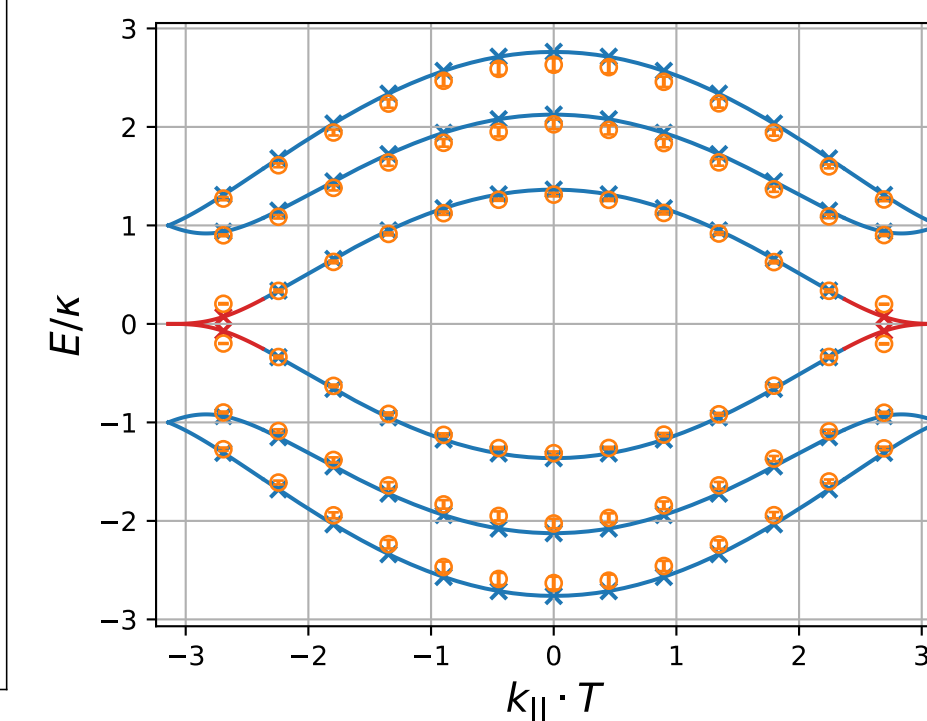
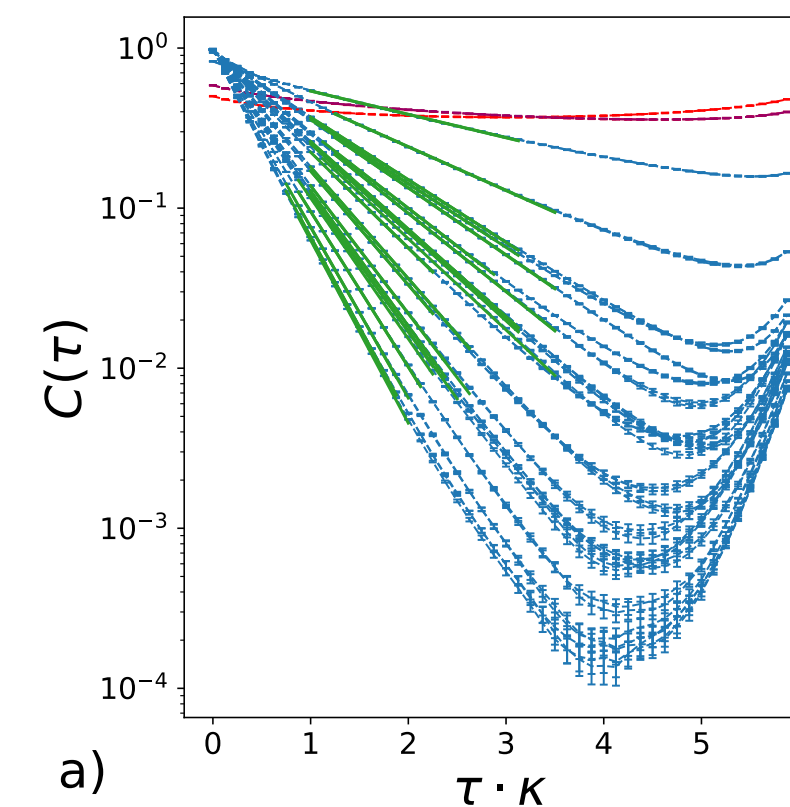
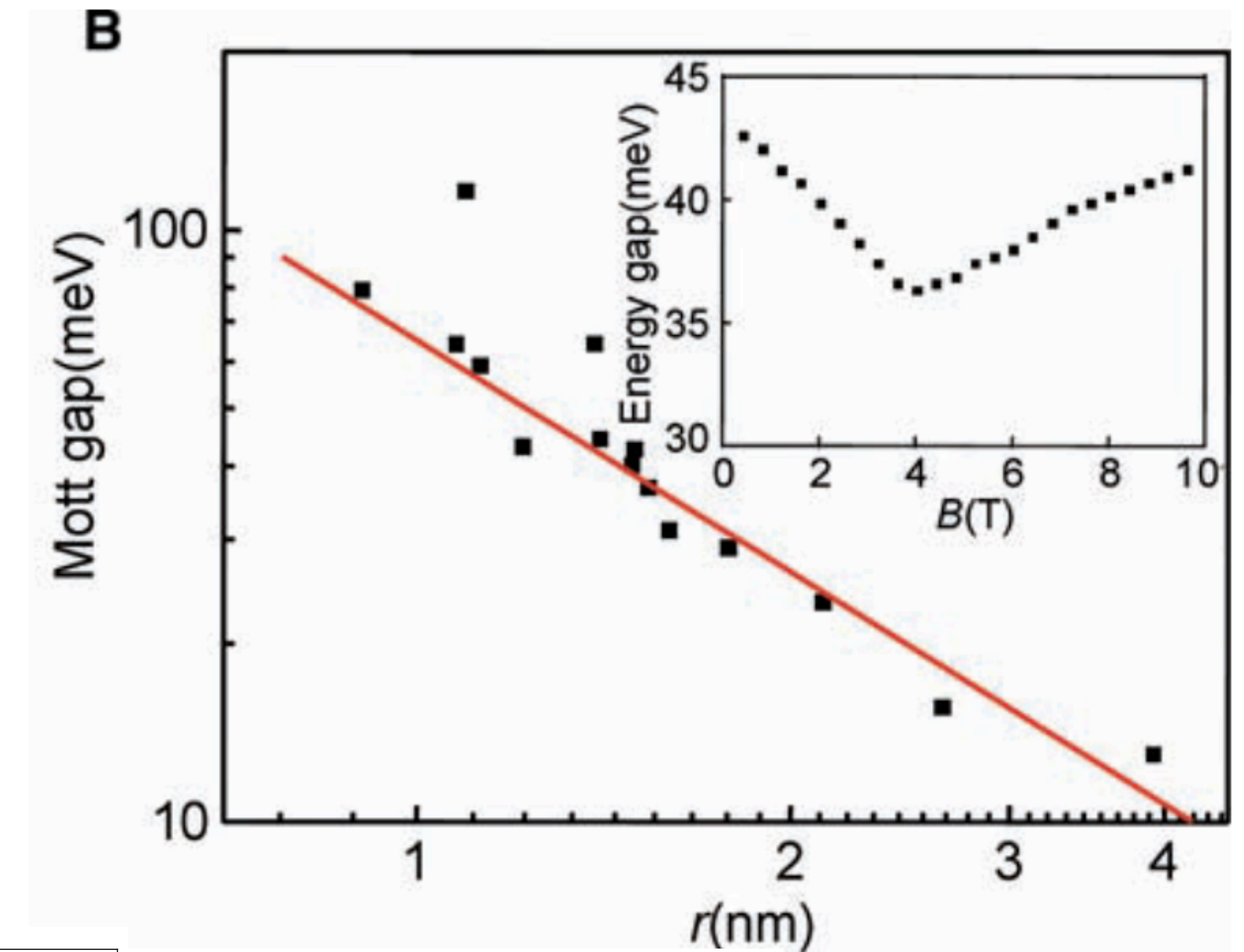




OUTLOOK

WHAT HUBBARD MODEL BEST DESCRIBES CARBON NANOSTRUCTURES?

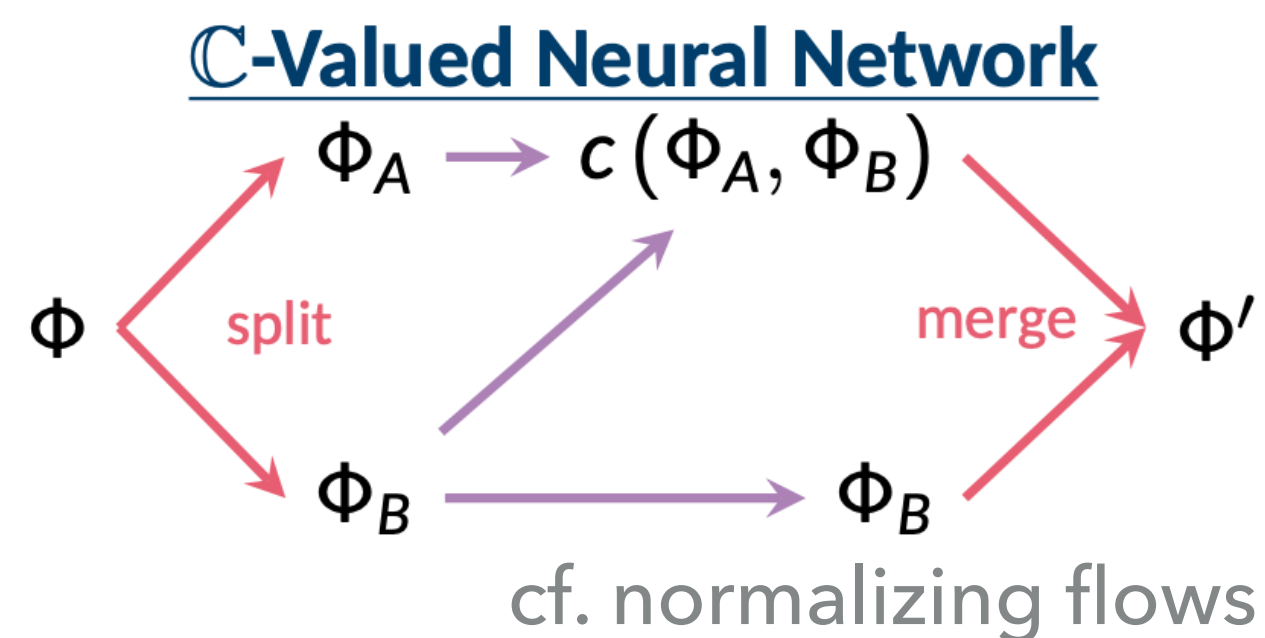
- ▶ Directly simulate nanotubes and reproduce the experimentally-observed gap's dependence on radius
- ▶ Extrapolate to $r = \infty$
- ▶ Are nonlocal interactions *really* needed?
Can we exclude a Hubbard description of graphene from first principles?
- ▶ Study 'the' Hubbard graphene, carbon nanostructures, ribbons, topological structures
- ▶ DOPE!



PERYLENE: MOLECULES IN SOLAR CELLS

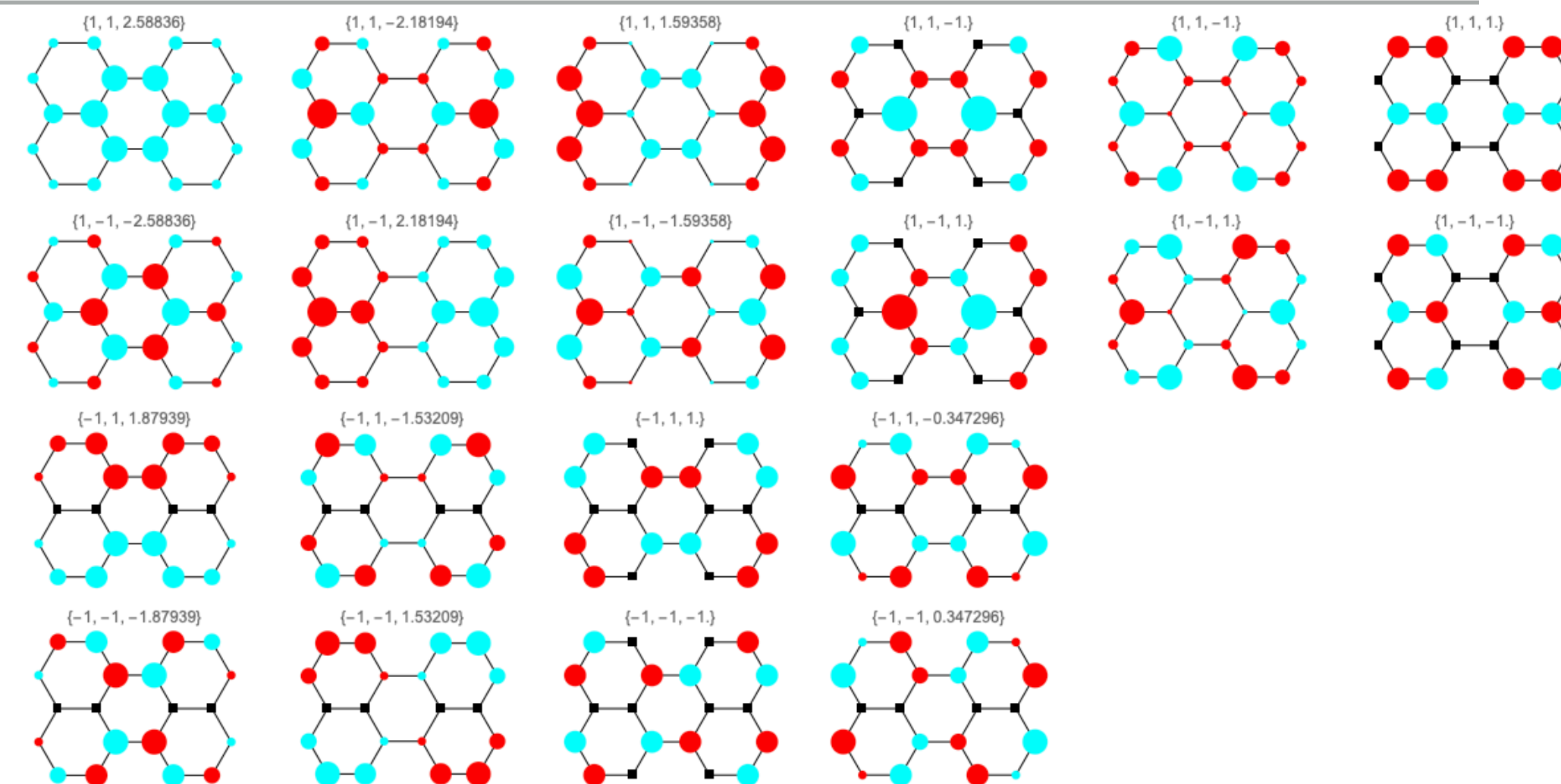
effort led by Marcel Rodekamp

- ▶ Current state of the art from chemistry is \sim DFT.
- ▶ Complex-valued, holomorphic, equivariant networks approximate the holomorphic flow

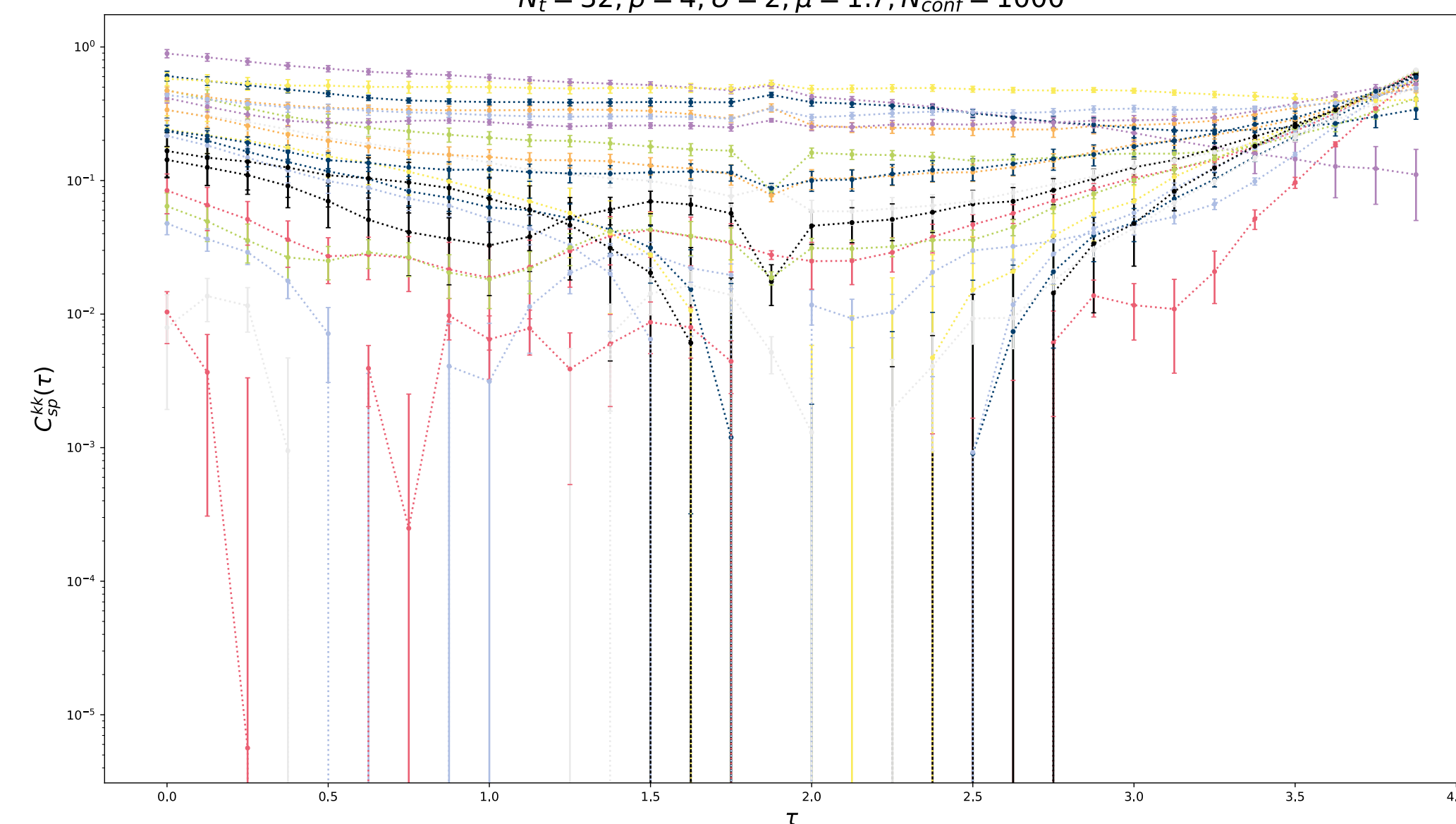


Alexandru, Bedaque, Lamm,
Lawrence, Physical Review D 96,
094505 (2017) 1709.01971
Albergo, Kanwar, Shanahan, PRD
100, 034515 1904.12072
Rodekamp, EB, Gäntgen et al.,
PRB 106(12):125139 2203.00390

- ▶ Can we get the electromagnetic responses with doped molecules?



$N_t = 32; \beta = 4; U = 2; \mu = 1.7, N_{conf} = 1000$

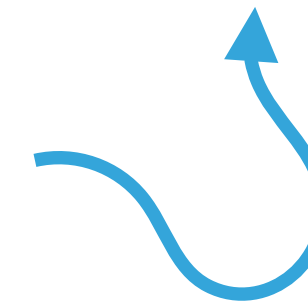


EQUIVARIANCE WITHOUT THINKING HARD ABOUT GROUP THEORY

$$NN : \phi \longrightarrow \phi$$

$$\text{Equivariance: } NN(T\phi) = T NN(\phi)$$

Time translation, spatial symmetries, ...



Only put 'conventional' ϕ through the ML

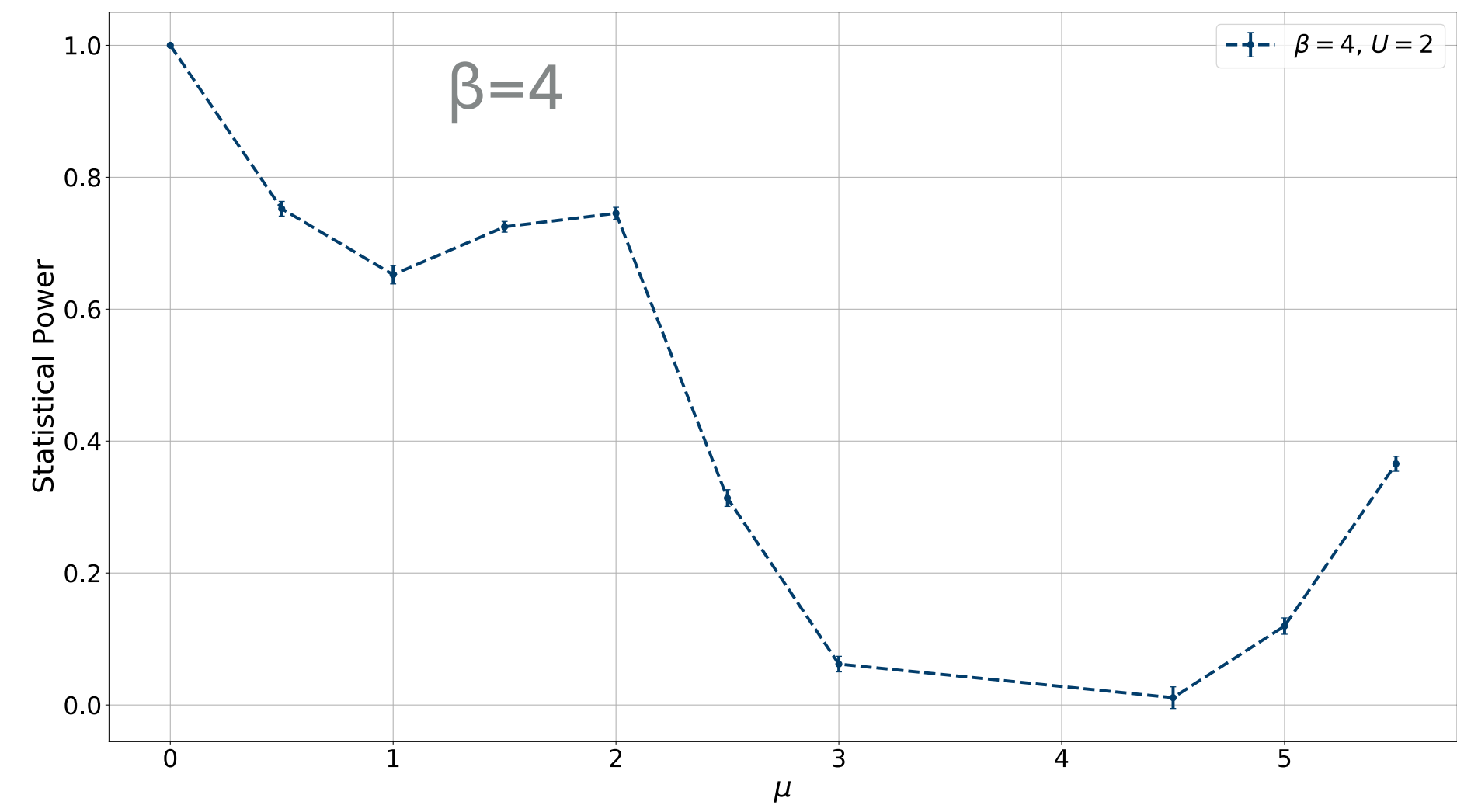
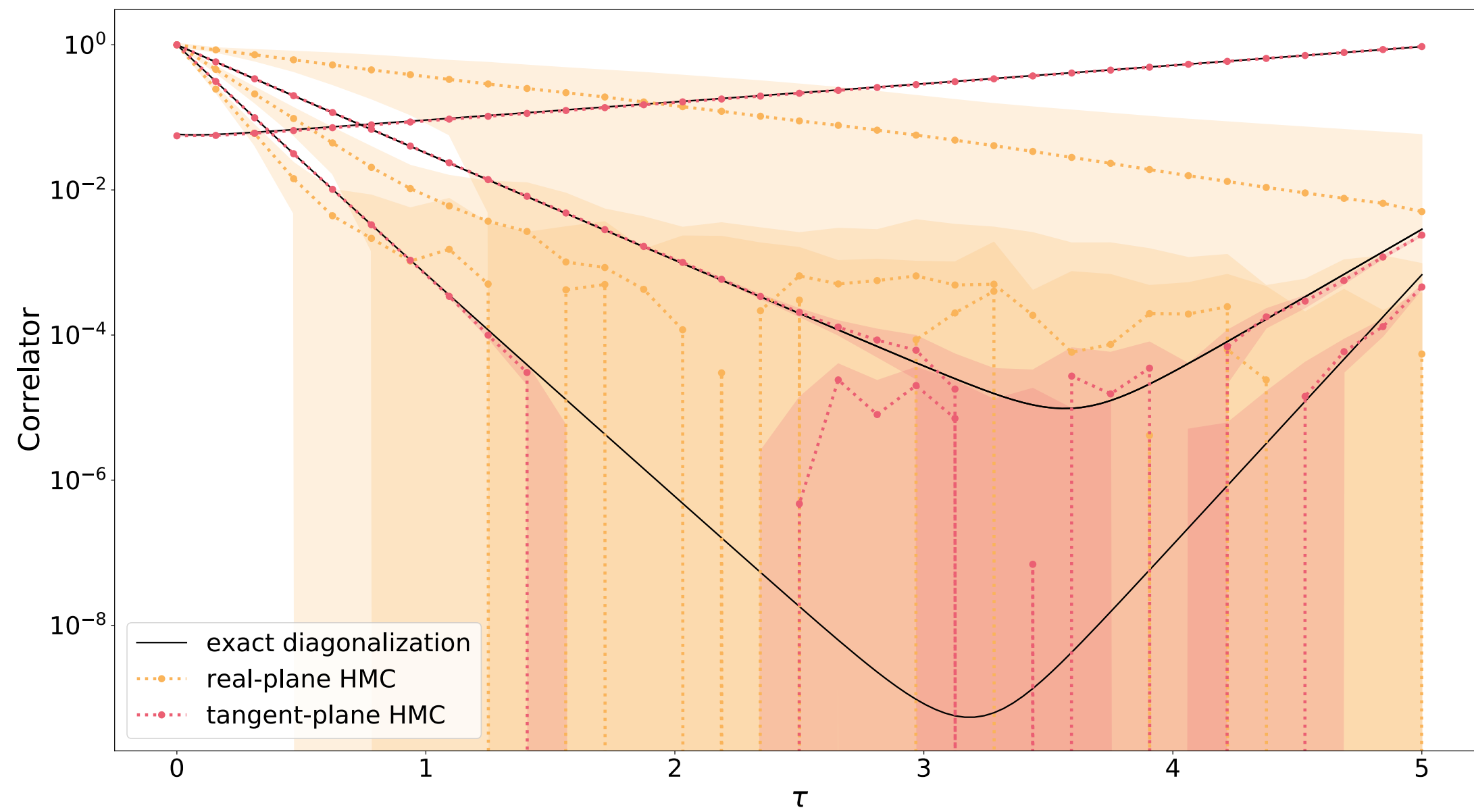
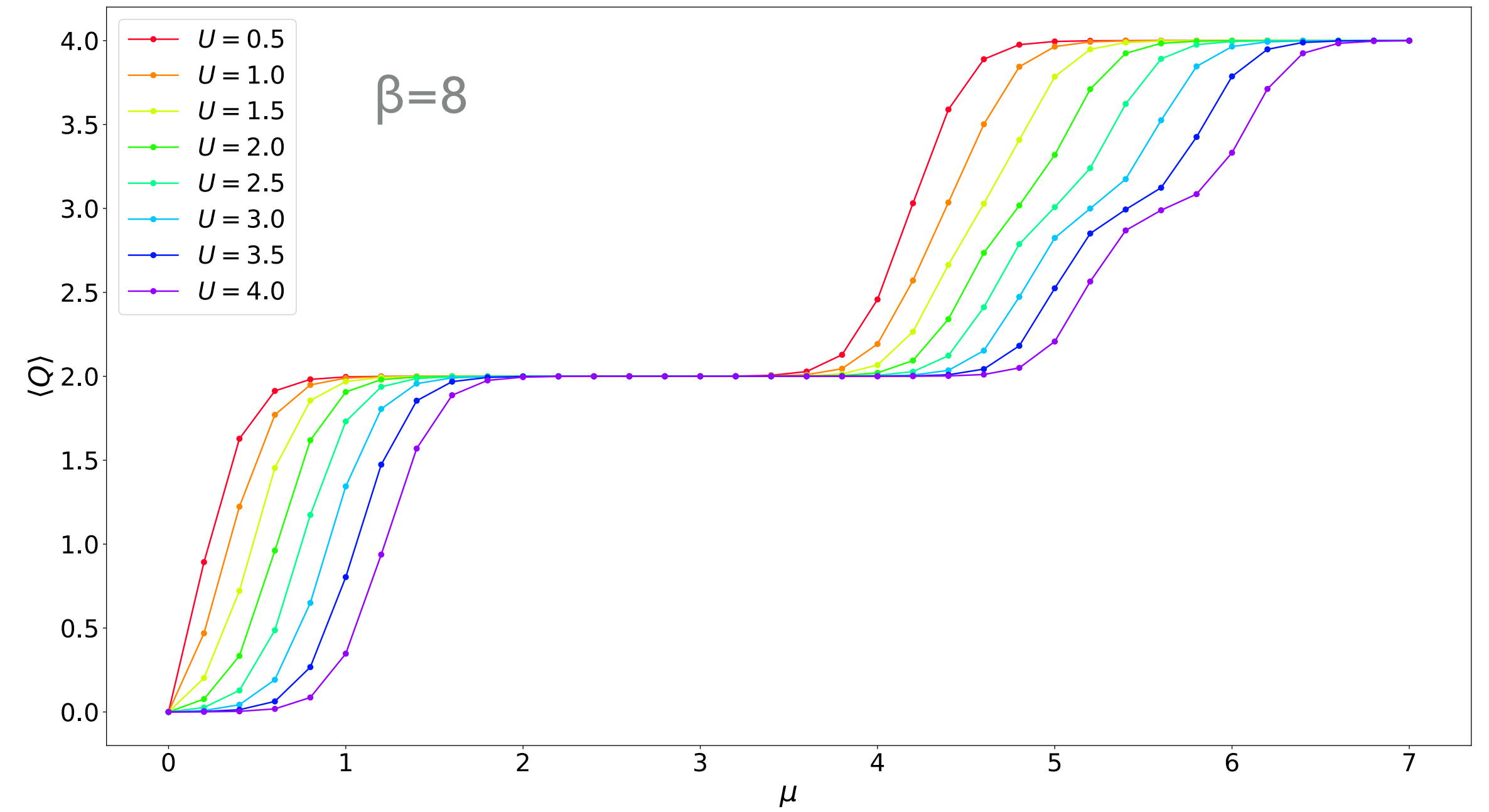
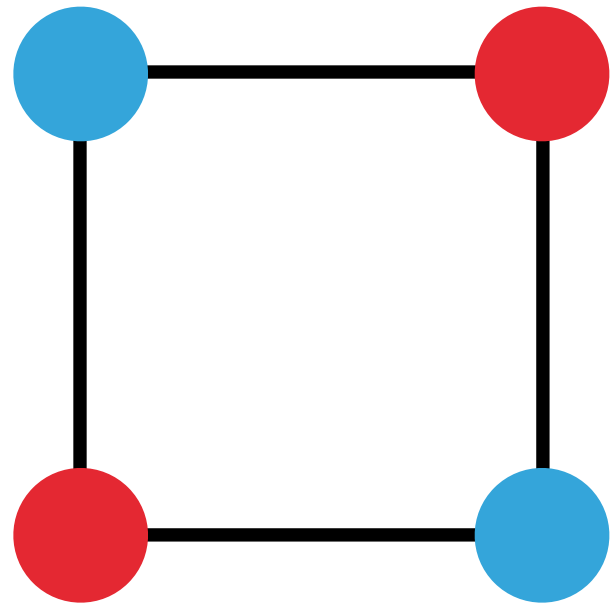
- Enforce a convention
- Apply network
- Undo convention
- Find t such that $\sum_x |\phi_{xt}|^2$ is maximal; shift by $-t$.
- Weights, biases, nonlinearities
- Shift back by $+t$.

The more prescriptive your convention the more equivariance you enjoy.

- Better for discrete / global symmetries!
- Annoying for gauge symmetries

SQUARE LATTICE: HIGH T_c SUPERCONDUCTORS?

Figures taken from C. Gäntgen's master's thesis



EWJ-80194 STEREO

M I S T R A I



FREDDIE HUBBARD

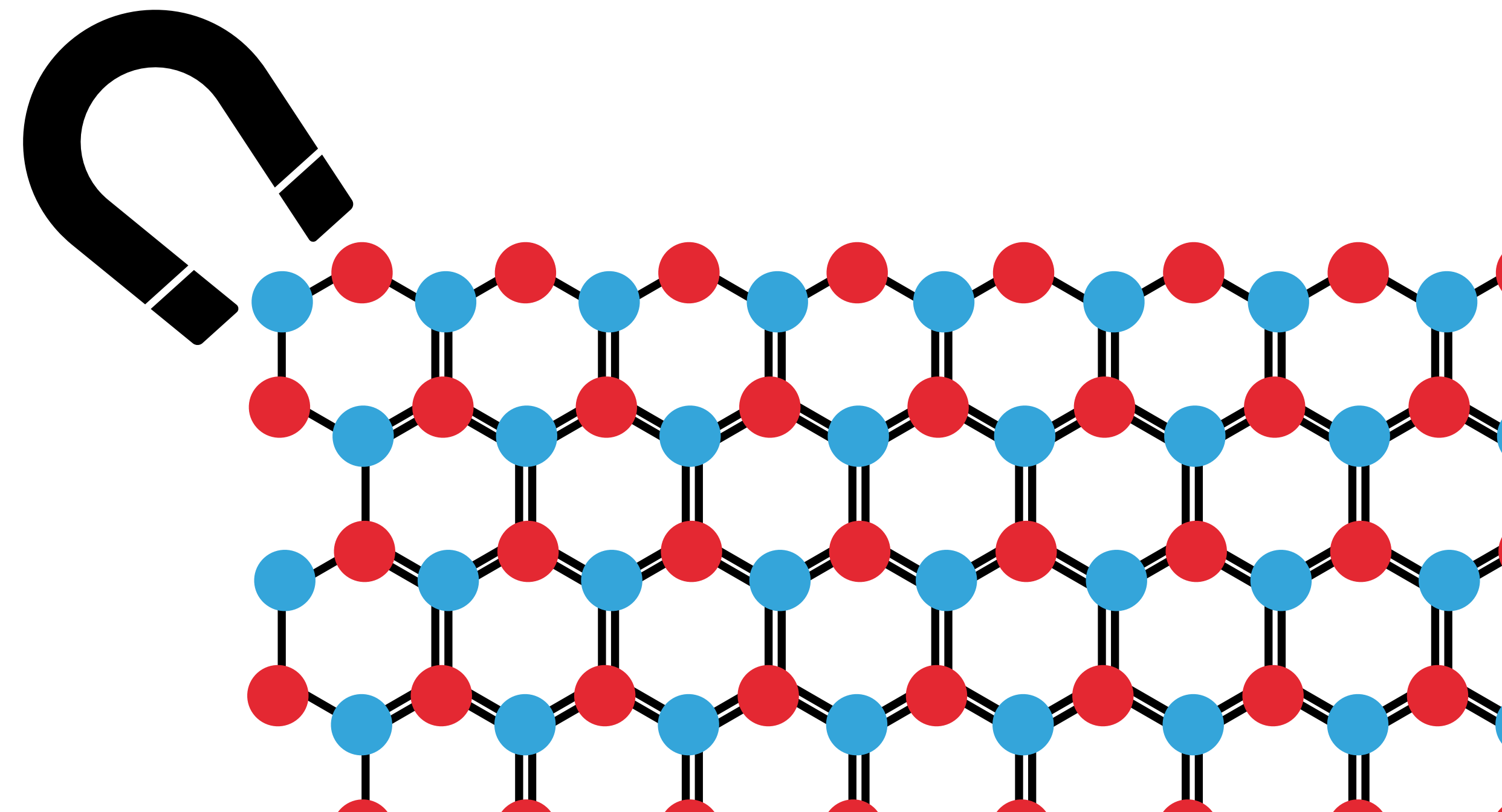
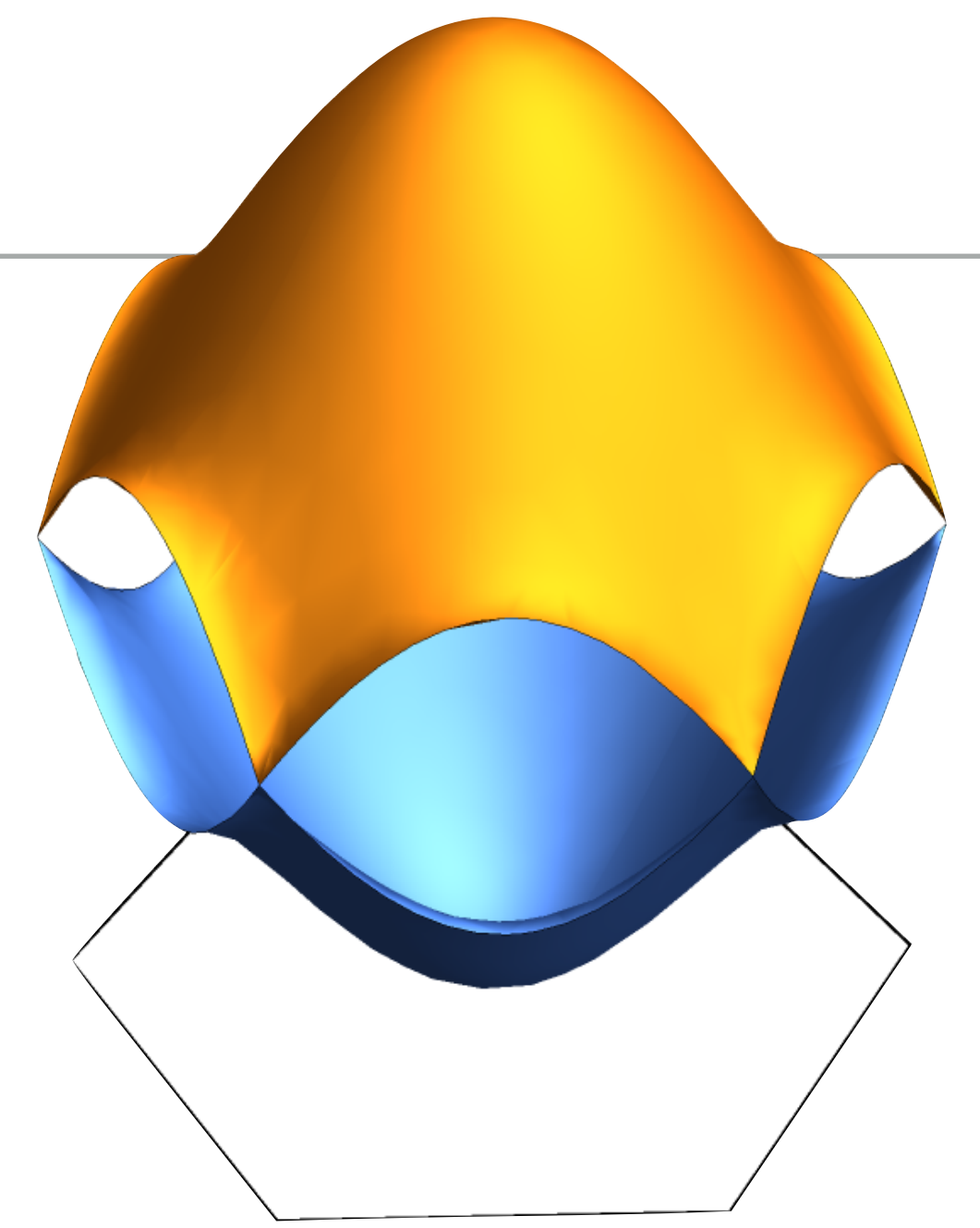
THE HUBBARD MODEL

BACKUP SLIDES

OTHER DIRECTIONS

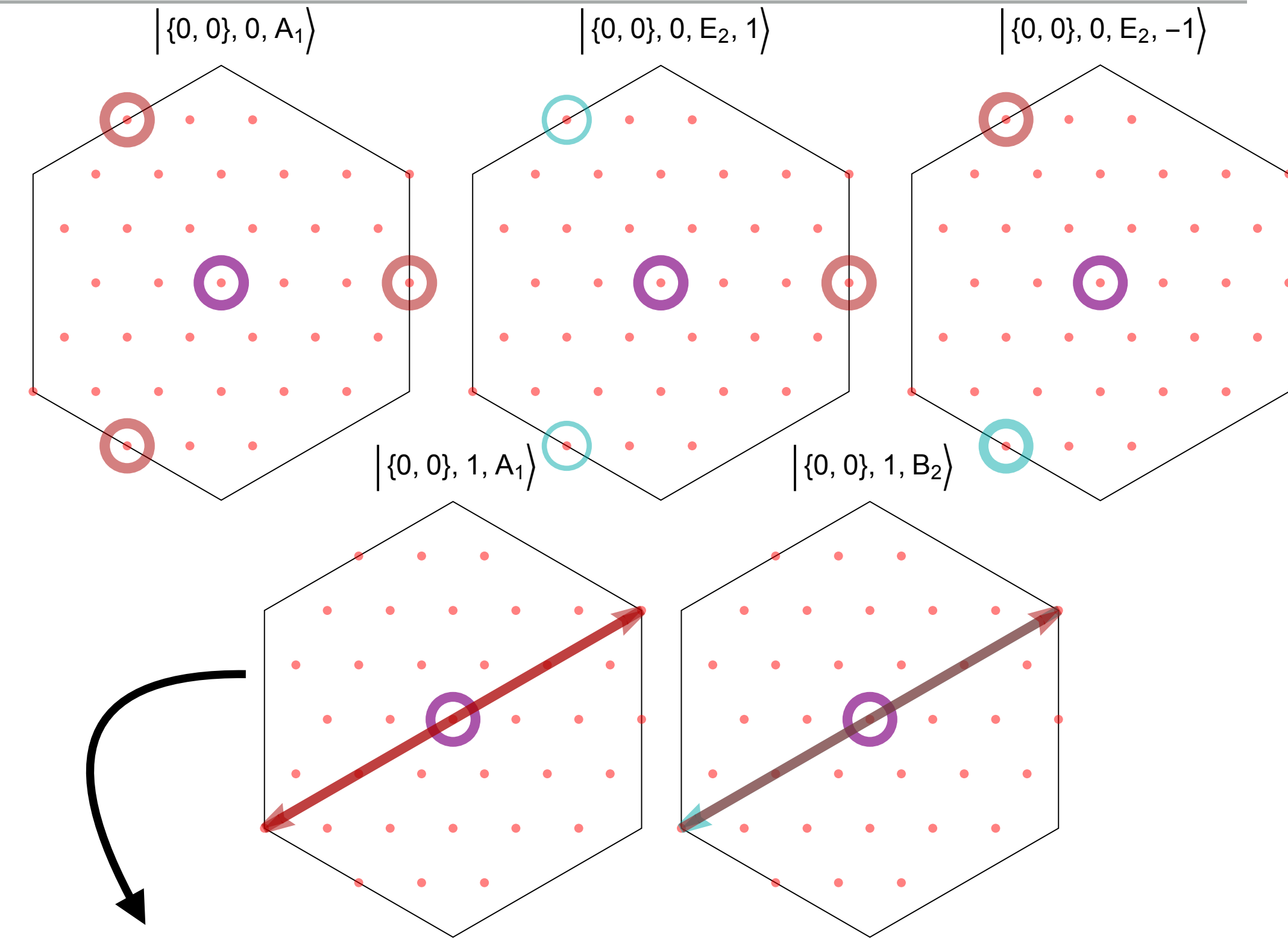
STRUCTURE FACTORS, MATRIX ELEMENTS

- ▶ Anomalous dimension
- ▶ Measure one-body spectrum, especially near K, K'
- ▶ Quasiparticle weight
- ▶ Measure momentum-dependence spin-spin structure factors
- ▶ Matrix elements!
Response to magnetic fields, ...

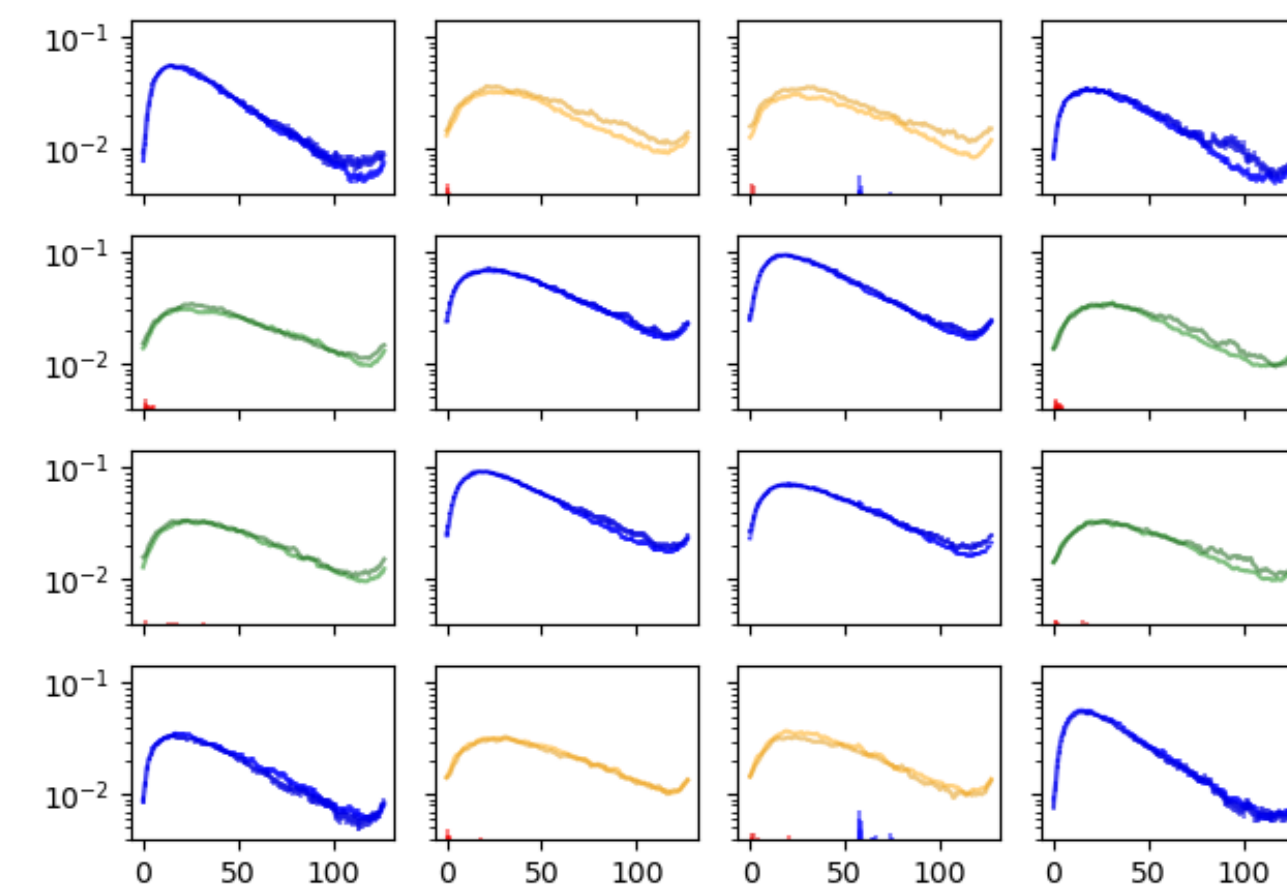


EXCITONS

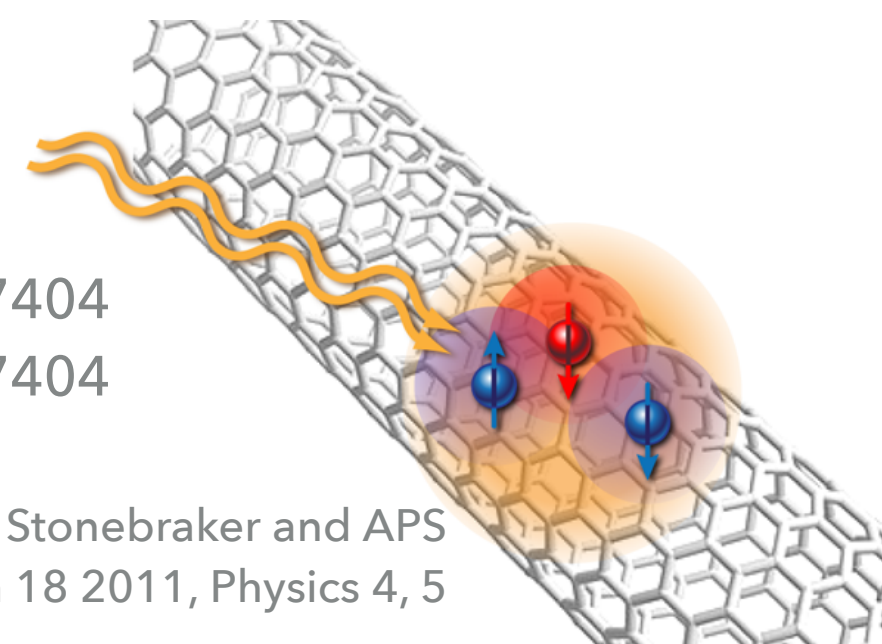
- ▶ Measure two-body energy levels
- ▶ Many states in each sector of total crystal momenta, little group
- ▶ Leverage the analog of the finite-volume LQCD formalism (Lüscher's method) for two-body phase shifts.
- ▶ Resonances? Bound states?
- ▶ ... trions?



$\kappa\beta=24$ $L=6$ $N_t=128$ $U/\kappa=3.00$



Matsunaga, Matsuda, and Kanemitsu PRL 106, 037404
10.1103/PhysRevLett.106.037404



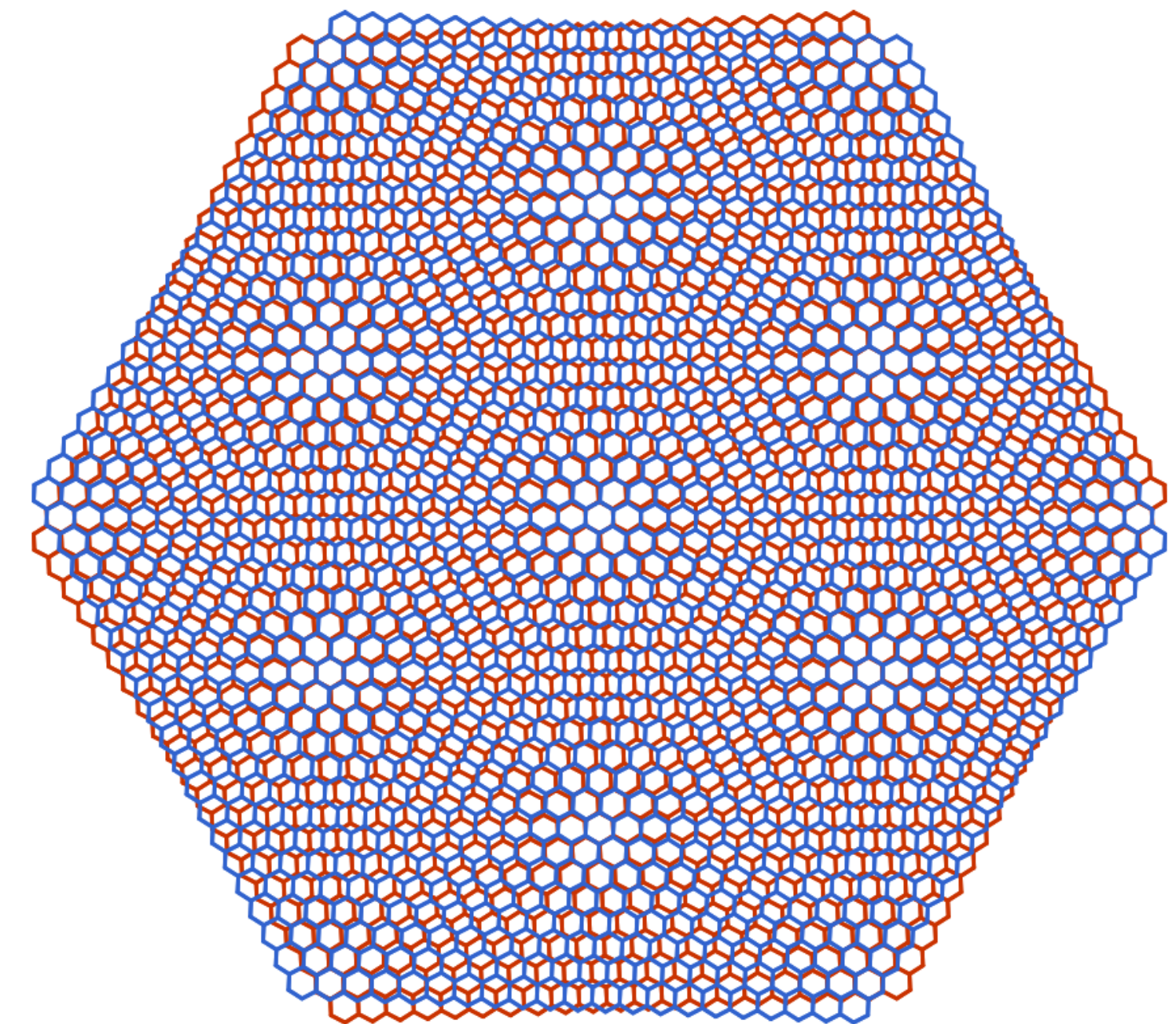
Alan Stonebraker and APS
Jan 18 2011, Physics 4, 5

OTHER MATERIALS / LATTICES

DIRECT EXPLORATION OF HUBBARD SYSTEMS

We can take the temporal continuum limit with $2 \times 10^2 \sim 20000$ ions !

- ▶ Add screened Coulomb
- ▶ Transition metal oxides (heavy fermions)
- ▶ Lanthanide materials (Kondo insulators)
- ▶ Perovskites (high- T_c superconductors) on a square / checkerboard lattice
- ▶ Magic-angle Moiré superlattices

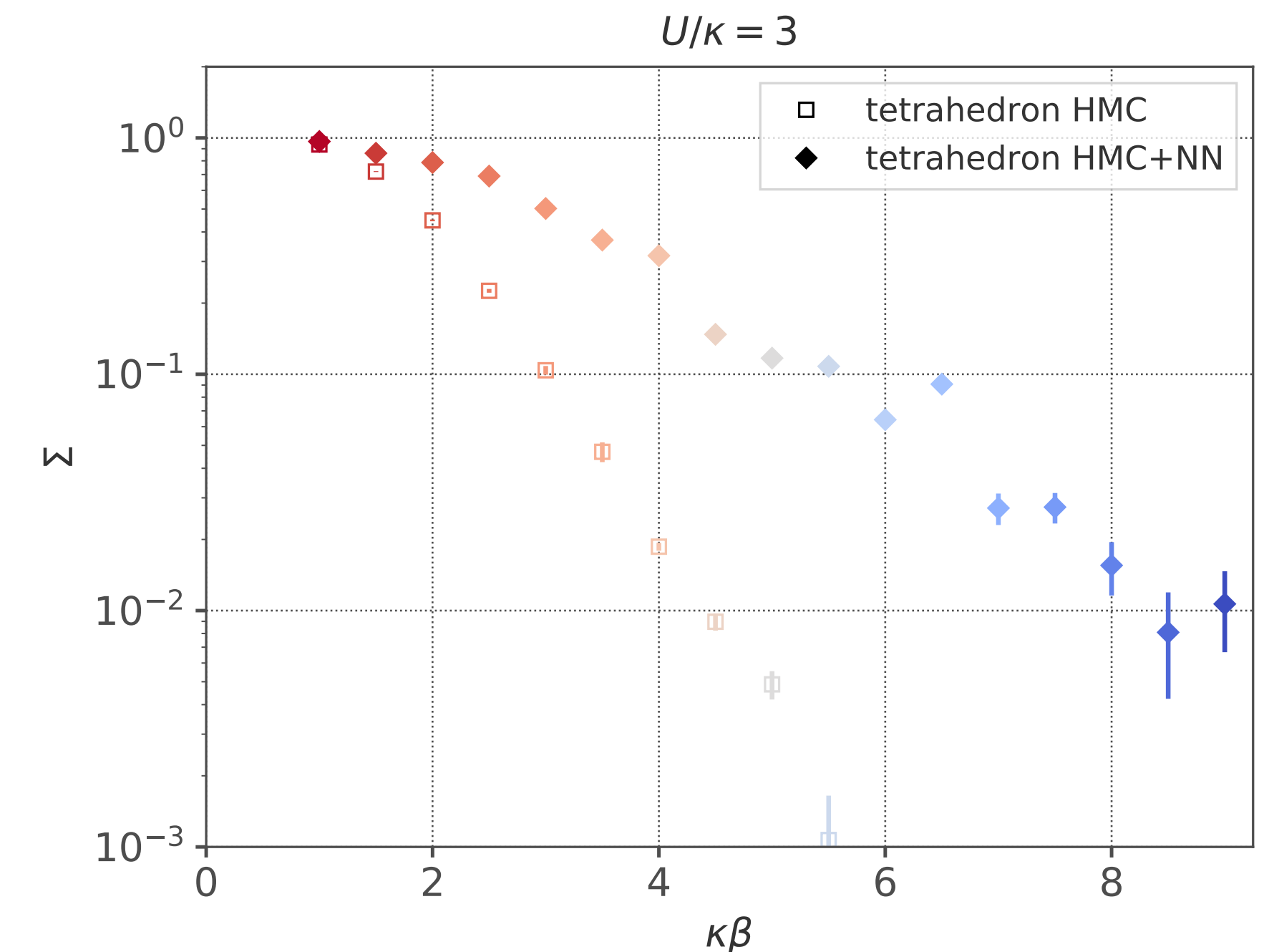


NEW ALGORITHMS UNLOCK NEW PHYSICS

- ▶ Non-bipartite, frustrated systems (fullerenes, triangular, kagome lattices)
- ▶ Break half-filling requirement (doping, semiconductors)
- ▶ Opportunities for machine learning, Lefschetz thimbles, ... (interdisciplinary applicability to QCD)
- ▶ Real-time physics via Schwinger-Keldysh?

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, -h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

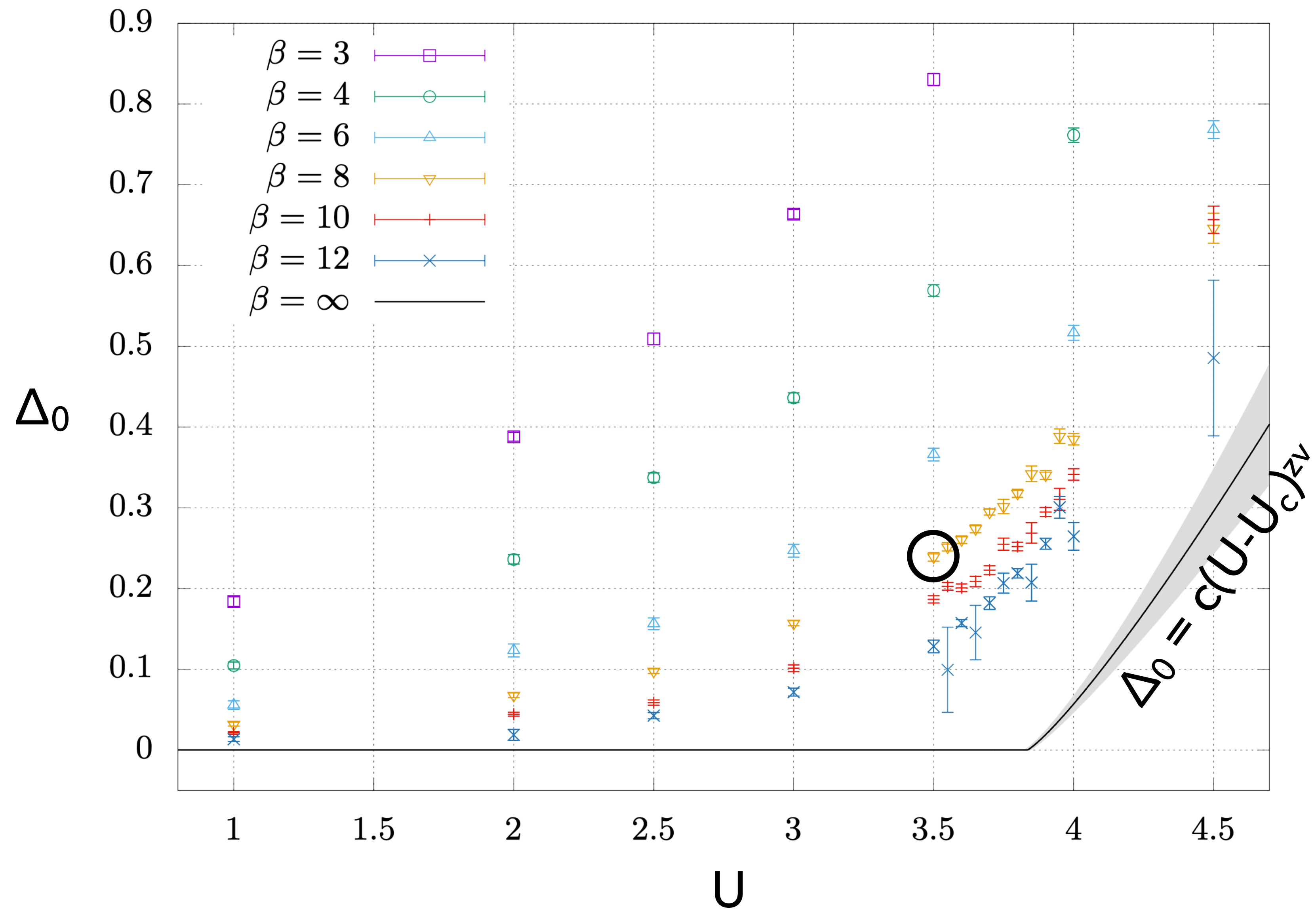
average phase = Σ



FINITE SIZE SCALING

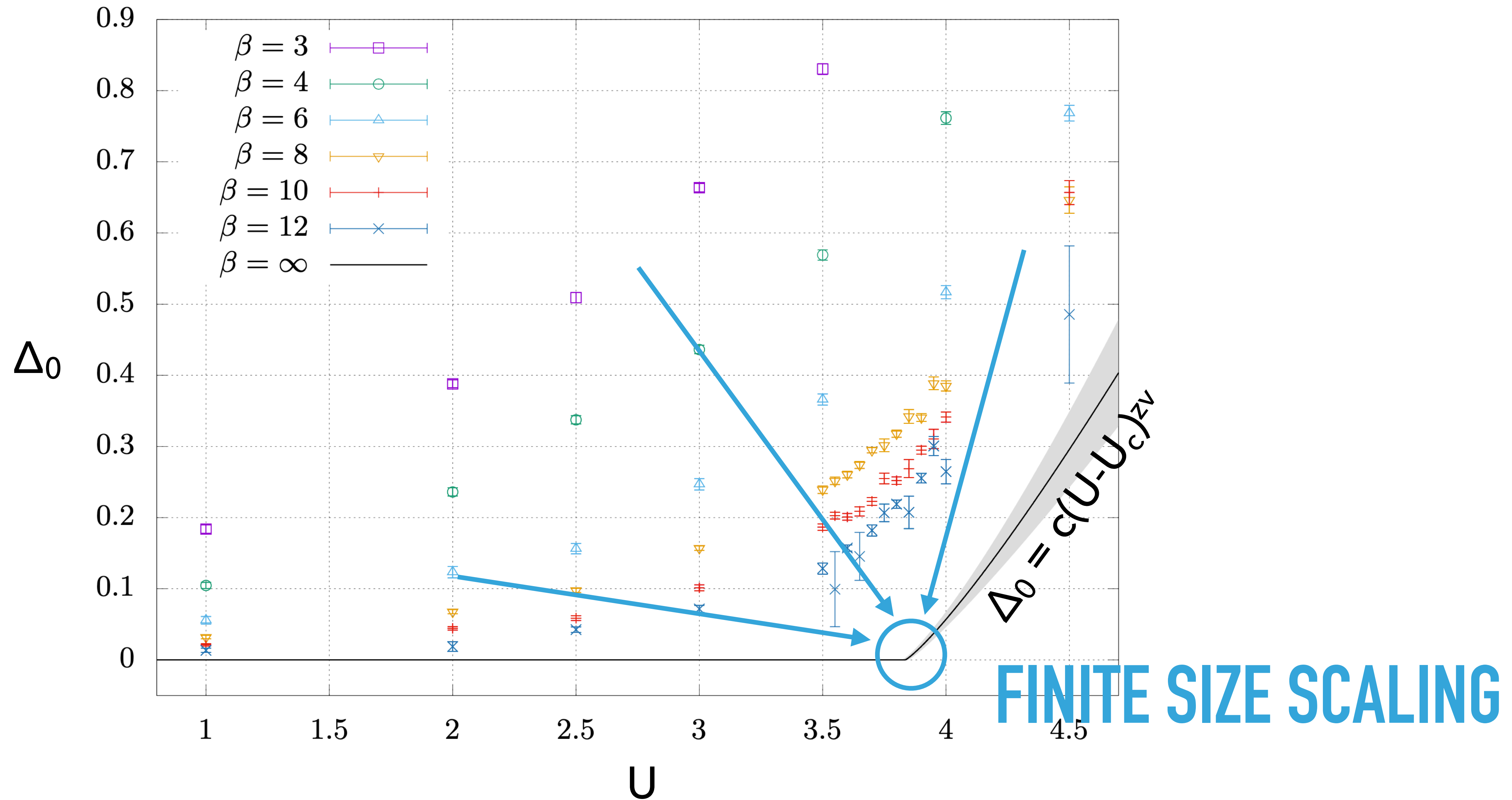
THERMODYNAMIC LIMIT

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112

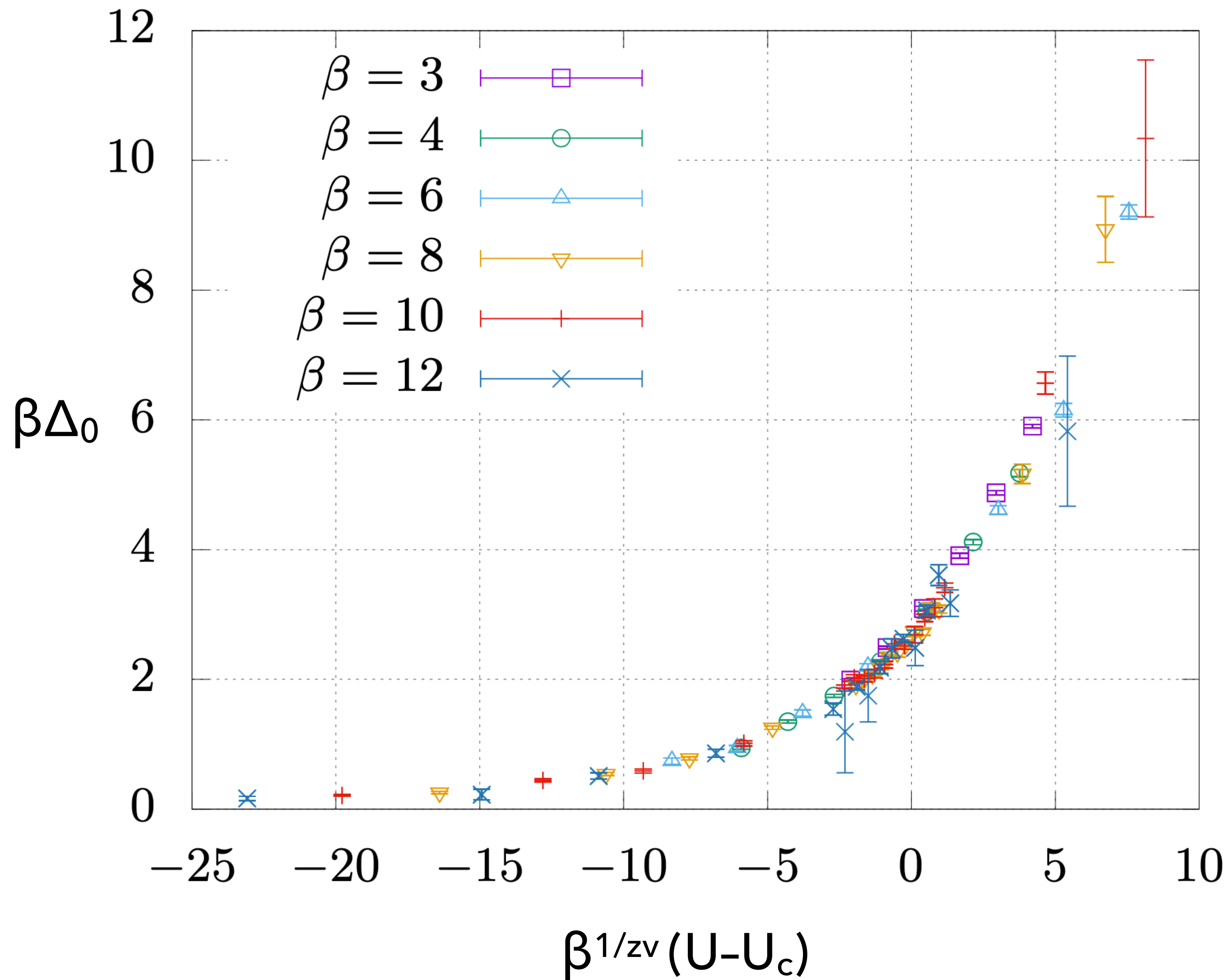


THERMODYNAMIC LIMIT

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112



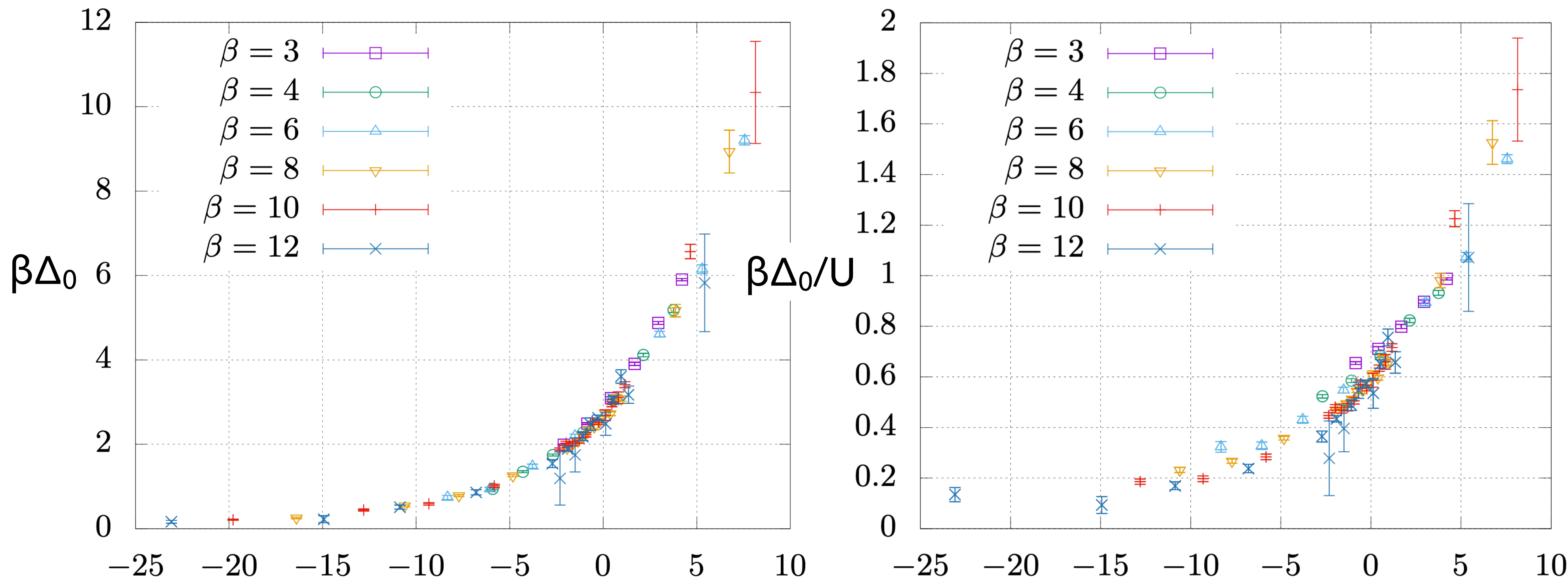
FINITE-SIZE SCALING



$U_c = 3.834(14)$
 $z\nu = 1.185(43)$

FINITE-SIZE SCALING MEAN-FIELD Δ/U

Assaad and Herbut, PRX 3, 031010 (2013) 1304.6340

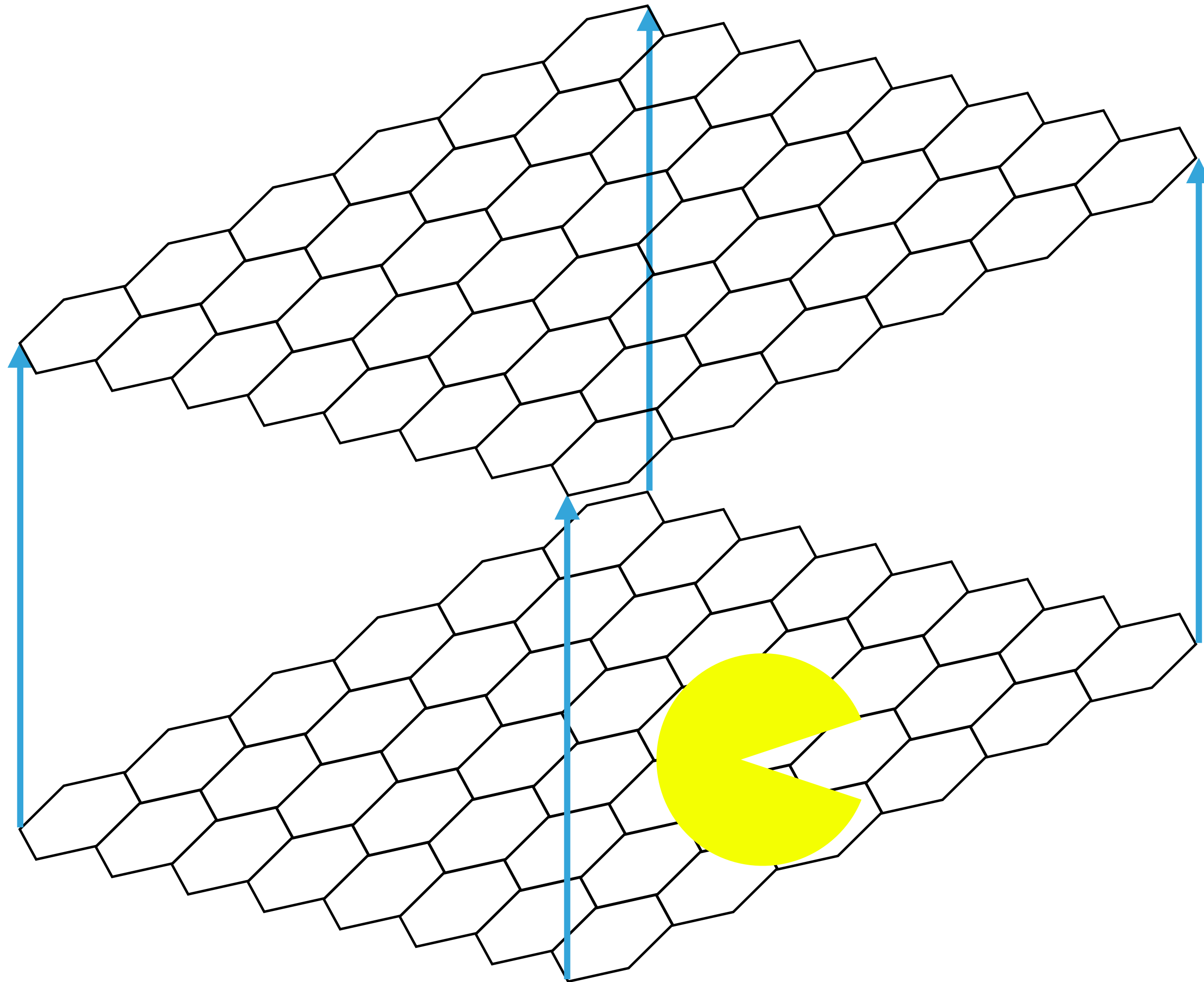


$U_c = 3.834(14)$
 $z\nu = 1.185(43)$
 $\tilde{\beta} = 1.095(37)$

$\tilde{\beta}$ is the critical exponent for $\langle m_s \rangle$

HMC + DISCRETIZATION

FORMULATING THE PATH INTEGRAL



$$\mathcal{Z} = \text{tr} [e^{-\beta H}]$$

Grassmannian coherent state identity

$$\mathbb{1} = \int \left[\prod_x d\psi_x^* d\psi_x d\eta_x^* d\eta_x \right] e^{-\sum_x \psi_x^* \psi_x + \eta_x^* \eta_x} |\psi, \eta\rangle \langle \psi, \eta|$$

$$H = \sum_{xy} (a_x^\dagger h_{xy} a_y + b_x^\dagger h_{xy} b_y) + \frac{1}{2} \sum_{xy} \rho_x V_{xy} \rho_y - \mu Q$$

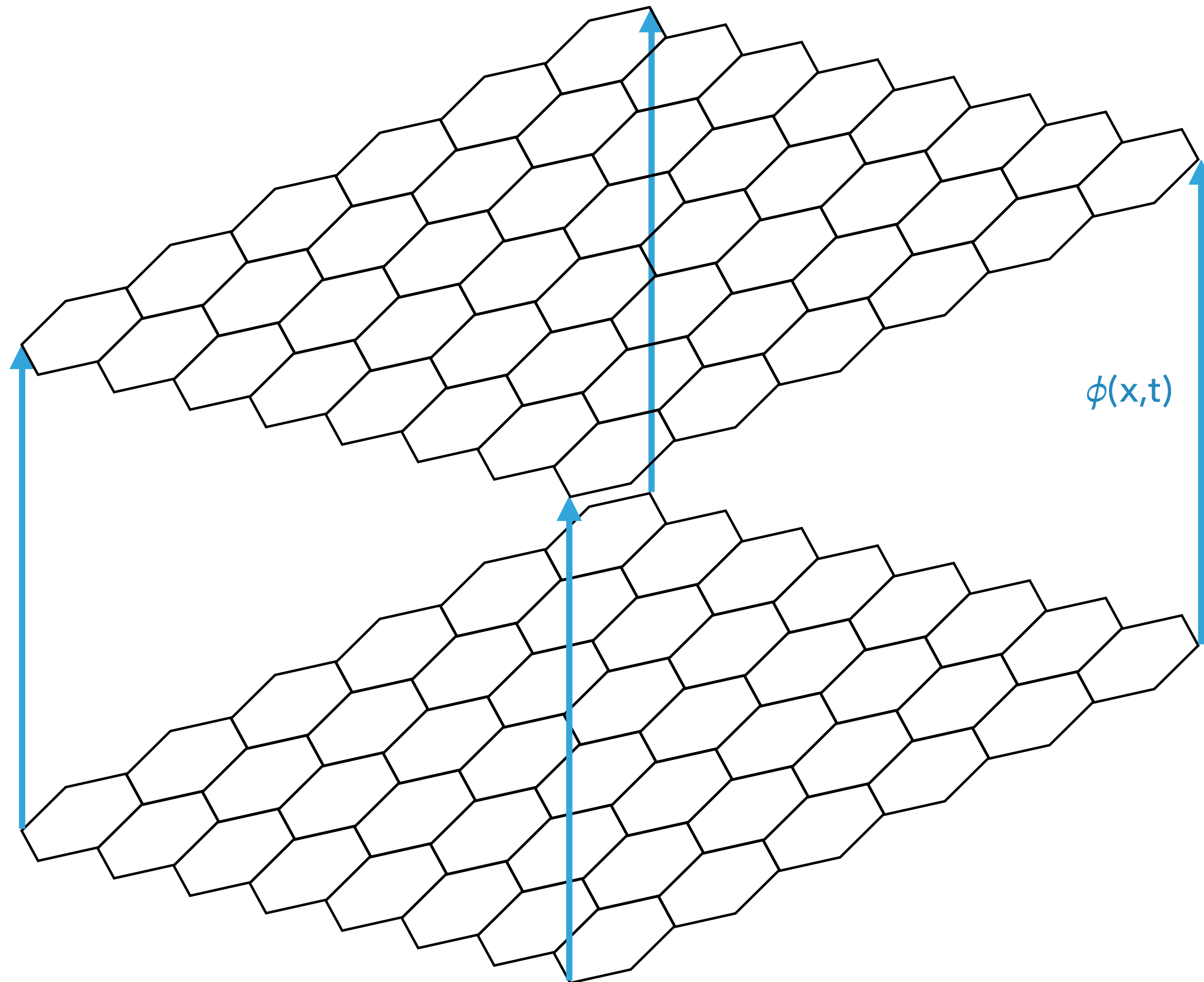
Linearize with Hubbard-Stratonovich transformation

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Easy to change the interaction

IMPORTANCE SAMPLING THE PATH INTEGRAL

Duane, Kennedy, Pendleton, and Roweth, PLB 195 216-222 (1987)
10.1016/0370-2693(87)91197-X



$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability

Generate an *ensemble* for fixed U, β, N_t, \dots

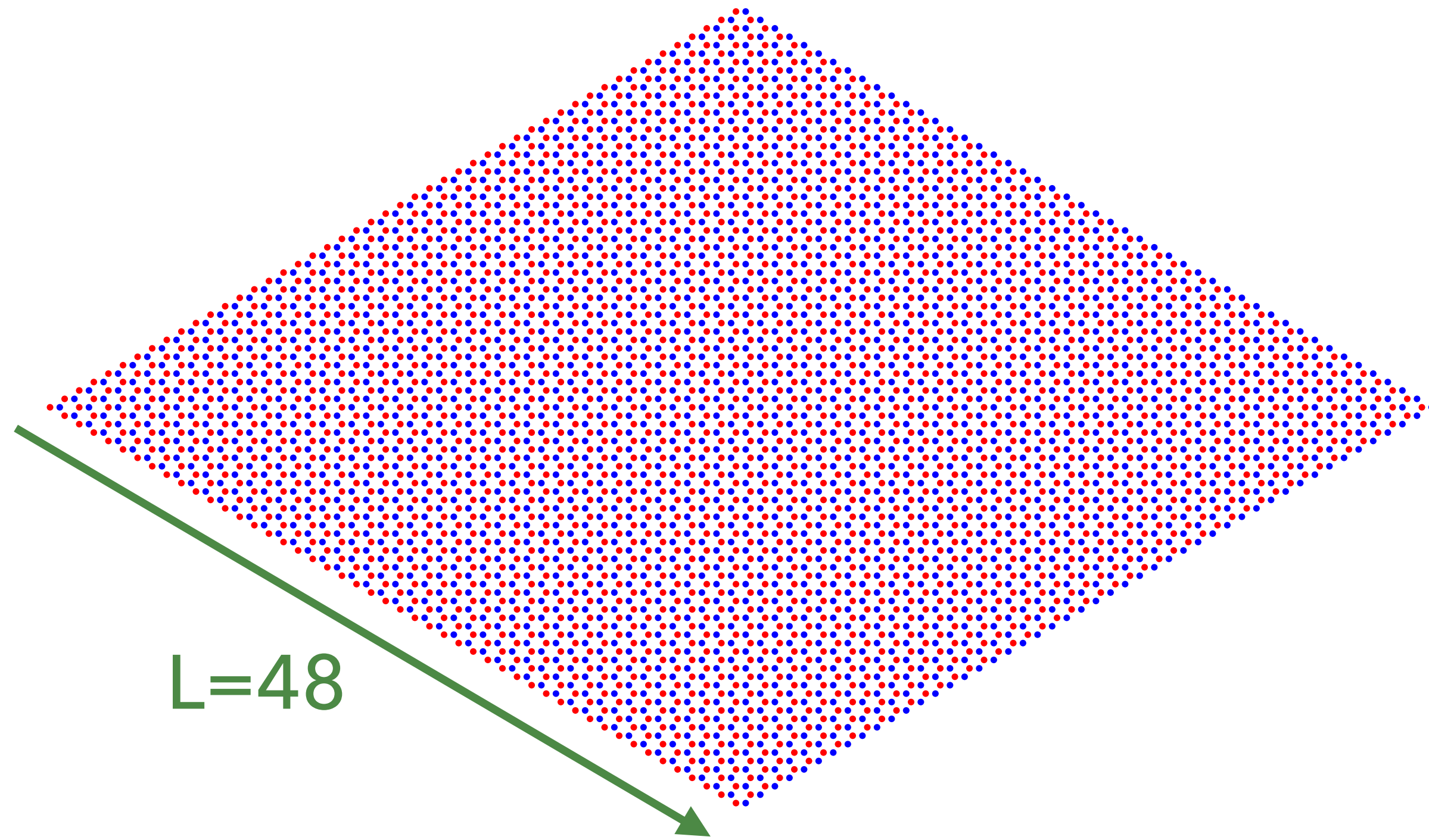
$$\{\phi_1, \phi_2, \dots, \phi_N\}$$

Hybrid Monte Carlo $\sim V^{5/4}$

Estimate any observable

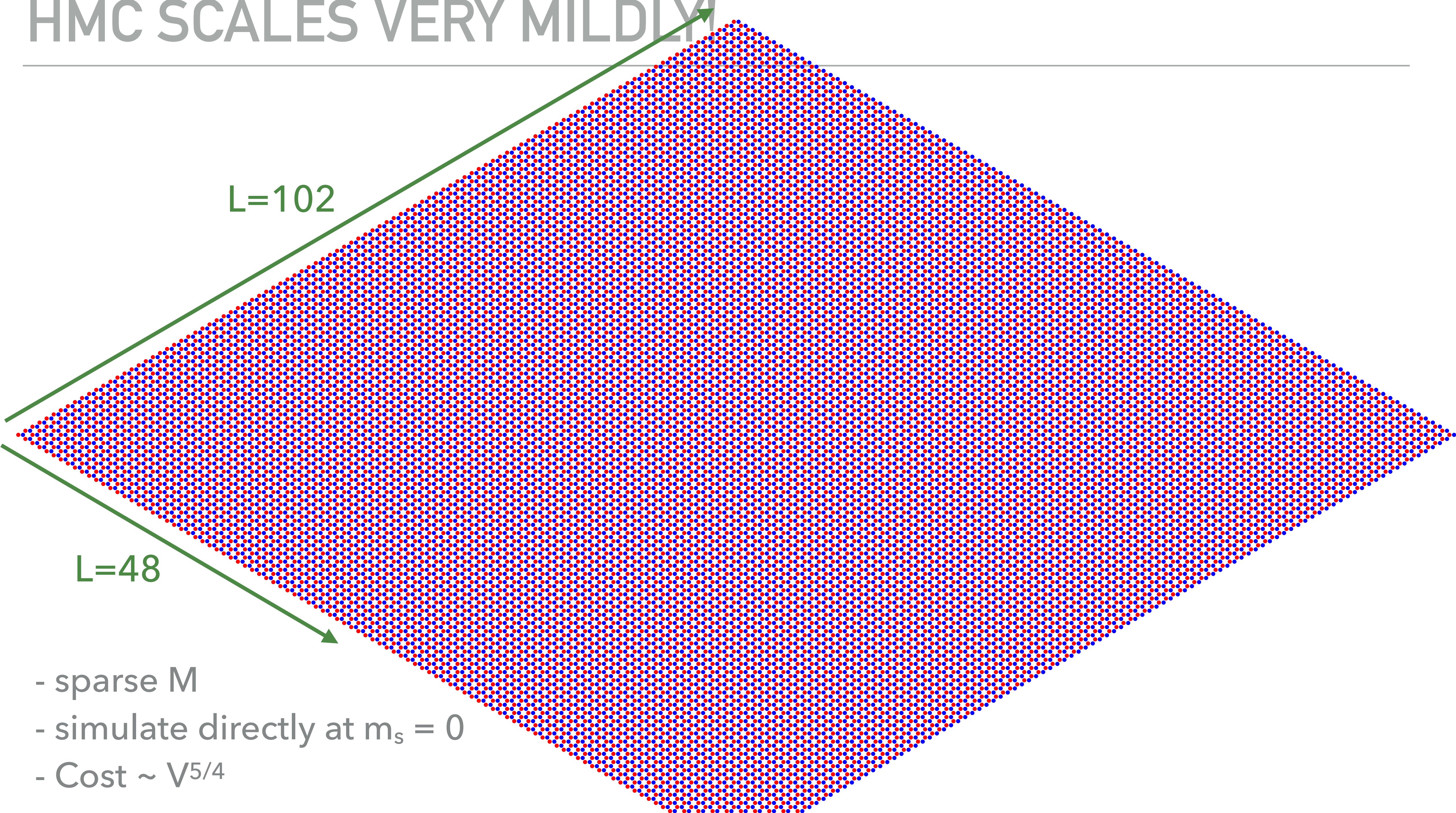
$$\begin{aligned} \langle \mathcal{O} \rangle &= \frac{1}{\mathcal{Z}} \int d\mathcal{Z} \mathcal{O}[\phi] \\ &= \frac{1}{N} \sum_i^N \mathcal{O}[\phi_i] \end{aligned}$$

HMC SCALES VERY MILDLY!



- sparse M
- simulate directly at $m_s = 0$
- Cost $\sim V^{5/4}$

HMC SCALES VERY MILDLY

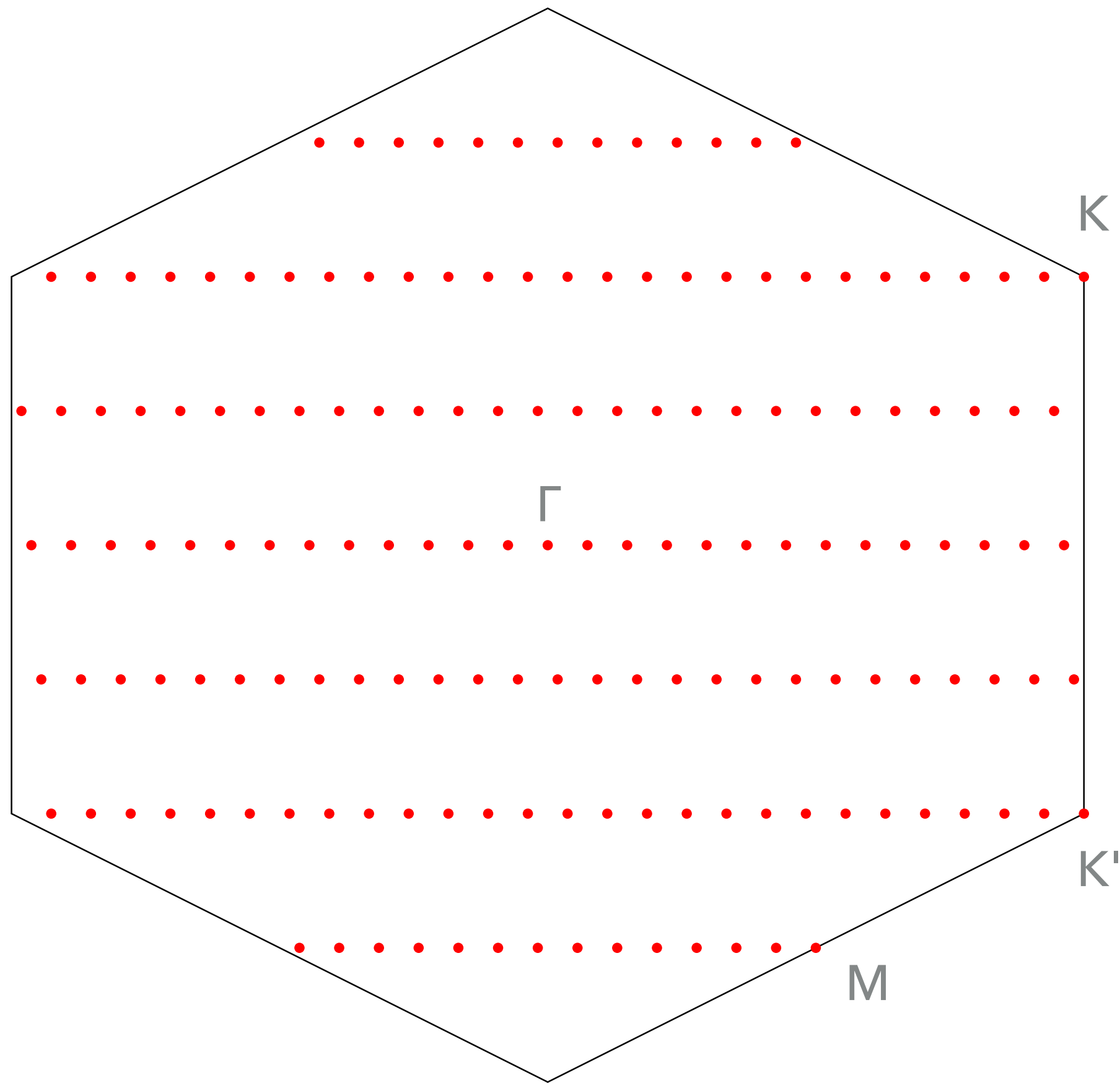


L=102

L=48

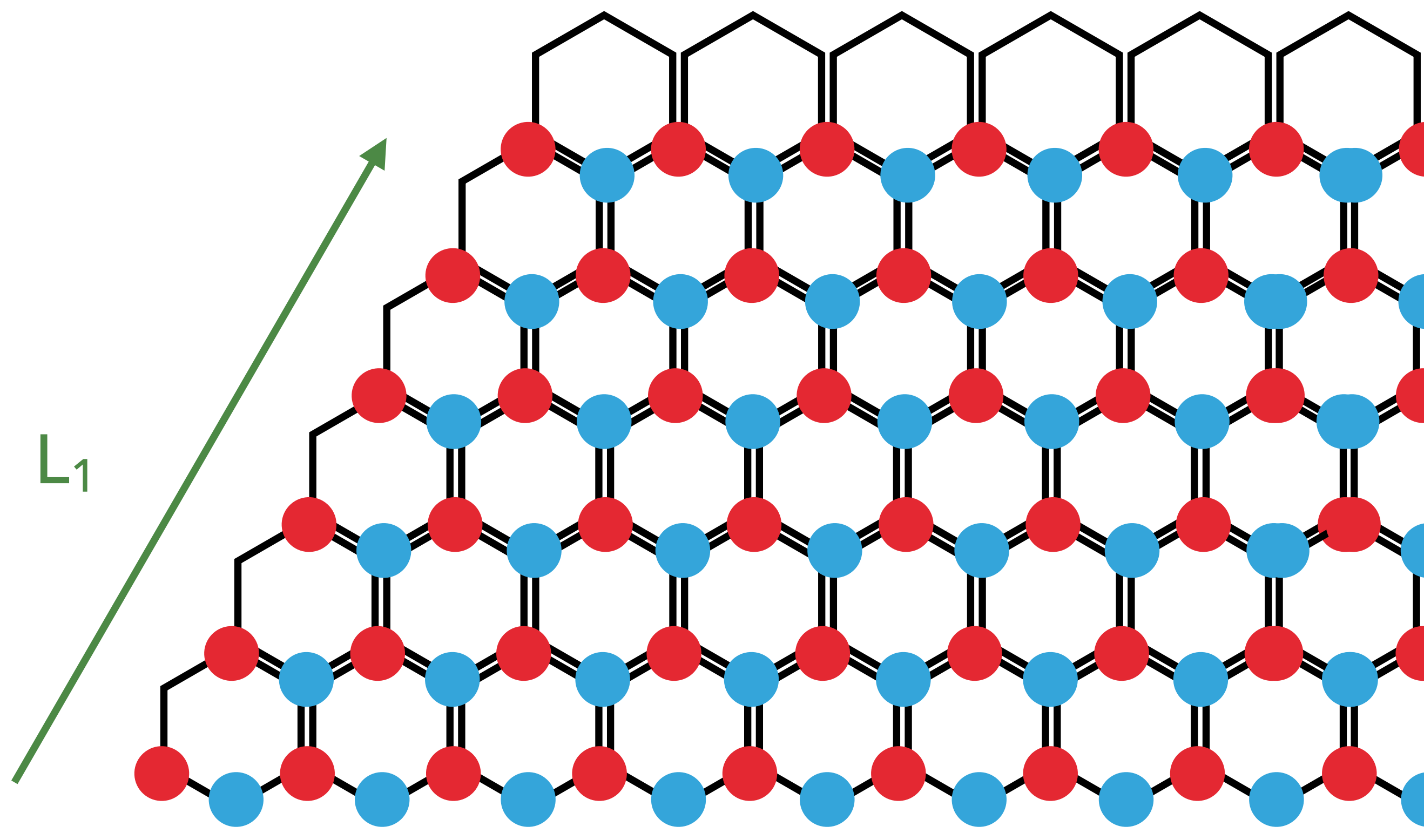
- sparse M
- simulate directly at $m_s = 0$
- Cost $\sim V^{5/4}$

BRILLOUIN ZONE



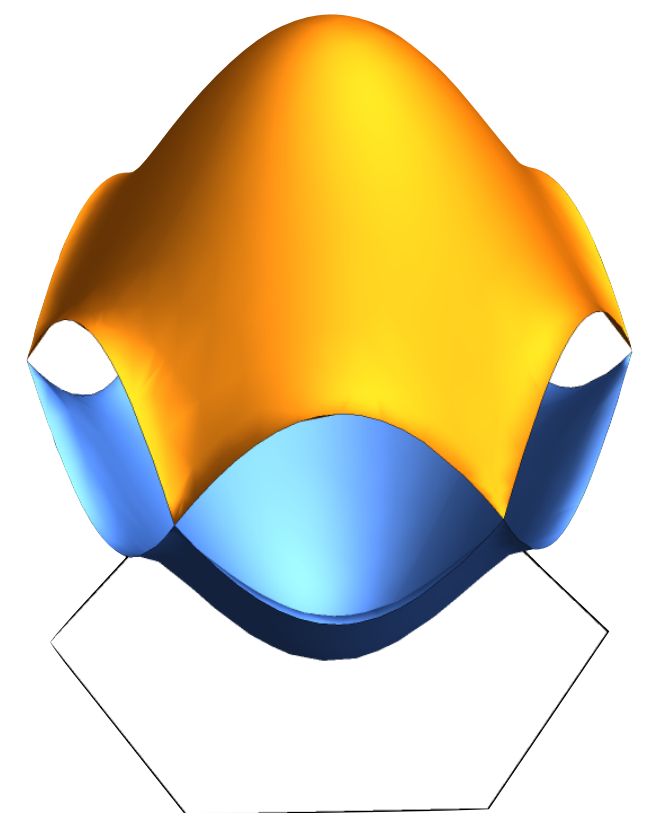
$$L_2 = 27$$

$L_1 = 6$



$$L_2$$

- $L_1 = L_2$ for graphene sheets. Can also make nanotubes
- Would-be Dirac points are directly accessible when $L_1, L_2 \equiv 0 \pmod{3}$



HMC + ERGODICITY

HYBRID MONTE CARLO

Multiply by $\mathbb{1} = \frac{1}{(2\pi)^{N/2}} \int \mathcal{D}p e^{-\frac{p^2}{2}}$

Sample p according to the gaussian

Evolve all the ϕ according to

$$\mathcal{H} = \frac{p^2}{2} + \frac{1}{2} \phi V^{-1} \phi - \log \det M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]$$

Accept / reject as in Metropolis-Hastings,
with \mathcal{H} as the energy

Refresh p according to the gaussian...

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability

Generate an *ensemble* for fixed U, β, N_t, \dots

$$\{\phi_1, \phi_2, \dots, \phi_N\}$$

Hybrid Monte Carlo

Estimate any observable

$$\begin{aligned} \langle \mathcal{O} \rangle &= \frac{1}{\mathcal{Z}} \int d\mathcal{Z} \mathcal{O}[\phi] \\ &= \frac{1}{N} \sum_i^N \mathcal{O}[\phi_i] + \mathcal{O}\left(\frac{1}{\sqrt{N}}\right) \end{aligned}$$

Multiply by $\mathbb{1} = \frac{1}{(2\pi)^{N/2}} \int \mathcal{D}p e^{-\frac{p^2}{2}}$

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability

Sample p according to the gaussian

Generate an *ensemble* for fixed U, β, N_t, \dots

Evolve all the ϕ according to

$$\{\phi_1, \phi_2, \dots, \phi_N\}$$

$$\mathcal{H} = \frac{p^2}{2} + \frac{1}{2} \phi V^{-1} \phi - \log \det M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]$$

Hybrid Monte Carlo

Accept / reject as in Metropolis-Hastings,
with \mathcal{H} as the energy

Estimate any observable

$$\begin{aligned} \langle \mathcal{O} \rangle &= \frac{1}{\mathcal{Z}} \int d\mathcal{Z} \mathcal{O}[\phi] \\ &= \frac{1}{N} \sum_i^N \mathcal{O}[\phi_i] + \mathcal{O} \left(\frac{1}{\sqrt{N}} \right) \end{aligned}$$

Refresh p according to the gaussian...

ERGODICITY OF HYBRID MONTE CARLO

Multiply by $\mathbb{1} = \frac{1}{(2\pi)^{N/2}} \int \mathcal{D}p e^{-\frac{p^2}{2}}$

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Sample p according to the gaussian

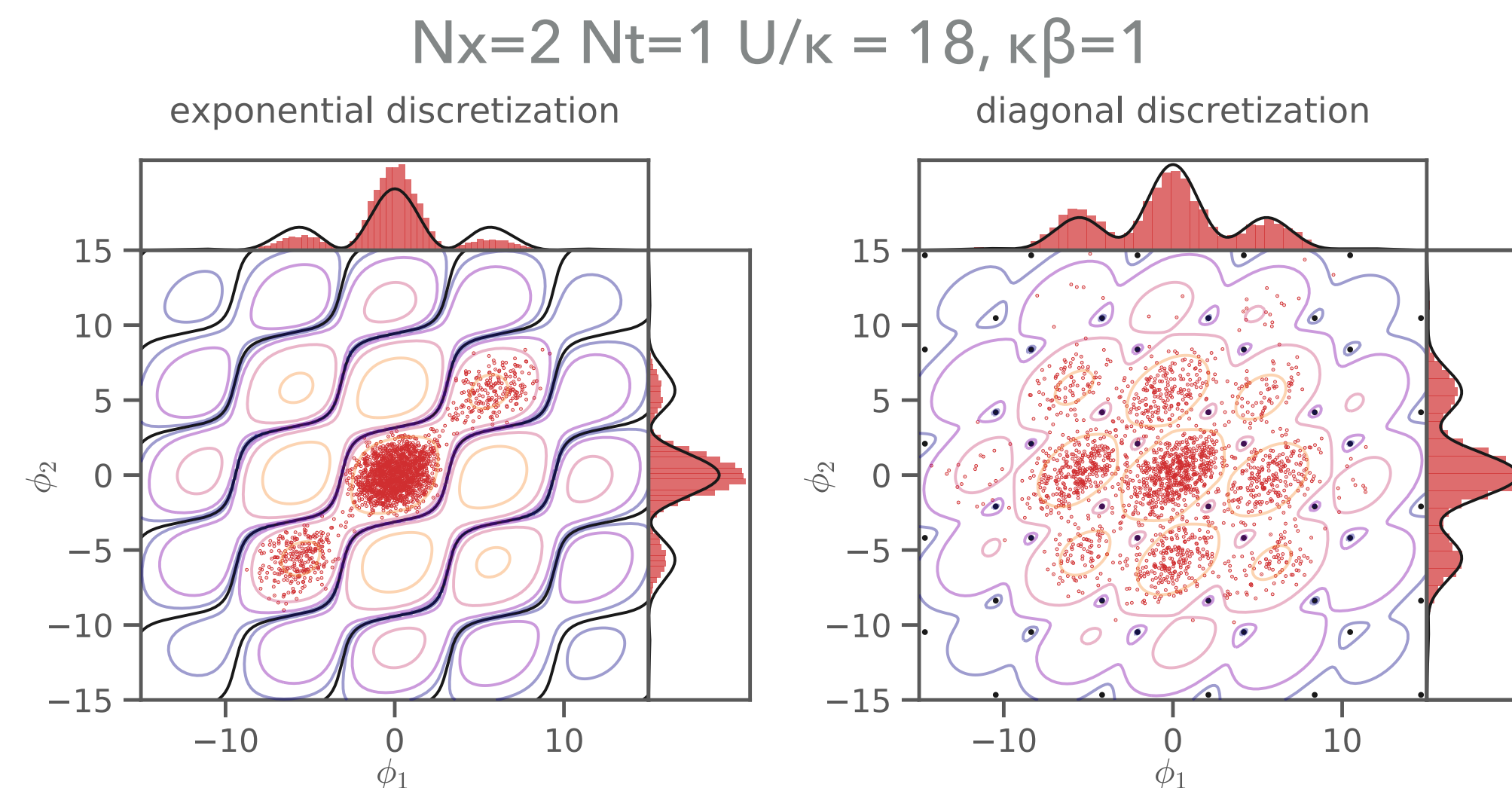
Evolve all the ϕ according to

$$\mathcal{H} = \frac{p^2}{2} + \frac{1}{2} \phi V^{-1} \phi - \log \det M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]$$

Potential

If the potential has infinite codimension-1 surfaces (the determinant vanishes), the evolution might not be able reach all ϕ , and the method has a bias.

- ▶ Depends on the discretization of the fermion operator M
- ▶ When $\mu \neq 0$ action is complex



By universality, barriers appear in the continuum limit

CHOICE OF DISCRETIZATION IS A BALANCING ACT

Exponential $M[\phi]_{x't',xt} = \delta_{x'x} \delta_{t't} - [e^h]_{x'x} e^{i\phi_{xt}} B_{t'} \delta_{t',t+1}$

Meng, Lang, Wessel, Assaad, Muramatsu, Nature 464 847-851 (2010) 10.1038/nature08942

(complex ϕ) Beyl, Goth, Assaad, PRB 97 085144 (2018) 10.1103/PhysRevB.97.085144

Ulybyshev and Valgushev, 1712.02188

Diagonal $M[\phi]_{x't',xt} = (\delta_{x'x} - h_{x'x}) \delta_{t't} - e^{i\phi_{xt}} B_{t'} \delta_{t',t+1}$

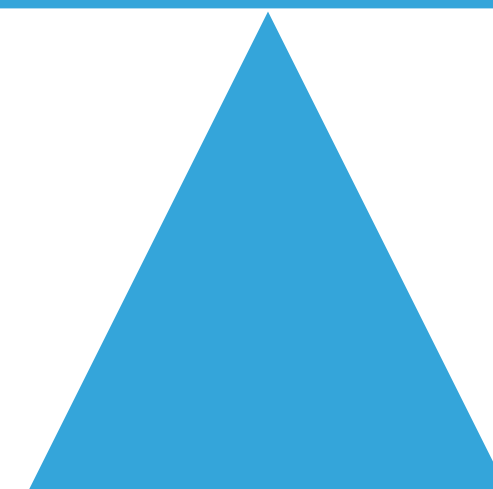
Brower, Rebbi, Schaich PoS LATTICE2011 056 (2011) 1204.5424

Luu and Lähde, PRB 93, 155106 (2016) 10.1103/PhysRevB.93.155106

EB, Körber, Krieg, Labus, Lähde, Luu, Lattice 2017 10.1051/epjconf/201817503009

Ergodicity

Chiral Symmetry



Beyl, Goth, Assaad (above)

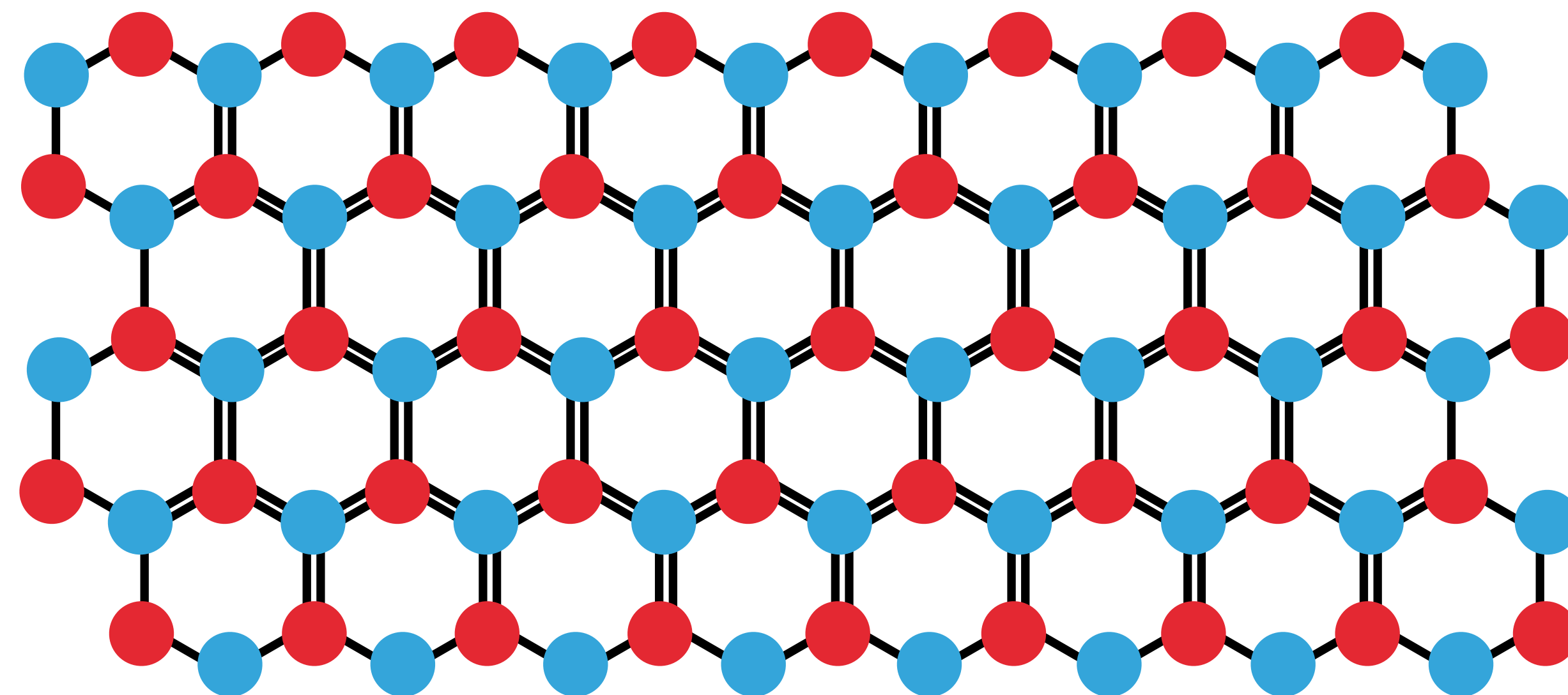
Wynen, EB, Körber, Lähde, Luu PRB 100 075141 10.1103/PhysRevB.100.075141

$$M_{\mathbf{x}'t', \mathbf{x}t} = \delta_{\mathbf{x}'\mathbf{x}} \left(\delta_{t', t-1} - e^{i\phi_{\mathbf{x}'t'}} \delta_{t', t} \right)$$

$$M_{\mathbf{x}'t', \mathbf{x}t} = \delta_{\mathbf{x}'\mathbf{x}} \left(\delta_{t', t} - e^{i\phi_{\mathbf{x}'t'}} \delta_{t', t+1} \right)$$

$$M_{\mathbf{x}'t', \mathbf{x}t} = -h_{\mathbf{x}', \mathbf{x}} \delta_{t' t}$$

$$M_{\mathbf{x}'t', \mathbf{x}t} = -h_{\mathbf{x}', \mathbf{x}} \delta_{t' t}$$



- time continuum limit $\delta \rightarrow 0$ is faster than forward (or backward) differencing.
- produces a dynamically-generated normal-ordering term
- simulate directly at $m_s = 0$
- HMC scales very mildly

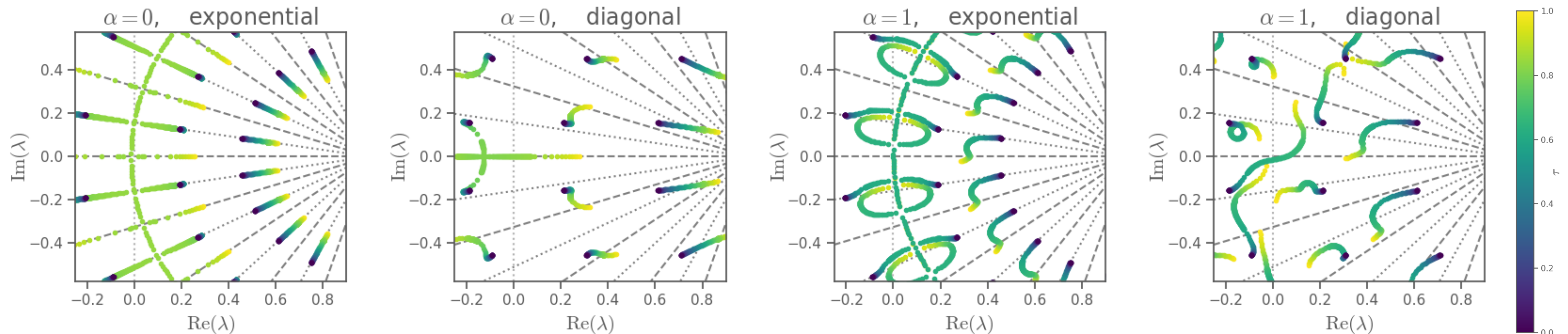
SYMMETRY CONSTRAINS M EIGENSPECTRUM

Exponential $M[\phi]_{x't',xt} = \delta_{x'x} \delta_{t't} - [e^h]_{x'x} e^{i\phi_{xt}} B_{t'} \delta_{t',t+1}$

Eigenvalue identity of the characteristic polynomial $P[\phi]$ $P[\phi](s) \sim P[\phi](1/s^*)^*$

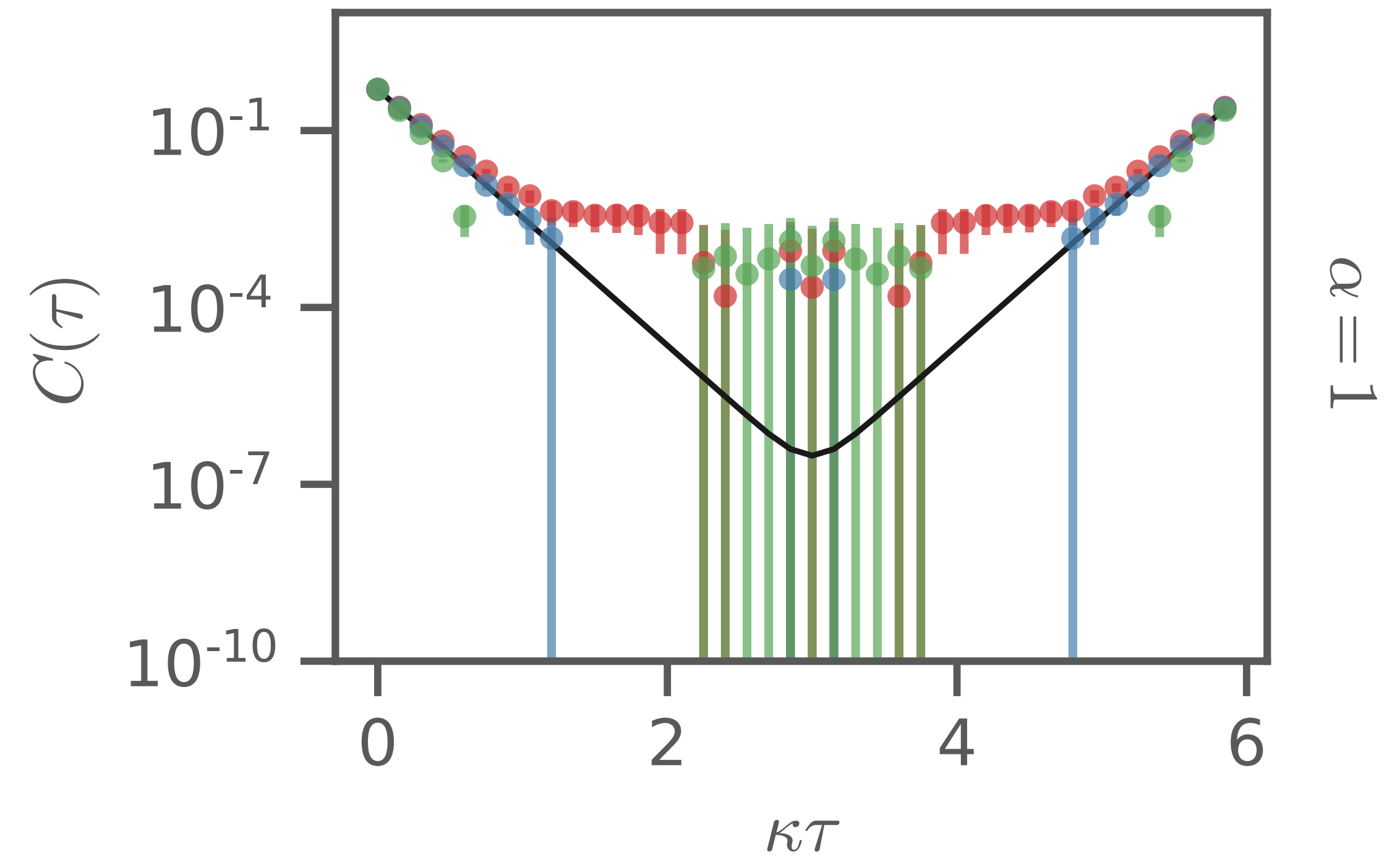
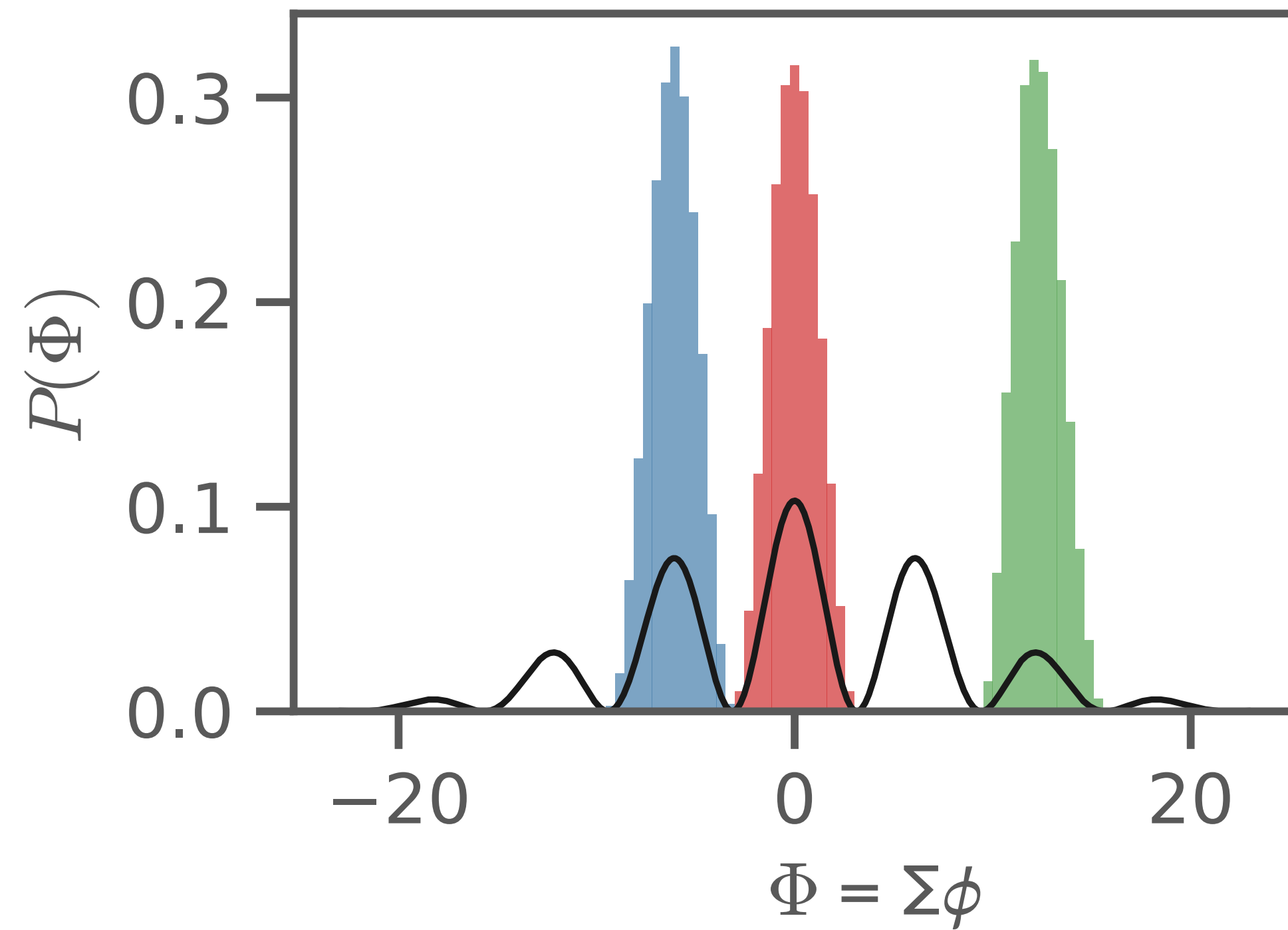
Diagonal $M[\phi]_{x't',xt} = (\delta_{x'x} - h_{x'x}) \delta_{t't} - e^{i\phi_{xt}} B_{t'} \delta_{t',t+1}$

No such identity



ERGODICITY PROBLEMS LEAD TO BIASES

$$N_x = 1, N_t = 40, U/\kappa = 10, \kappa\beta = 6$$



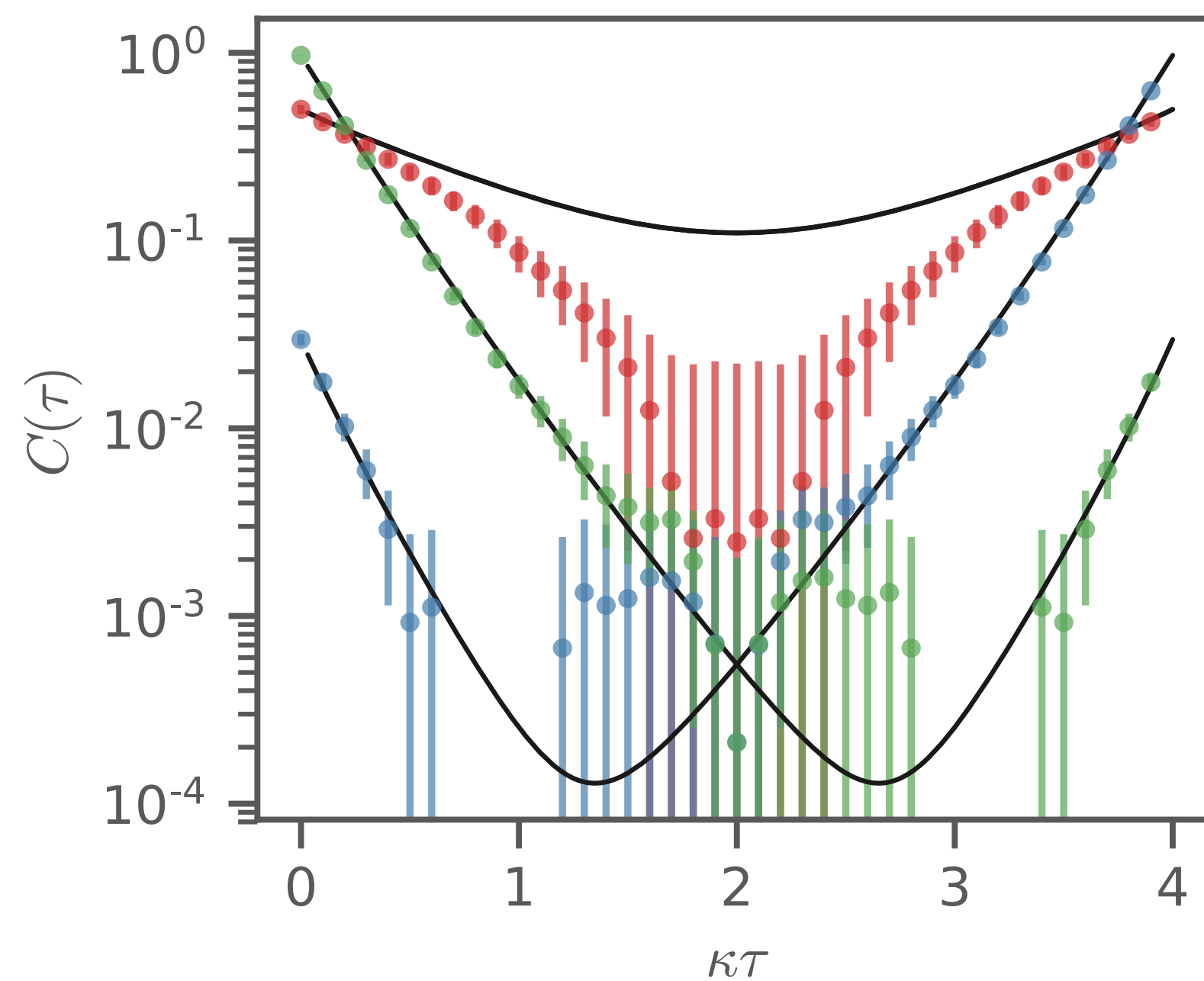
ERGODICITY PROBLEMS LEAD TO BIASES

Exponential $M[\phi]_{x't',xt} = \delta_{x'x} \delta_{t't} - [e^h]_{x'x} e^{i\phi_{xt}} B_{t'} \delta_{t',t+1}$

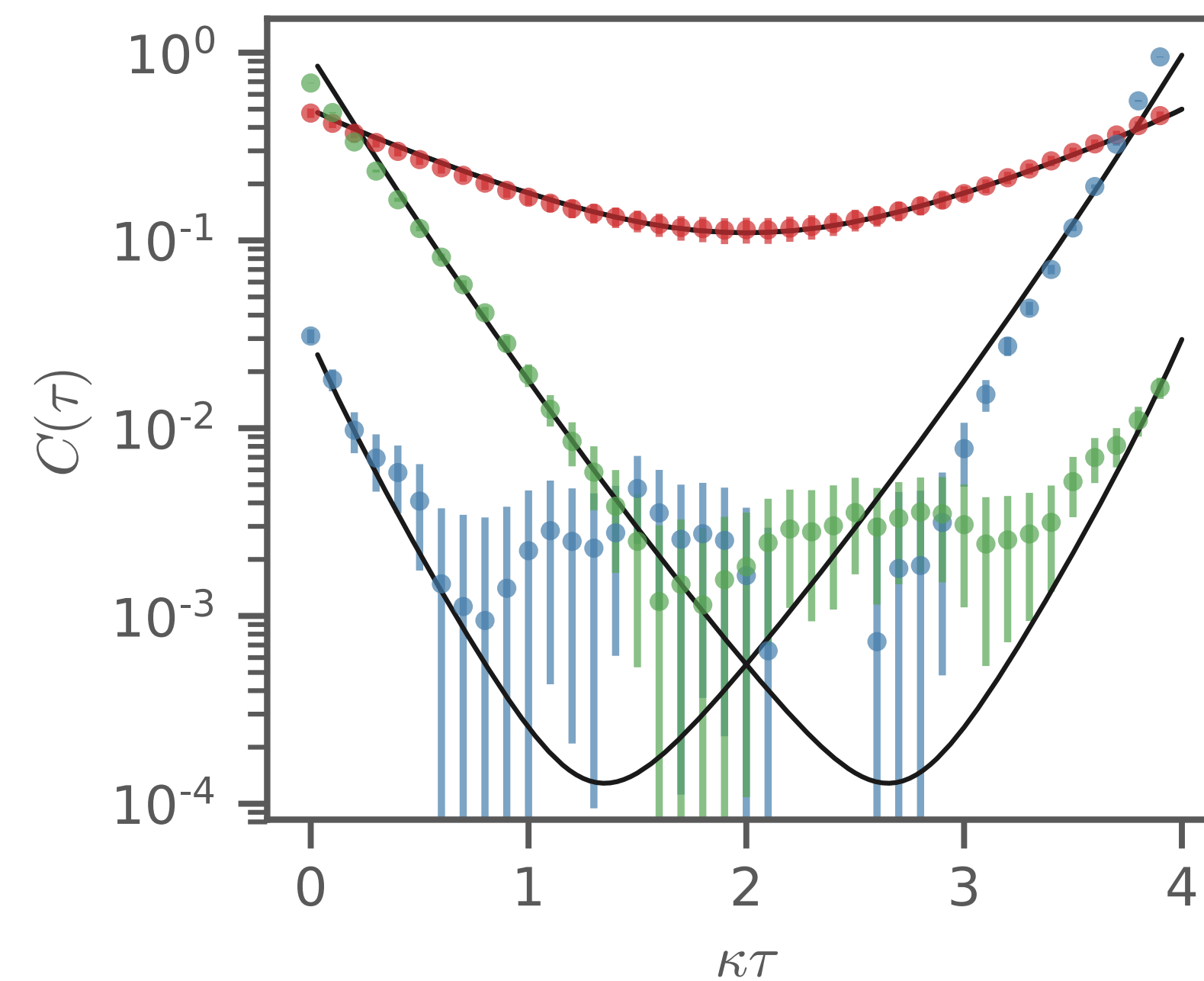
Diagonal $M[\phi]_{x't',xt} = (\delta_{x'x} - h_{x'x}) \delta_{t't} - e^{i\phi_{xt}} B_{t'} \delta_{t',t+1}$

$N_x = 4$ (square), $N_t = 40$, $U/\kappa = 4$, $\kappa\beta = 4$

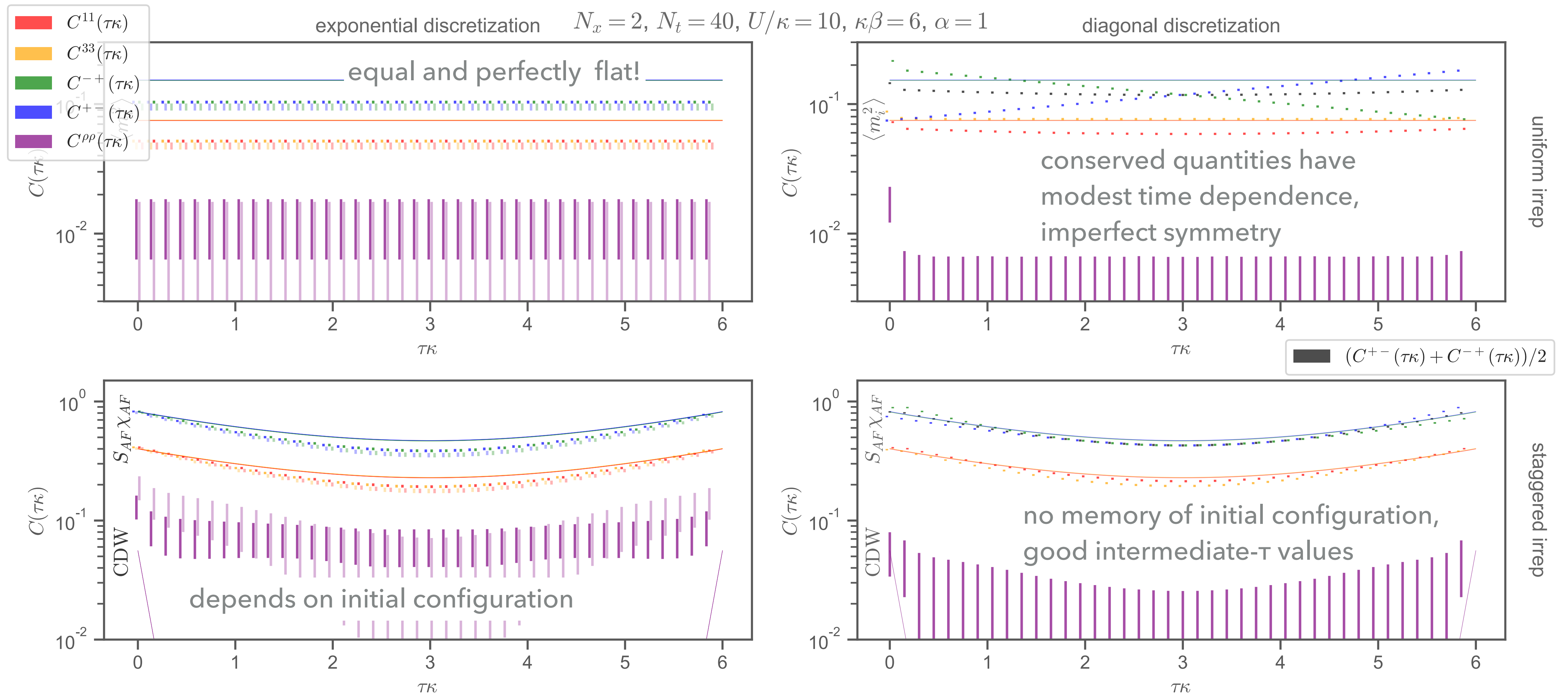
exponential discretization



diagonal discretization



CHOICE OF DISCRETIZATION IS A BALANCING ACT



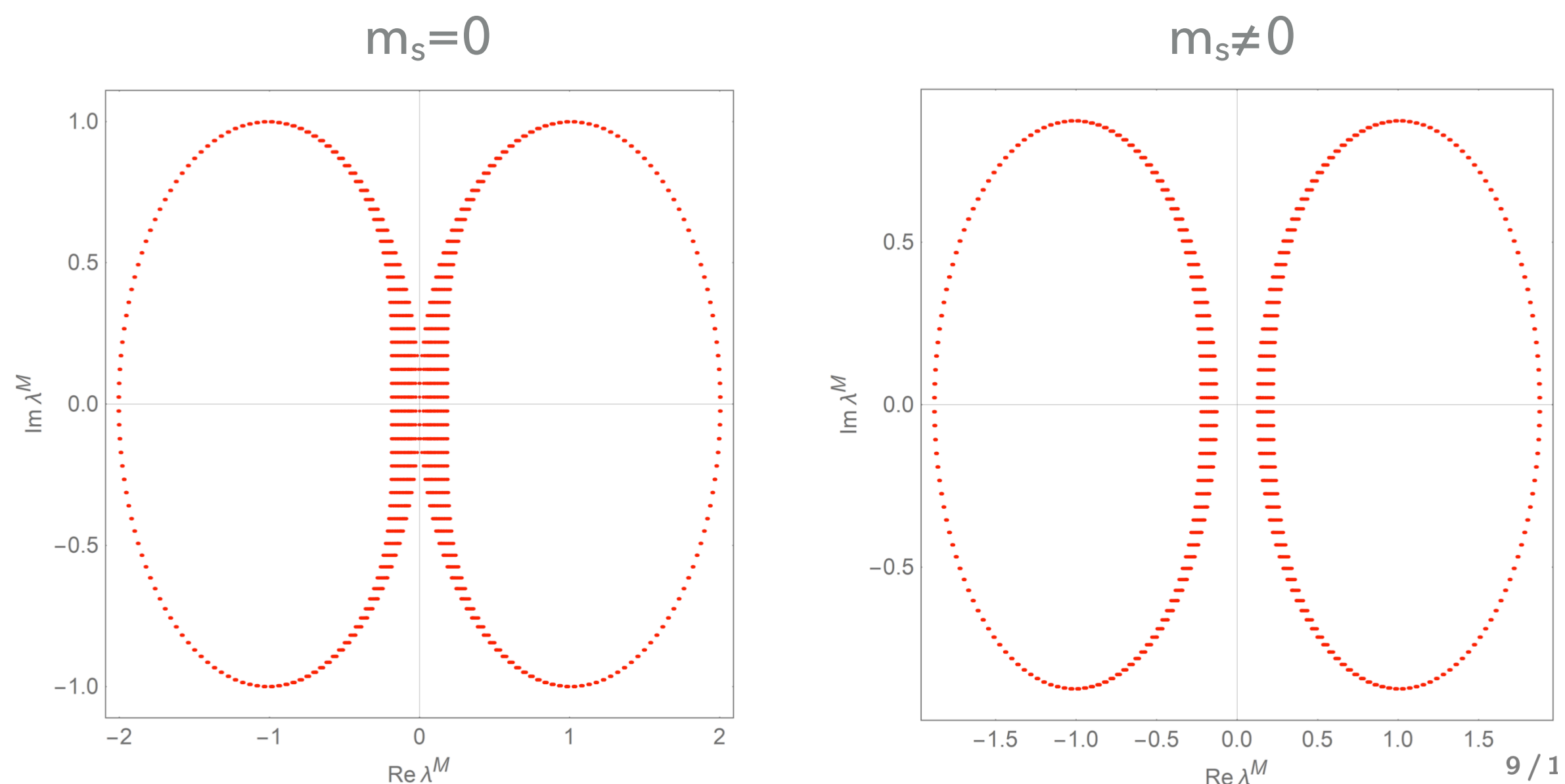
HMC + SOLVERS

ACCELERATING HYBRID MONTE CARLO

Multiply by $\mathbb{1} = \frac{\det(M[m_s]M^\dagger[m_s])}{\det(M[m_s]M^\dagger[m_s])}$

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det(M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability



Generate an ensemble for fixed U, β, N_t, \dots

$$\{\phi_1, \phi_2, \dots, \phi_N\}$$

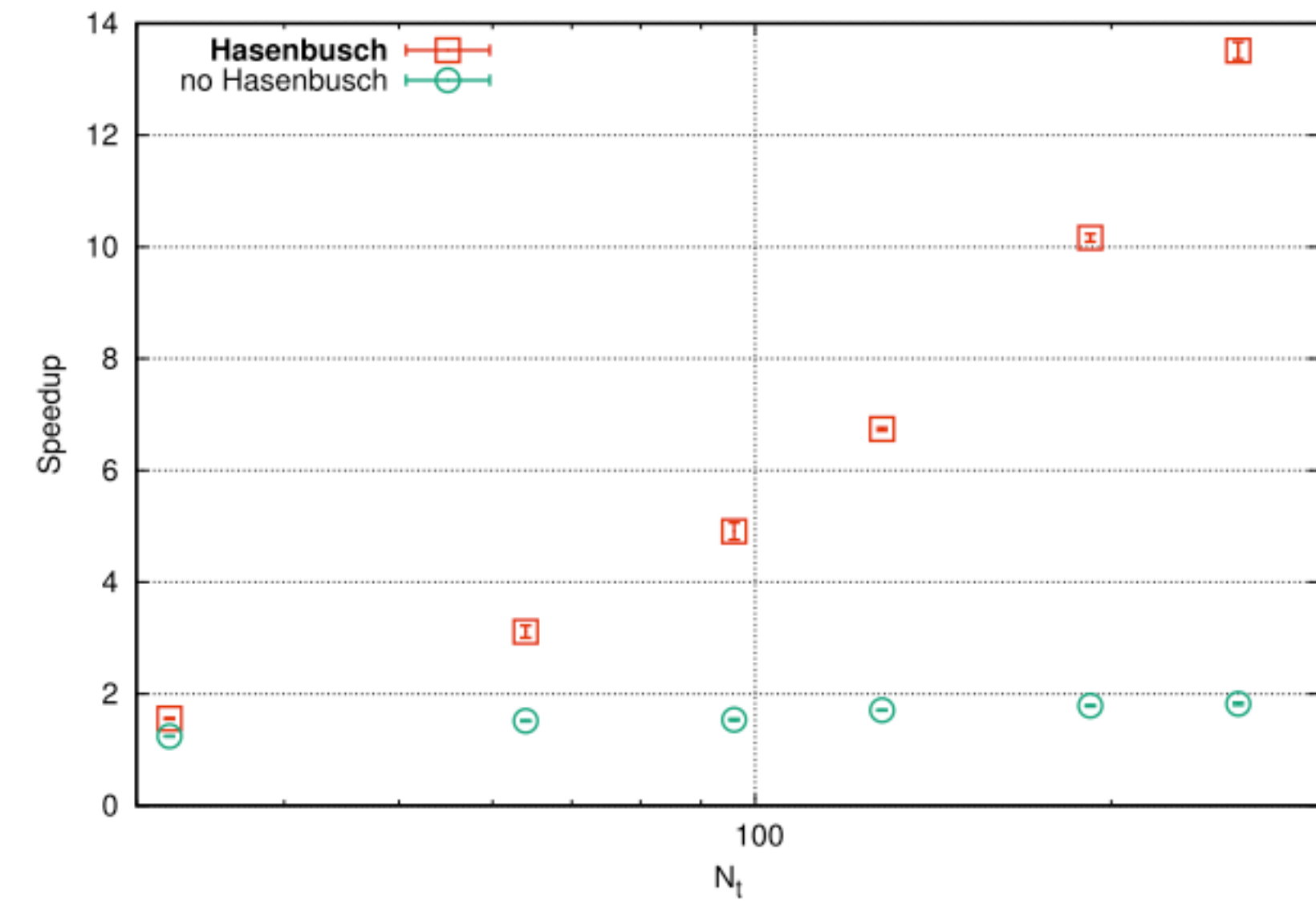
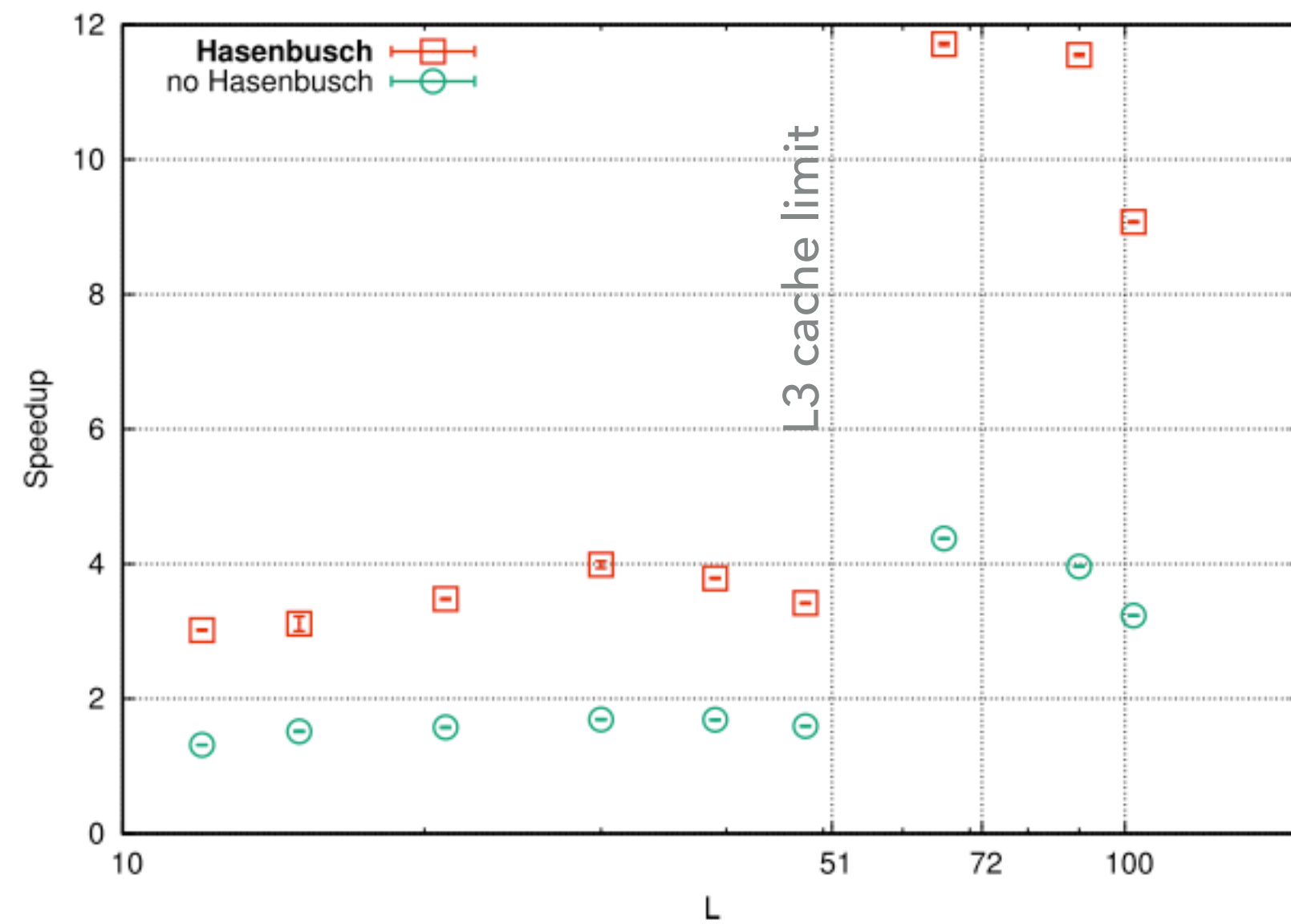
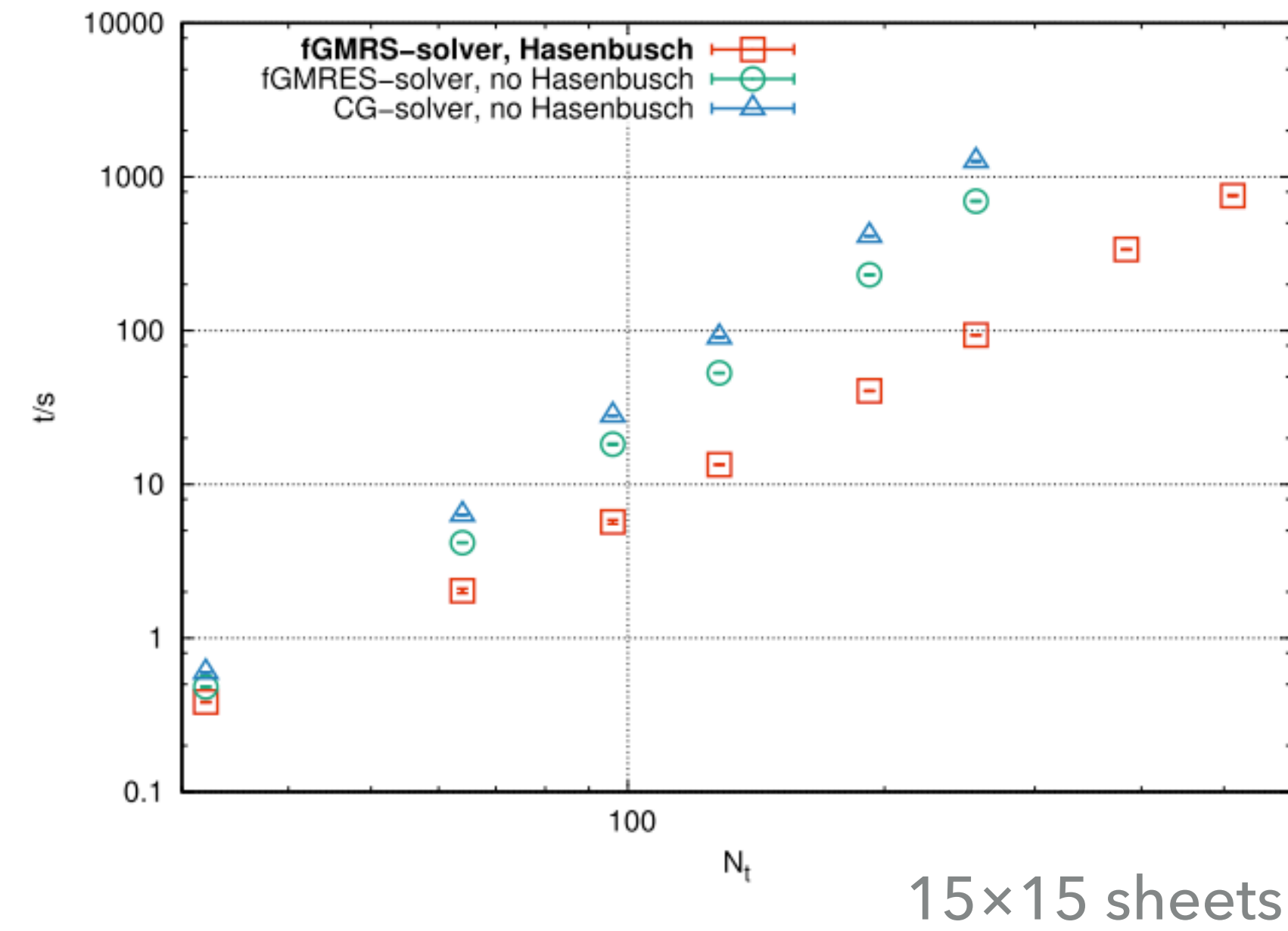
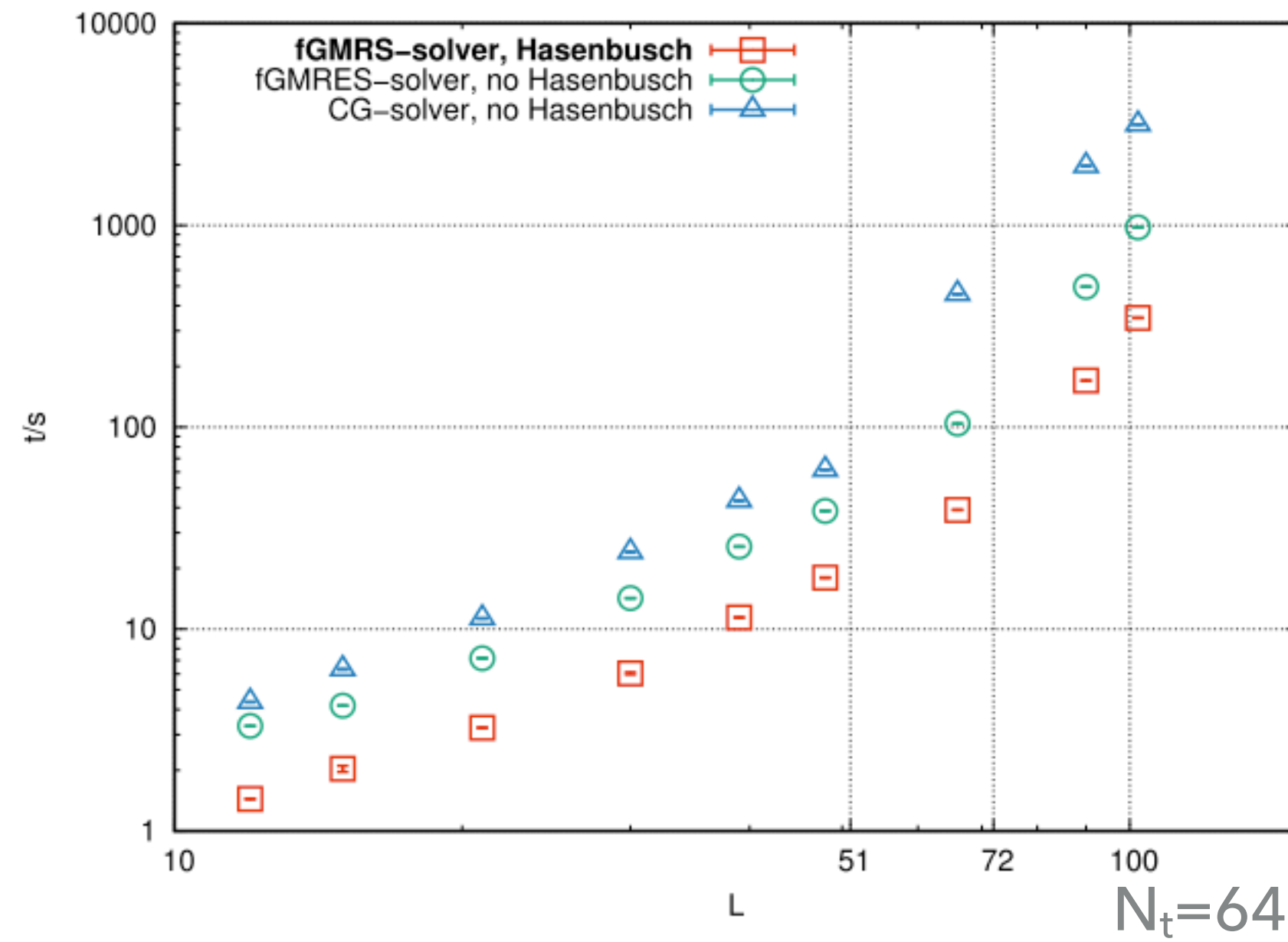
Hybrid Monte Carlo

Estimate any observable

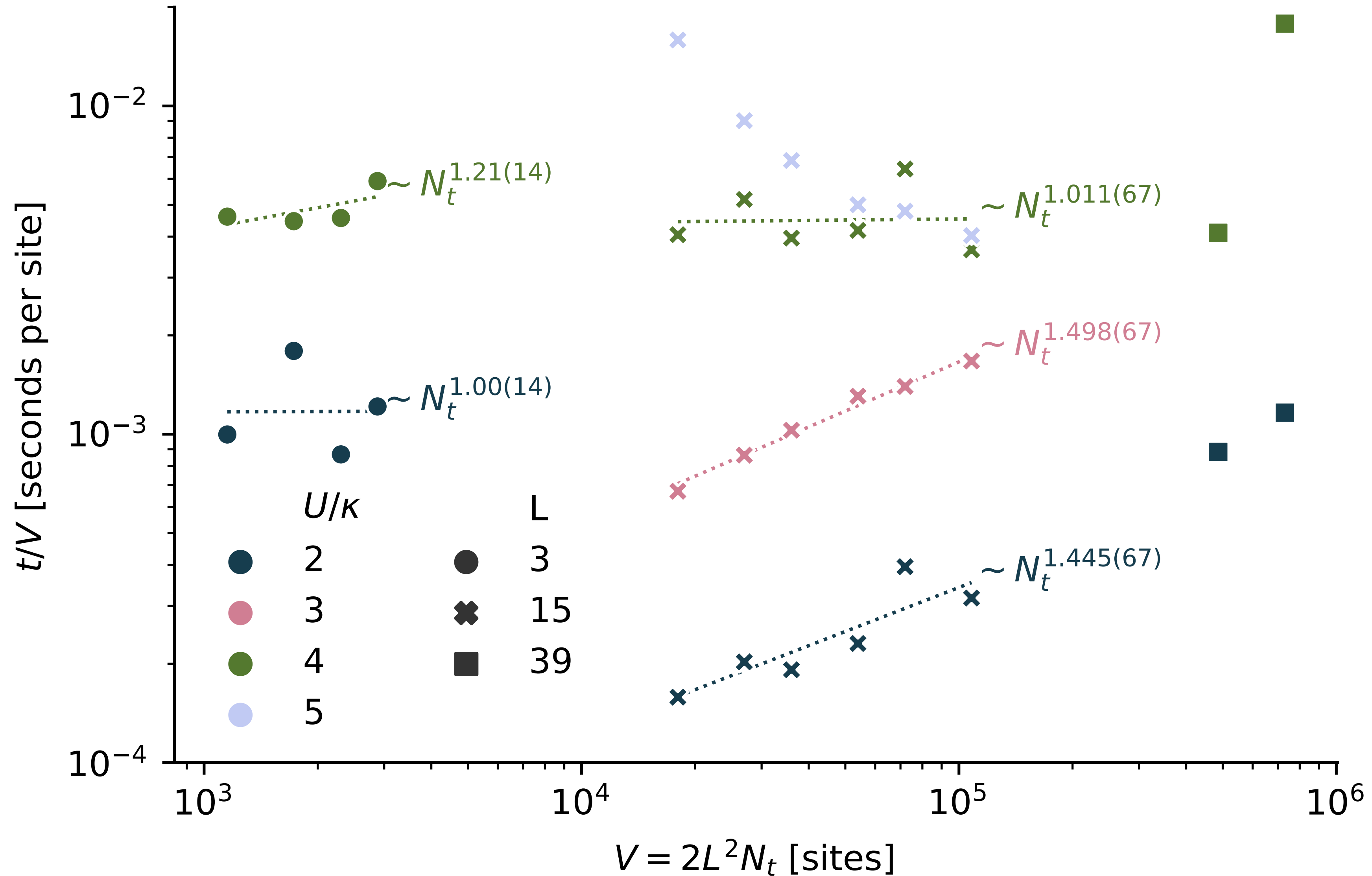
$$\begin{aligned} \langle \mathcal{O} \rangle &= \frac{1}{\mathcal{Z}} \int d\mathcal{Z} \mathcal{O}[\phi] \\ &= \frac{1}{N} \sum_i^N \mathcal{O}[\phi_i] + \mathcal{O}\left(\frac{1}{\sqrt{N}}\right) \end{aligned}$$

- Condition number is much better
- $\det(m_s = 0) / \det(m_s \neq 0)$ is close to 1

SCALING OF HASENBUSCH SOLVER



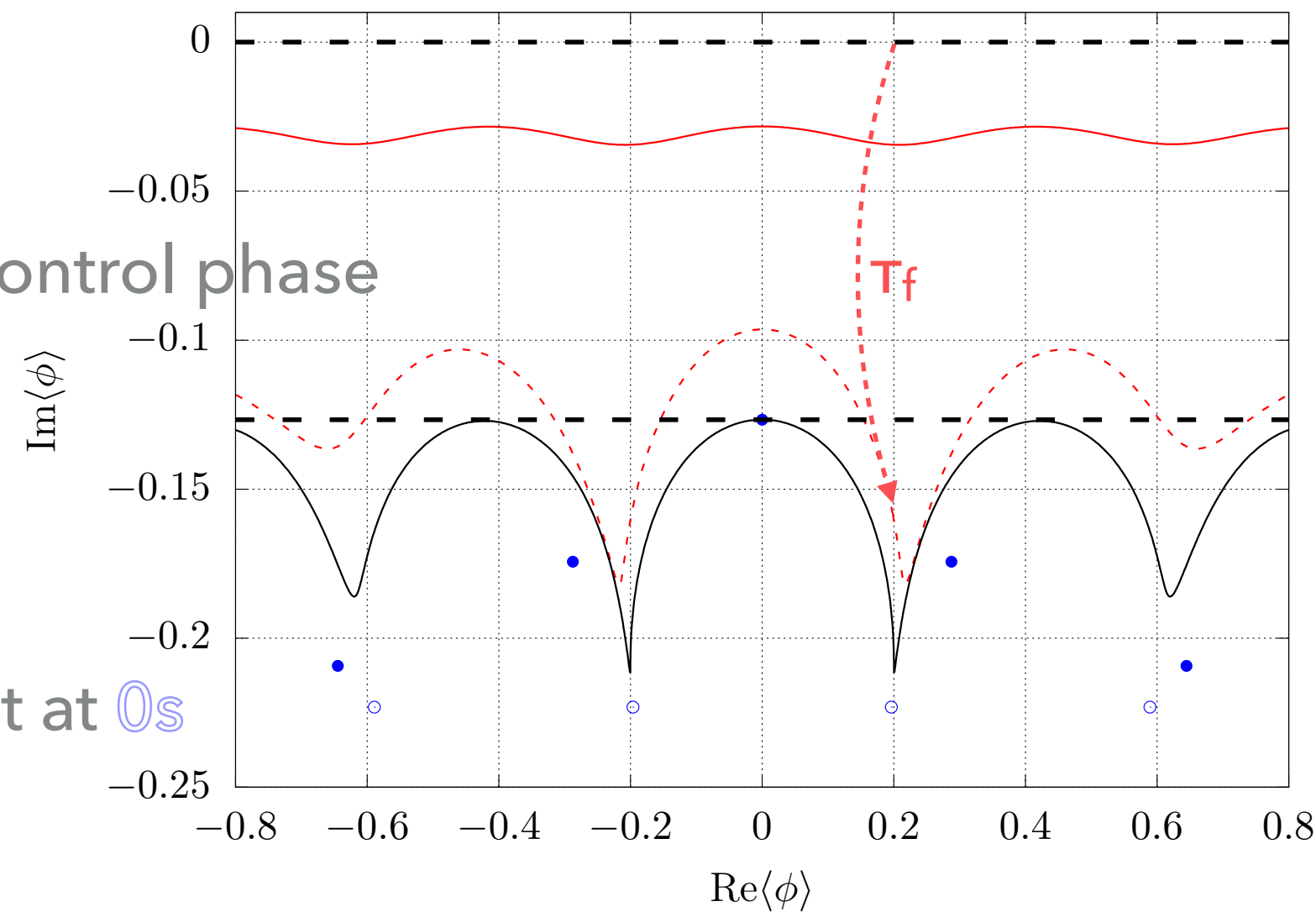
HMC IS EXPECTED TO SCALE LIKE $V^{5/4}$



FLOW + ERGODICITY

LEFSCHETZ THIMBLES HAVE CONSTANT SIGN

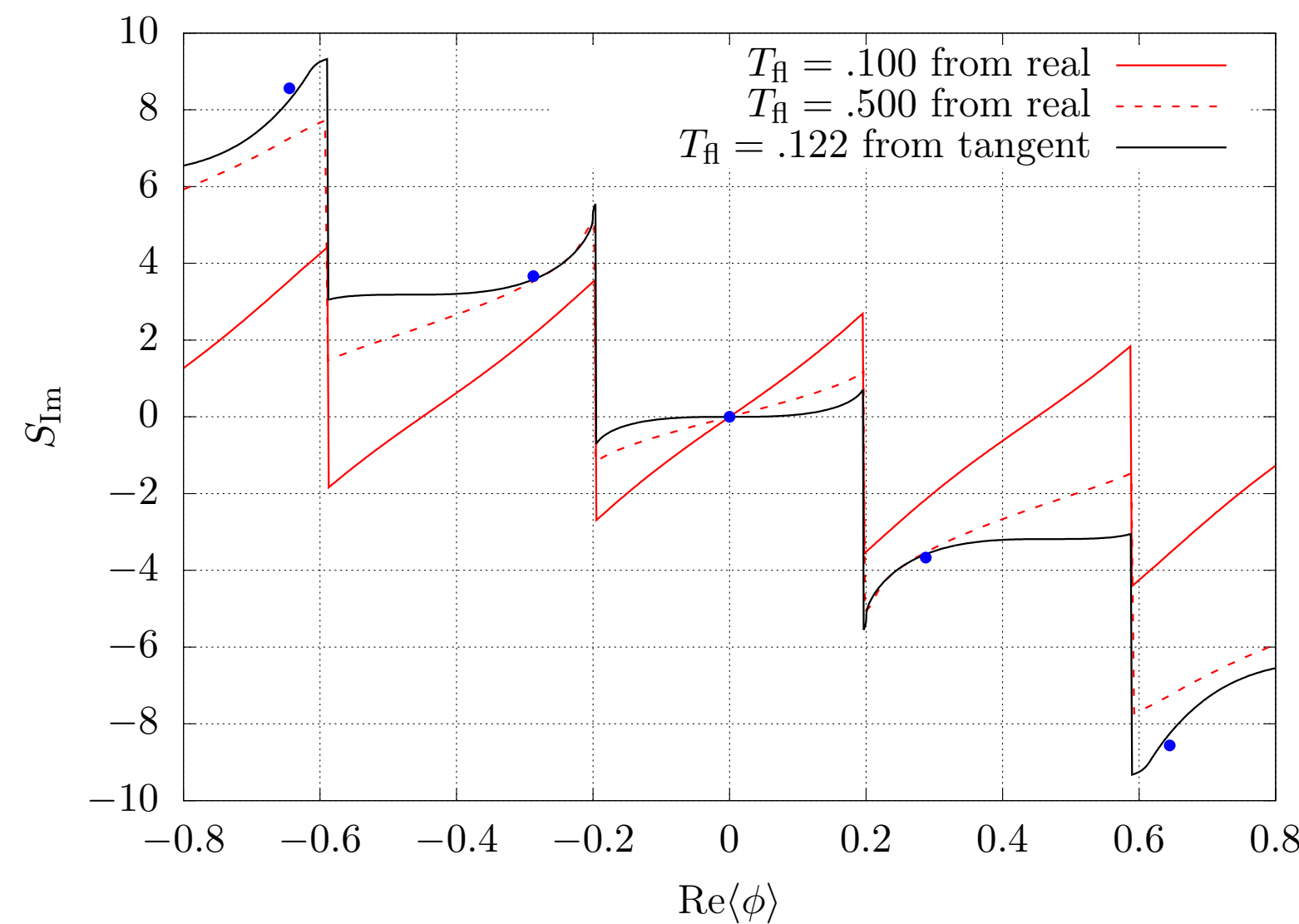
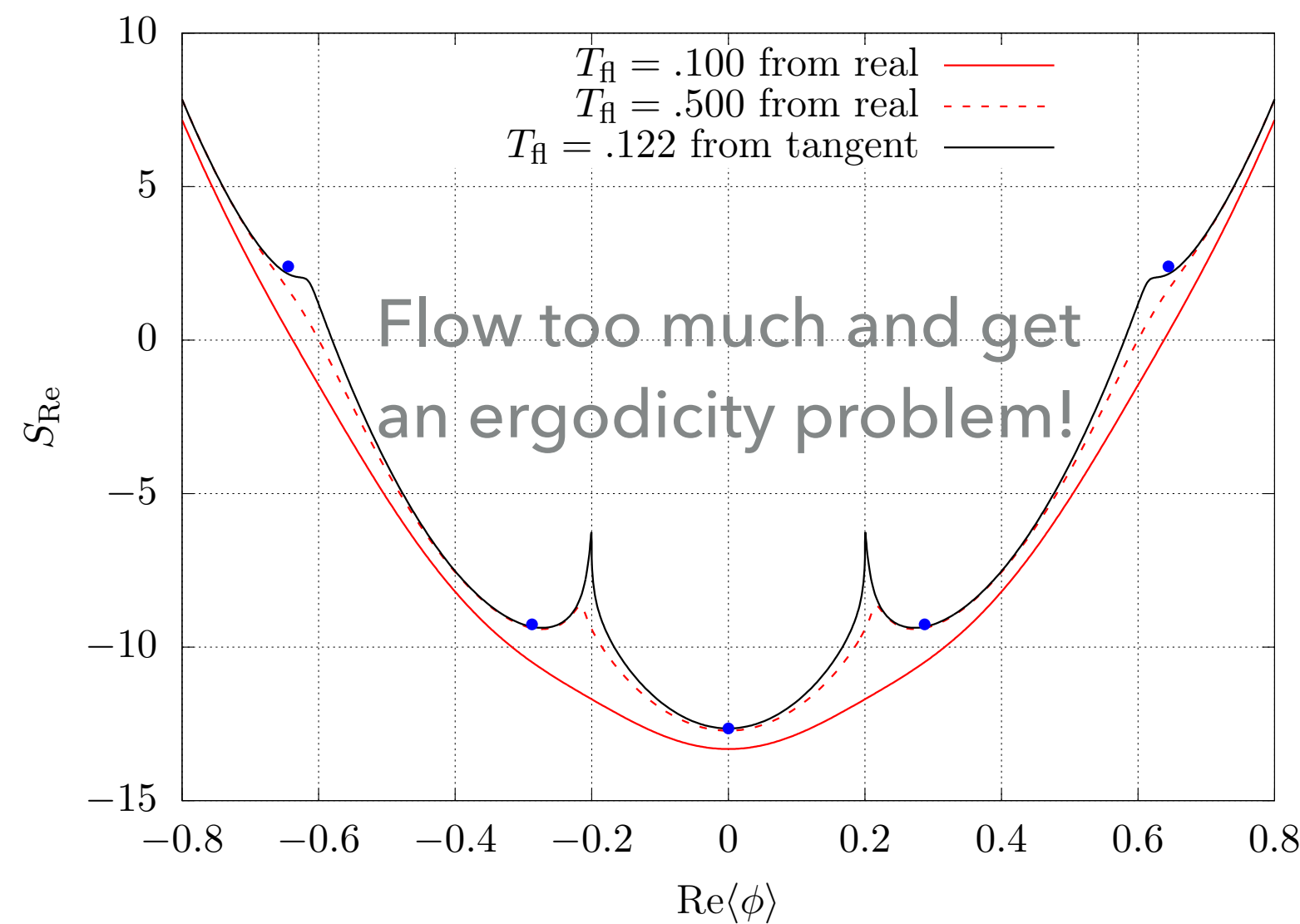
Critical points control phase



Thimbles meet at \mathcal{O}_S

Complexify ϕ , try to integrate on steepest-descent analogues called *Lefschetz thimbles*, fixed points of holomorphic flow

$$\frac{d\phi^R}{d\tau_f} = \pm \frac{\partial S^R}{\partial \phi_i^R} = \pm \frac{\partial S^I}{\partial \phi_i^I} \quad \frac{d\phi^I}{d\tau_f} = \pm \frac{\partial S^R}{\partial \phi_i^I} = \mp \frac{\partial S^I}{\partial \phi_i^R}$$



As you flow to the thimbles, phases become constant.

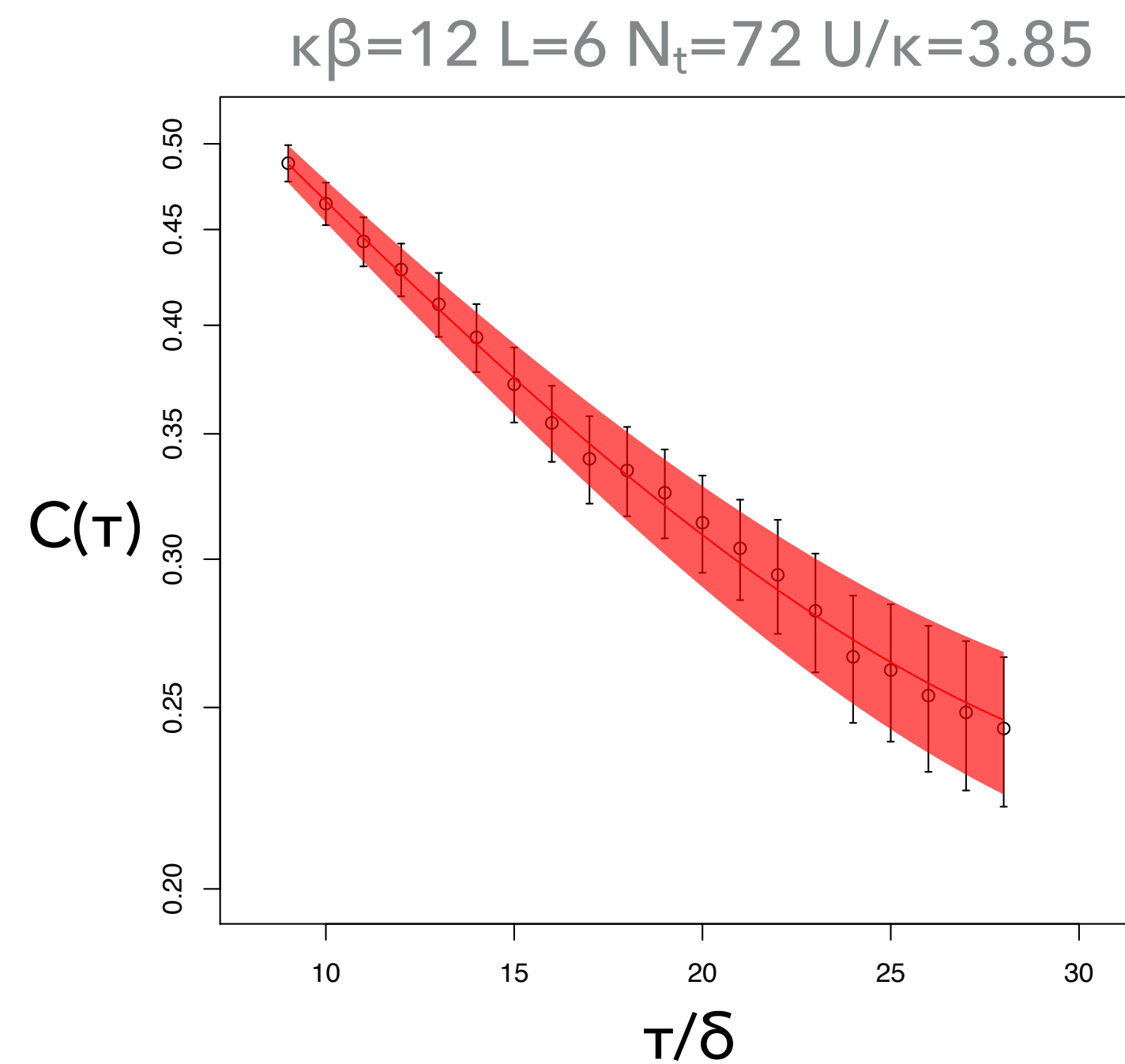
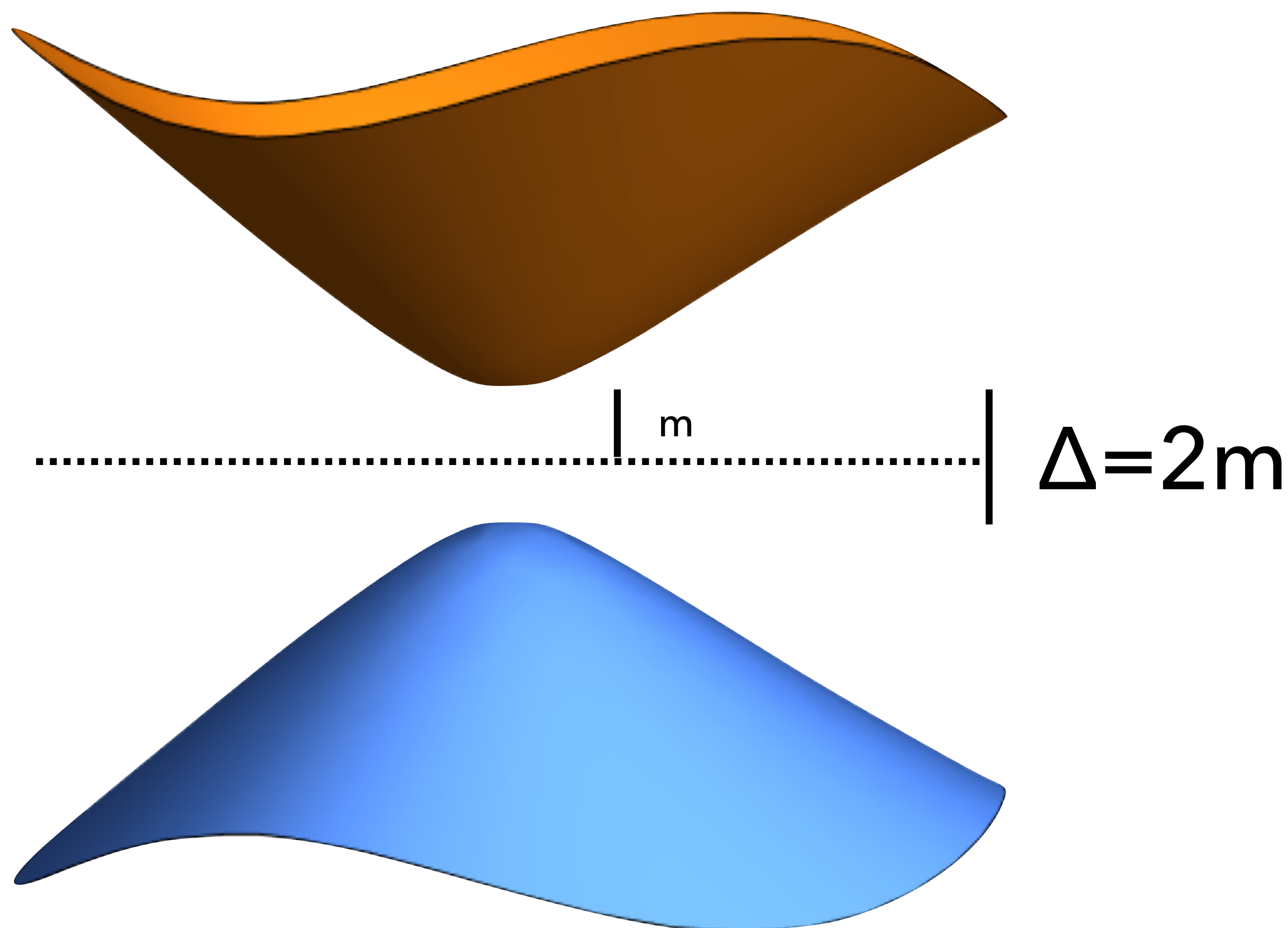
RESULTS

EFFECTIVE MASS

$$C_k^{AB}(\tau) = \left\langle \left(\sum_y e^{-iky} a_y^{A\dagger}(\tau) \right) \left(\sum_x e^{ikx} a_x^B(0) \right) \right\rangle = \frac{1}{\mathcal{Z}} \text{tr} \left[a_k^{A\dagger} e^{-H\tau} a_k^B e^{-H(\beta-\tau)} \right] = \frac{1}{\mathcal{Z}} \sum_{\alpha n} \langle \alpha | a_k^{A\dagger} | n \rangle \langle n | a_k^B | \alpha \rangle e^{-(E_n - E_\alpha)\tau} e^{-E_\alpha \beta}$$

$$\lim_{\tau \rightarrow \infty} -\partial_t \ln \lim_{\beta \rightarrow \infty} C_k^{AB}(\tau) = (E_1 - E_\Omega) = m$$

$$C_k^{AB}(\tau) = \left\langle \sum_t (M_{AB}^{-1})_{k,t+\tau;k,t} \right\rangle$$

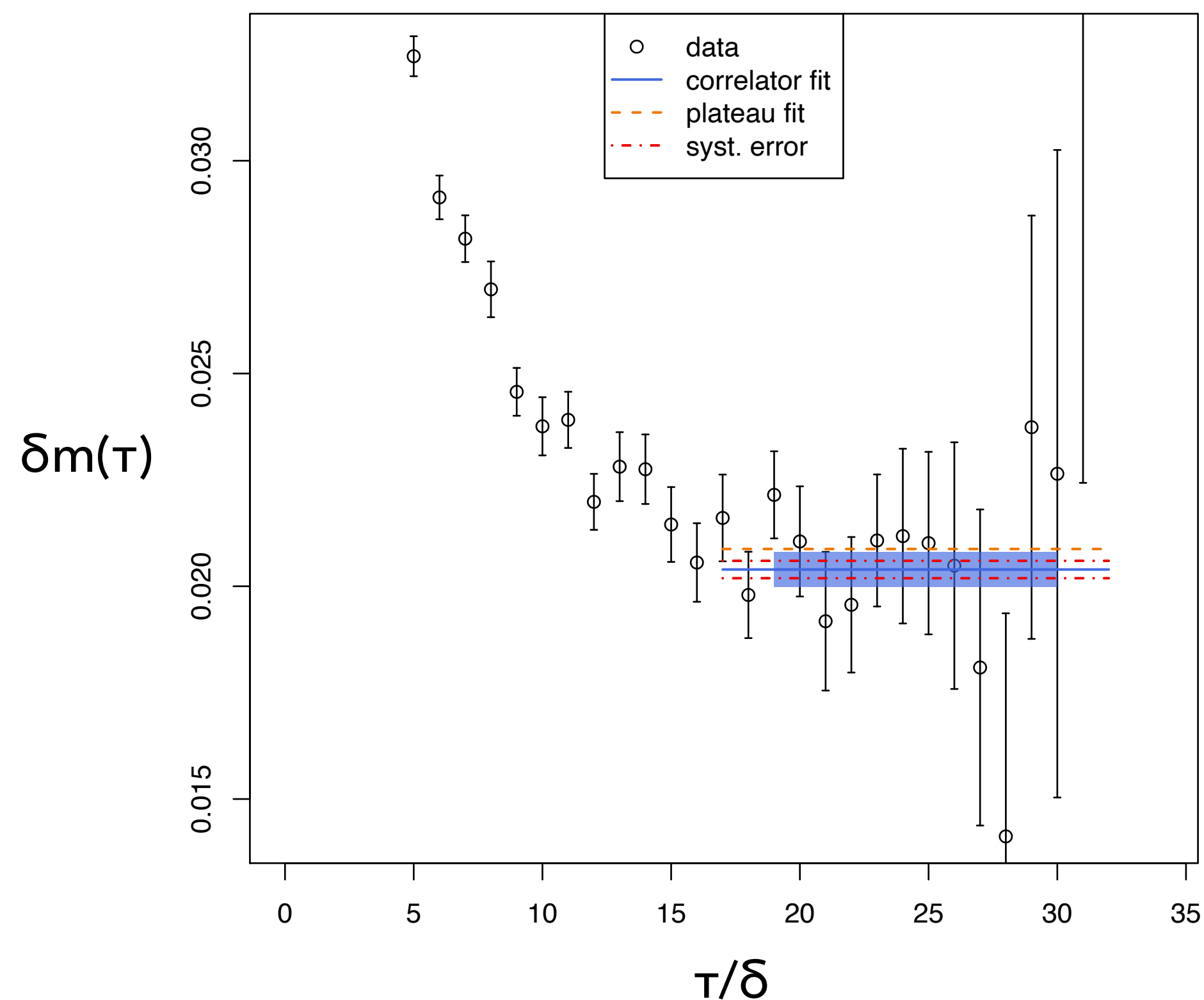


$$C_k^{AB}(\tau) = \left\langle \left(\sum_y e^{-iky} a_y^{A\dagger}(\tau) \right) \left(\sum_x e^{ikx} a_x^B(0) \right) \right\rangle = \frac{1}{\mathcal{Z}} \text{tr} \left[a_k^{A\dagger} e^{-H\tau} a_k^B e^{-H(\beta-\tau)} \right] = \frac{1}{\mathcal{Z}} \sum_{\alpha n} \langle \alpha | a_k^{A\dagger} | n \rangle \langle n | a_k^B | \alpha \rangle e^{-(E_n - E_\alpha)\tau} e^{-E_\alpha \beta}$$

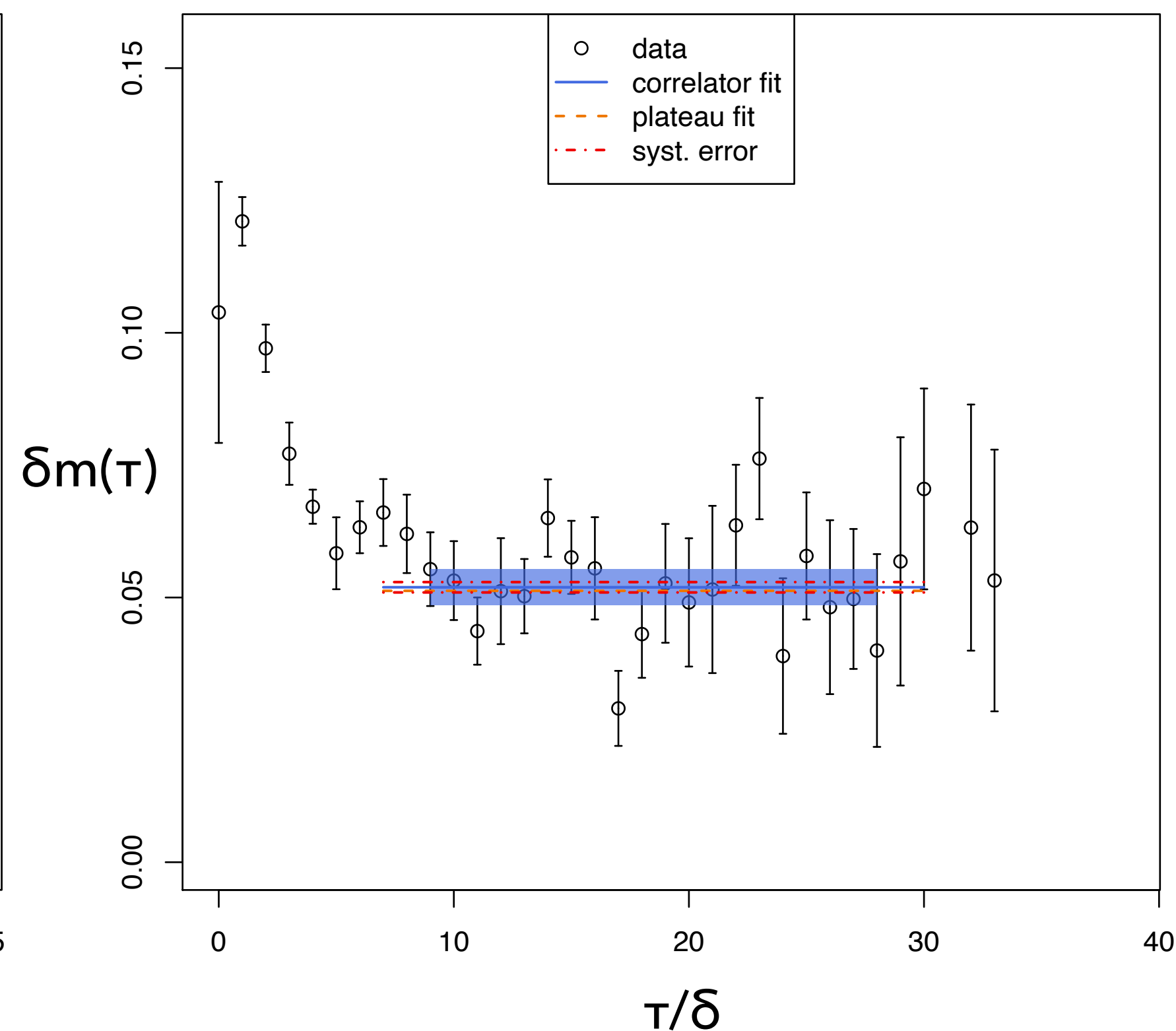
$$C_k^{AB}(\tau) = \left\langle \sum_t (M_{AB}^{-1})_{k,t+\tau;k,t} \right\rangle$$

$$\lim_{\tau \rightarrow \infty} -\partial_t \ln \lim_{\beta \rightarrow \infty} C_k^{AB}(\tau) = (E_1 - E_\Omega) = m$$

$\kappa\beta=8$ $L=15$ $N_t=64$ $U/\kappa=3.5$

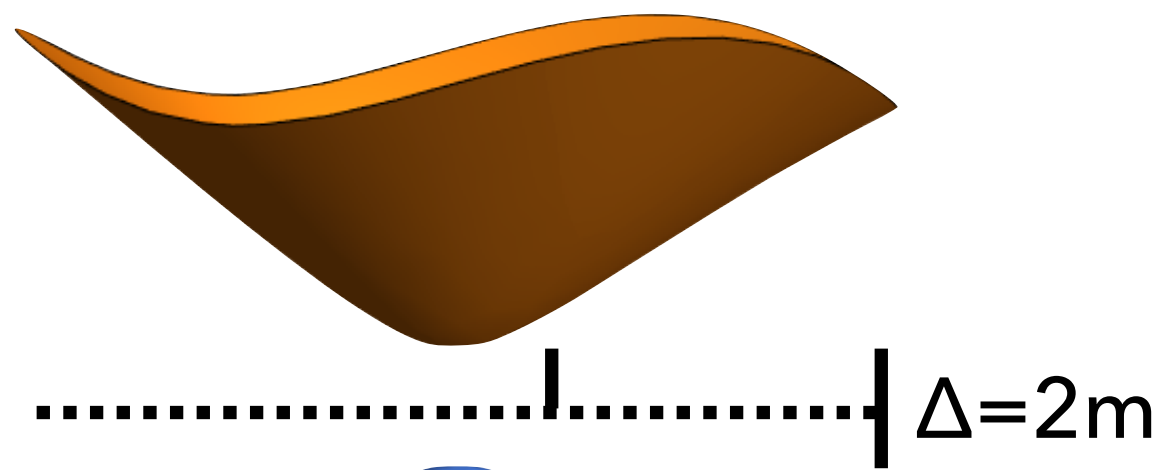


$\kappa\beta=12$ $L=6$ $N_t=72$ $U/\kappa=3.85$

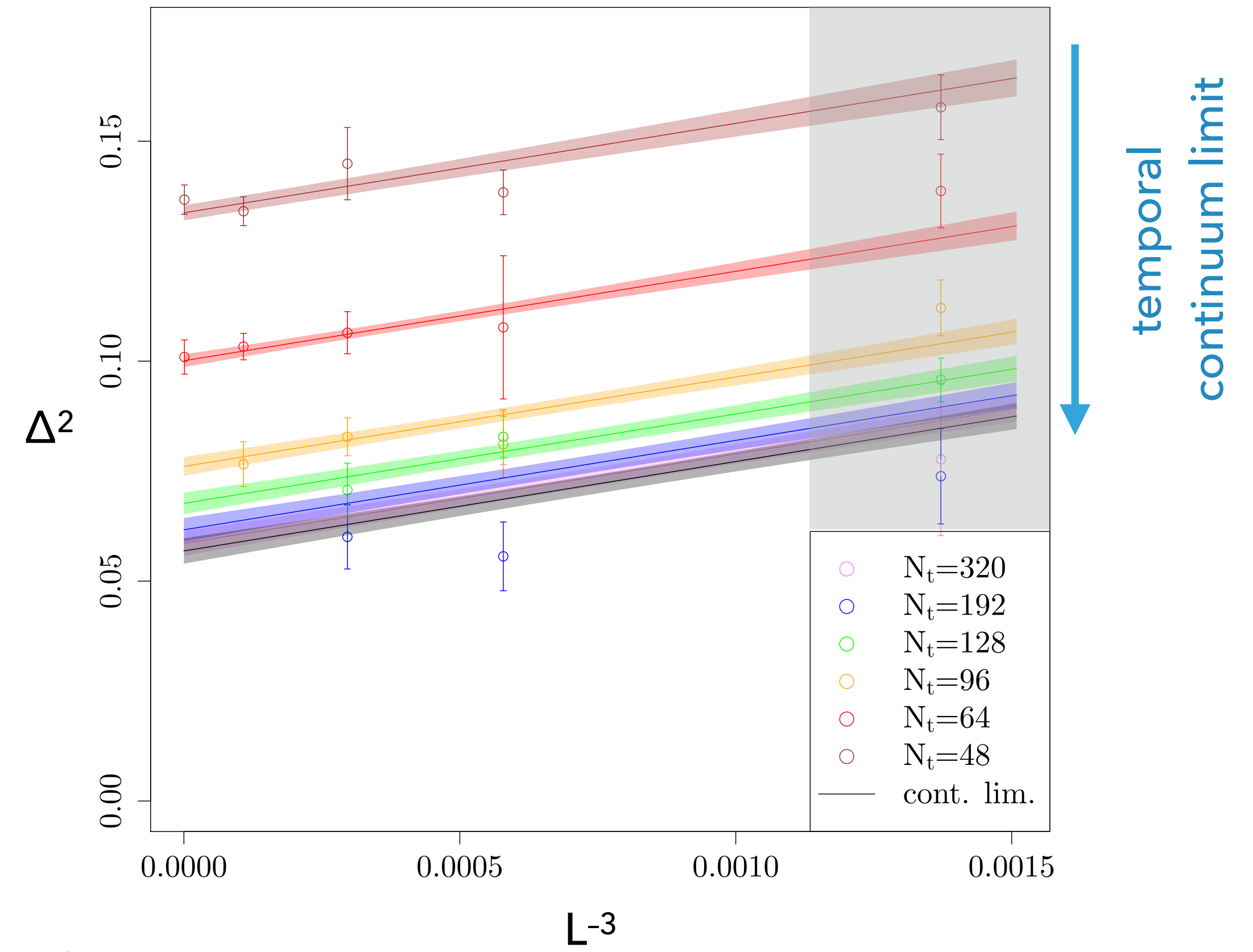
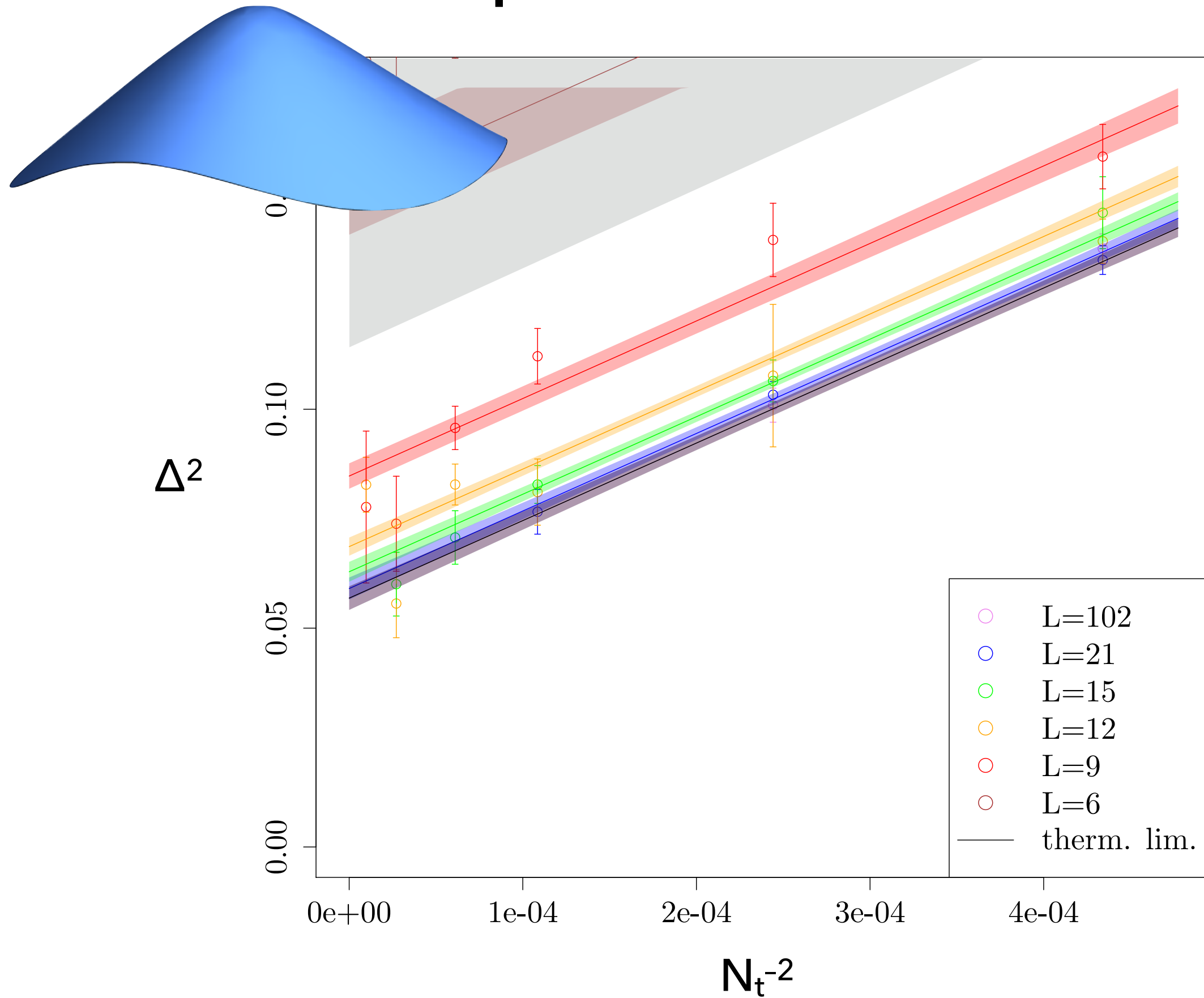


SYSTEMATIC EXTRAPOLATION

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112



$$\Delta^2(L, N_t) = \Delta_0^2 + \frac{c_0}{N_t^2} + \frac{c_1}{L^3} + \dots$$



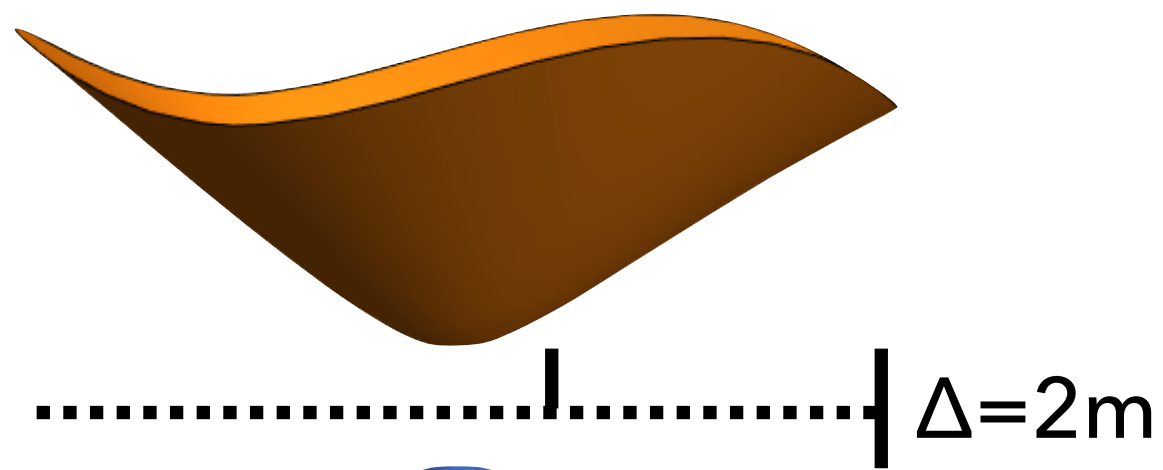
$\kappa\beta=8 \quad U/\kappa=3.5$

← temporal continuum limit

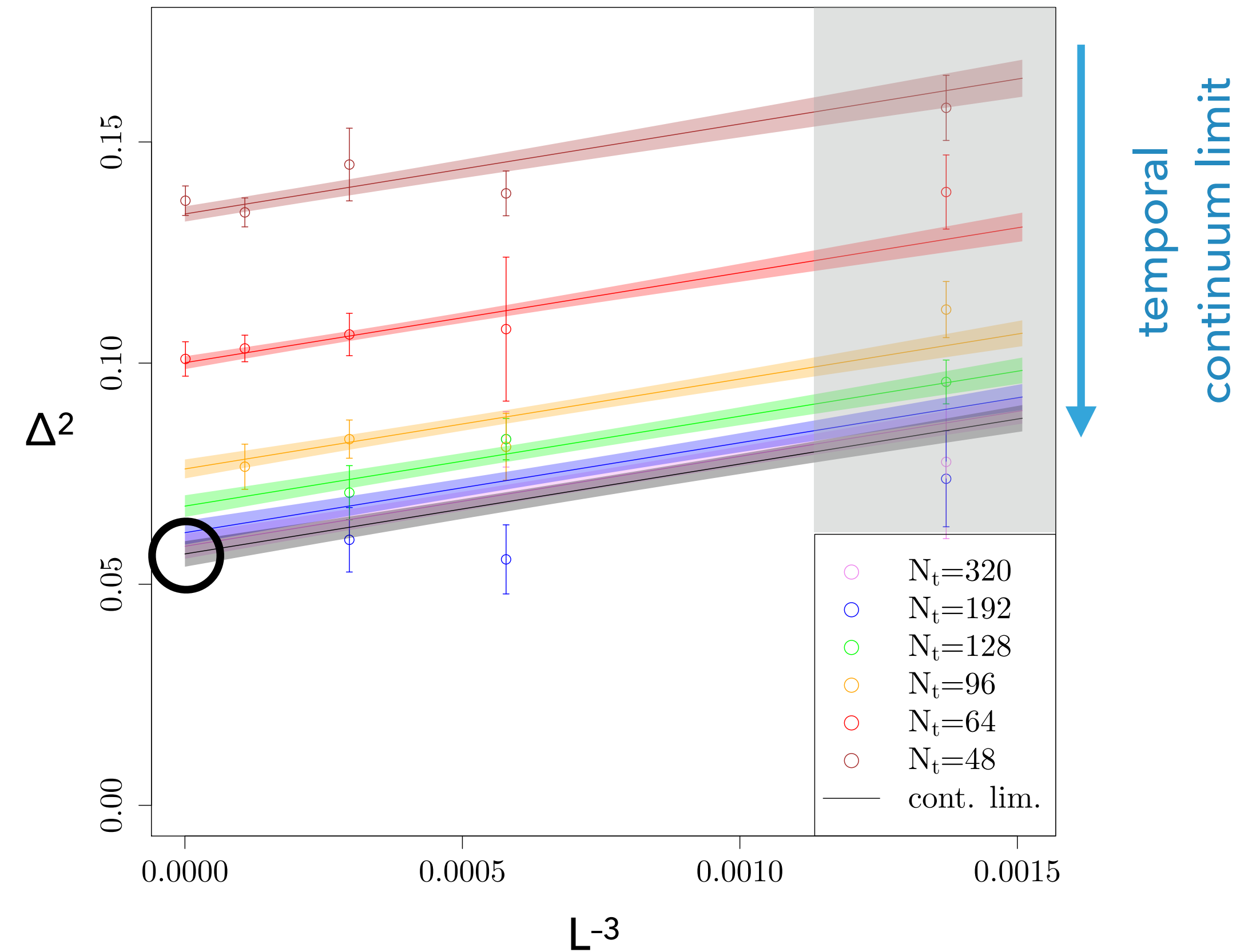
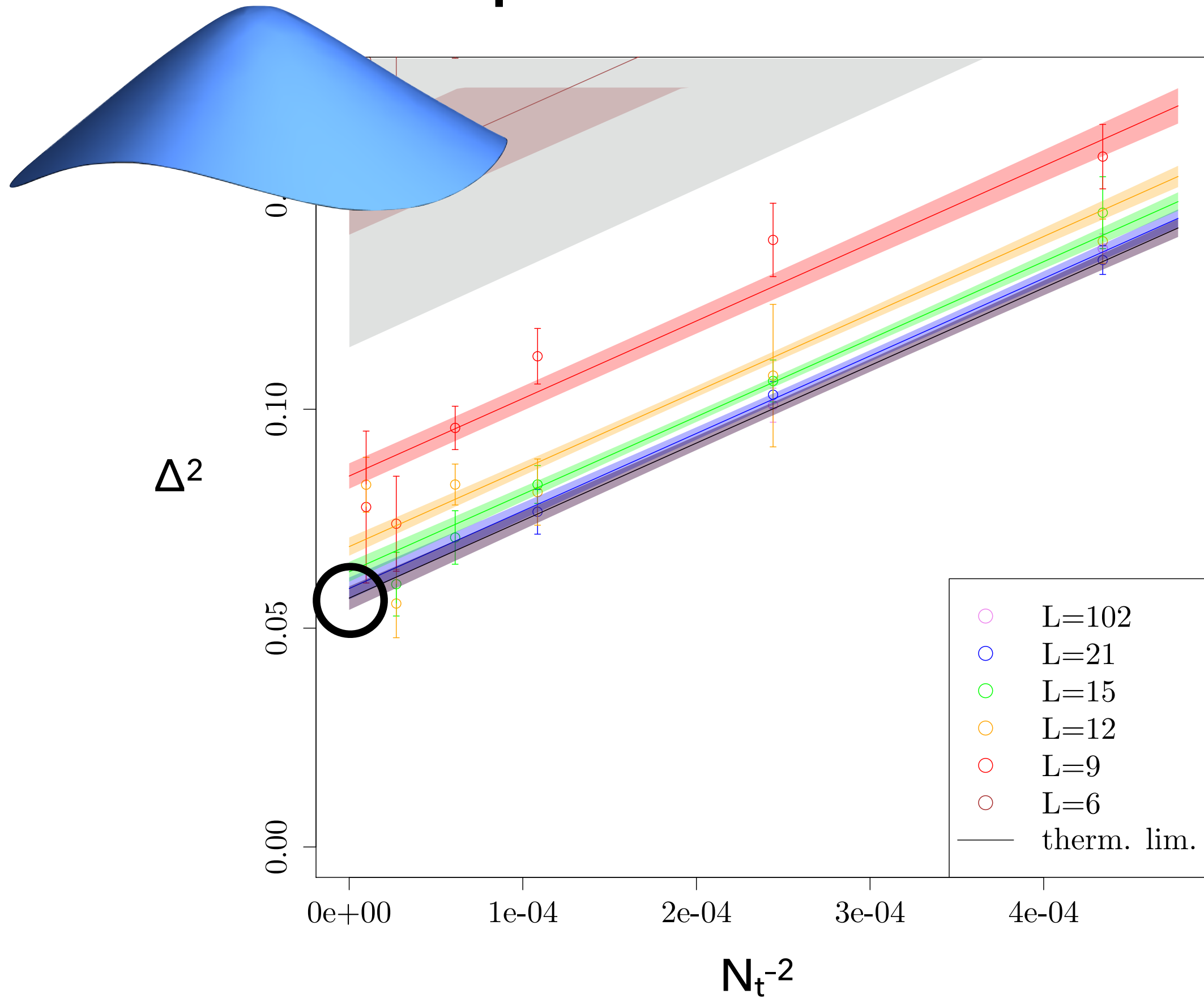
← infinite volume limit

SYSTEMATIC EXTRAPOLATION

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112



$$\Delta^2(L, N_t) = \Delta_0^2 + \frac{c_0}{N_t^2} + \frac{c_1}{L^3} + \dots$$



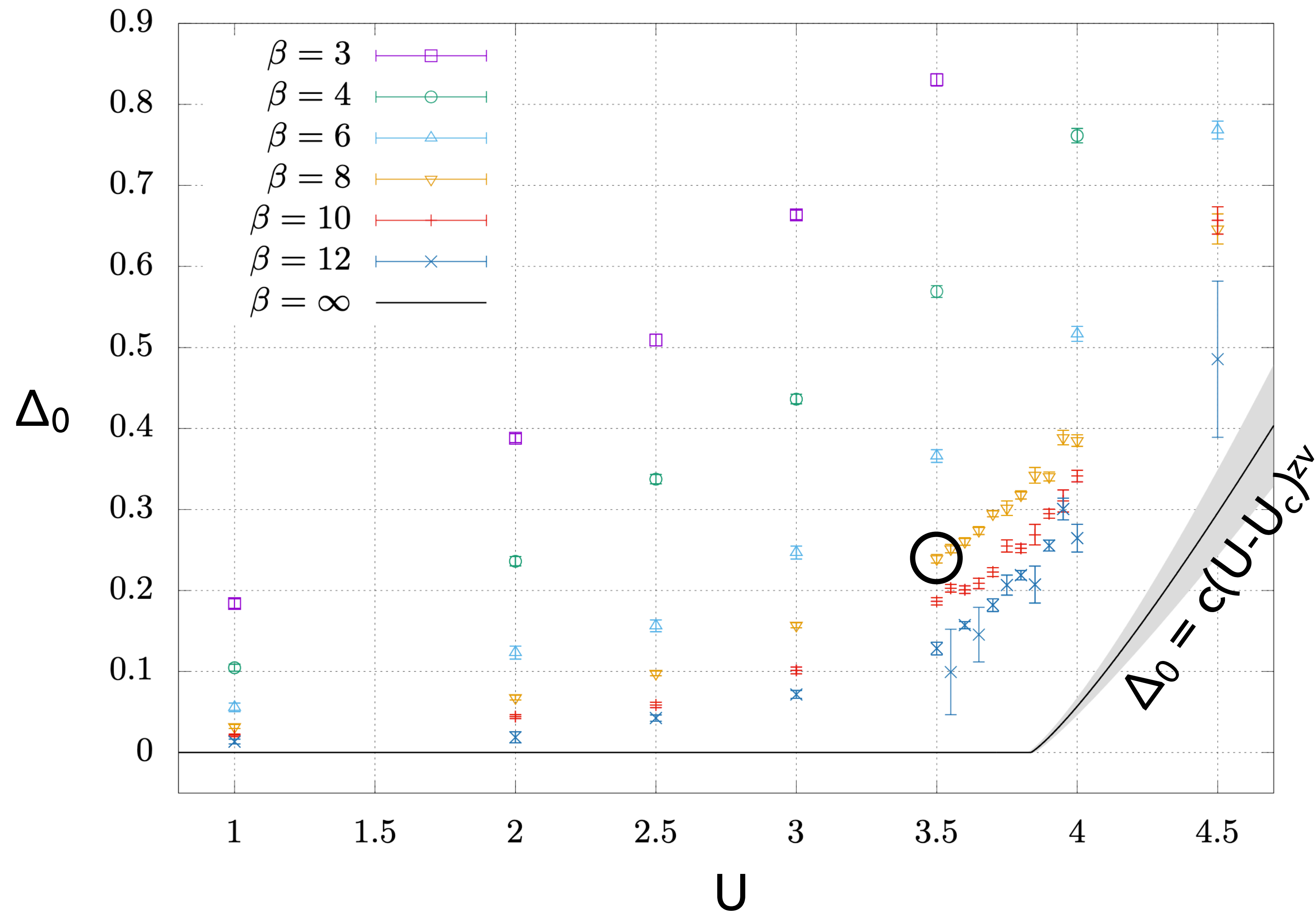
$\kappa\beta=8 \quad U/\kappa=3.5$

temporal continuum limit

infinite volume limit

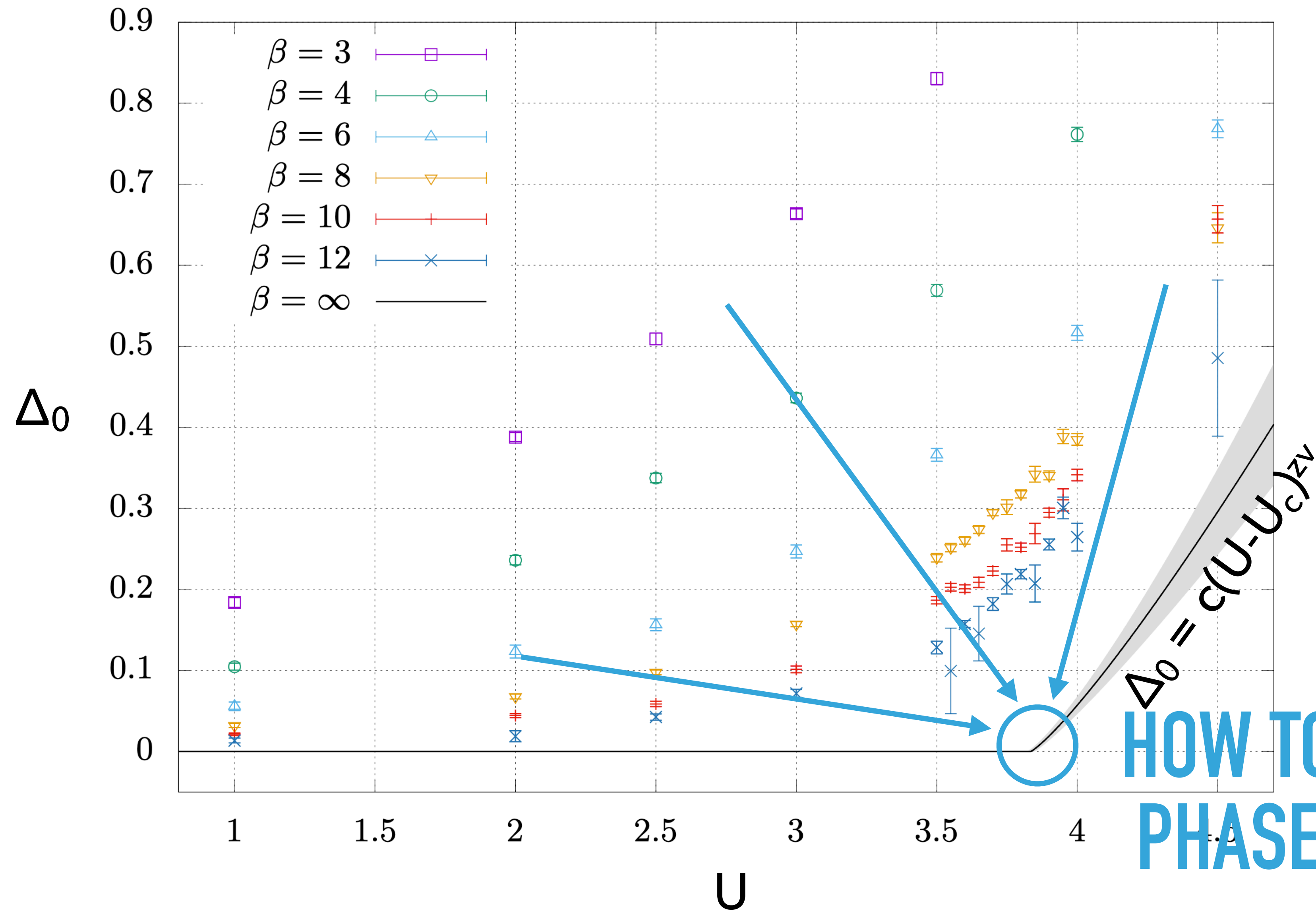
THERMODYNAMIC LIMIT

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112



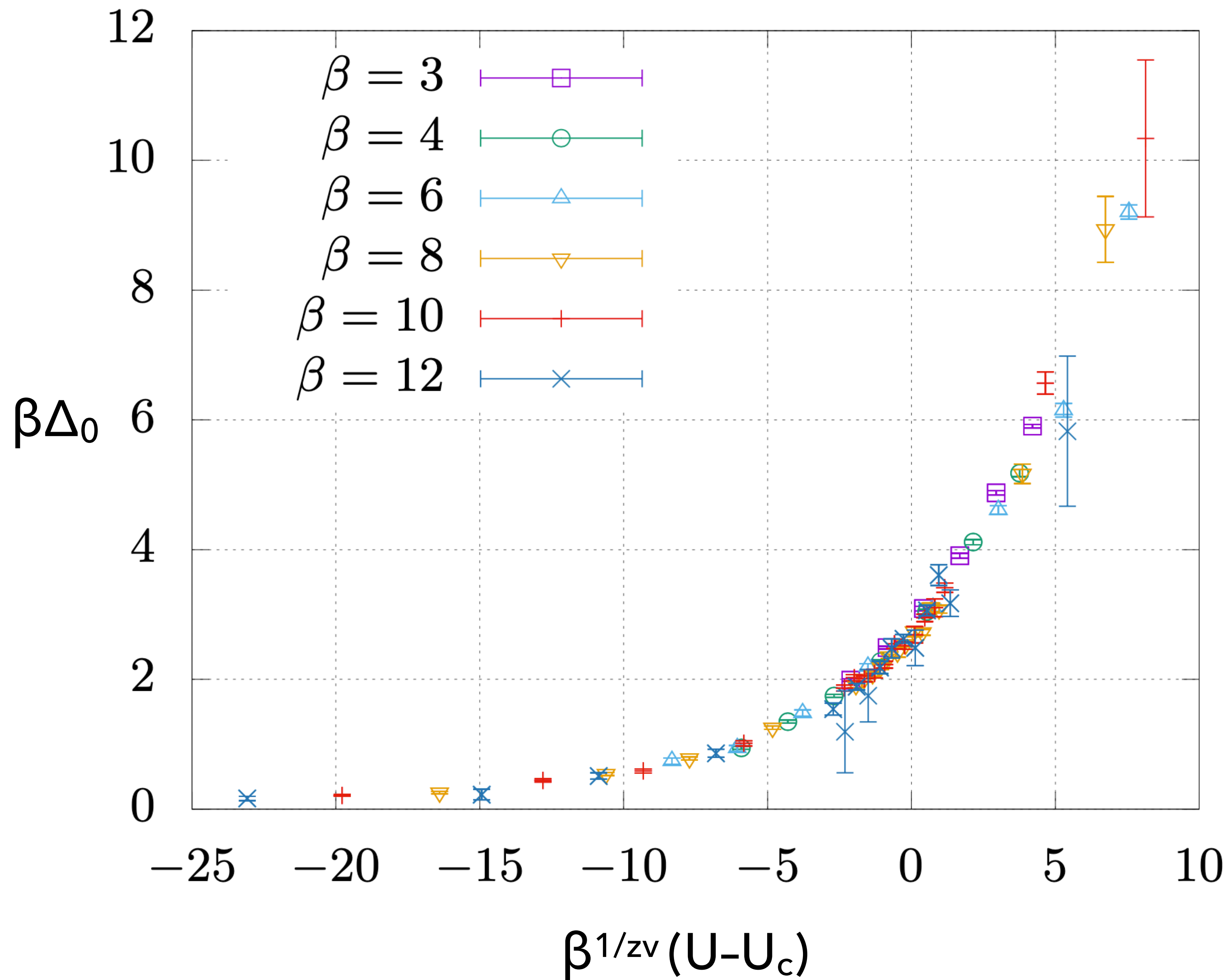
THERMODYNAMIC LIMIT

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112



HOW TO EXTRACT THE PHASE TRANSITION?

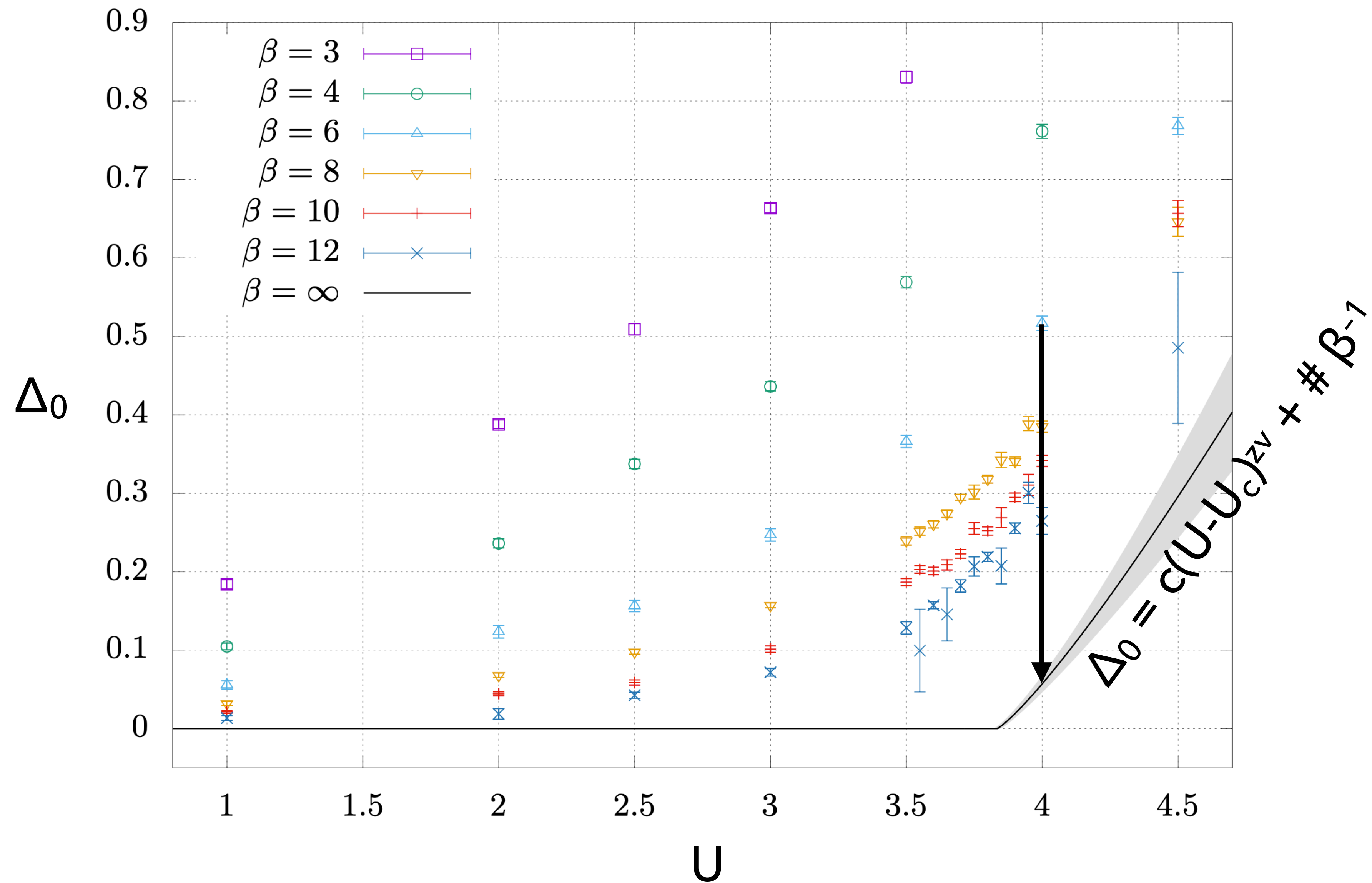
FINITE-SIZE SCALING



$U_c = 3.834(14)$
 $z\nu = 1.185(43)$

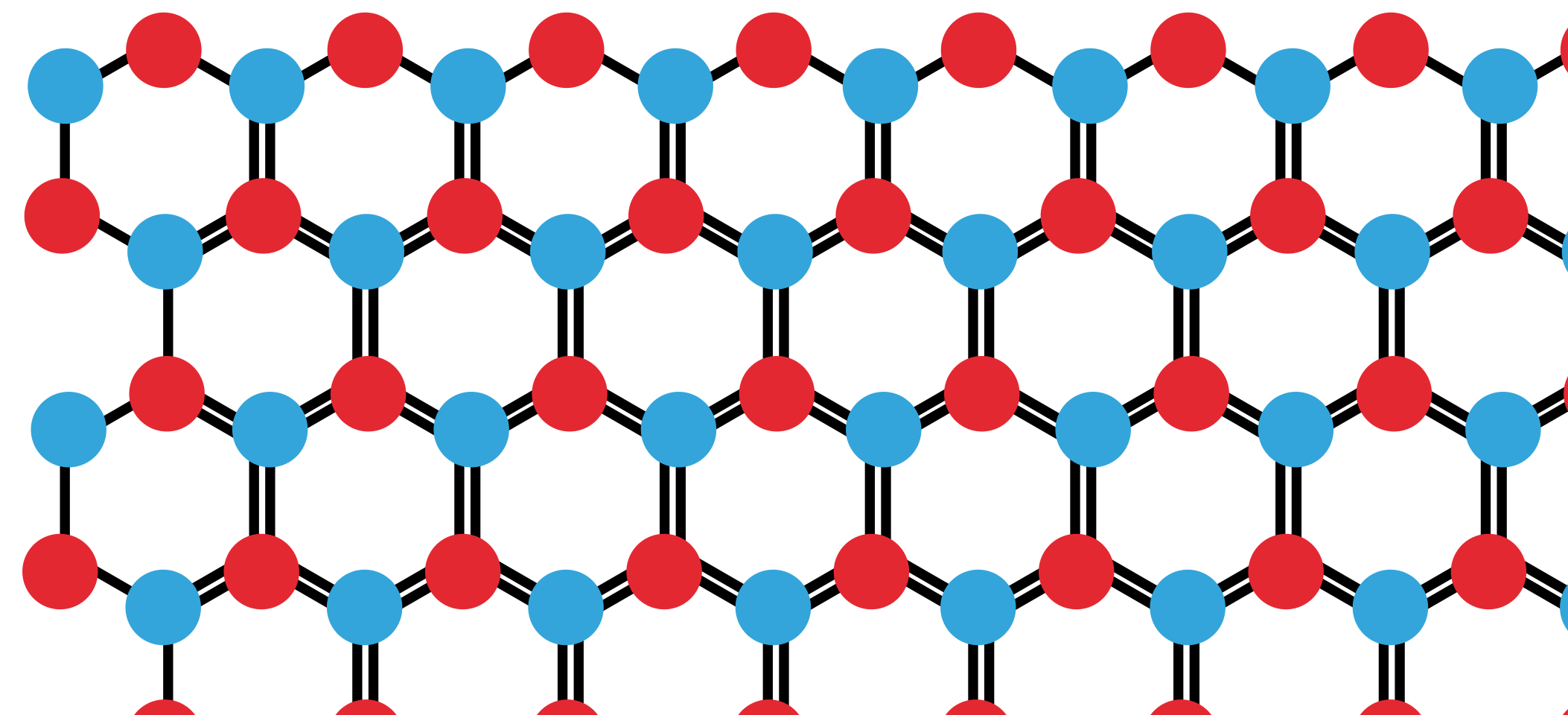
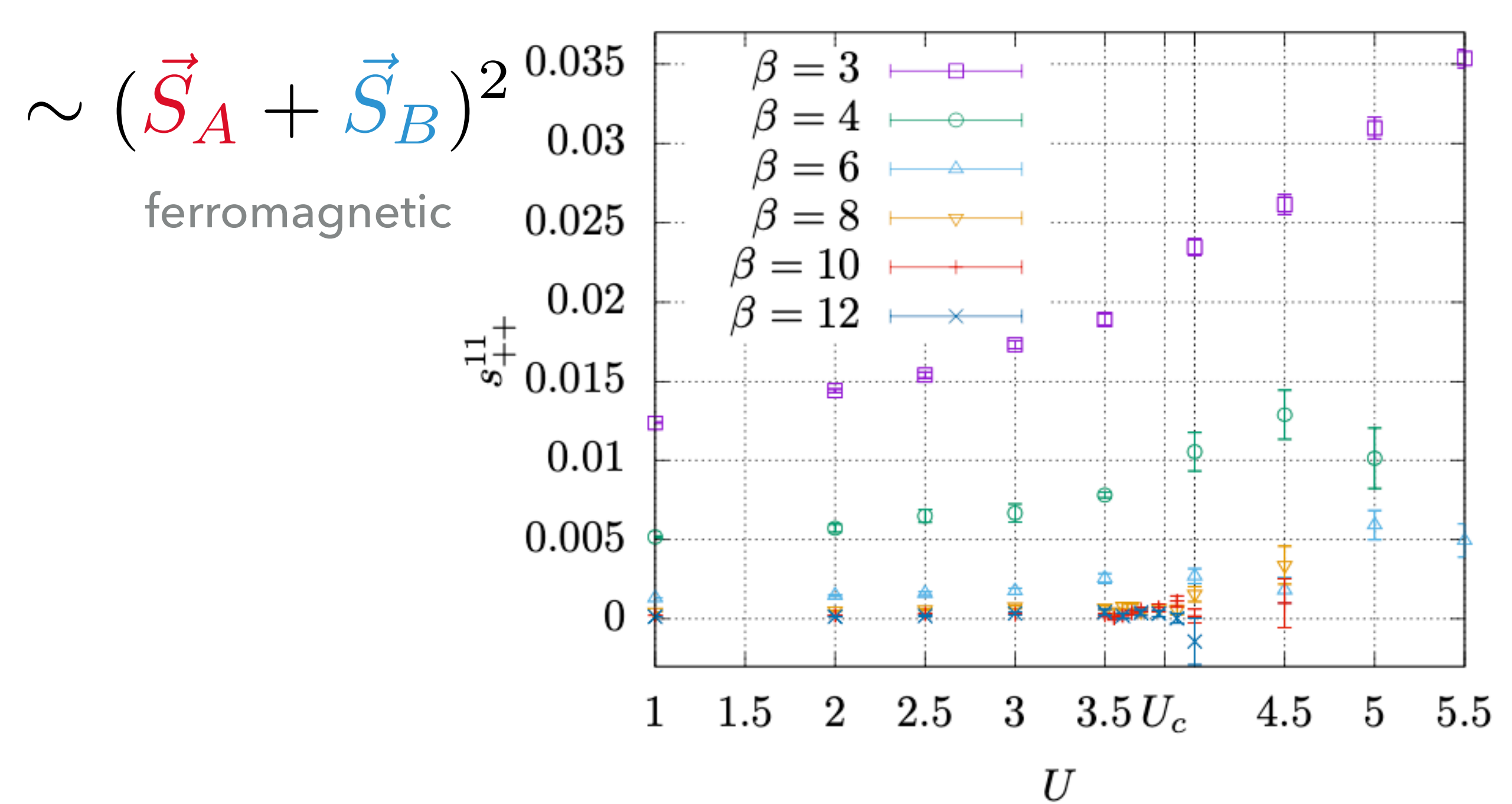
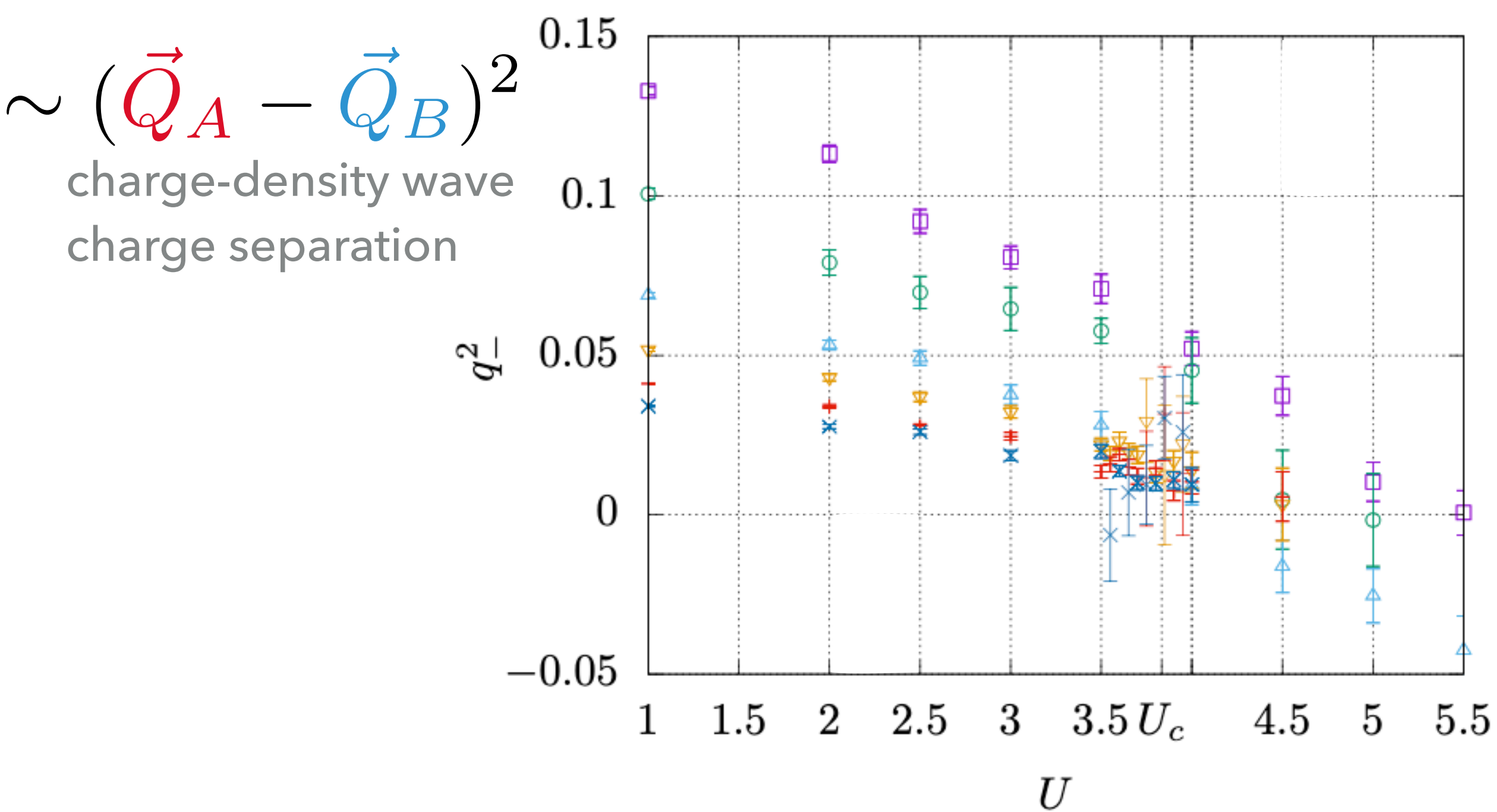
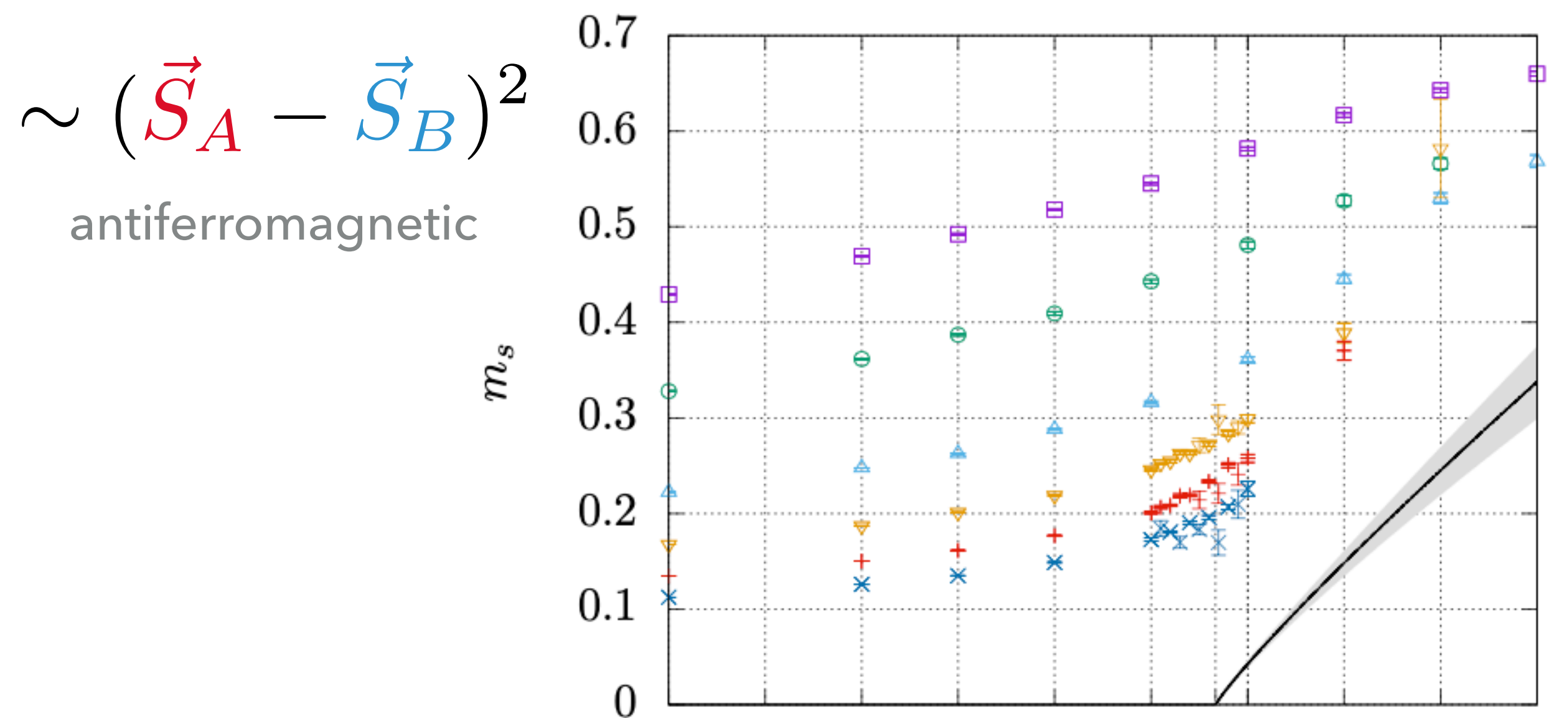
ZERO-TEMPERATURE LIMIT

Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2005.11112

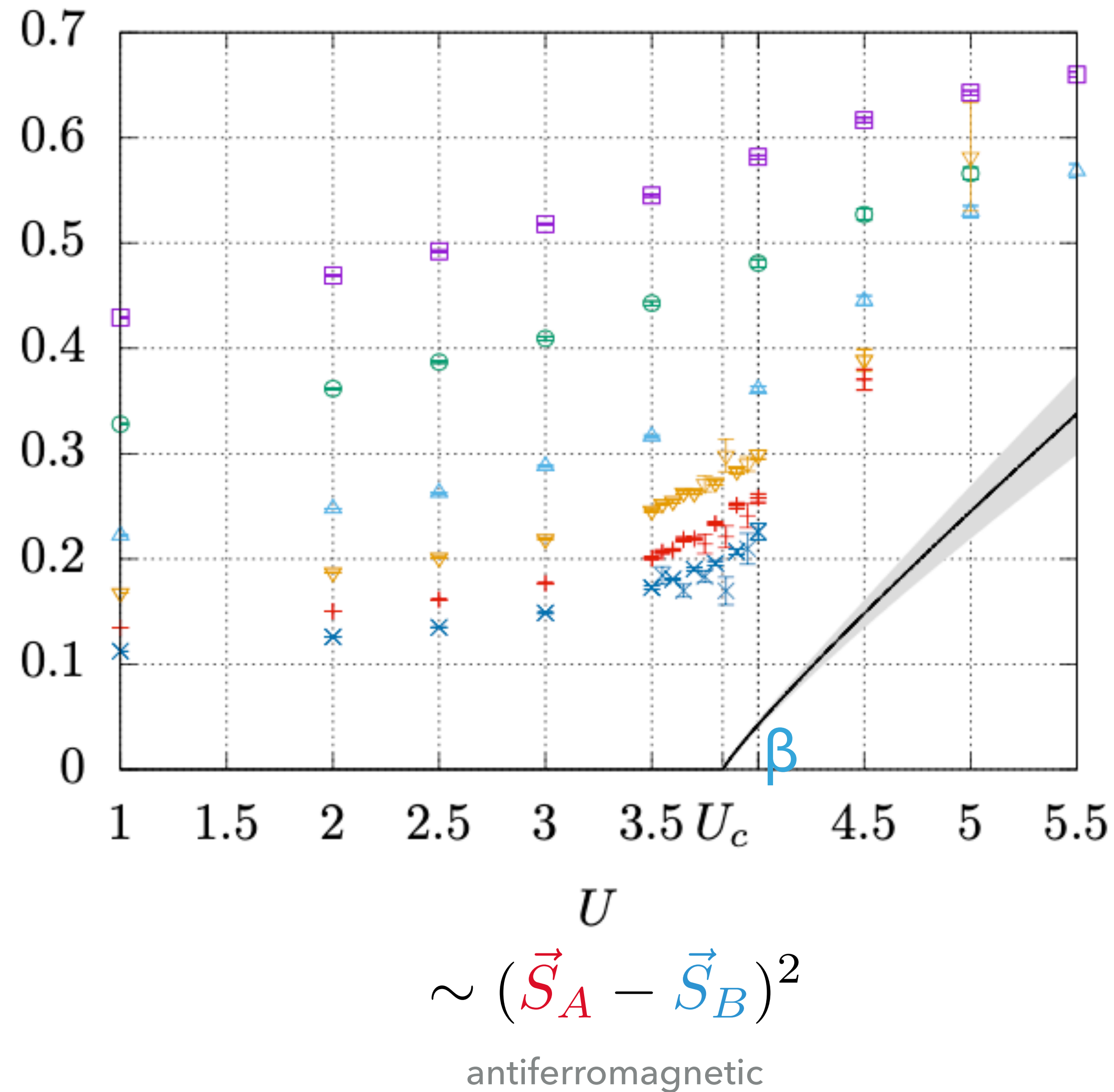
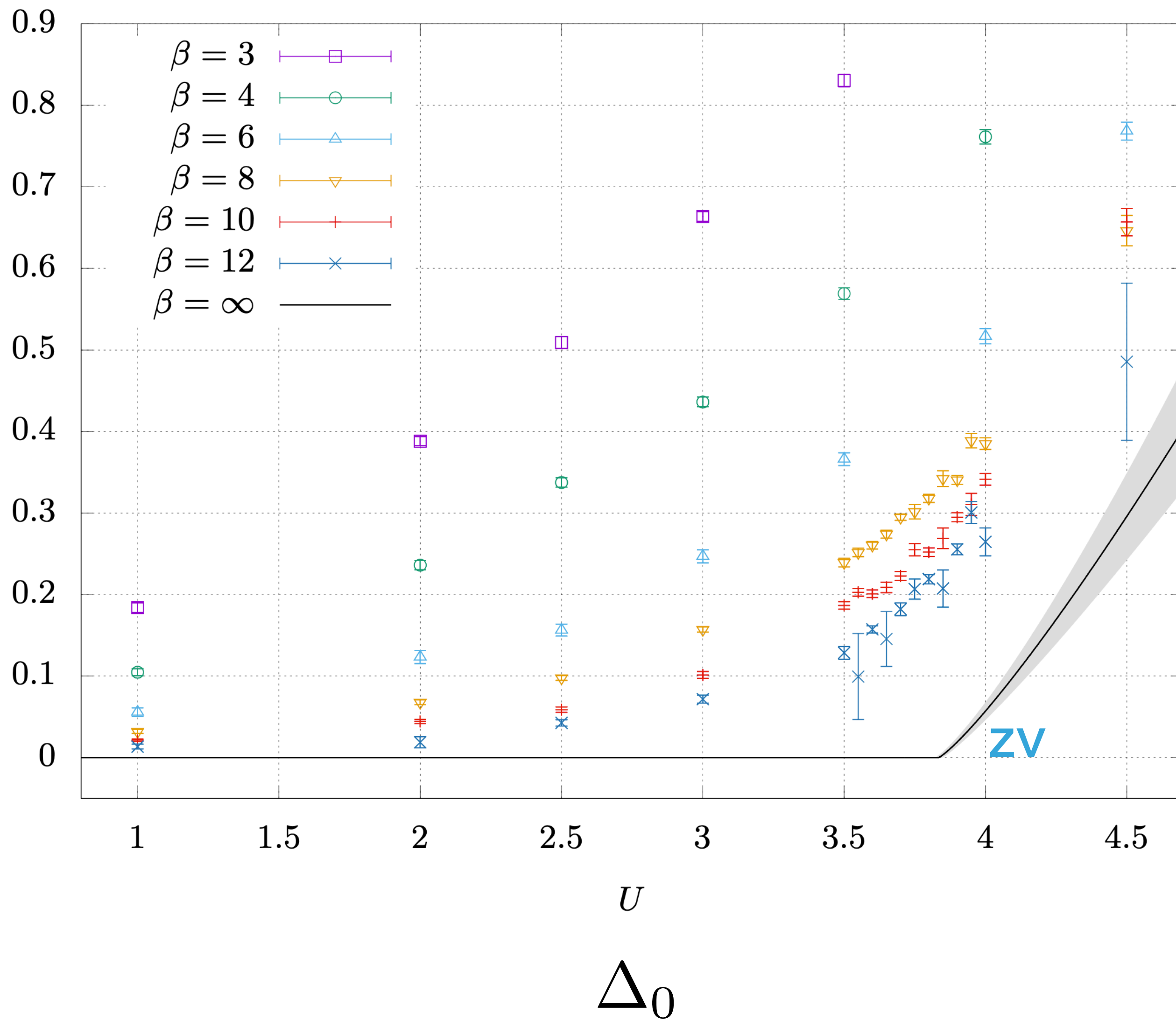


ZERO-TEMPERATURE LIMIT

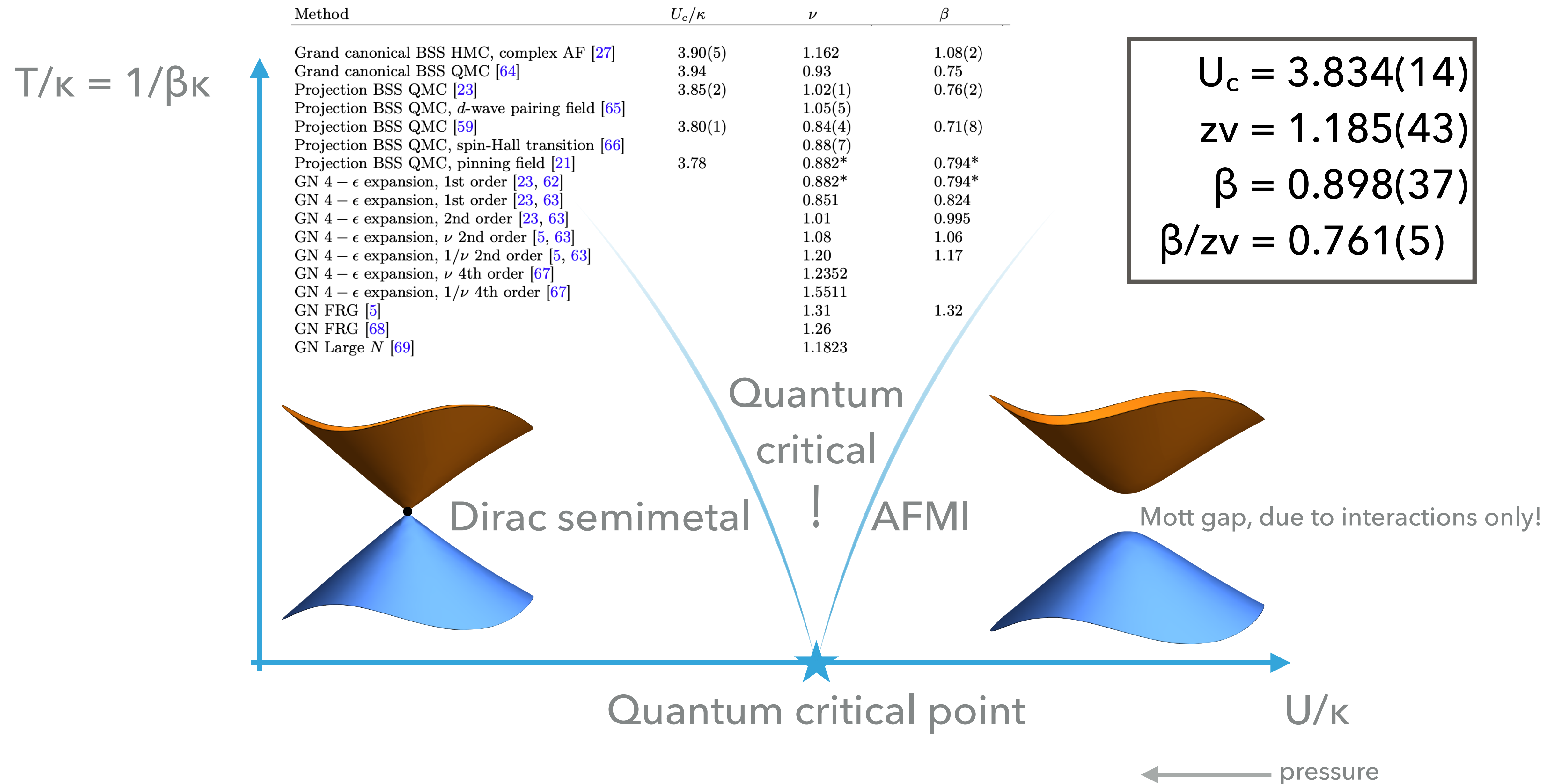
Ostmeyer, EB, Krieg, Lähde, Luu, Urbach 2105.06936



THE PHASE TRANSITION



PHASE DIAGRAM



SIGN PROBLEMS

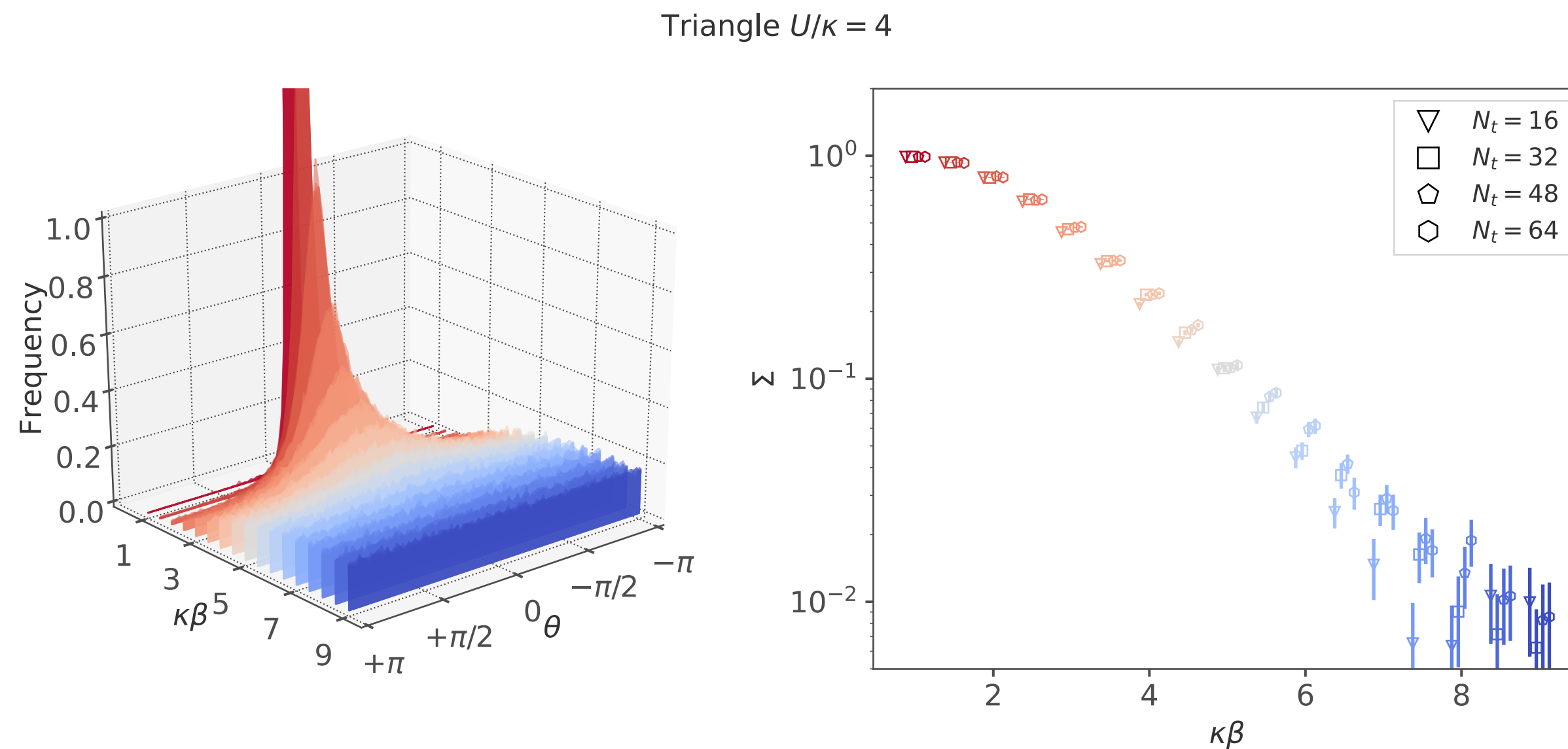


COMPLEX WEIGHTS LEAD TO SIGN PROBLEMS

The average sign, or *statistical power* Σ

$$\Sigma = \left| \langle e^{-iS^I} \rangle_R \right|$$

is the ratio of two partition functions \sim differences of free energies
so is extensive in spacetime volume \rightarrow exponentially decays



$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Probability?

Generate an *ensemble* for fixed U, β, N_t, \dots

$$\{\phi_1, \phi_2, \dots, \phi_N\}$$

Hybrid Monte Carlo

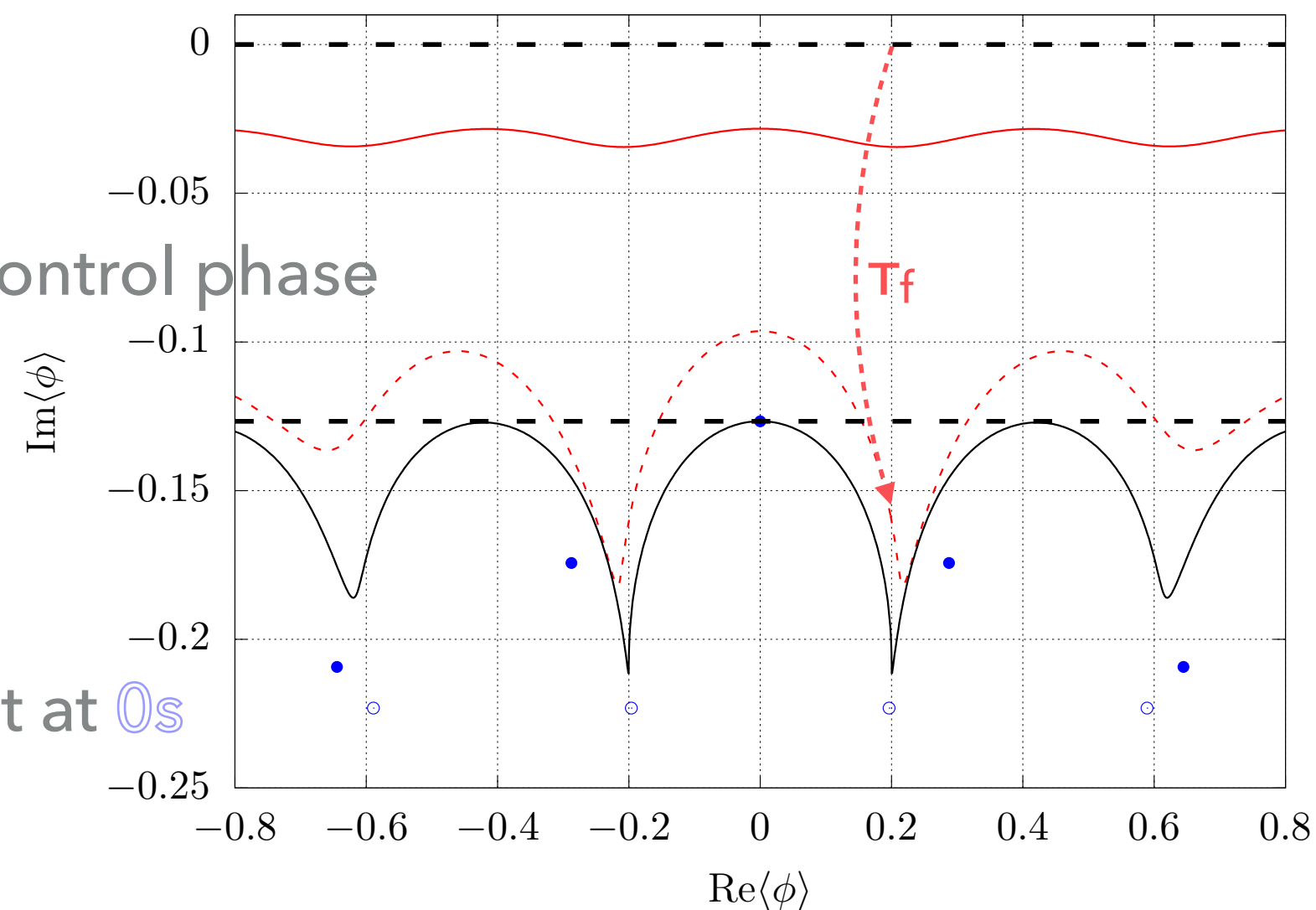
Estimate any observable

$$\langle \mathcal{O} \rangle = \frac{\langle \mathcal{O} e^{-iS^I} \rangle_R}{\langle e^{-iS^I} \rangle_R}$$

Importance sample according to the real part of the weight

LEFSCHETZ THIMBLES HAVE CONSTANT SIGN

Critical points control phase

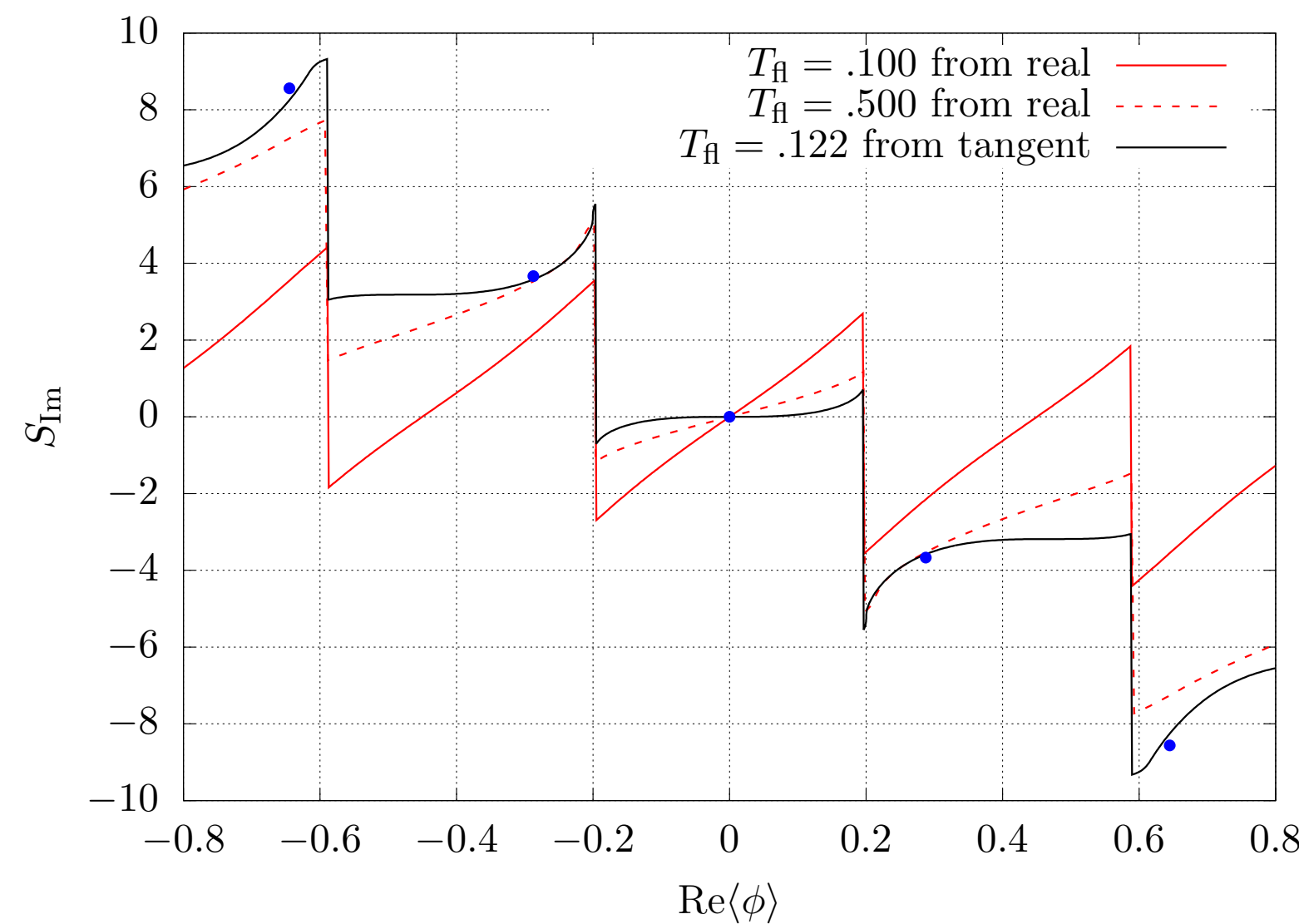
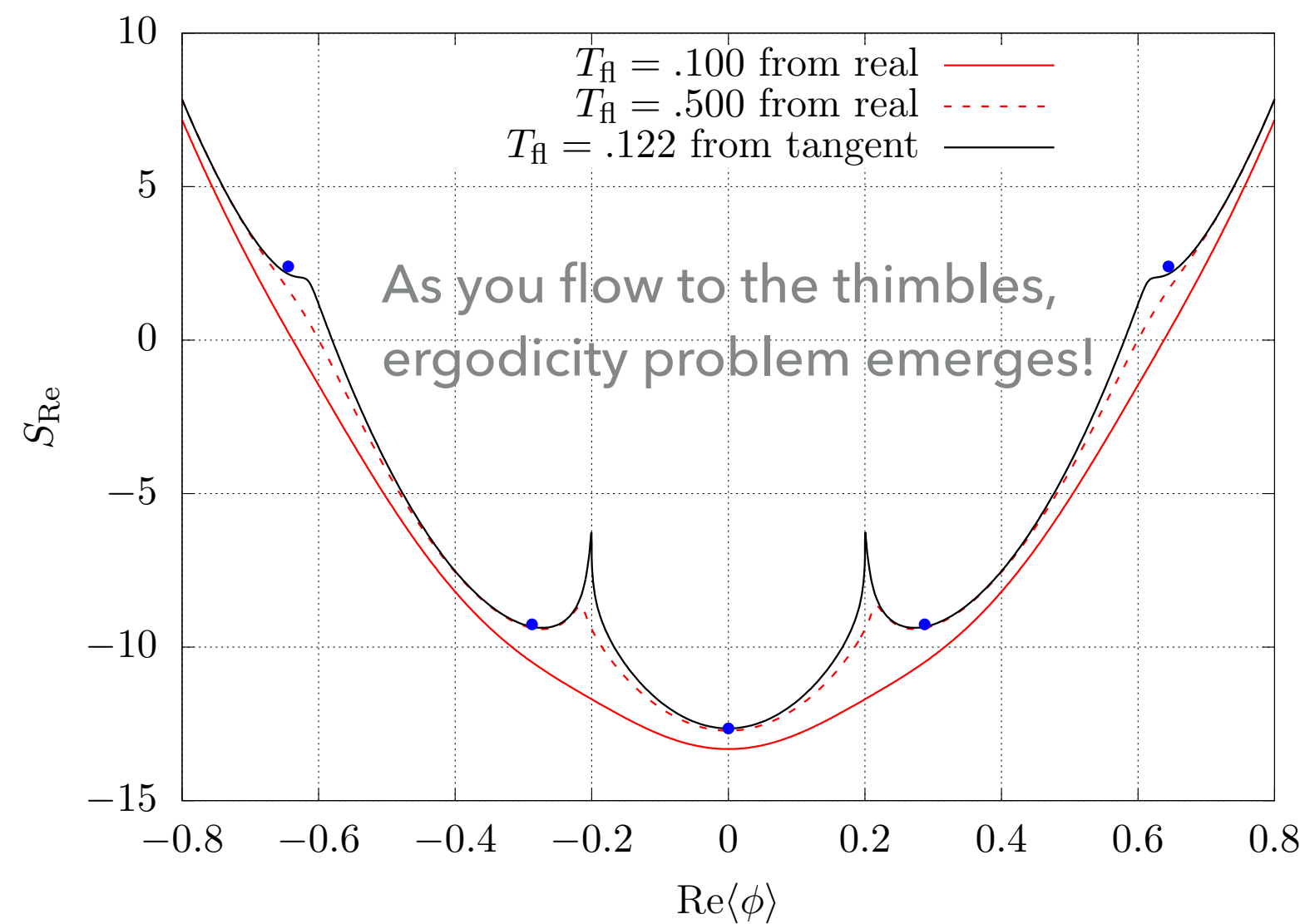


Thimbles meet at \mathcal{O}_S

$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

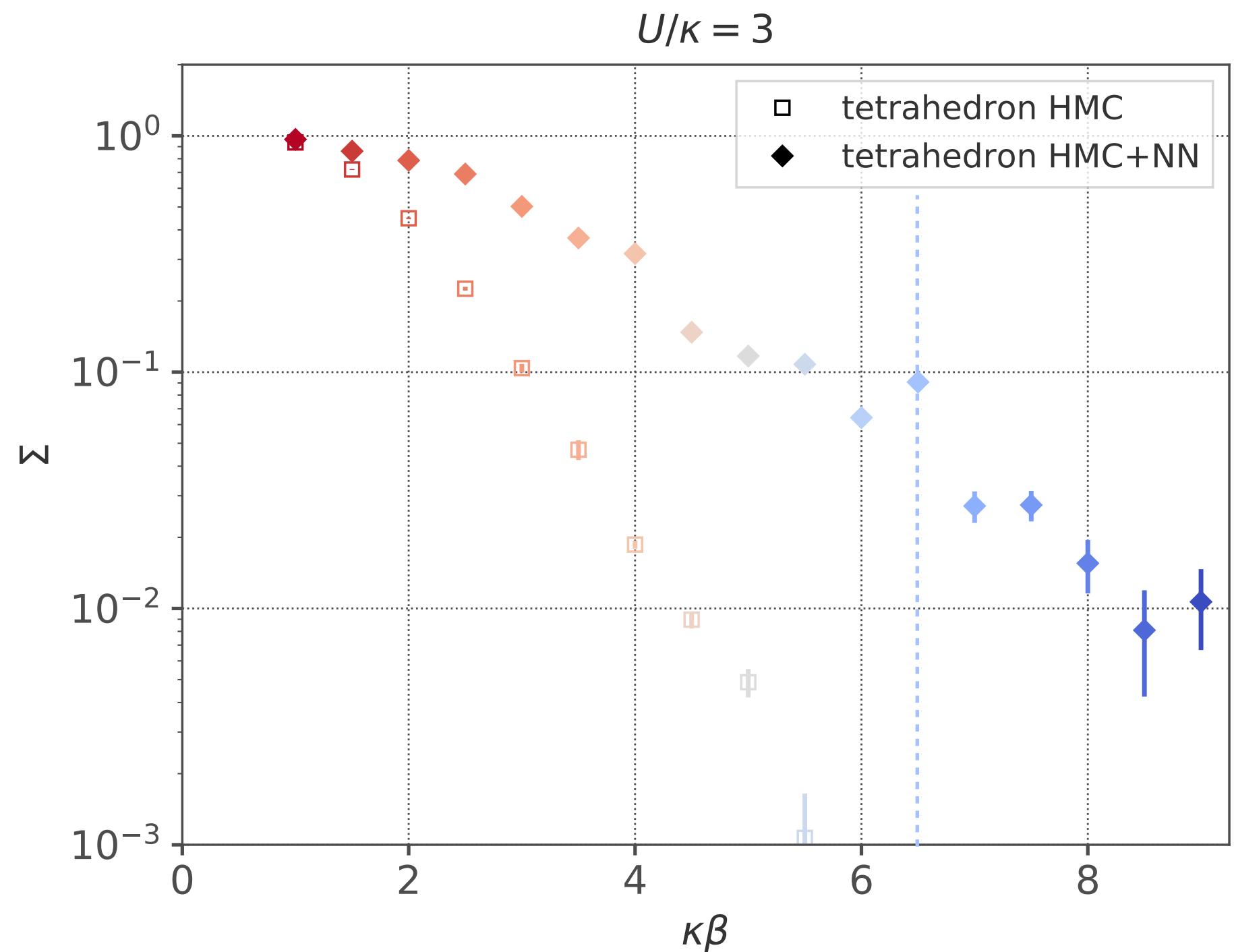
Complexify ϕ , integrate on steepest-descent analogues called *Lefschetz thimbles*, fixed points of holomorphic flow

$$\frac{d\phi^R}{d\tau_f} = \pm \frac{\partial S^R}{\partial \phi_i^R} = \pm \frac{\partial S^I}{\partial \phi_i^I} \quad \frac{d\phi^I}{d\tau_f} = \pm \frac{\partial S^R}{\partial \phi_i^I} = \mp \frac{\partial S^I}{\partial \phi_i^R}$$



As you flow to the thimbles, phases become constant.

MACHINE LEARNING FOR FAST FLOWS



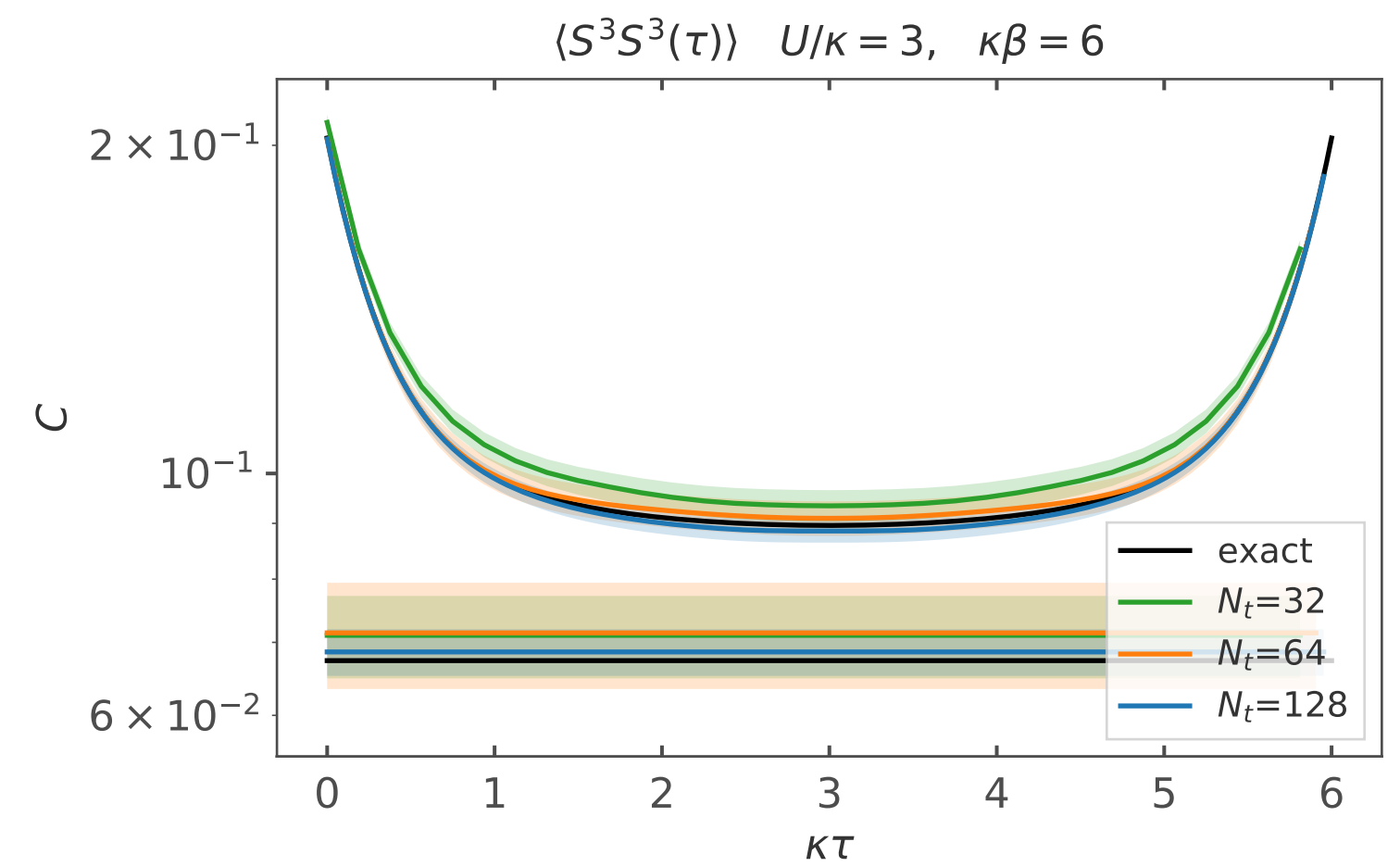
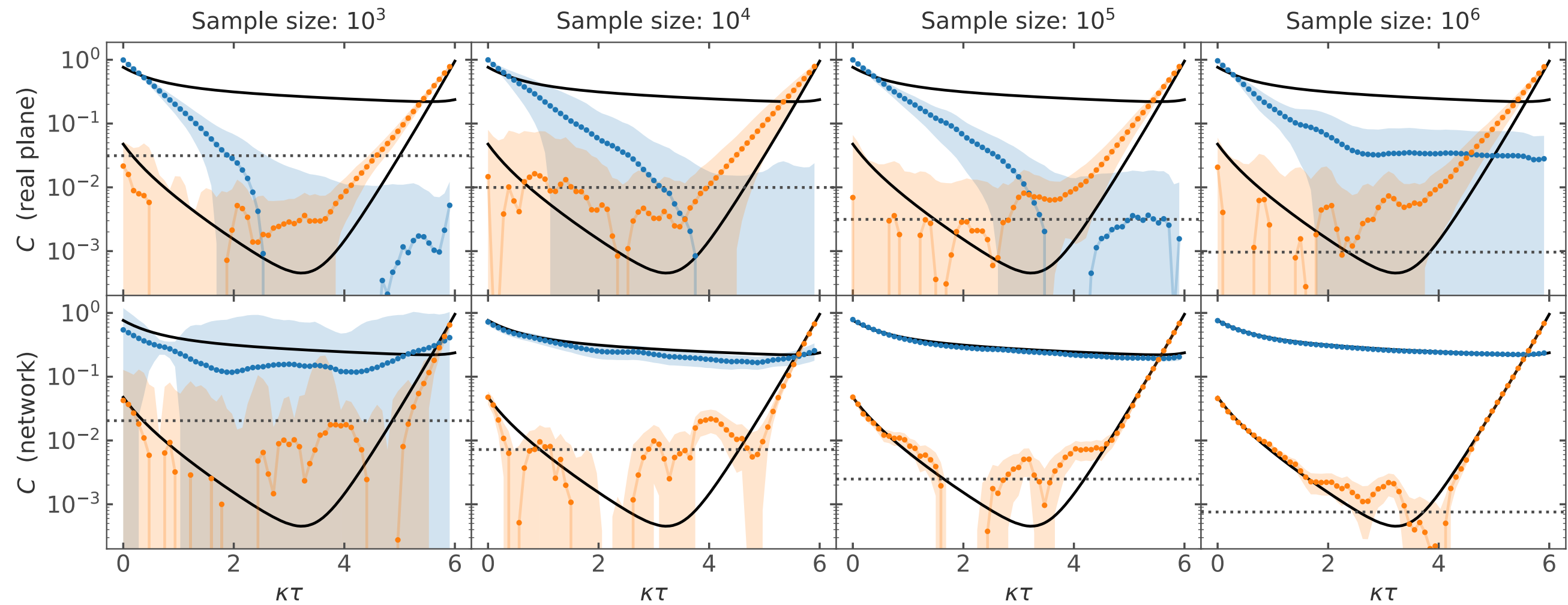
$$\mathcal{Z} \propto \int \mathcal{D}\phi \det (M[\phi, h, \mu] M^\dagger[\phi, h, -\mu]) e^{-\frac{1}{2} \sum_{xyt} \phi_{xt} V_{xy}^{-1} \phi_{yt}}$$

Complexify ϕ , integrate on steepest-descent analogues called *Lefschetz thimbles*, fixed points of holomorphic flow

$$\frac{d\phi^R}{d\tau_f} = \pm \frac{\partial S^R}{\partial \phi_i^R} = \pm \frac{\partial S^I}{\partial \phi_i^I} \quad \frac{d\phi^I}{d\tau_f} = \pm \frac{\partial S^R}{\partial \phi_i^I} = \mp \frac{\partial S^I}{\partial \phi_i^R}$$

Flow is expensive and slow during HMC.
Use neural networks to approximate the thimble!

$\langle a_\tau a^\dagger \rangle$



SIGN PROBLEMS ABOUND

THE MOST INTERESTING PHYSICS REQUIRES NEW TOOLS

- ▶ Finite μ

Ulybyshev, Winterowd, Zafeiropolous PRD 101, 014508 (2020) 10.1103/PhysRevD.101.014508 1906.07678

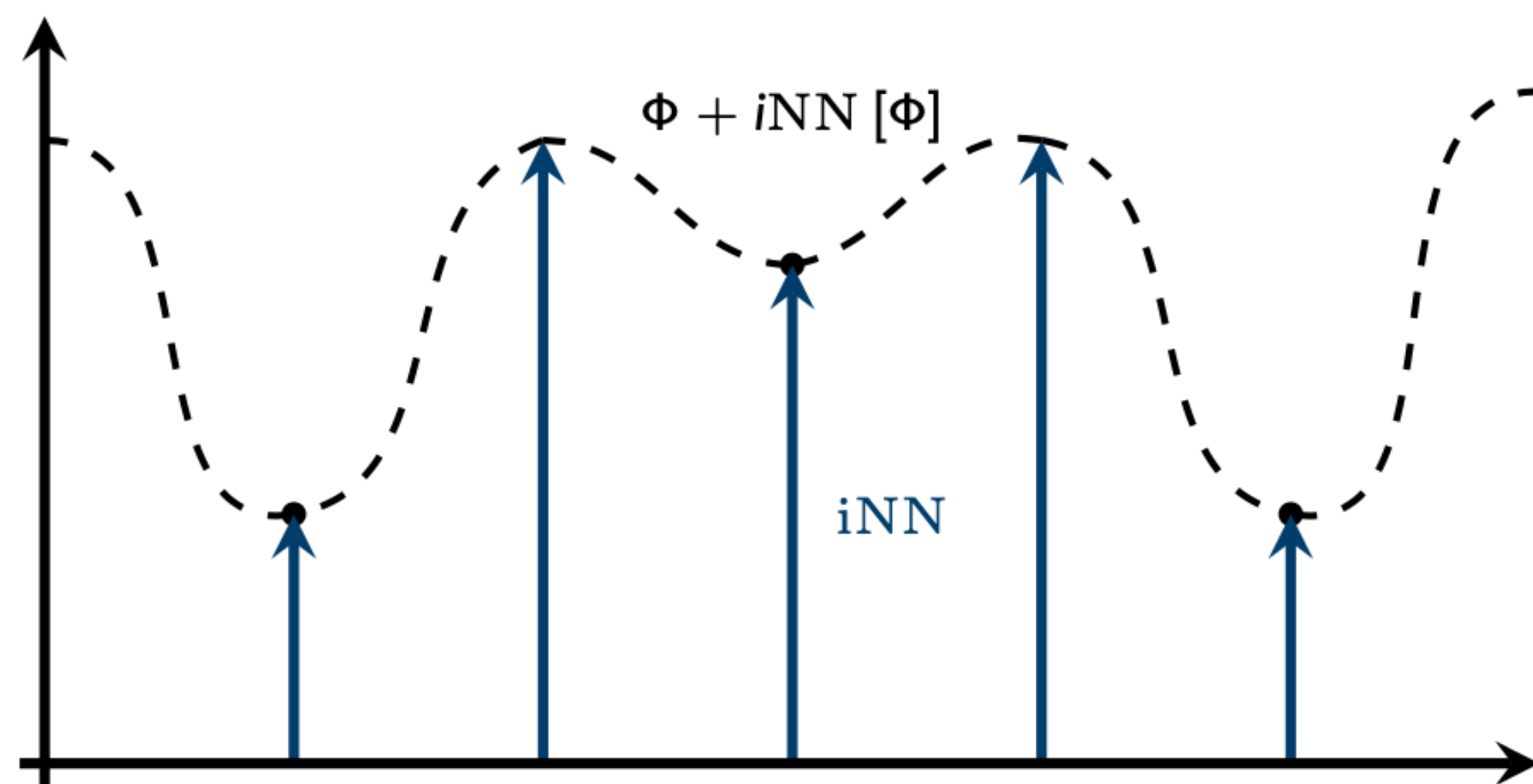
- ▶ Non-bipartite lattices, fullerenes, ...

Wynen, EB, Krieg, Luu, Ostmeyer 2006.11221

- ▶ Real-time

- ▶ Develop tools for QCD!

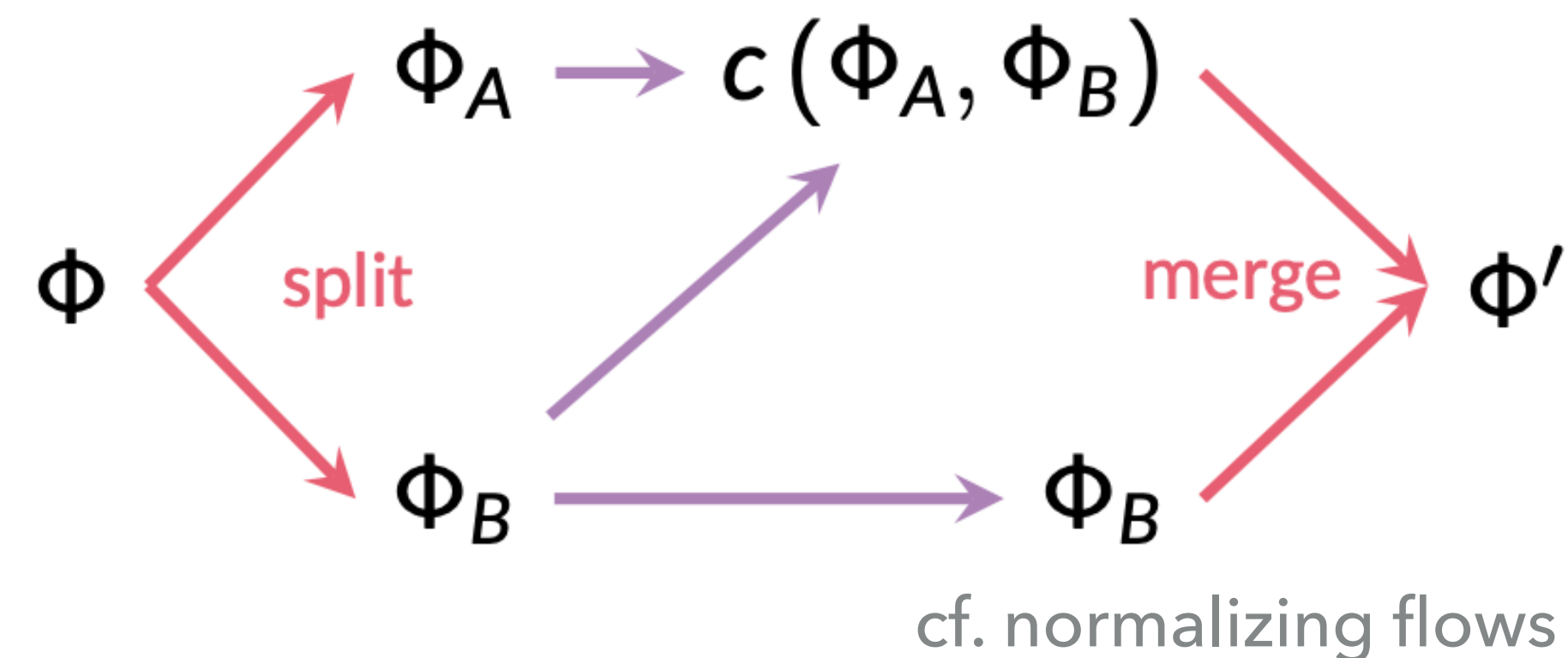
\mathbb{R} -Valued Neural Network



Train the network so that

- Input of the network is $\text{Re}(\Phi)$ at the end of the flow
- Output of the network is $\text{Im}(\Phi)$ at the end of the flow

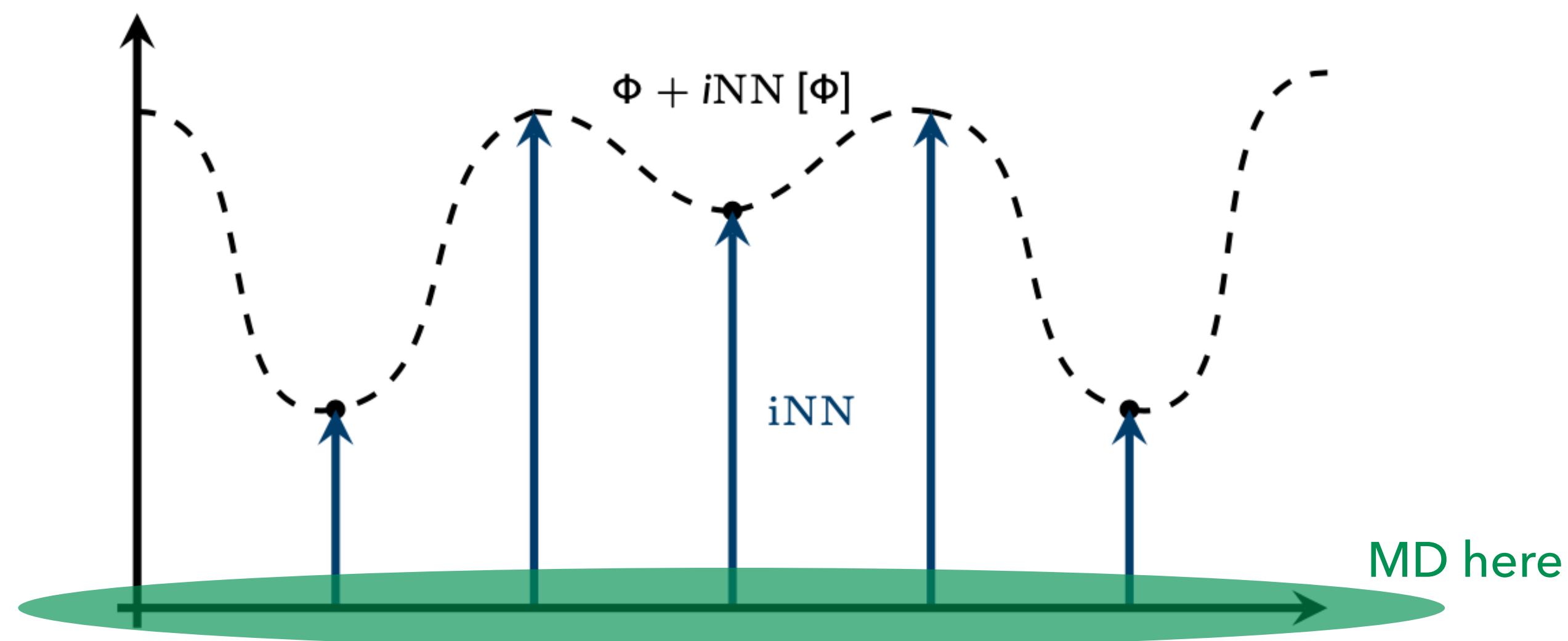
\mathbb{C} -Valued Neural Network



Train the network so that

- Input of the network is Φ at the start of the flow
- Output of the network is Φ at the end of the flow

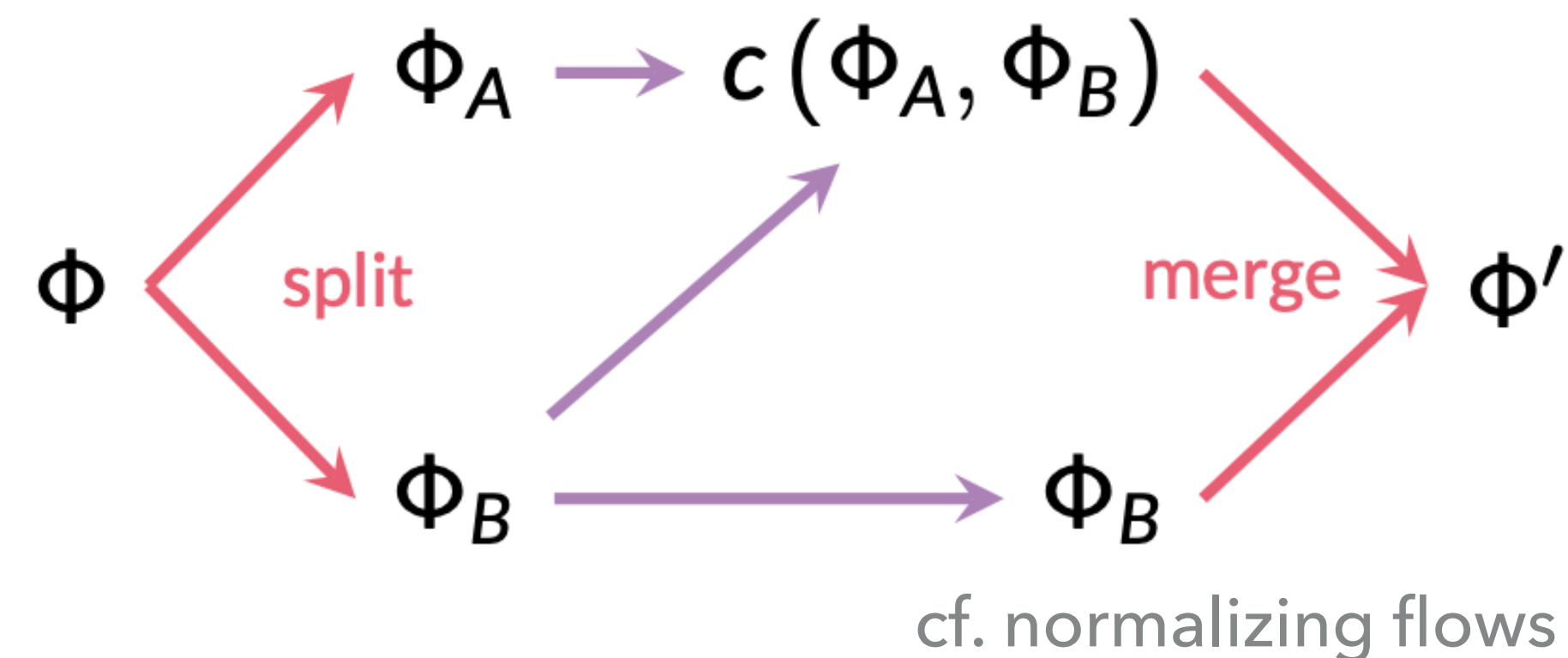
\mathbb{R} -Valued Neural Network



Train the network so that

- Input of the network is $\text{Re}(\Phi)$ at the end of the flow
- Output of the network is $\text{Im}(\Phi)$ at the end of the flow

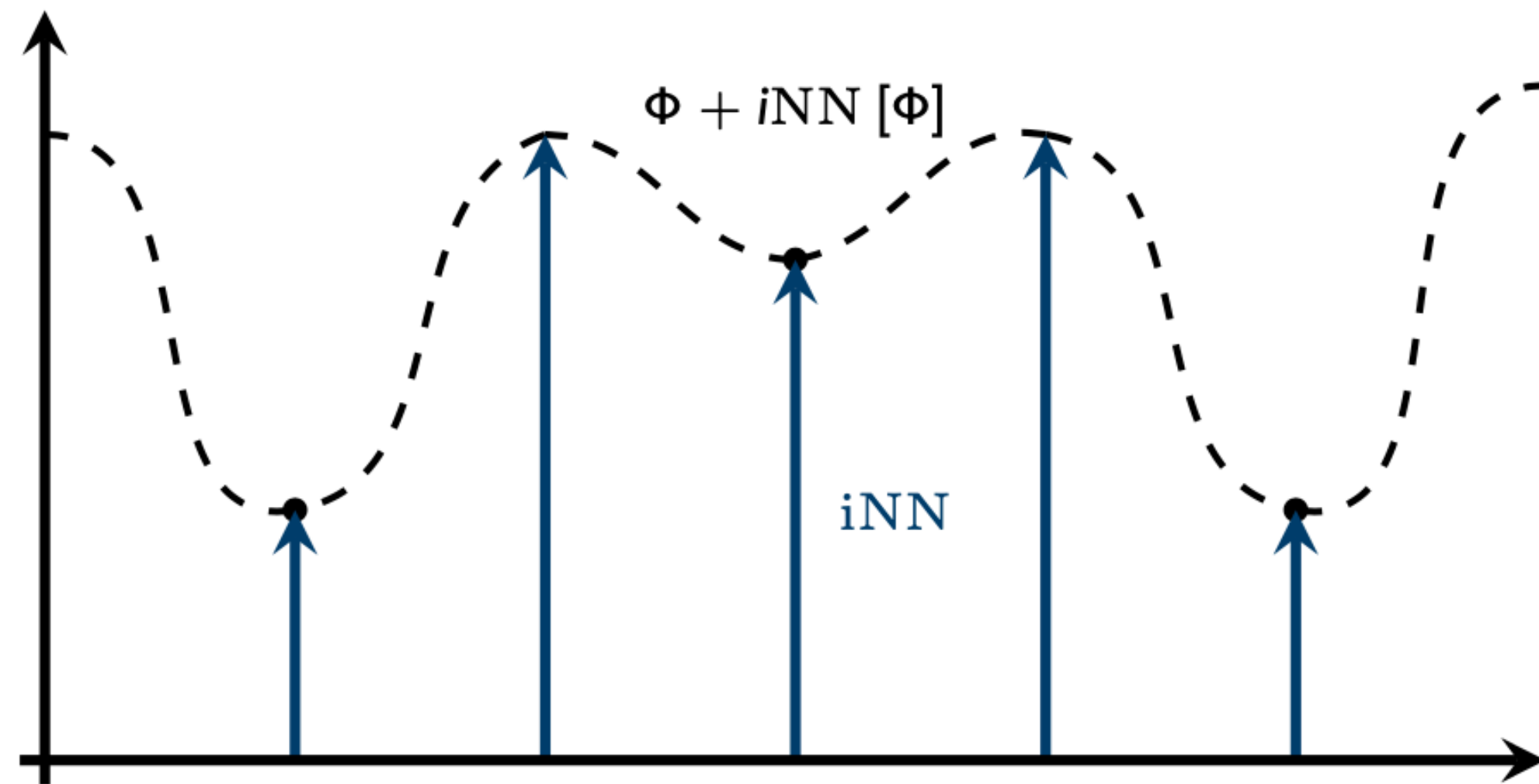
\mathbb{C} -Valued Neural Network



Train the network so that

- Input of the network is Φ at the start of the flow
- Output of the network is Φ at the end of the flow

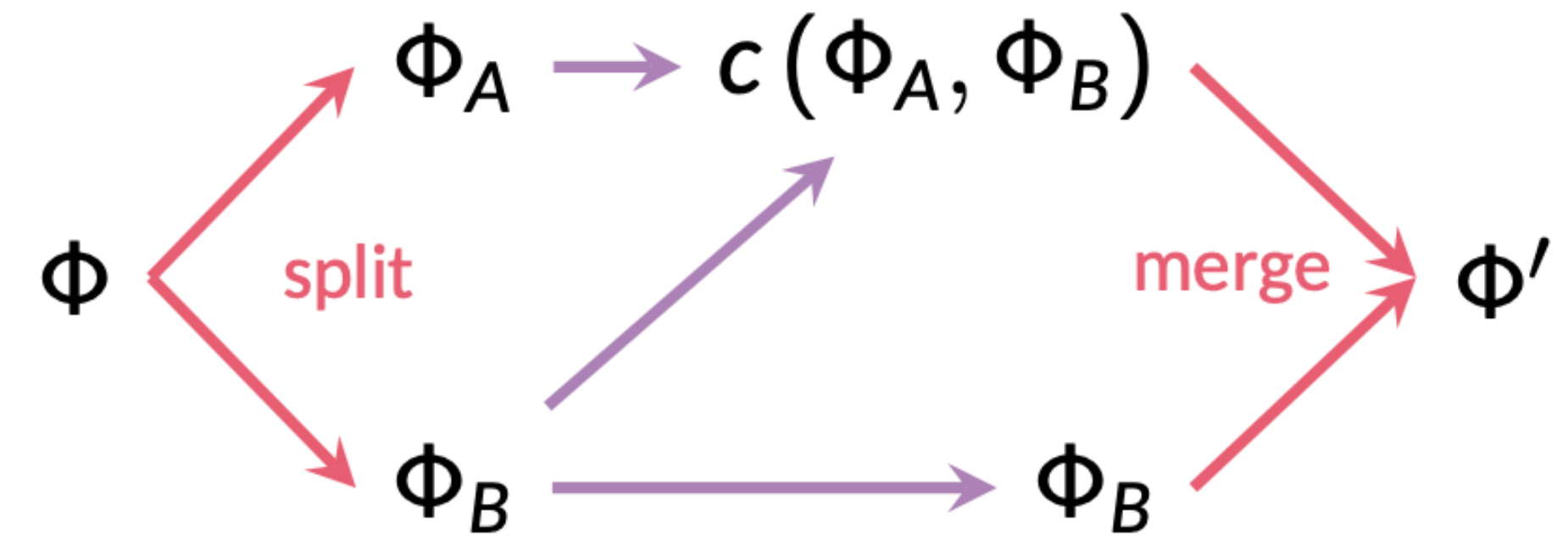
\mathbb{R} -Valued Neural Network



$$\mathbb{1} + i \begin{pmatrix} \frac{\partial \text{NN}[\phi]_0}{\partial \phi_0} & \dots & \frac{\partial \text{NN}[\phi]_{|\Lambda|}}{\partial \phi_0} \\ \vdots & \ddots & \vdots \\ \frac{\partial \text{NN}[\phi]_0}{\partial \phi_{|\Lambda|}} & \dots & \frac{\partial \text{NN}[\phi]_{|\Lambda|}}{\partial \phi_{|\Lambda|}} \end{pmatrix}$$

Jacobian is horrible: V^3

\mathbb{C} -Valued Neural Network

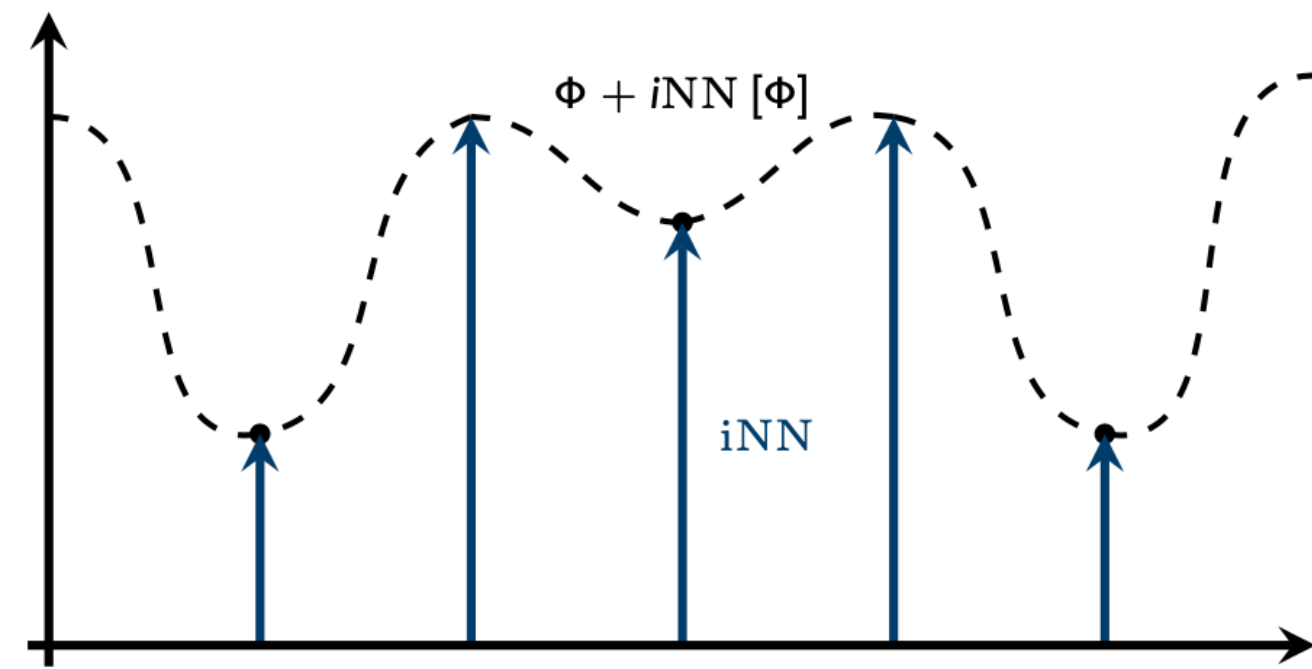


$$\begin{pmatrix} \frac{\partial c(\phi_A, \phi_B)}{\partial \phi_A} & \mathbf{0} \\ \star & \mathbb{1} \end{pmatrix}$$

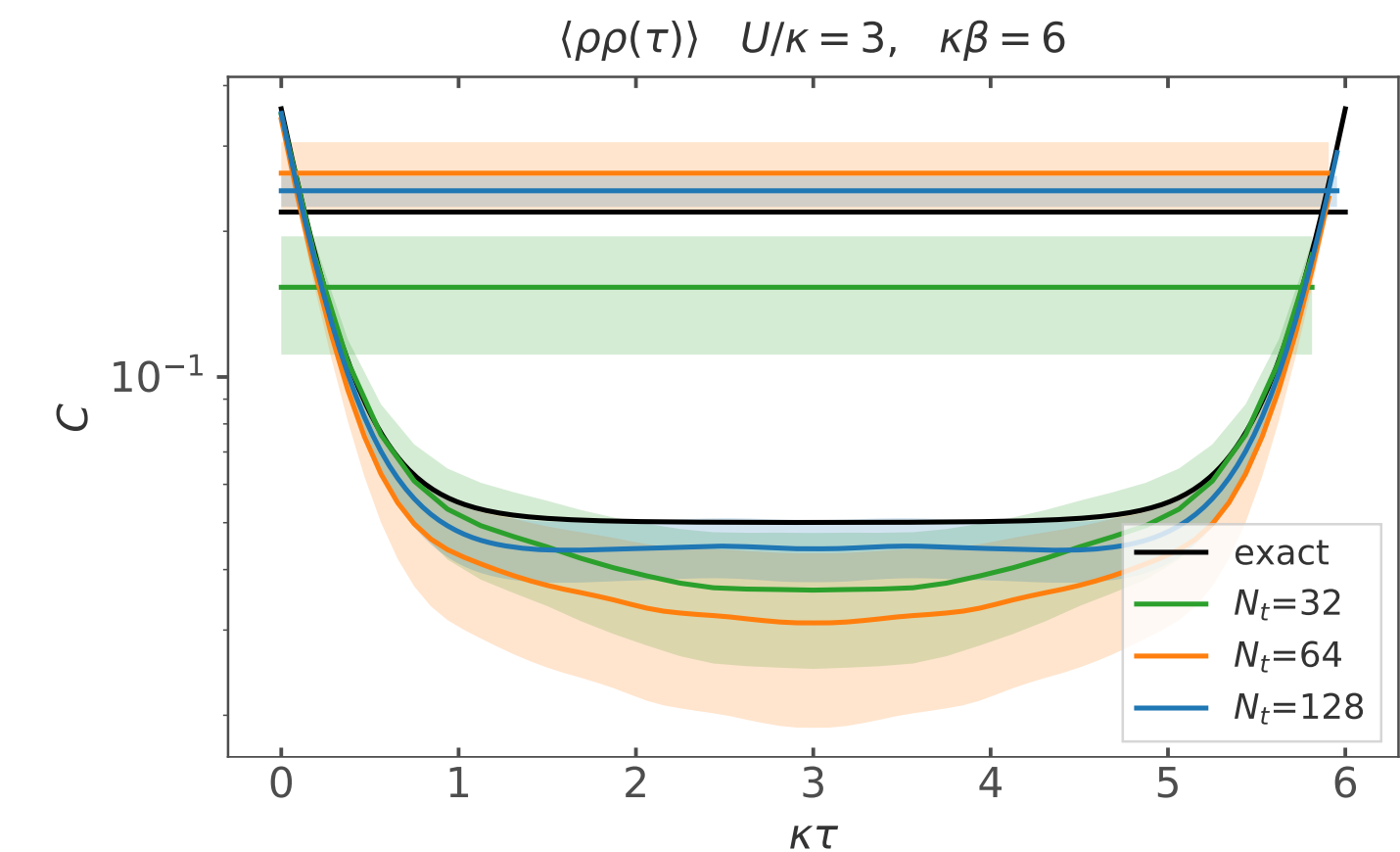
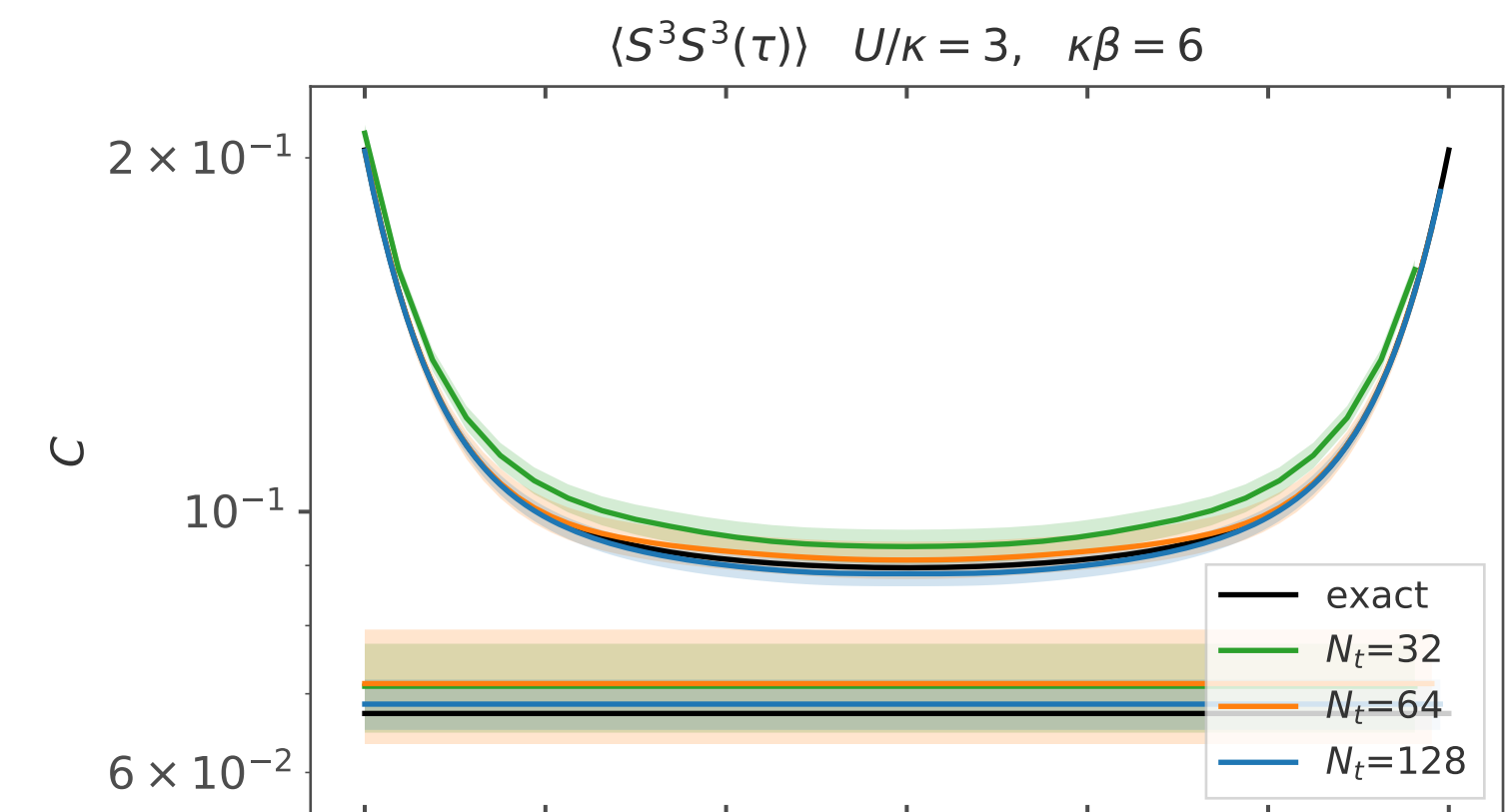
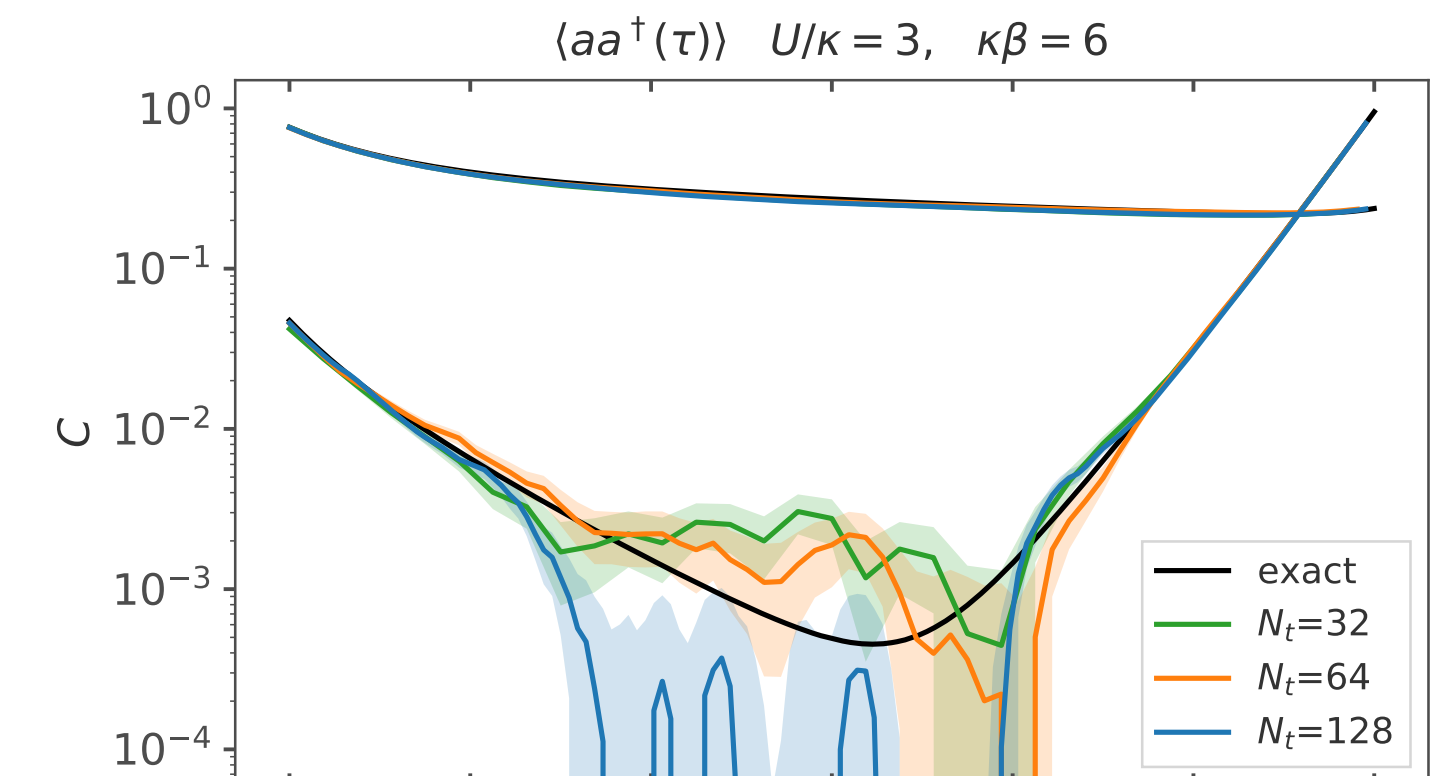
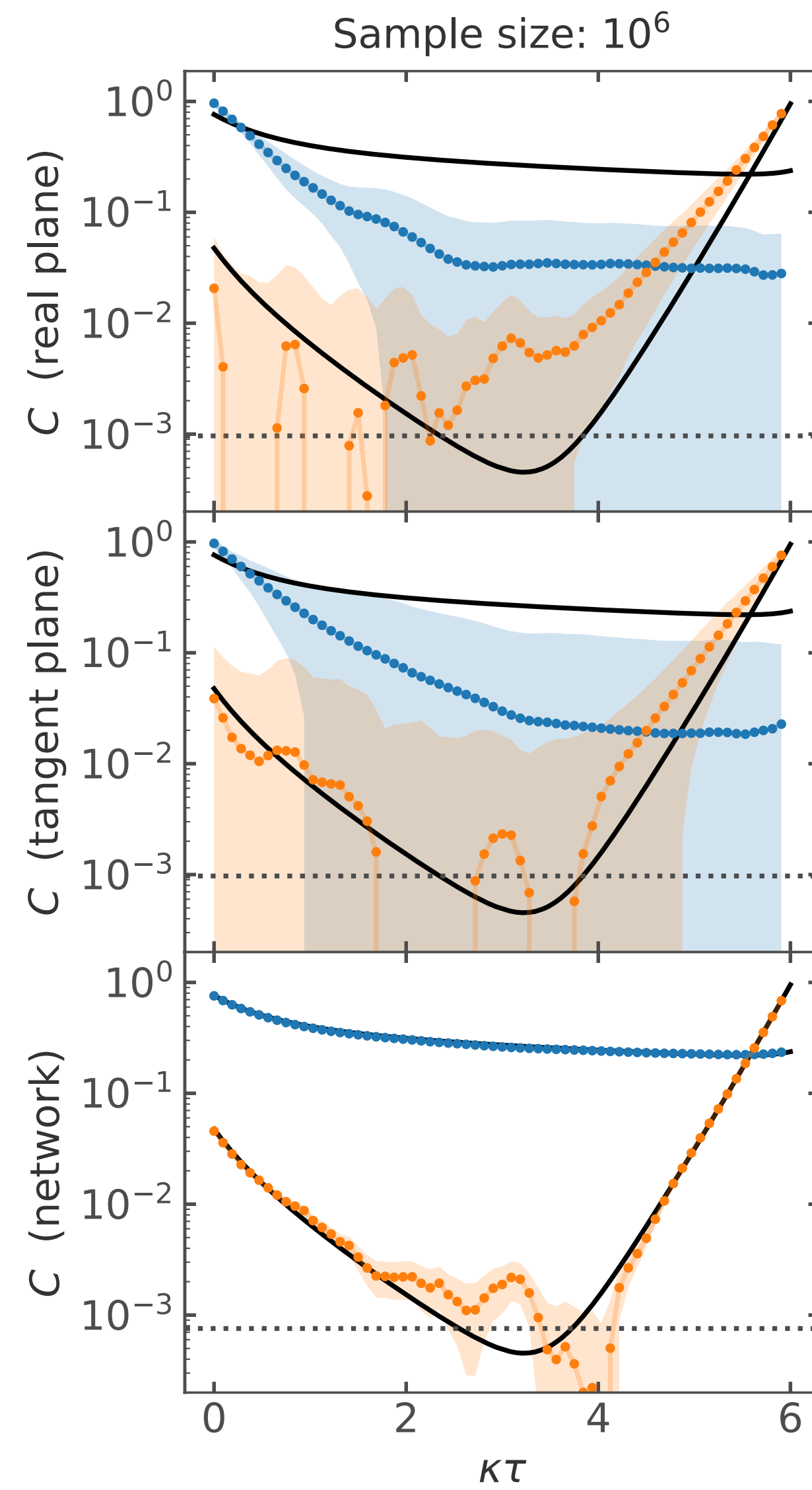
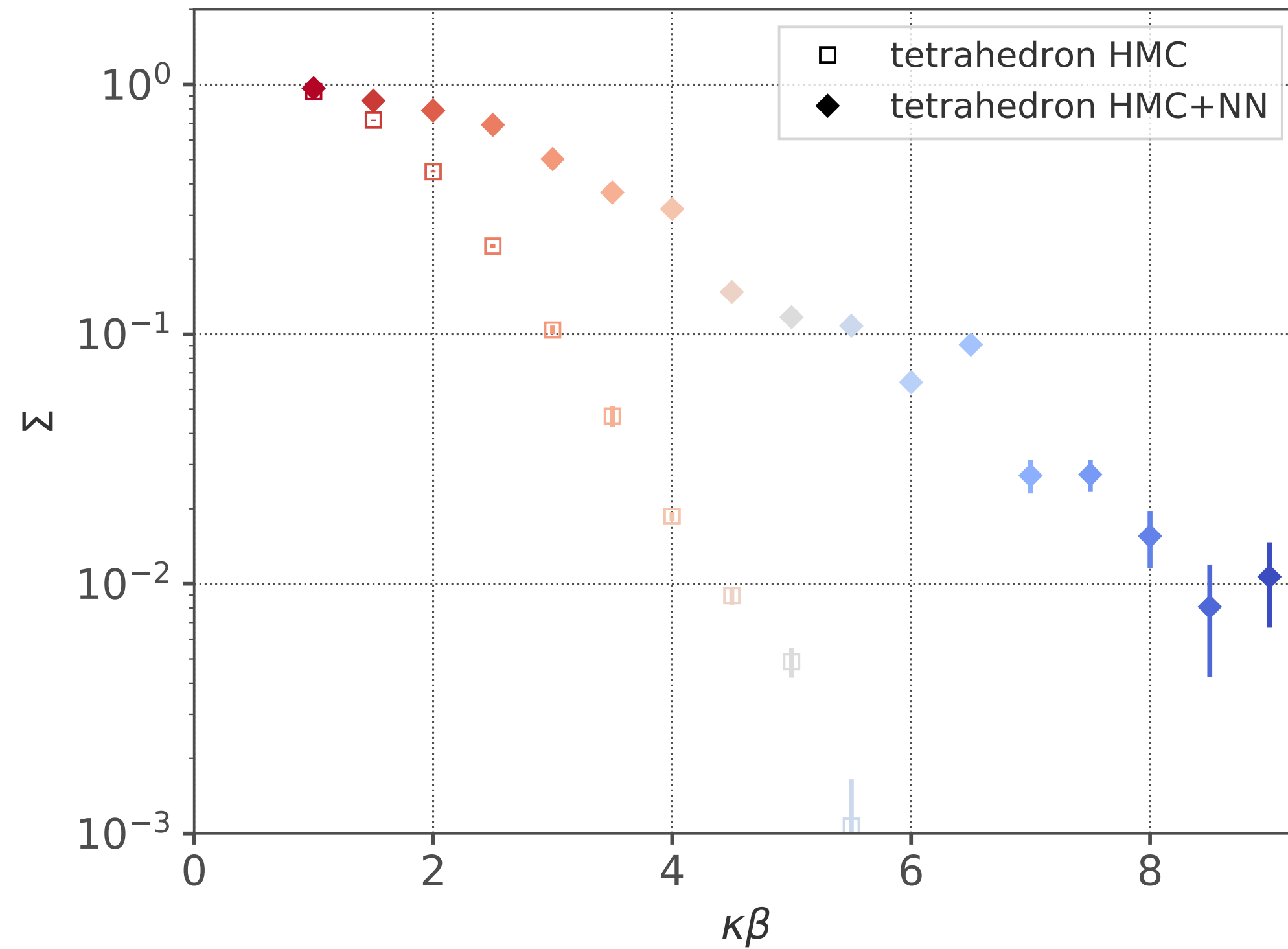
Jacobian is easy: V^1

NONBIPARTITE EXAMPLES

Wynen, EB, Krieg, Luu, Ostmeyer 2006.11221
 Rodekamp, EB, Gäntgen, Krieg, Luu, Ostmeyer 2203.00390

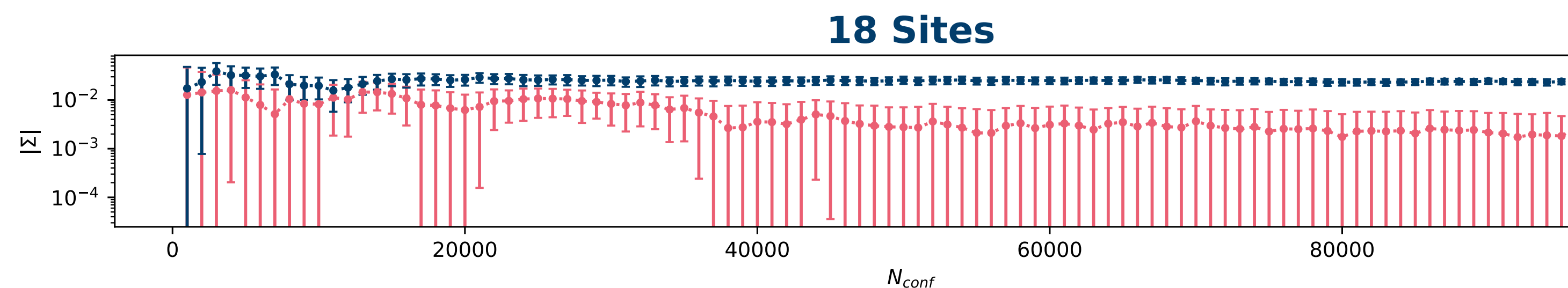
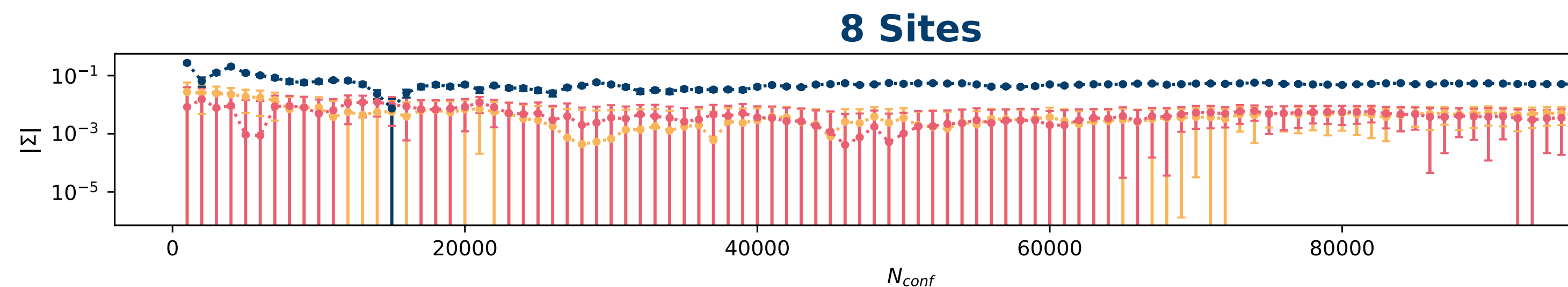
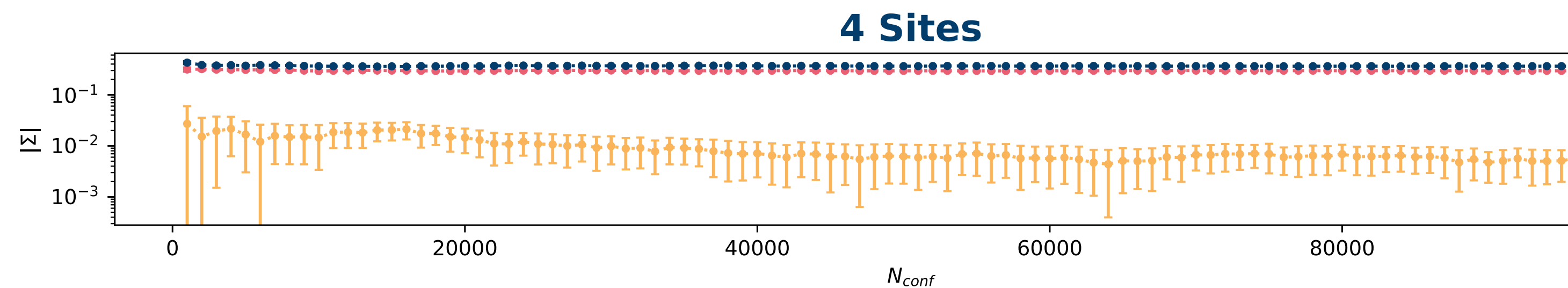
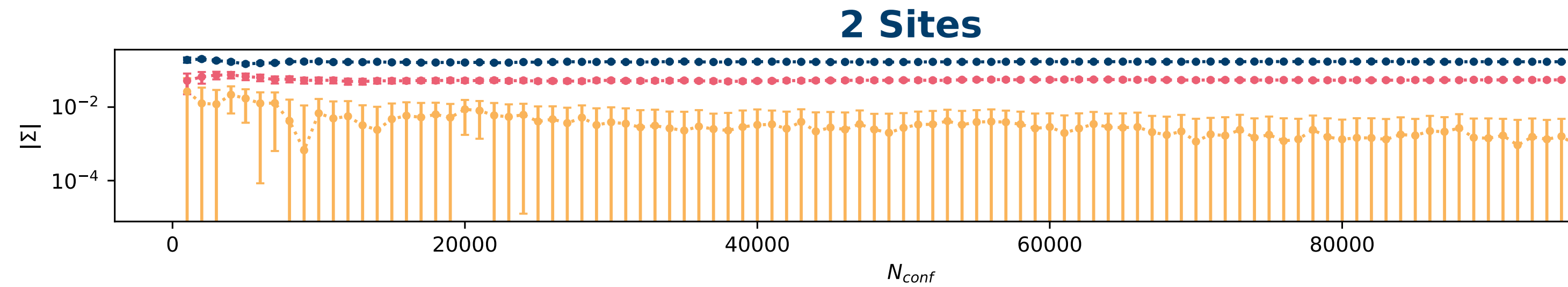
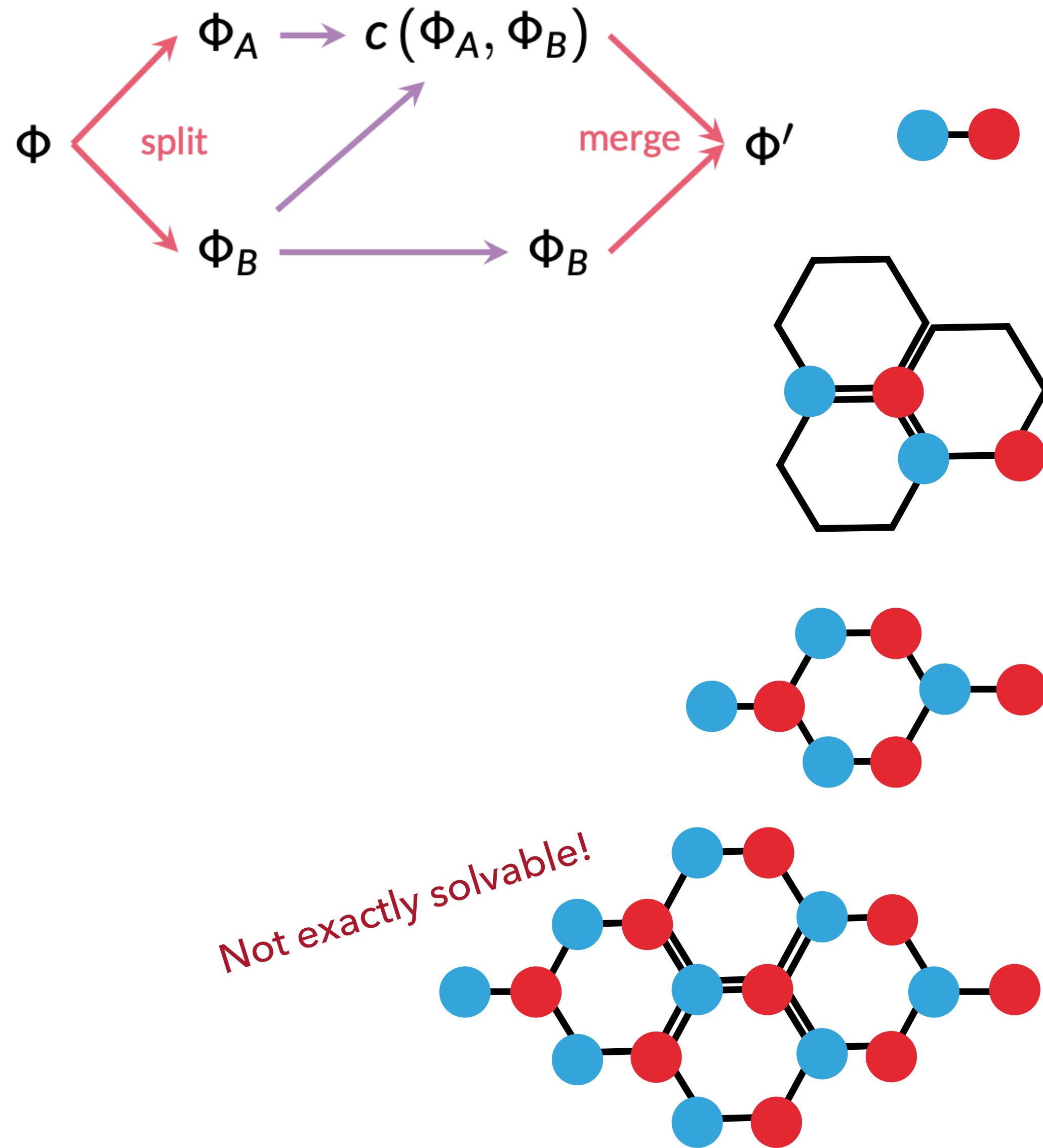


$U/\kappa = 3$



DOPED EXAMPLES

Wynen, EB, Krieg, Luu, Ostmeyer 2006.11221
 Rodekamp, EB, Gäntgen, Krieg, Luu, Ostmeyer 2003.00390



Real Plane HMC Tangent Plane HMC ML HMC

DOPED EXAMPLES

