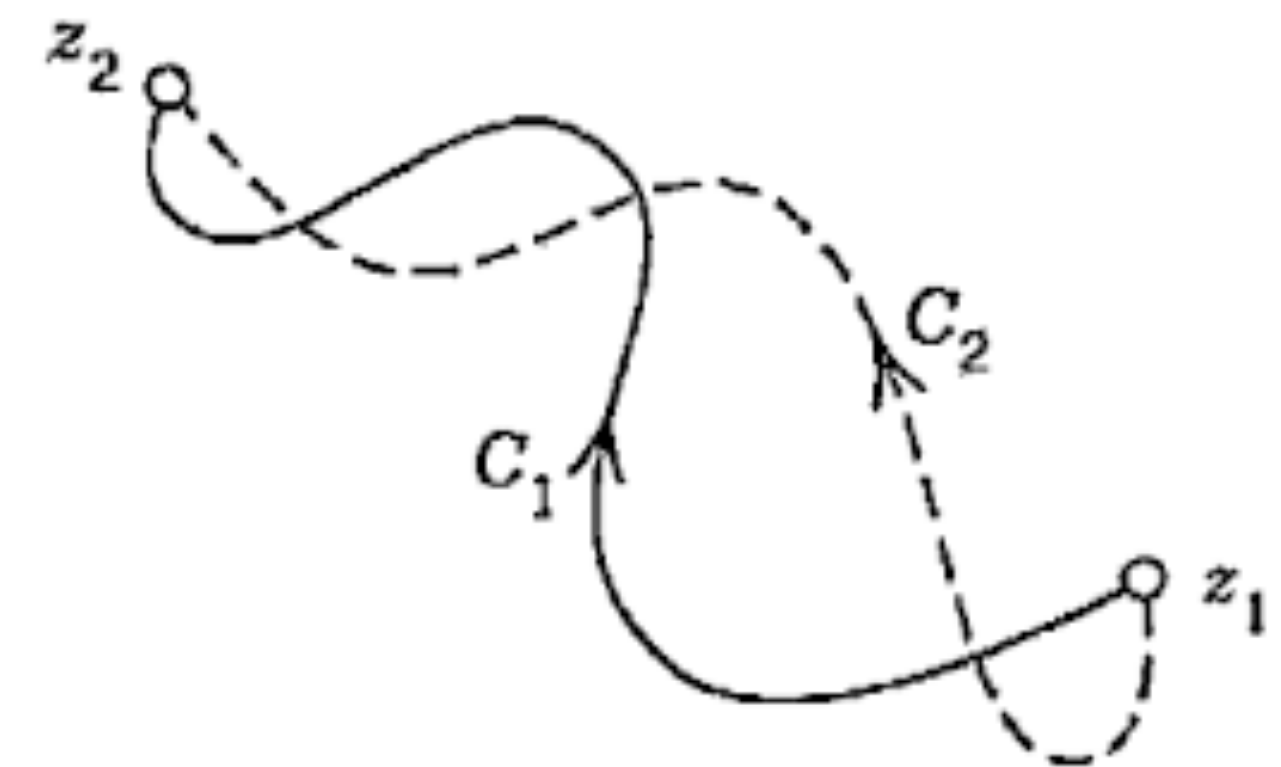


Neural-network contour deformations

for the

signal-to-noise problem



Detmold, GK, Wagman [PRD98 \(2018\) 074511](#), [PoS LATTICE2018 176](#)

Detmold, GK, Wagman, Warrington [PRD102 \(2020\) 014514](#)

Detmold, GK, Lamm, Wagman, Warrington [PRD103 \(2021\) 094517](#)

Detmold, GK, Lin, Shanahan, Wagman [2309.00600](#), [NeurIPS ML4PS \(2023\)](#)

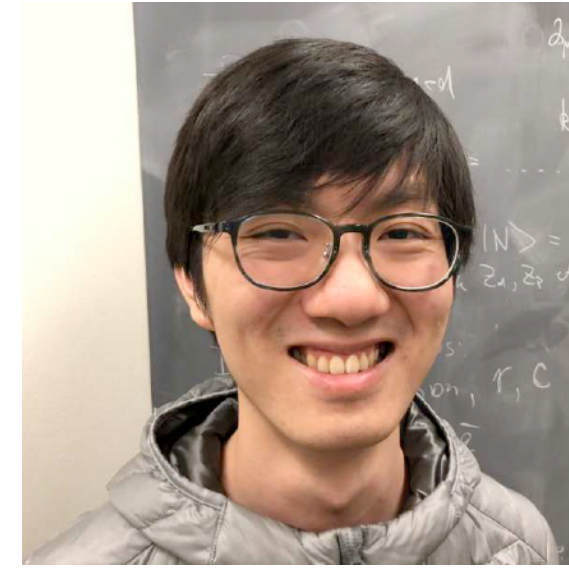
Gurtej Kanwar

Institute for Theoretical Physics, AEC, U. Bern

July 24-26, 2024
ML-LFT Workshop, Swansea



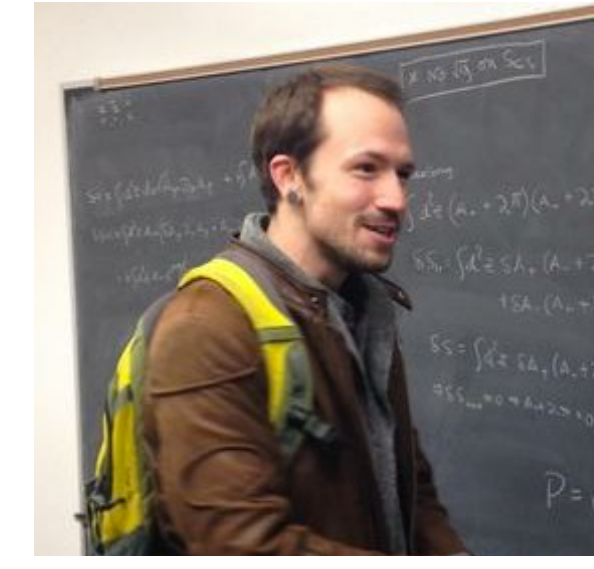
Will Detmold



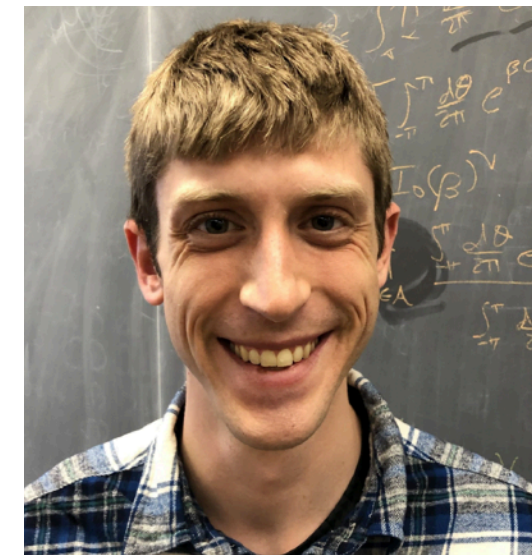
Yin Lin



Phiala Shanahan



Neill Warrington



Mike Wagman



Hank Lamm

LQFT correlation functions

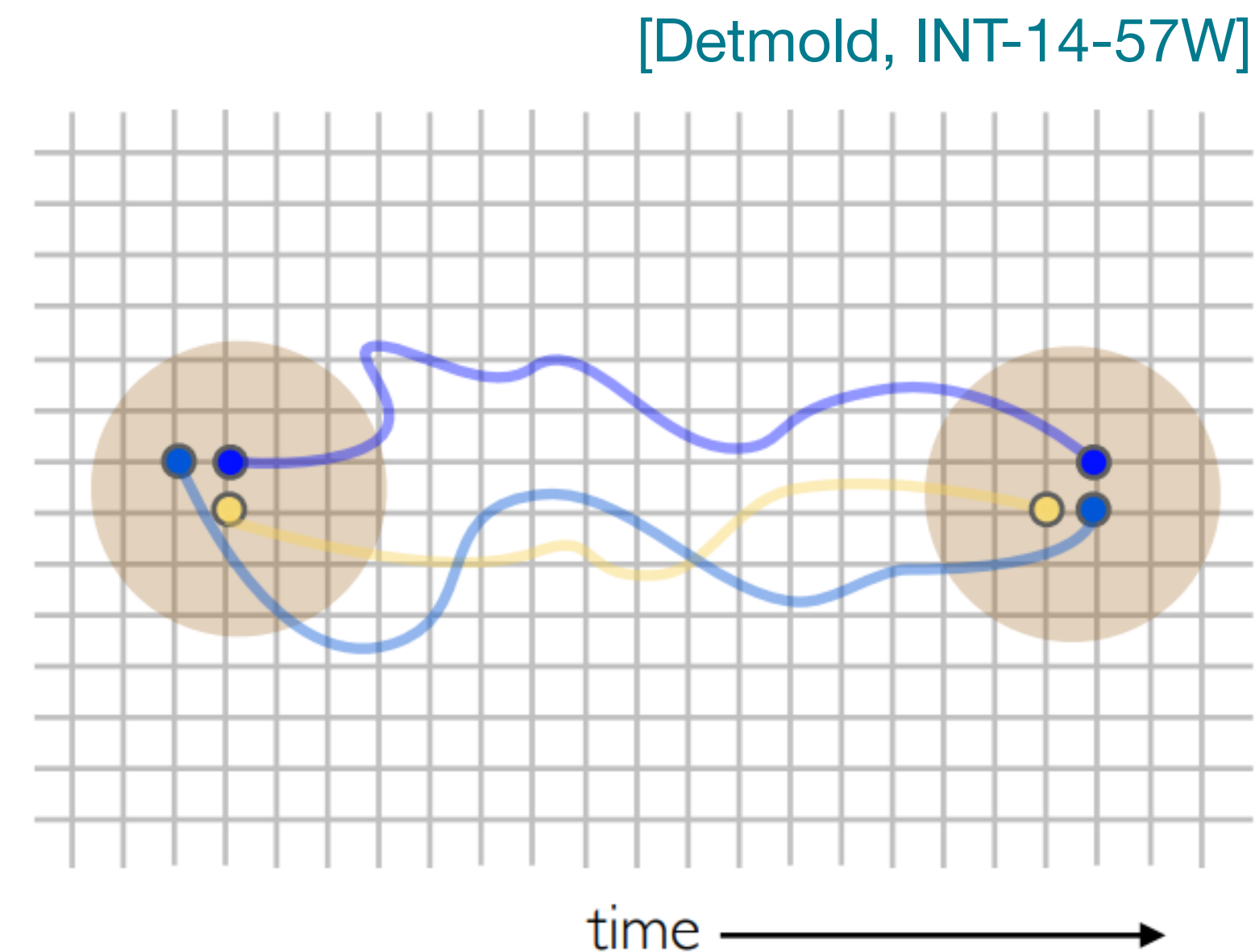
Imaginary-time correlation functions inform us of the spectrum of the theory

$$\langle \mathcal{A}(t) \mathcal{A}^\dagger(0) \rangle = \sum_n Z_n e^{-E_n t} \xrightarrow{t\Delta E \gg 1} Z_0 e^{-E_0 t}$$

Operators designed to create/
annihilate state(s) of interest

Ground state energy (e.g.
particle mass)

Matrix elements, form factors, etc. accessible via additional operator insertions.



E.g. for the nucleon in lattice QCD

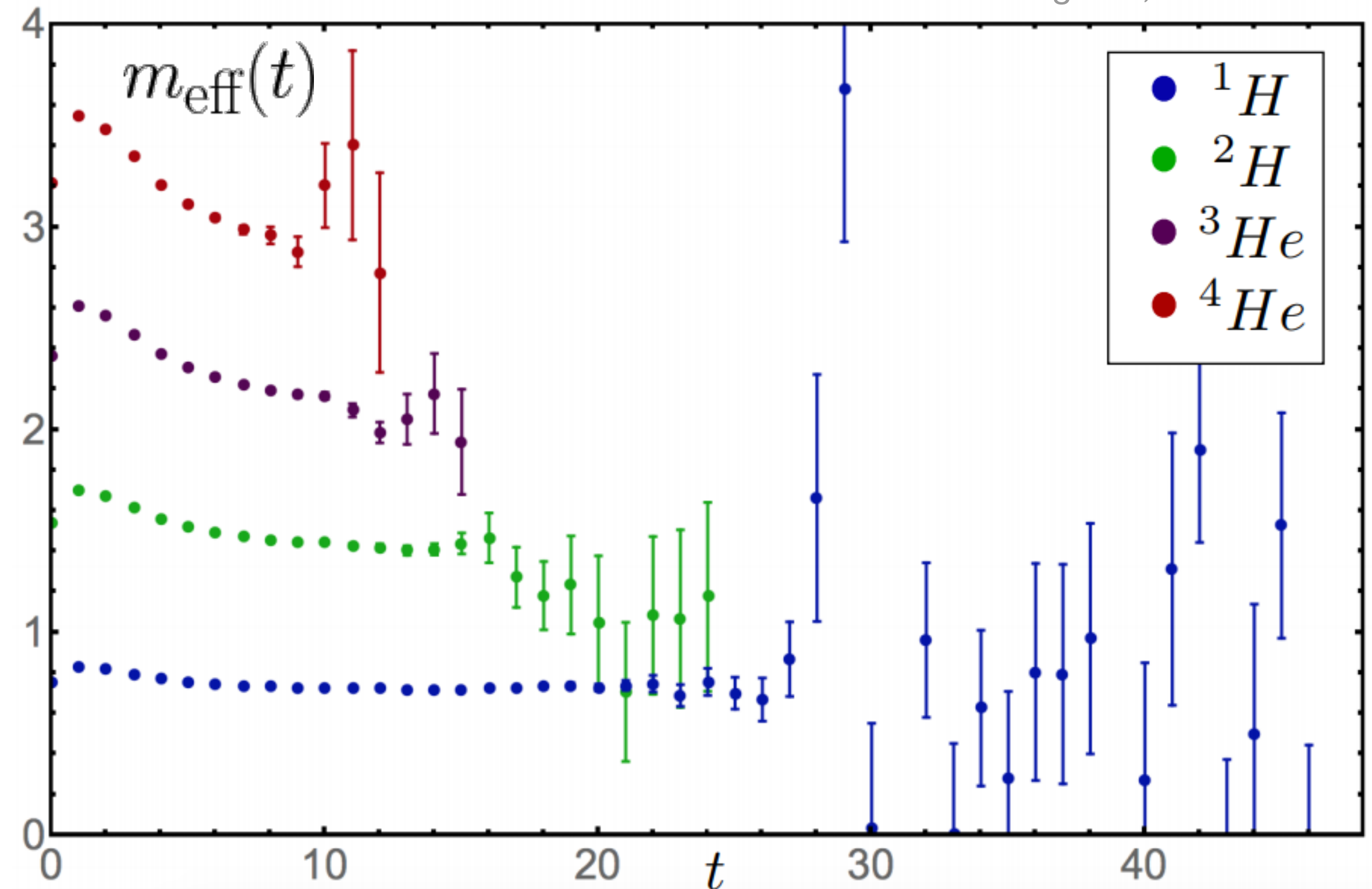
Noise problems in correlation functions

$$\begin{aligned} m_{\text{eff}}(t) &= -\partial_t \log \langle \mathcal{A}(t) \mathcal{A}^\dagger(0) \rangle \\ &= -\partial_t \log \left[\sum_n Z_n e^{-E_n t} \right] \xrightarrow{t\Delta E \gg 1} E_0 \end{aligned}$$

Must find **plateau***

- Small t : Excited-state contamination
- Large t : **Signal-to-noise ratio vanishes exponentially!**

Wagman, Lattice 2018



* Or regime dominated by only a few states

Noise problem = sign problem

$$C(t) \equiv \mathcal{A}(t)\mathcal{A}^\dagger(0)$$

Intuitively a noise problem implies ...

$$\text{Err} \sim \sqrt{\langle |C(t)|^2 \rangle}$$

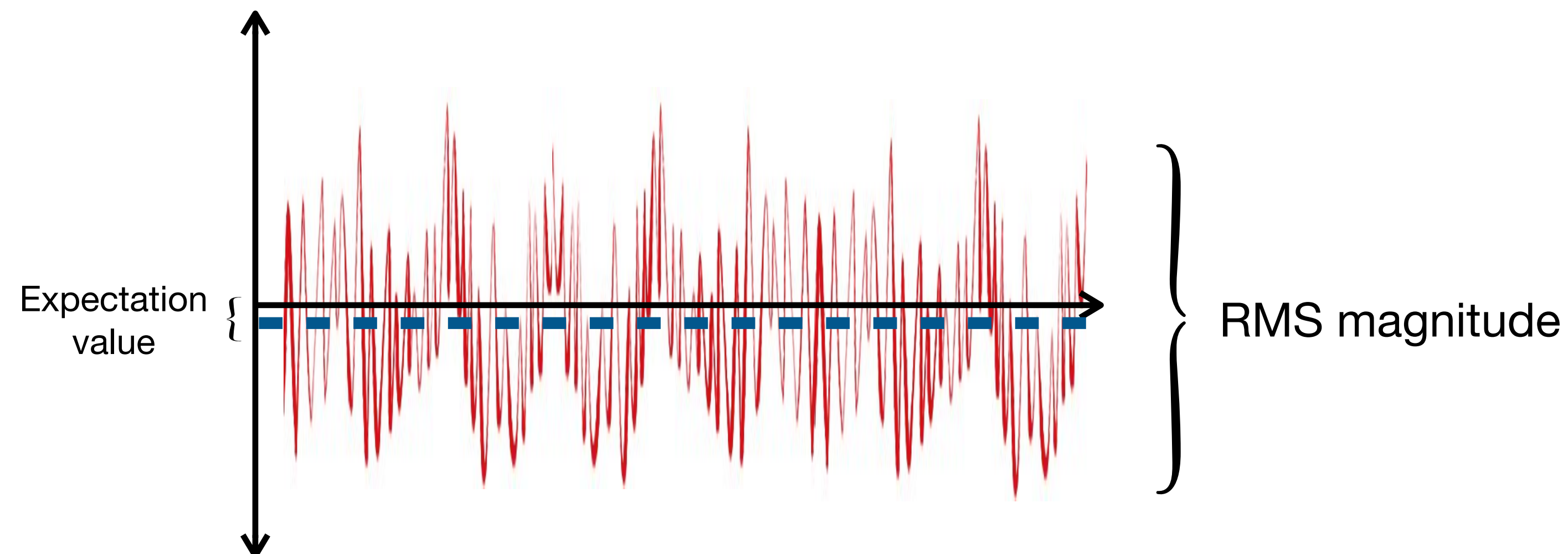
The RMS magnitude ...

\gg

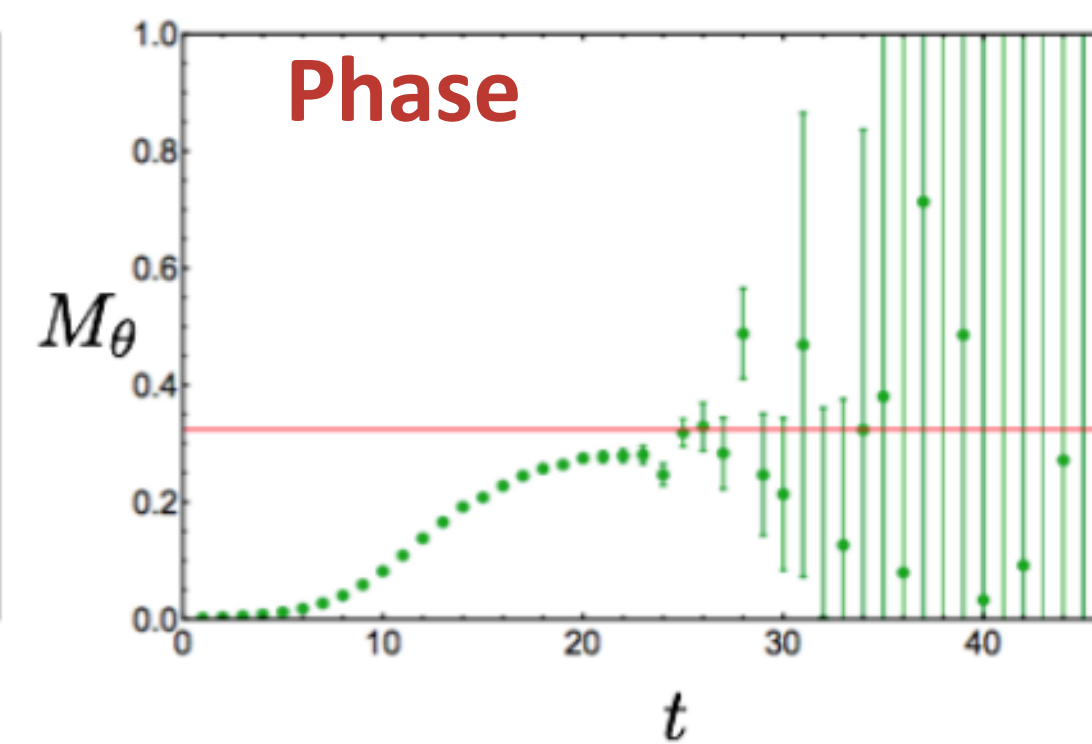
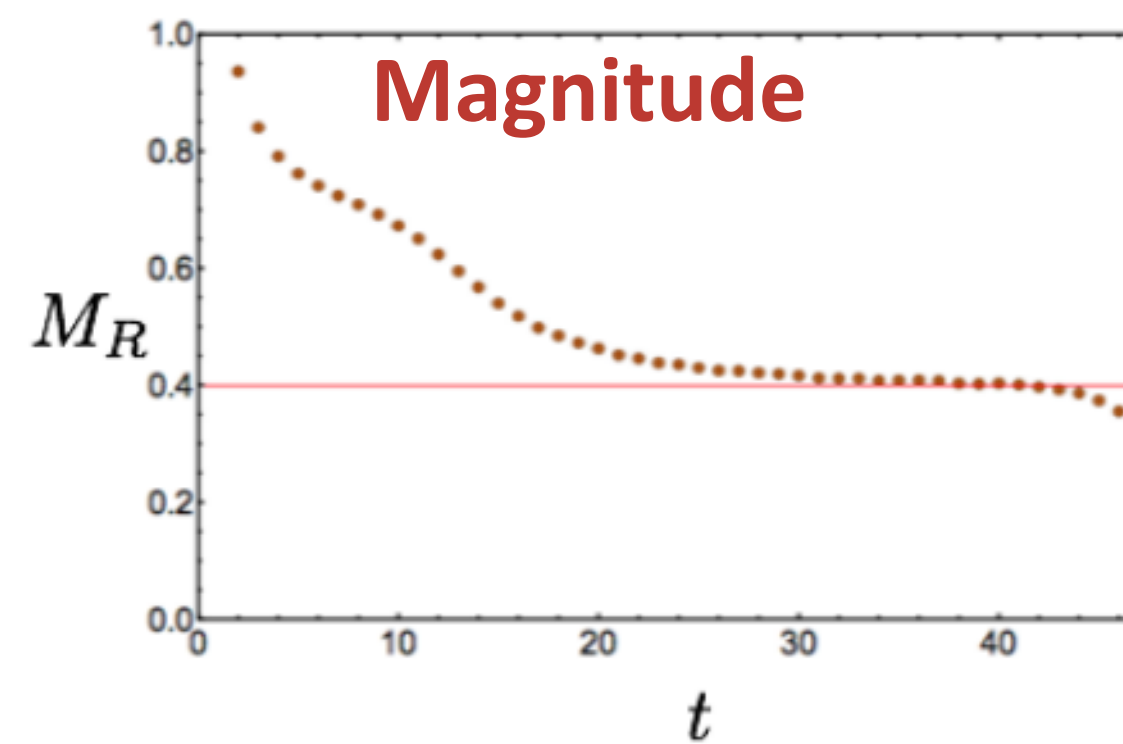
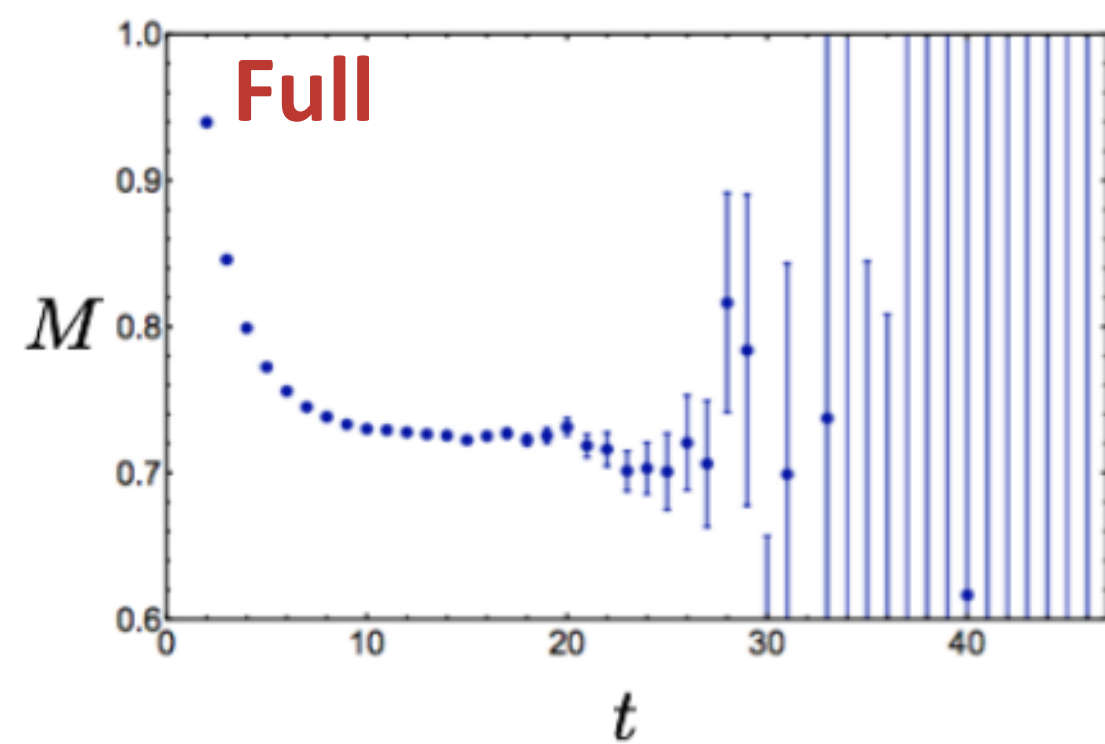
$$|\langle C(t) \rangle|$$

... is exponentially larger than the expectation value.

... which is a sign problem! (if the magnitude is well concentrated)



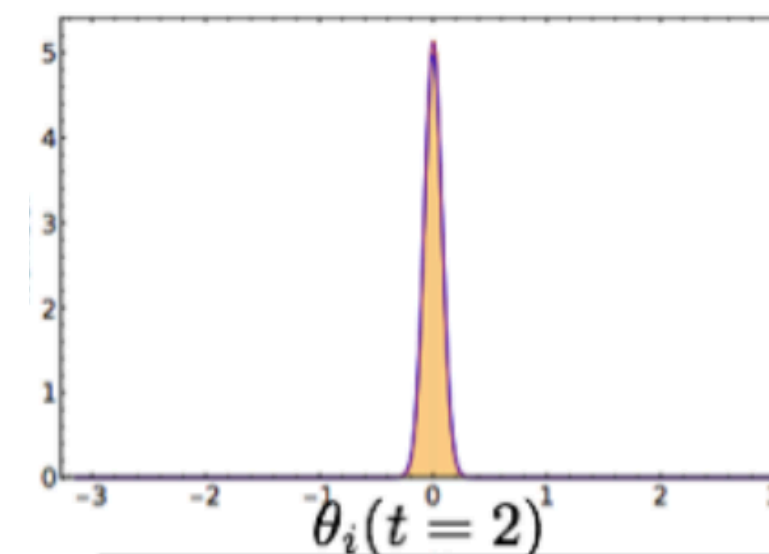
Noise problem = sign problem



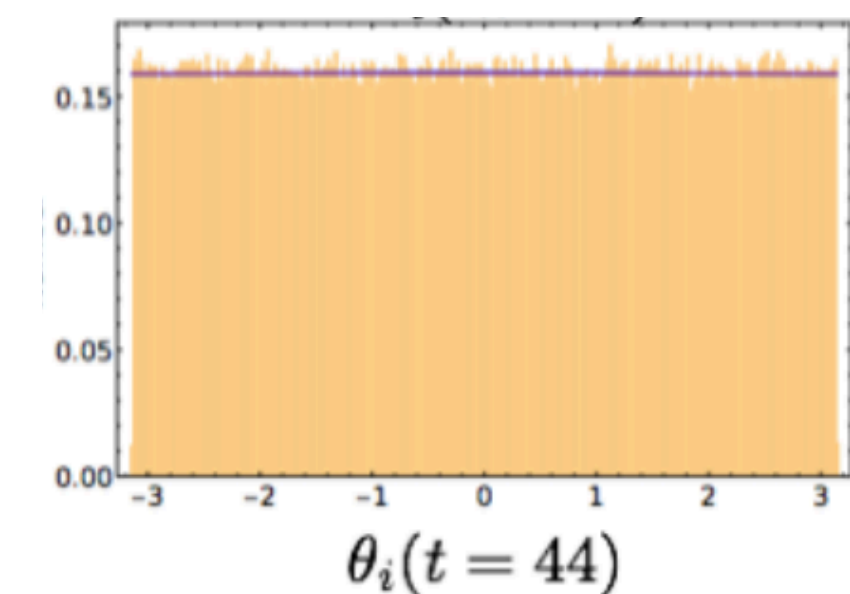
Noise in nucleon effective mass carried by complex phase

[Wagman & Savage, PRD96 (2017) 114508]

Distribution of complex phases \sim uniform at large baryon correlator separations.



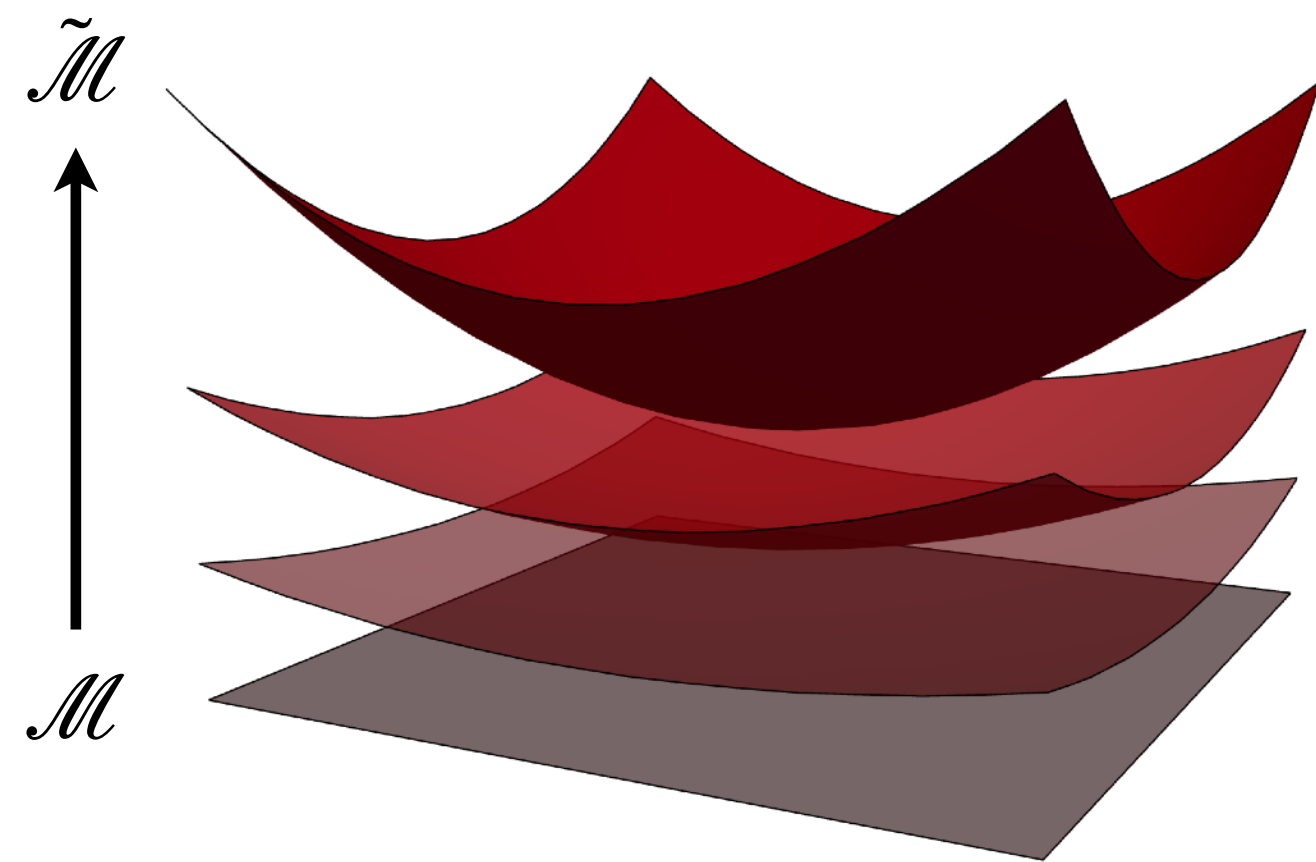
vs.



Path integral deformations

Lattice integrands are often holomorphic, allowing the integration contour to be deformed without bias.

$$\langle \mathcal{O} \rangle = \frac{1}{Z} \int_{\mathcal{M}} e^{-S(\phi)} \mathcal{O}(\phi) = \frac{1}{Z} \int_{\tilde{\mathcal{M}}} e^{-S(\tilde{\phi})} \mathcal{O}(\tilde{\phi})$$



Connection to
Lefschetz thimbles,
Complex Langevin

Many works applying this to oscillatory path integrals originating from complex $S(\phi)$ (e.g. non-zero density, real time)

Cristoforetti, et al. PRD86(074506), PRD88(051501), PRD89(114505)

Aarts PRD88(094501)

Alexandru, et al. PRD93(014504), JHEP05(053), PRD96(094505), PRD98(054514), PRD98(034506), PRD97(094510), PRL121(191602)

Fujii, et al. JHEP12(125)

Tanizaki, et al. NJP18(033002)

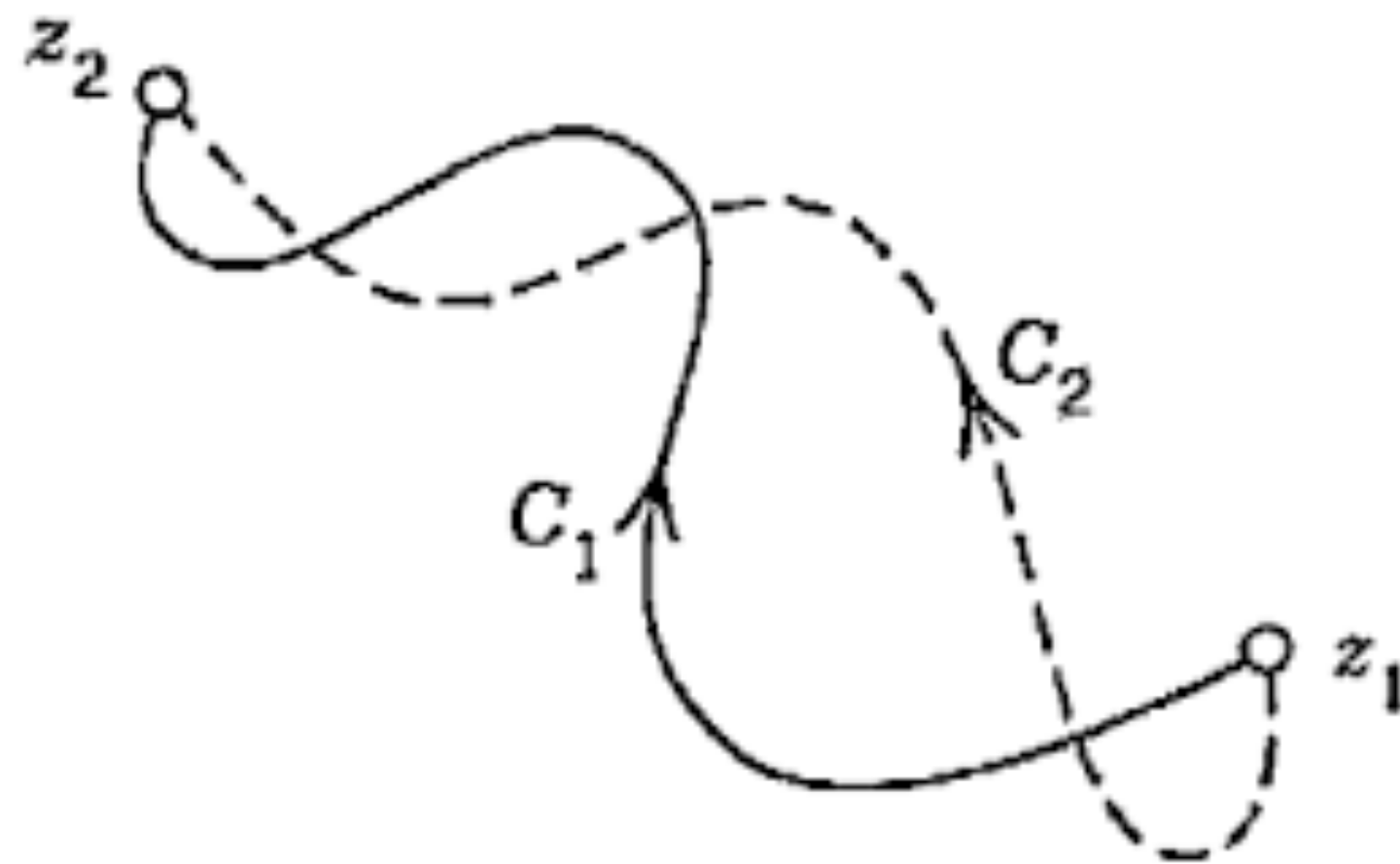
Mori, et al. PTEP2018(023B04), PRD99(014033)

Alexandru, et al. PRL117(081602), PRD95(114501)

Mou, et al. JHEP11(135)

...

Can we leverage **contour deformations** to shrink $\text{Var}[\mathcal{O}]$ while preserving $\langle \mathcal{O} \rangle$?



Case study: Gaussian StN problem

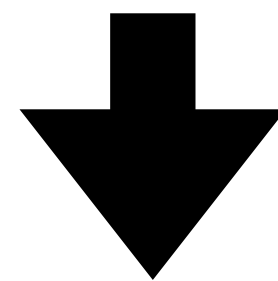
Toy example: Simple observable in a Gaussian “theory”

$$\langle e^{ikx} \rangle = \frac{1}{Z} \int_{-\infty}^{\infty} dx [e^{ikx}] e^{-x^2/2}$$

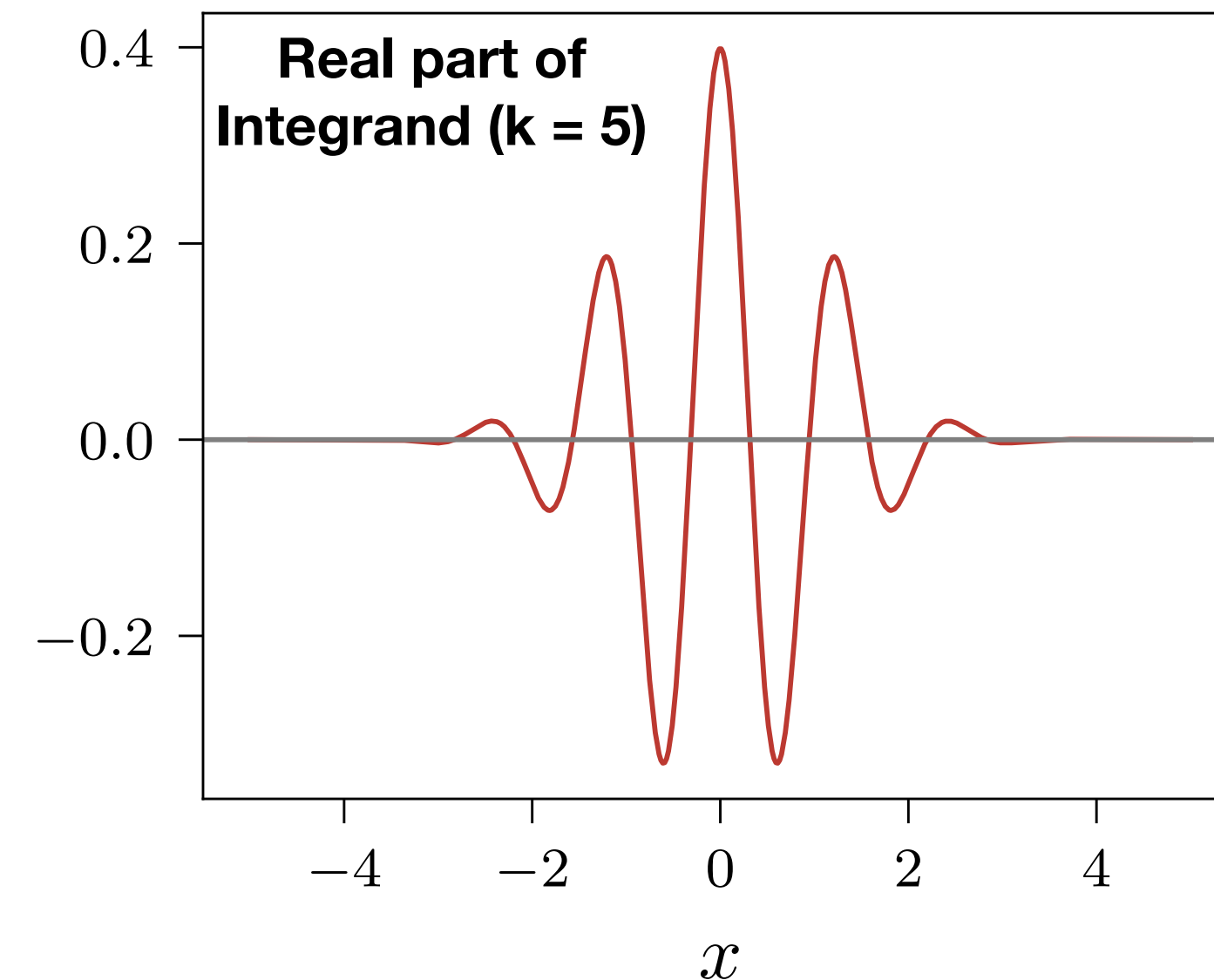
Monte Carlo approach: Sample $p(x) = e^{-x^2/2}/Z$, measure e^{ikx}

Signal: $\langle e^{ikx} \rangle = e^{-k^2/2}$

Variance: $\langle |e^{ikx}|^2 \rangle = 1$



StN and sign problem!

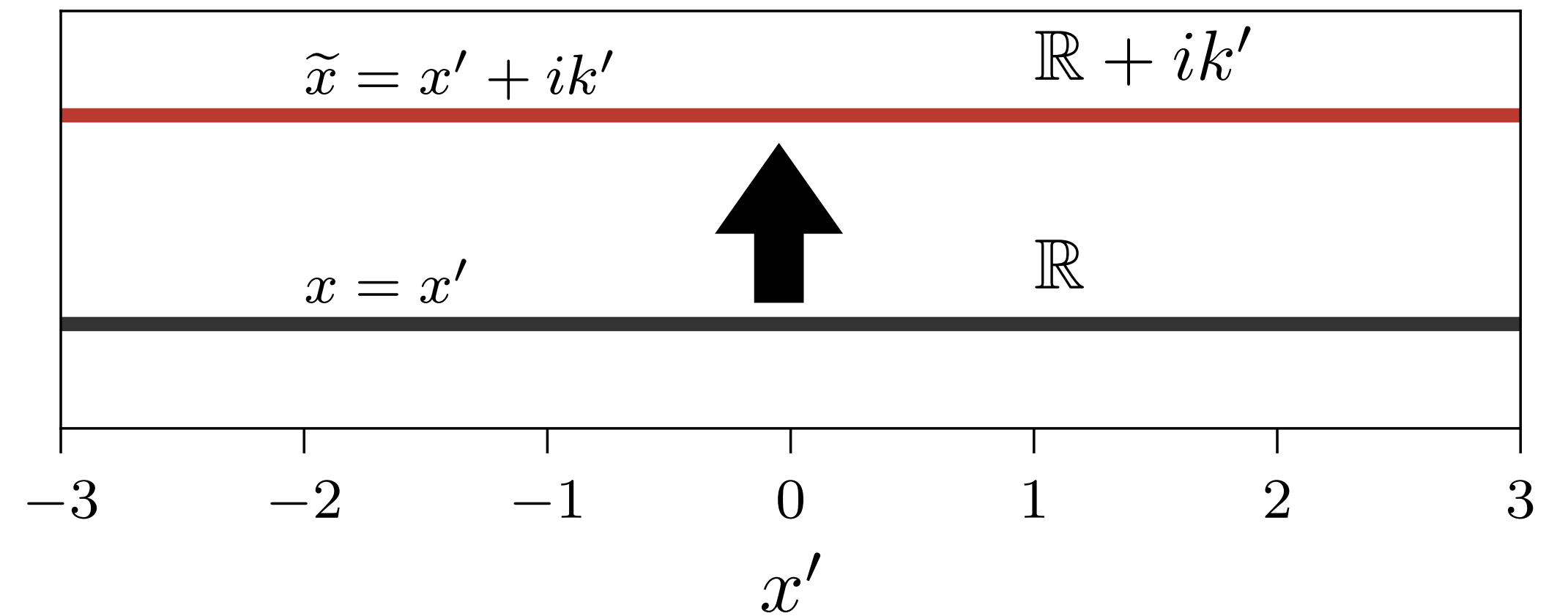


Case study: Gaussian StN problem

Deformation approach:

$$\begin{aligned}
 \langle e^{ikx} \rangle &= \frac{1}{Z} \int_{\mathbb{R}} dx [e^{ikx}] e^{-\frac{1}{2}x^2} && \text{1. Analytically continue \& deform contour} \\
 &= \frac{1}{Z} \int_{\mathbb{R}+ik'} d\tilde{x} [e^{ik\tilde{x}}] e^{-\frac{1}{2}\tilde{x}^2} \\
 &= \frac{1}{Z} \int_{-\infty}^{\infty} dx' [e^{ikx'-kk'}] e^{-\frac{1}{2}x'^2 - ix'k' + \frac{1}{2}k'^2} && \text{2. Give coordinates to new contour} \\
 &= \left\langle e^{ix'(k-k')} e^{\frac{1}{2}k'^2 - kk'} \right\rangle && \text{3. New observable w.r.t original MC weights}
 \end{aligned}$$

Result: $\langle e^{ikx} \rangle = \left\langle e^{ix'(k-k')} e^{\frac{1}{2}k'^2 - kk'} \right\rangle$

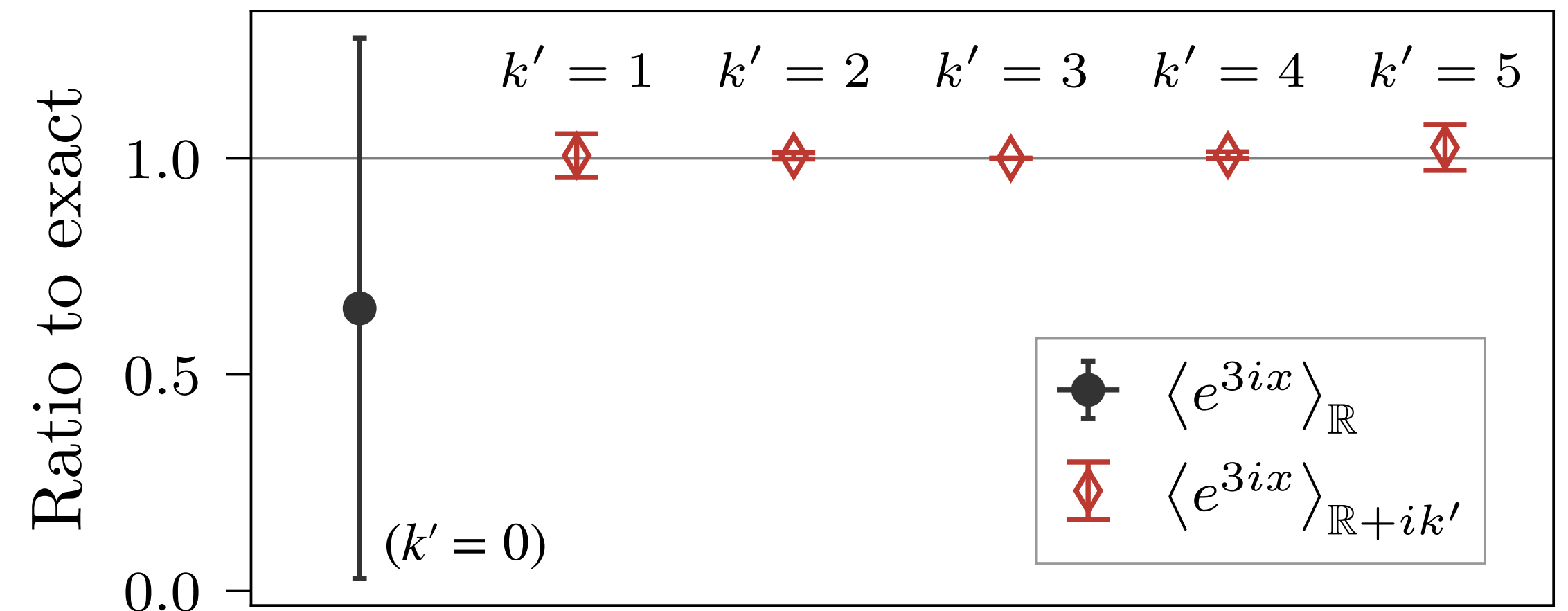


Case study: Gaussian StN problem

Less severe sign problem: Deformed observable $[e^{ikx' - kk'}]$ has smaller magnitude for $kk' > 0$.

Exactness preserved: Anti-correlated phase fluctuations from the deformed action $e^{-\frac{1}{2}x'^2 - ix'k' + \frac{1}{2}k'^2}$.

$$\text{Result: } \langle e^{ikx} \rangle = \left\langle e^{ix'(k-k')} e^{\frac{1}{2}k'^2 - kk'} \right\rangle$$

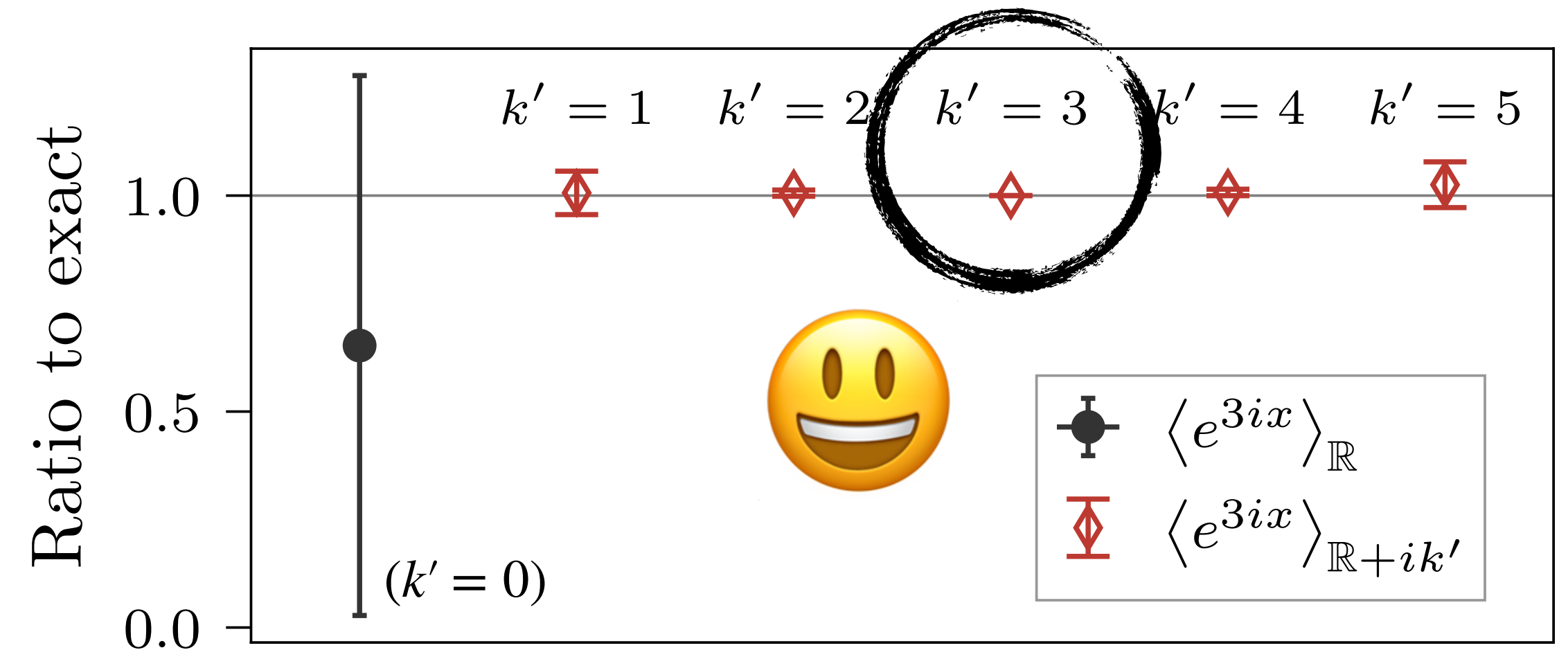


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$$\text{Result: } \langle e^{ikx} \rangle = \left\langle e^{ix'(k-k')} e^{\frac{1}{2}k'^2 - kk'} \right\rangle$$



Integral deformations for noisy observables

Detmold, GK, Wagman, Warrington PRD102 (2020) 014514

- Deformed path integral defines a **modified observable**:

$$\mathcal{Q}(\phi) \equiv \det J(\phi) e^{-[S(\tilde{\phi}(\phi)) - S(\phi)]} \mathcal{O}(\tilde{\phi}(\phi))$$

- Identical Monte Carlo expectation values $\langle \cdot \rangle$, different variance:

$$\begin{aligned} \langle \mathcal{Q}(\phi) \rangle &= \langle \mathcal{O}(\phi) \rangle \\ \text{Var}[\mathcal{Q}(\phi)] &\neq \text{Var}[\mathcal{O}(\phi)] \end{aligned}$$

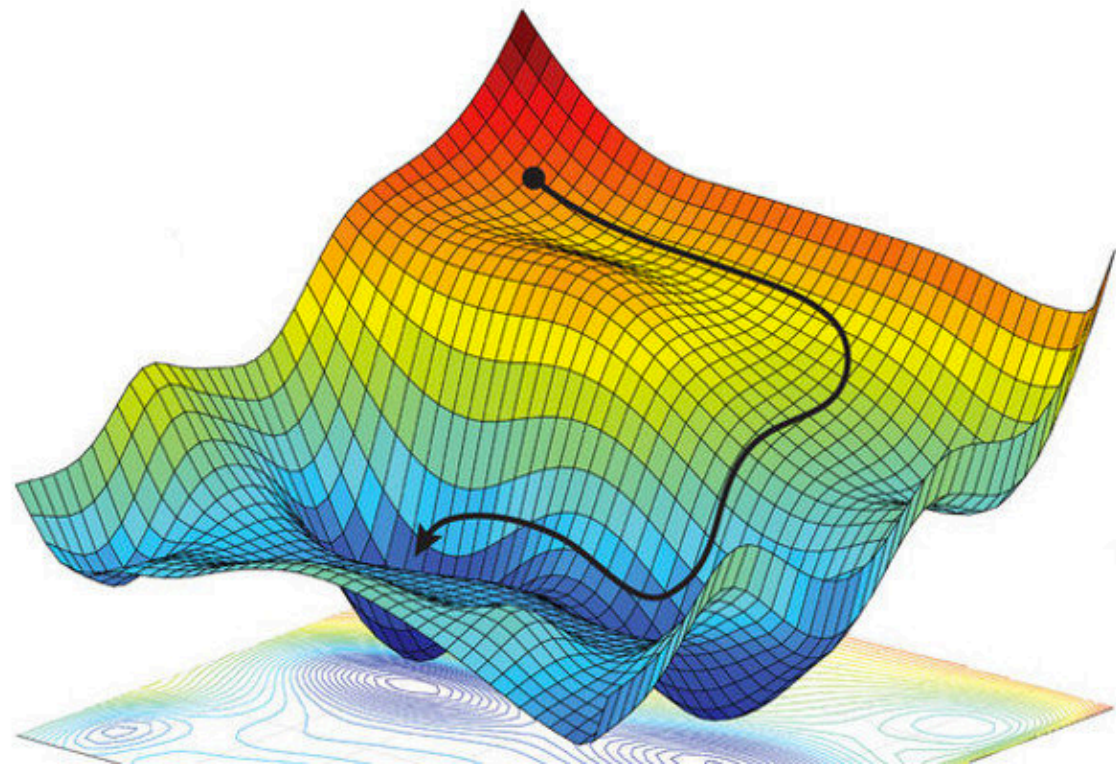
Variance may be reduced by a good choice of deformation

Learning the integration contour

The choice of $f : \phi \mapsto \tilde{\phi}$ defines $\tilde{\mathcal{M}}$, $\mathcal{Q}(\phi)$, and the variance.

Parameterize $f(\phi; \omega)$ then **minimize variance**.

- Caveat: Complex analyticity
- Caveat: $SU(N)$ variables



[Image credit: 1805.04829]

Insight: gradients of variance w.r.t. ω can be defined using **original Monte Carlo ensemble**.

$$\begin{aligned} \nabla_{\vec{\omega}} \text{Var}[\text{Re } \mathcal{Q}] &= \langle \nabla_{\vec{\omega}} (\text{Re } \mathcal{Q})^2 \rangle = 2 \langle \text{Re } \mathcal{Q} \text{ Re } \nabla_{\vec{\omega}} \mathcal{Q} \rangle \\ &= 2 \left\langle (\text{Re } \mathcal{Q}) \text{Re} \left(\mathcal{Q} \left[-\nabla_{\vec{\omega}} S_{\text{eff}} + \frac{\nabla_{\vec{\omega}} \mathcal{O}(\tilde{U})}{\mathcal{O}(\tilde{U})} \right] \right) \right\rangle \end{aligned}$$

Analytic continuation & holomorphy

Write Boltzmann weight e^{-S} and observable \mathcal{O} in terms of **real field variables...**

- For $SU(N)$, we use angular parameters $\Omega \equiv (\phi_1, \dots, \theta_1, \dots)$,
 $\phi_i \in [0, 2\pi]$ and $\theta_i \in [0, \pi/2]$
 $(N^2 - 1)$ angles

Bronzan PRD38 (1988) 1994

SU(3) parameterization

... then **analytically continue**.

- Complexified angular params extends
 $SU(N) \rightarrow SL(N, \mathbb{C})$

- Adjoints should be rewritten $U^\dagger \rightarrow U^{-1}$

$$U(\Omega) = \begin{pmatrix} c_1 c_2 e^{i\phi_1} & s_1 e^{i\phi_3} & c_1 s_2 e^{i\phi_4} \\ s_2 s_3 e^{-i(\phi_4 + \phi_5)} & c_1 c_3 e^{i\phi_2} & -c_2 s_3 e^{-i(\phi_1 + \phi_5)} \\ s_1 c_2 c_3 e^{i(\phi_1 + \phi_2 - \phi_3)} & & s_1 s_2 c_3 e^{i(\phi_2 - \phi_3 + \phi_4)} \\ -s_1 c_2 s_3 e^{i(\phi_1 - \phi_3 + \phi_5)} & c_1 s_3 e^{i\phi_5} & c_2 c_3 e^{-i(\phi_1 + \phi_2)} \\ s_2 c_3 e^{-i(\phi_2 + \phi_4)} & & s_1 s_2 s_3 e^{-i(\phi_3 - \phi_4 - \phi_5)} \end{pmatrix}$$

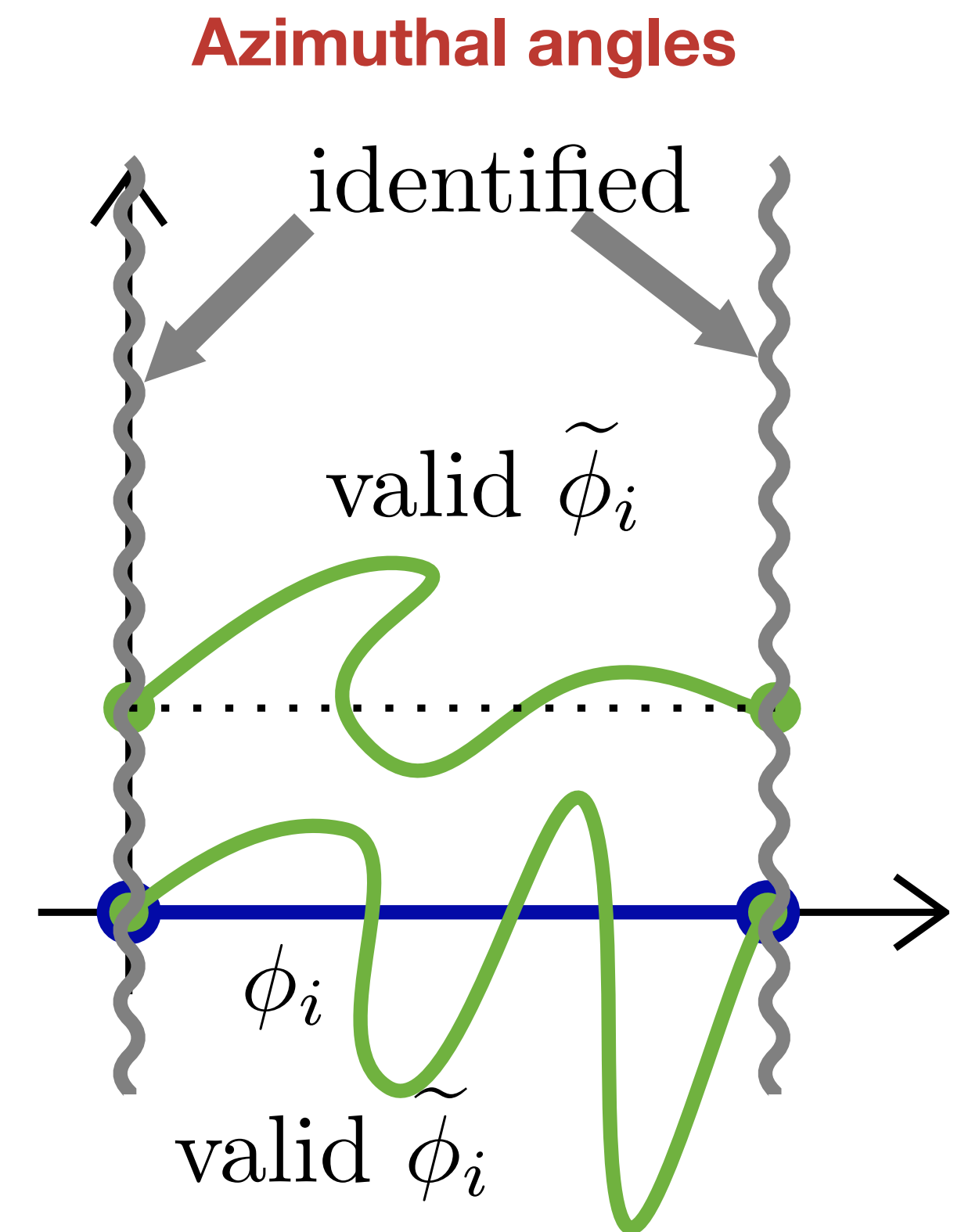
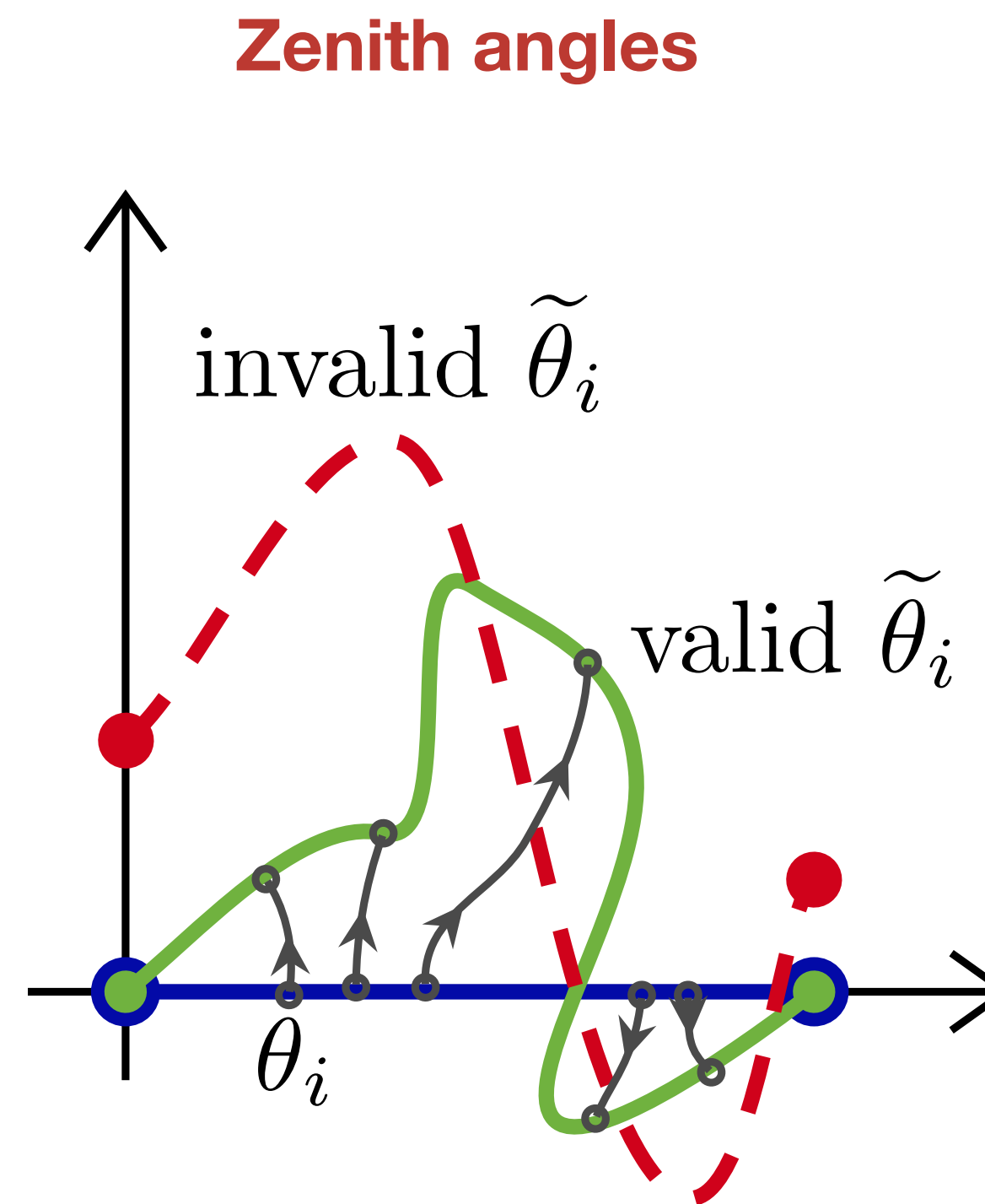
where $s_i \equiv \sin(\theta_i)$, $c_i \equiv \cos(\theta_i)$.

Deforming angular variables

Angular parameterization of $SU(N)$ has two types of angles:

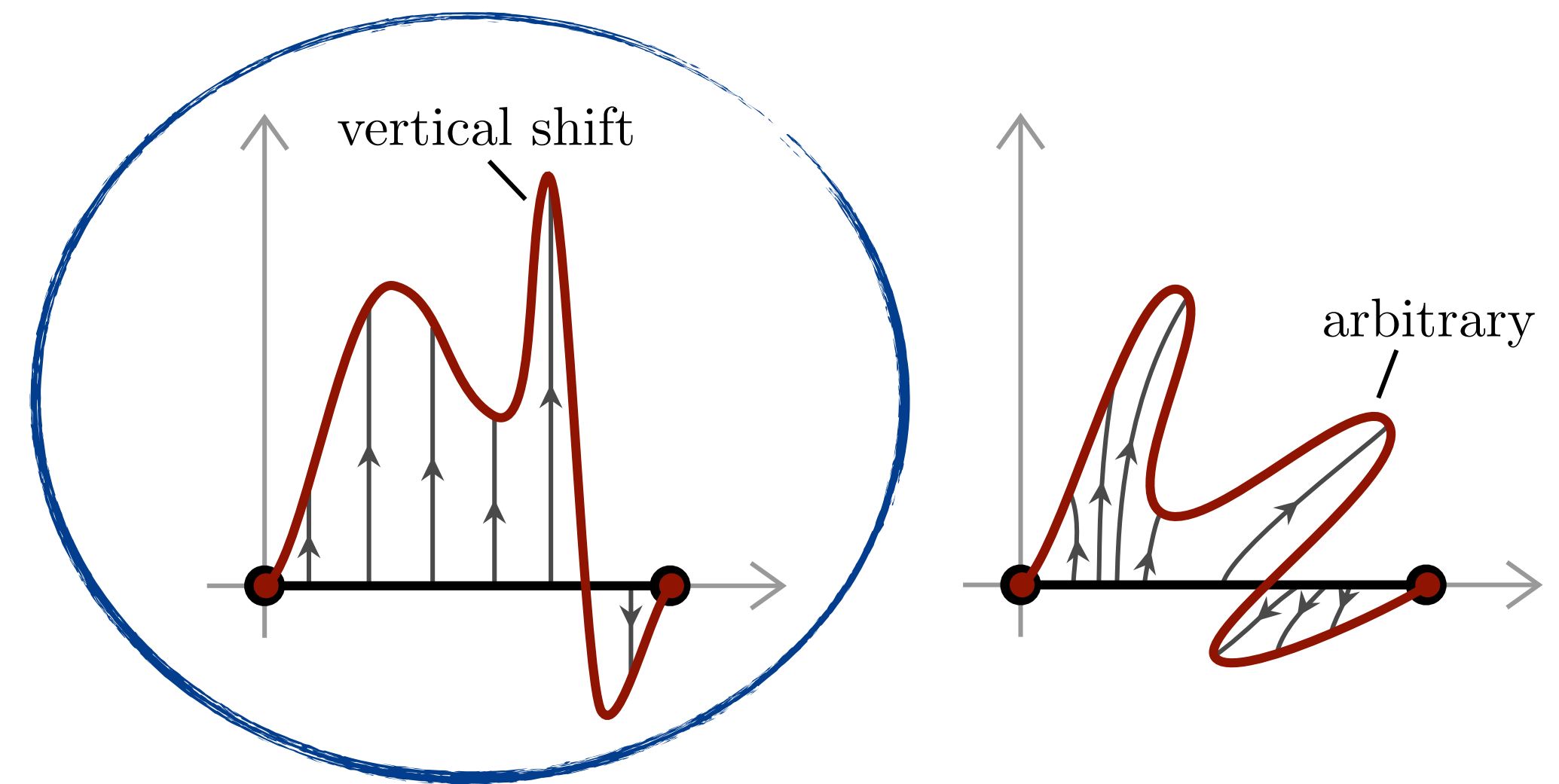
- **Azimuthal** angles $\phi_j \in [0, 2\pi]$
- **Zenith** angles $\theta_i \in [0, \pi/2]$

In each case, must deal appropriately with **endpoints**.



Deformation

Vertical deformations $\tilde{\Omega}_x = \Omega_x + if(\Omega)$



Fourier series definition of $f(\Omega)$, using a subset of all possible terms

$$\tilde{\phi}_x^a = \phi_x^a + ik_0^{x;\phi^a} + i \sum_{y \leq x} f_{\phi^a}(\Omega_y; \kappa^{xy}, \lambda^{xy}, \chi^{xy}, \zeta^{xy}),$$

$$\tilde{\theta}_x^a = \theta_x^a + i \sum_{y \leq x} f_{\theta^a}(\Omega_y; \kappa^{xy}, \lambda^{xy}, \chi^{xy}).$$

Original
real part

Parameterized
imaginary shift

$$f_{\theta^a} = \sum_{m=1}^{\Lambda} \kappa_m^{xy;a} \sin(2m\theta_y^a) \left\{ 1 + \sum_{n=1}^{\Lambda} \left[\sum_{\substack{r \neq a \\ r=1}}^3 \lambda_{mn}^{xy;ar} \sin(2n\theta_y^r) + \sum_{s=1}^5 \eta_{mn}^{xy;as} \sin(n\phi_y^s + \chi_{mn}^{xy;as}) \right] \right\},$$

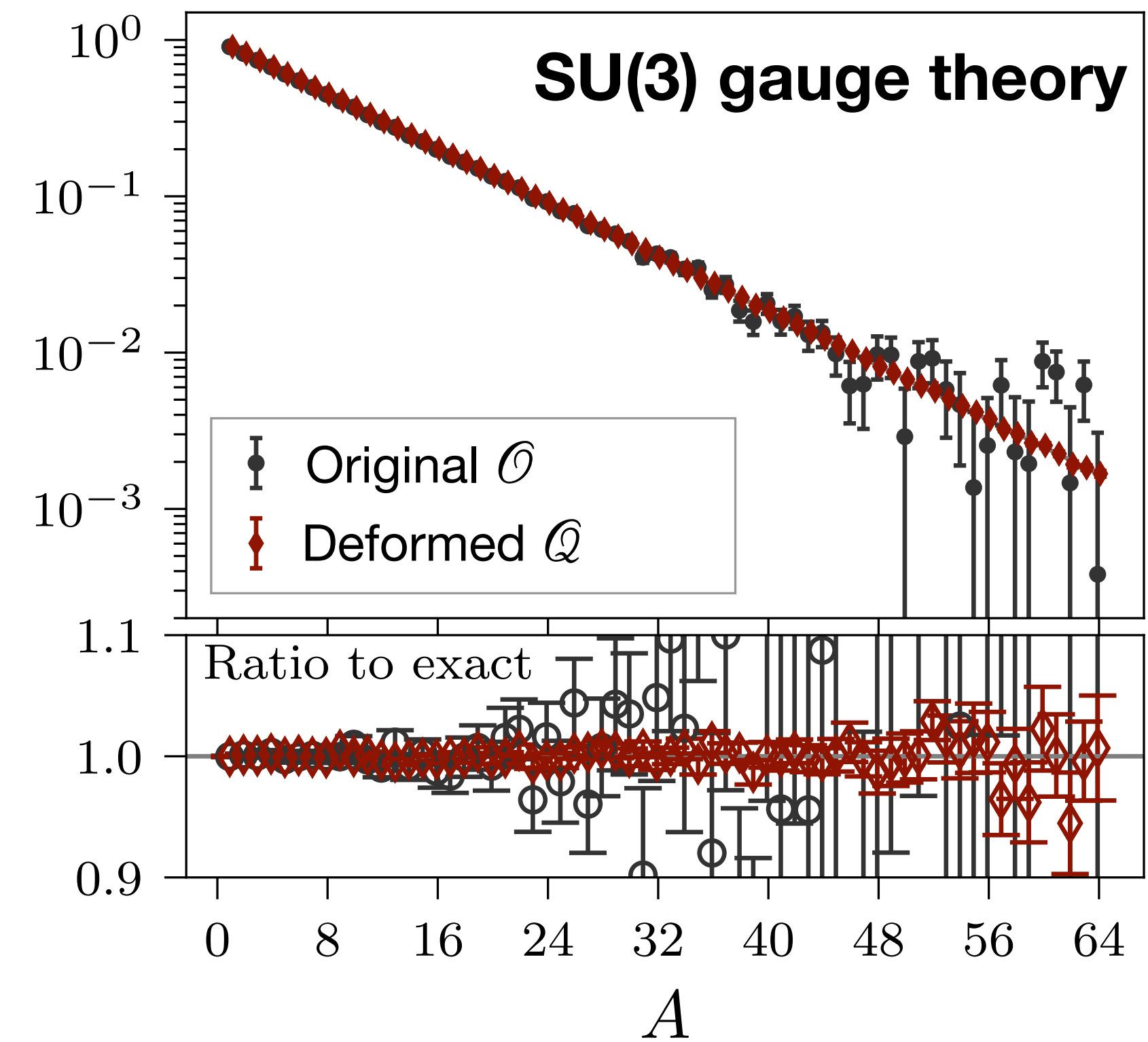
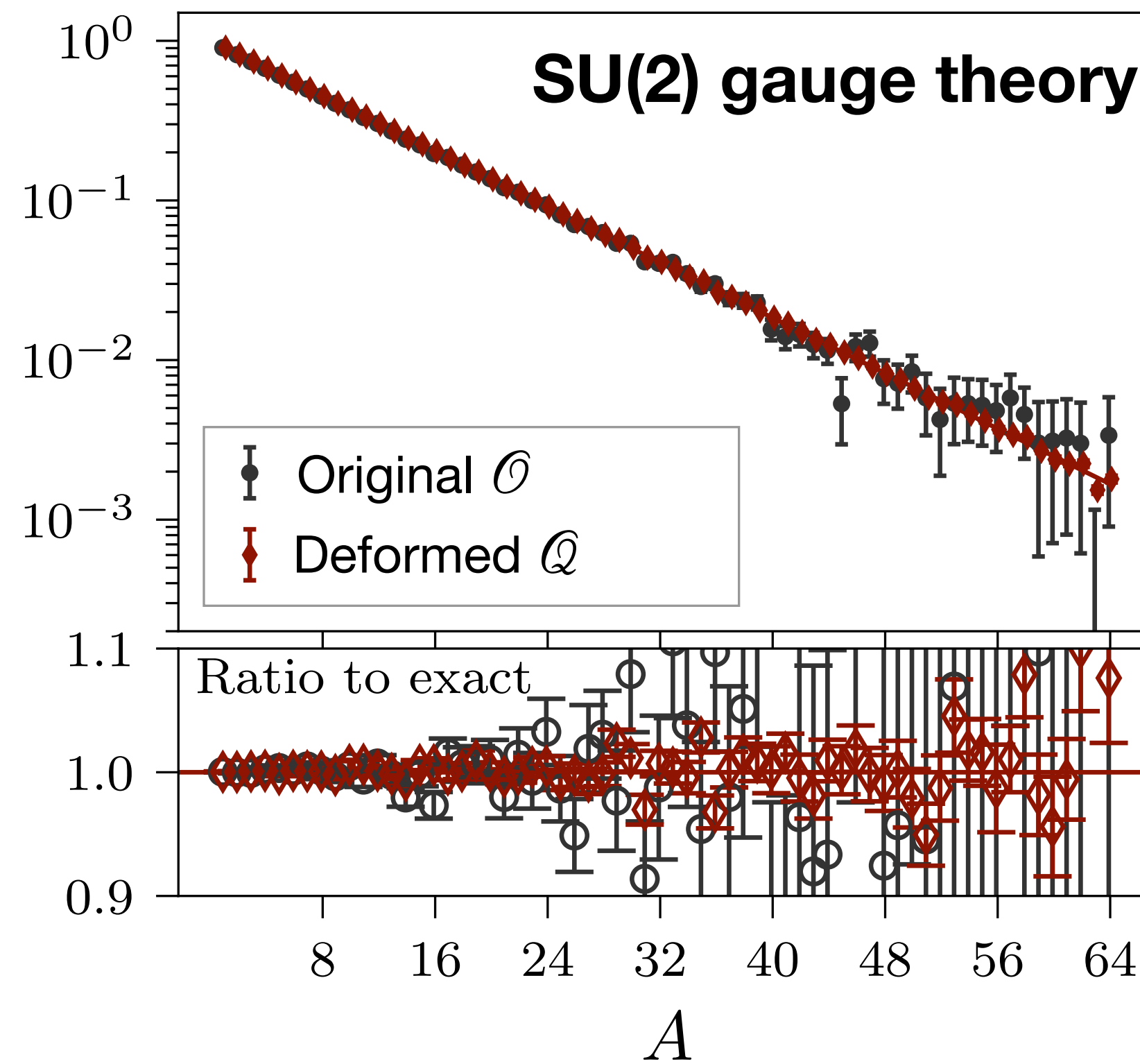
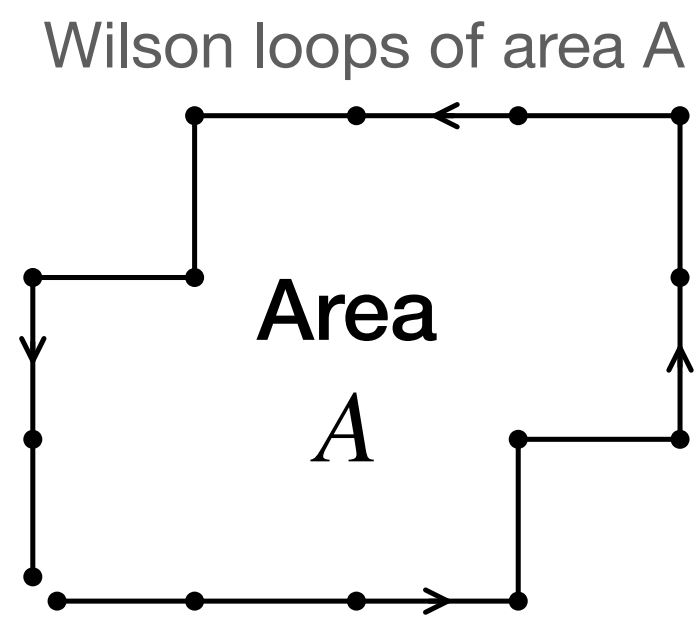
$$f_{\phi^a} = \sum_{m=1}^{\Lambda} \kappa_m^{xy;a} \sin(m\phi_y^a + \zeta_m^{xy;a}) \left\{ 1 + \sum_{n=1}^{\Lambda} \left[\sum_{r=1}^3 \lambda_{mn}^{xy;ar} \sin(2n\theta_y^r) + \sum_{\substack{s \neq a \\ s=1}}^5 \eta_{mn}^{xy;as} \sin(n\phi_y^s + \chi_{mn}^{xy;as}) \right] \right\}.$$

Triangular Jacobian: $f(\Omega)$ only allowed to depend on $y \leq x$. Jacobian determinant calculable in $O(V)$.

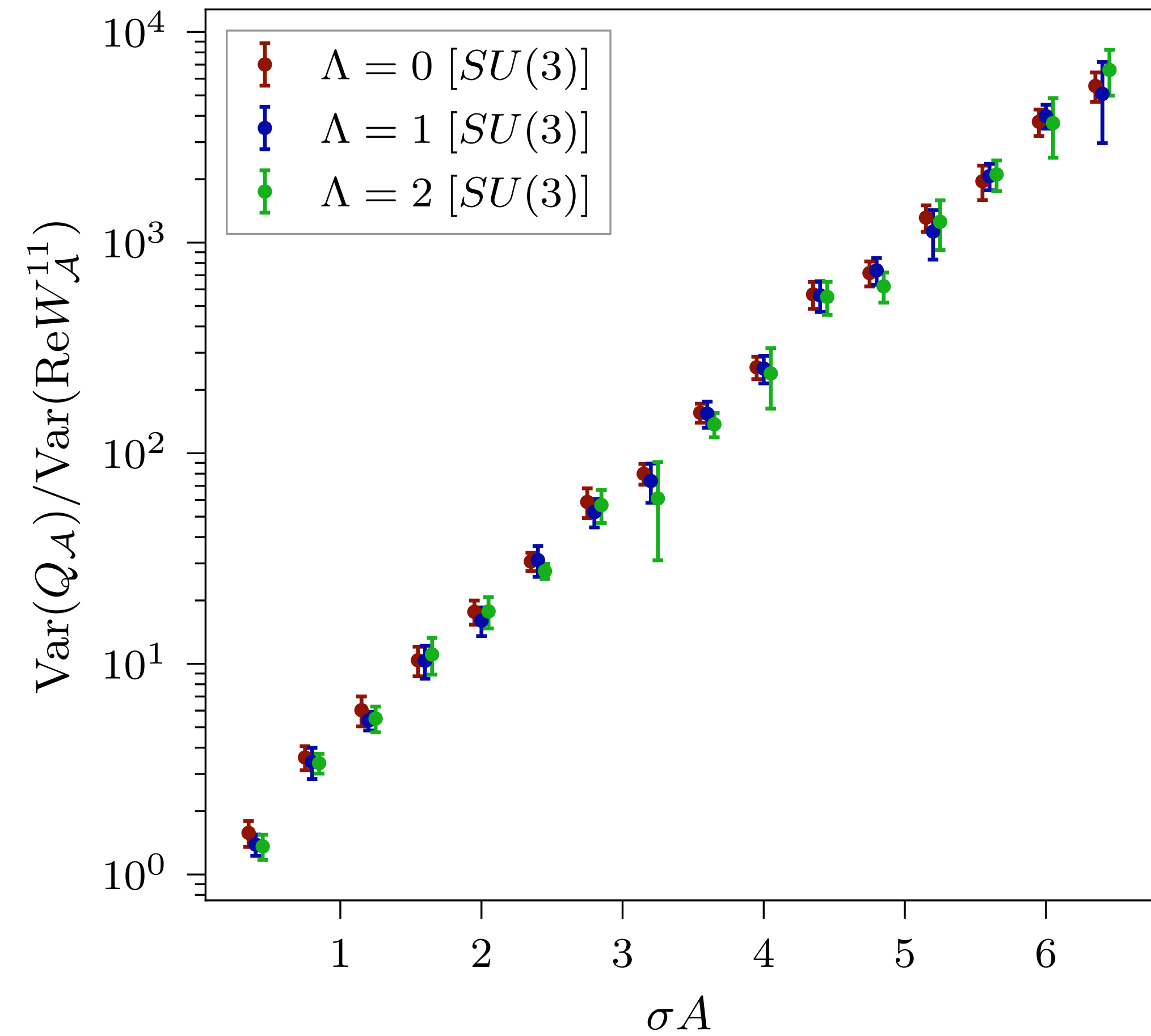
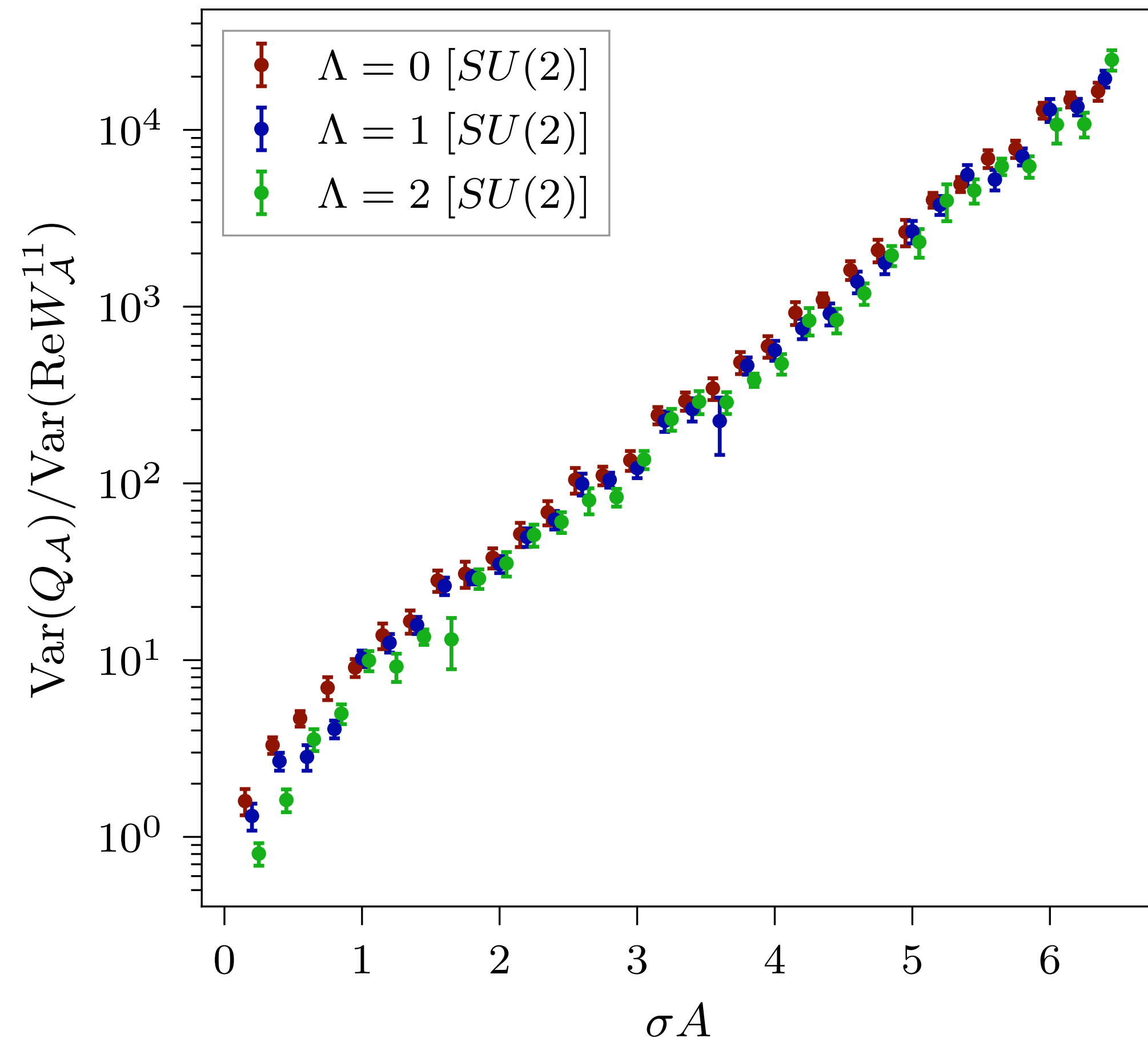
This is key for scalability!

Deformations crush 1+1D noise problems

Detmold, GK, Lamm, Wagman, Warrington PRD103 (2021) 094517

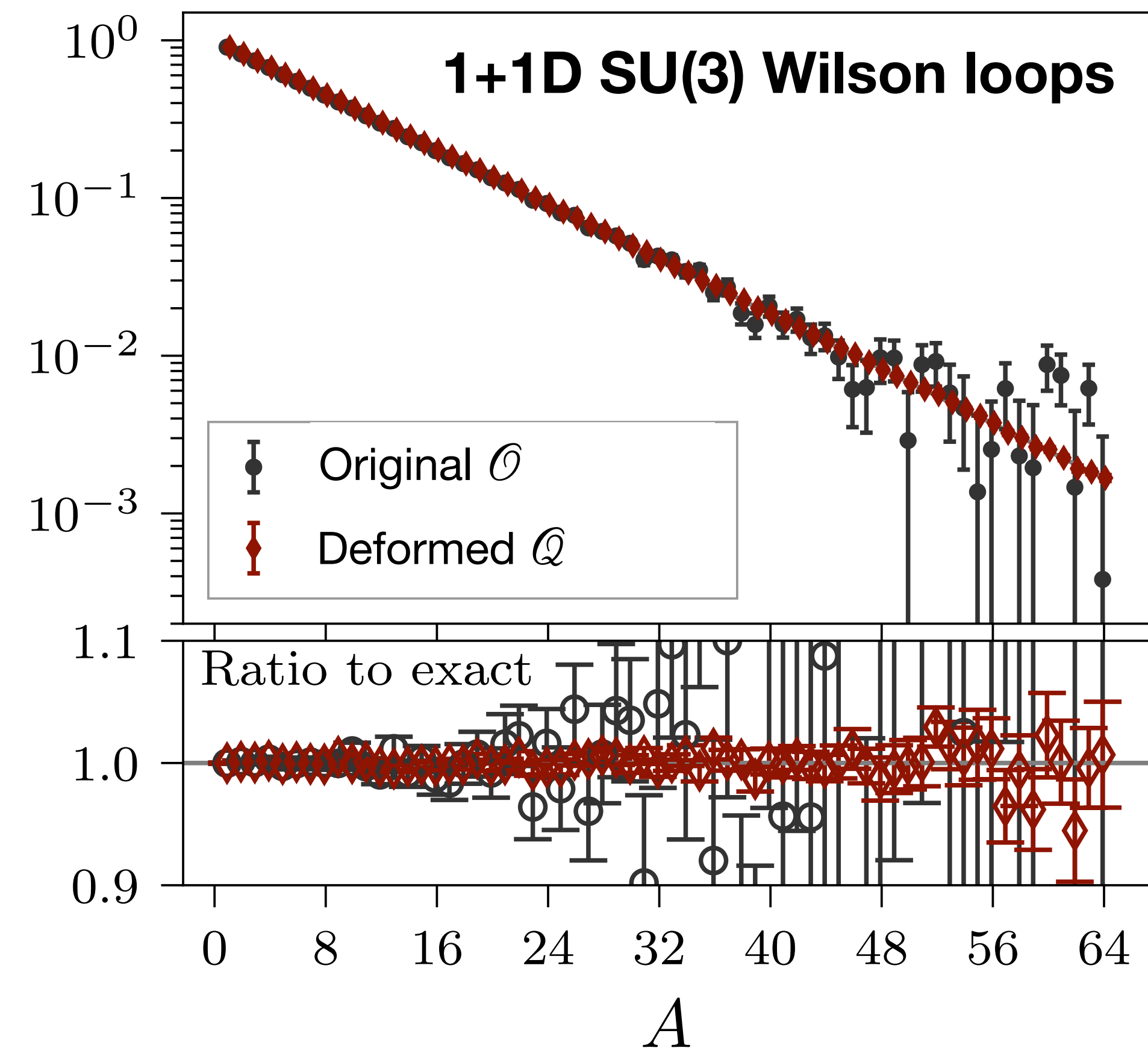


No scaling with Fourier cutoff



Restricted “constant shift” deformations

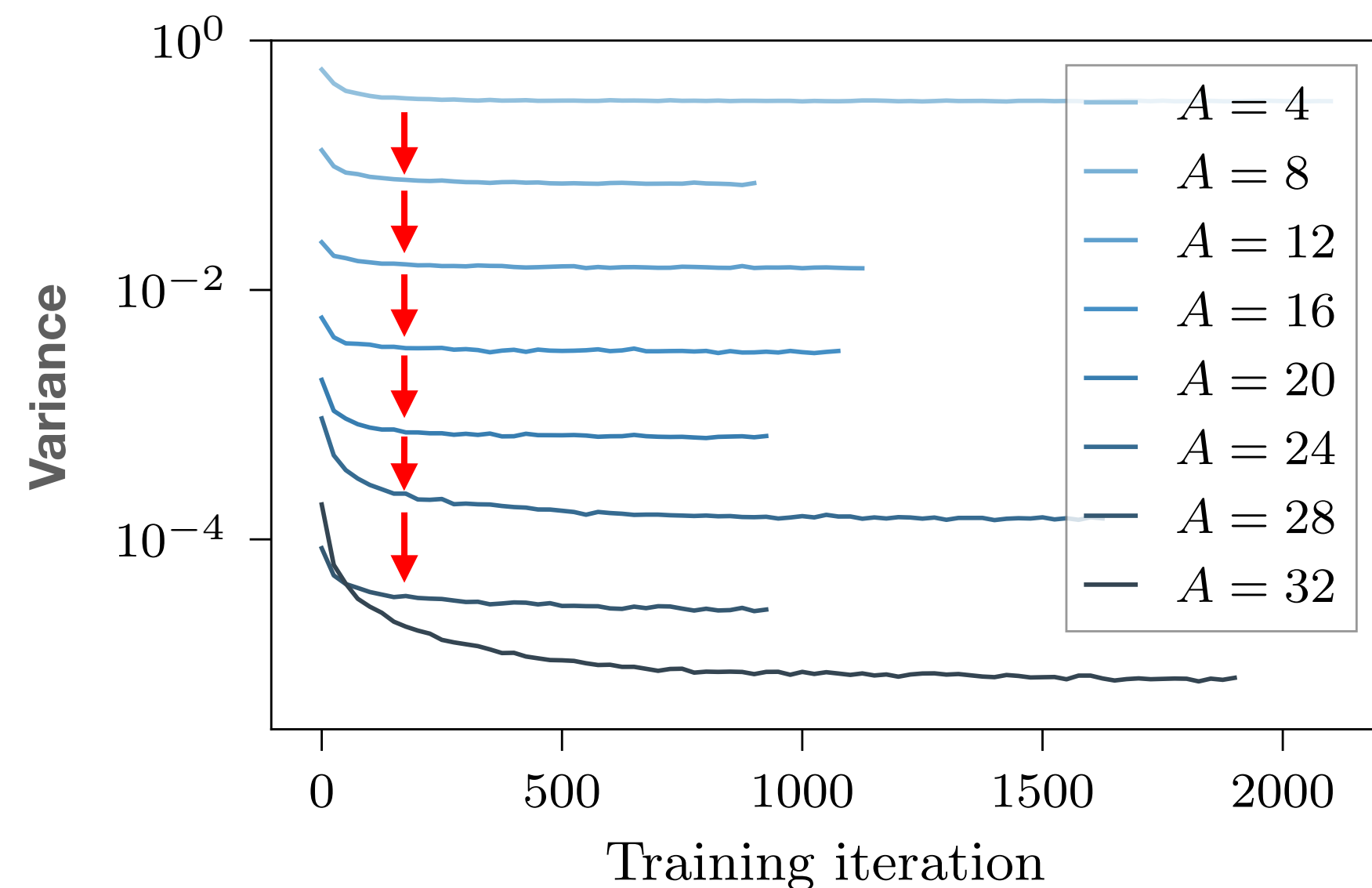
- Deform only periodic angular variables $\phi_i \in [0, 2\pi]$
- $\tilde{\phi}_i = \phi_i + i\lambda$
- Field-independent, but spacetime dependent $\implies O(V)$ learnable parameters



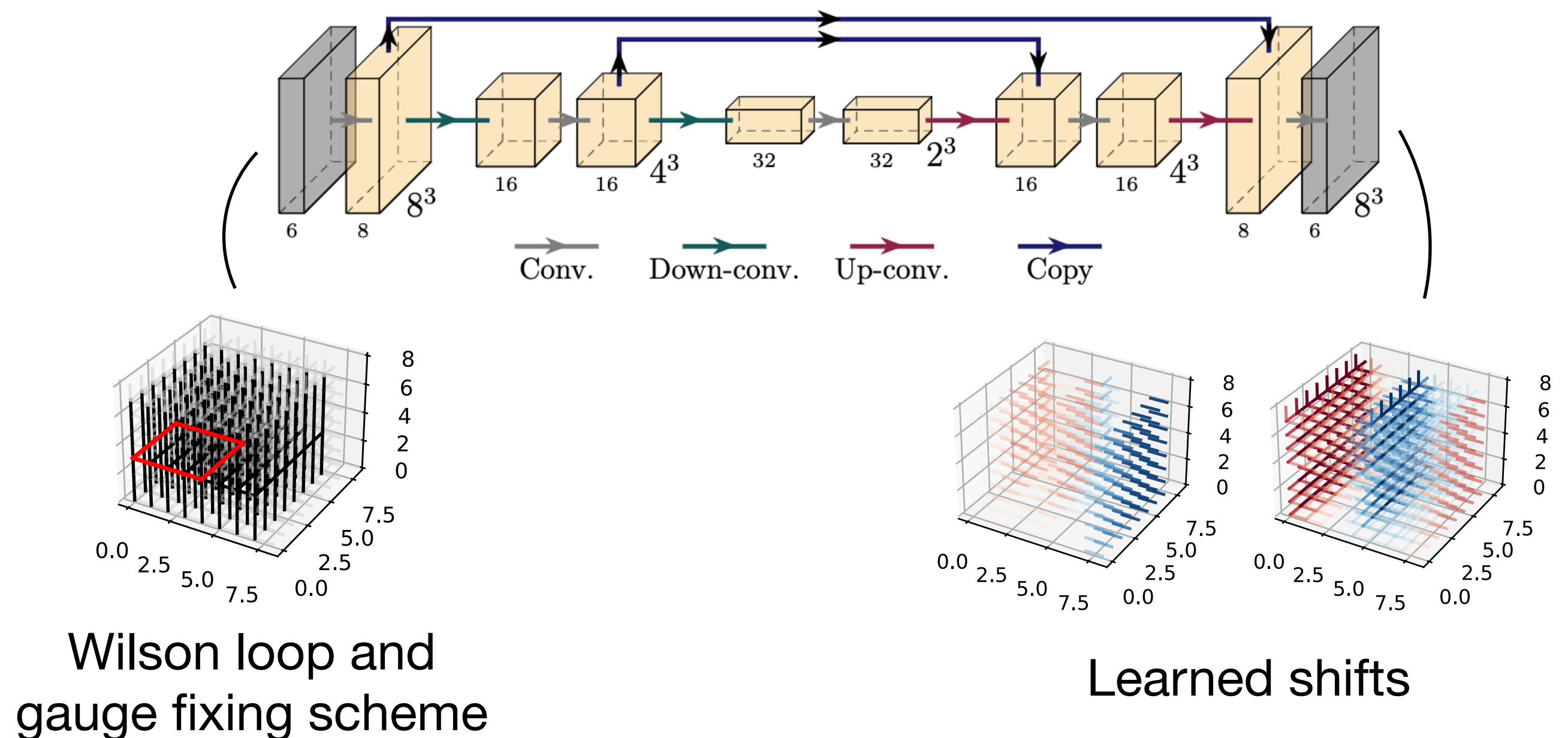
Direct vs. U-net parameterization

Direct parameterization with $O(V)$ learnable parameters leads to **overtraining** in typical problems!

1. Transfer learning

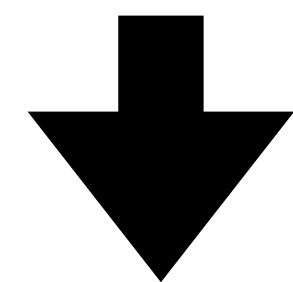
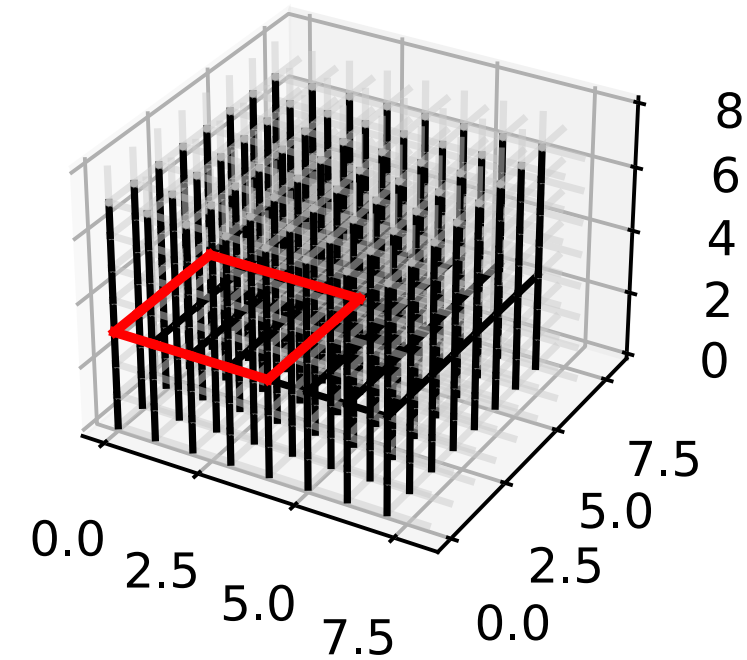


2. Indirect U-net parameterization



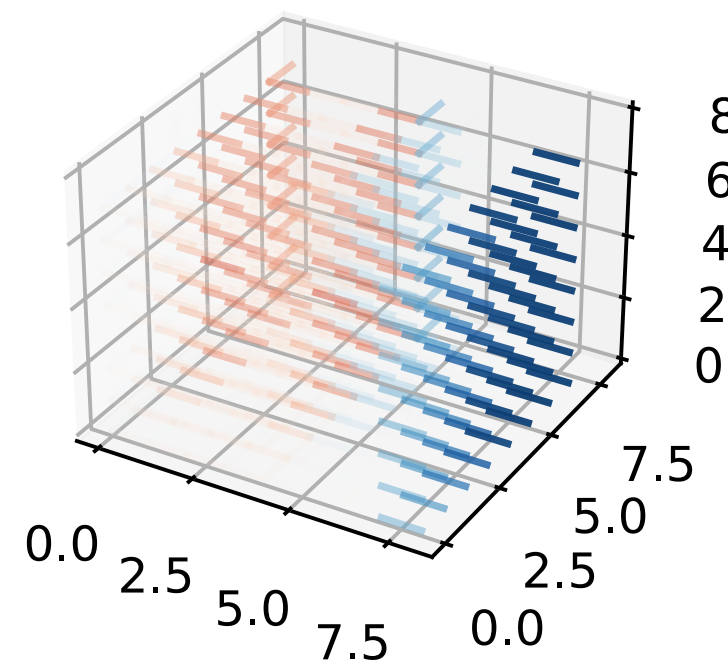
U-nets allow training for 2+1D problems

Wilson loop and gauge fixing scheme

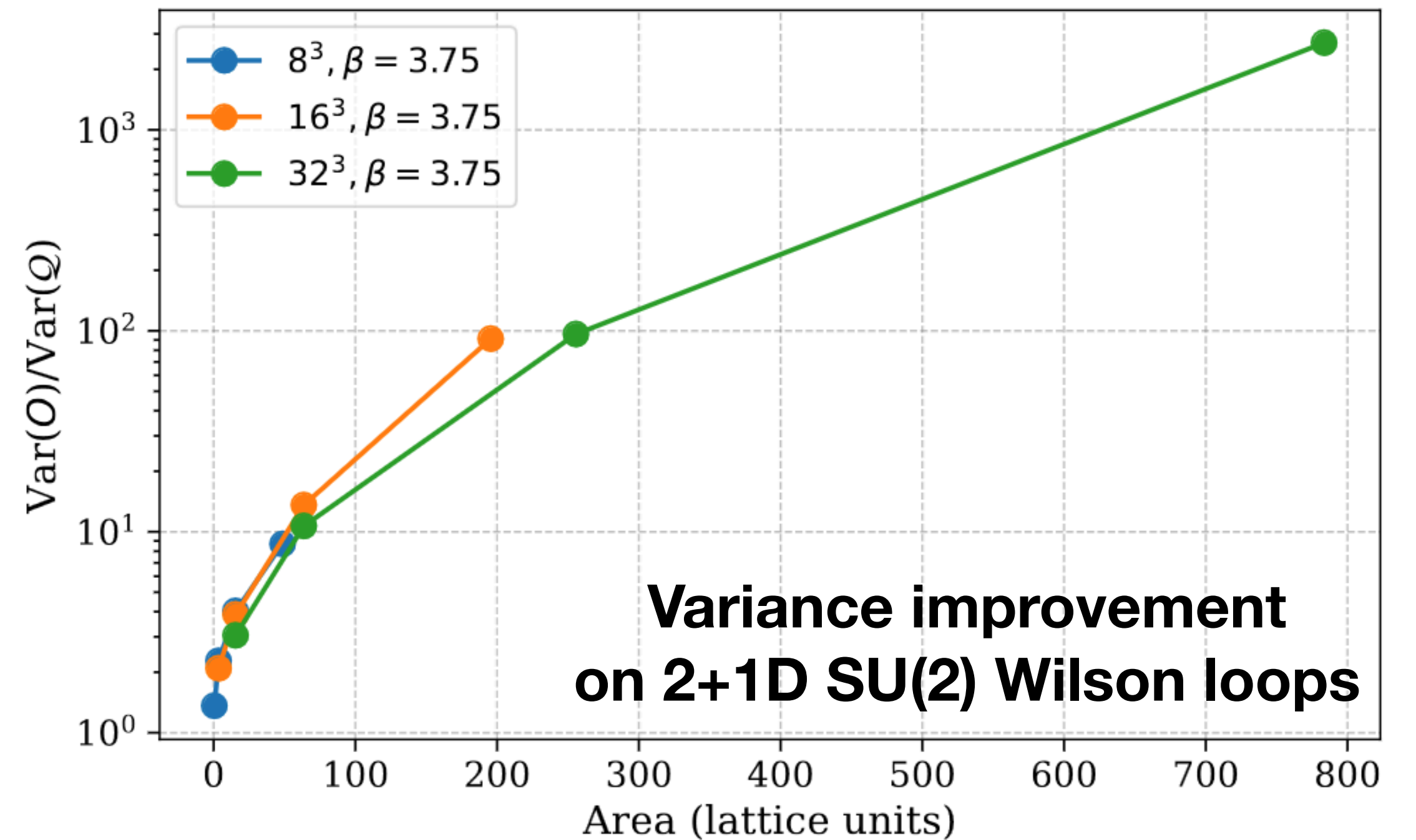
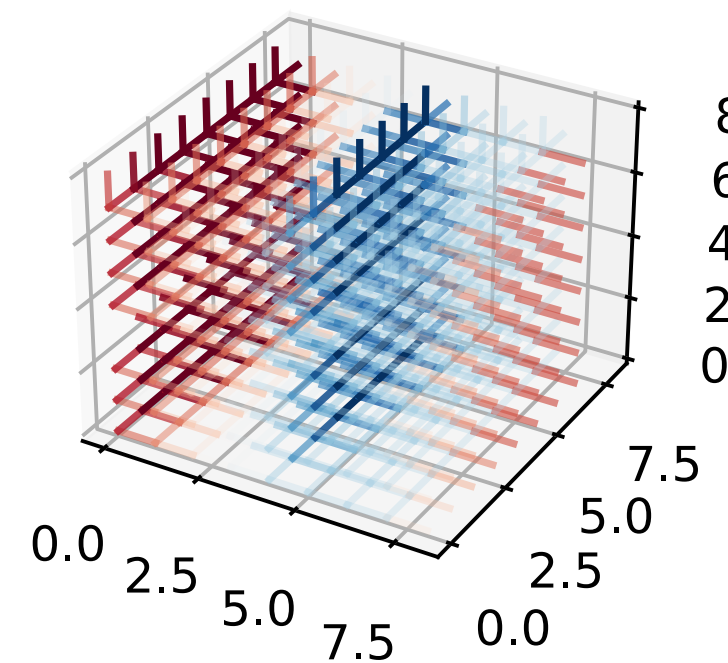


Learned shifts

Phi1 shift



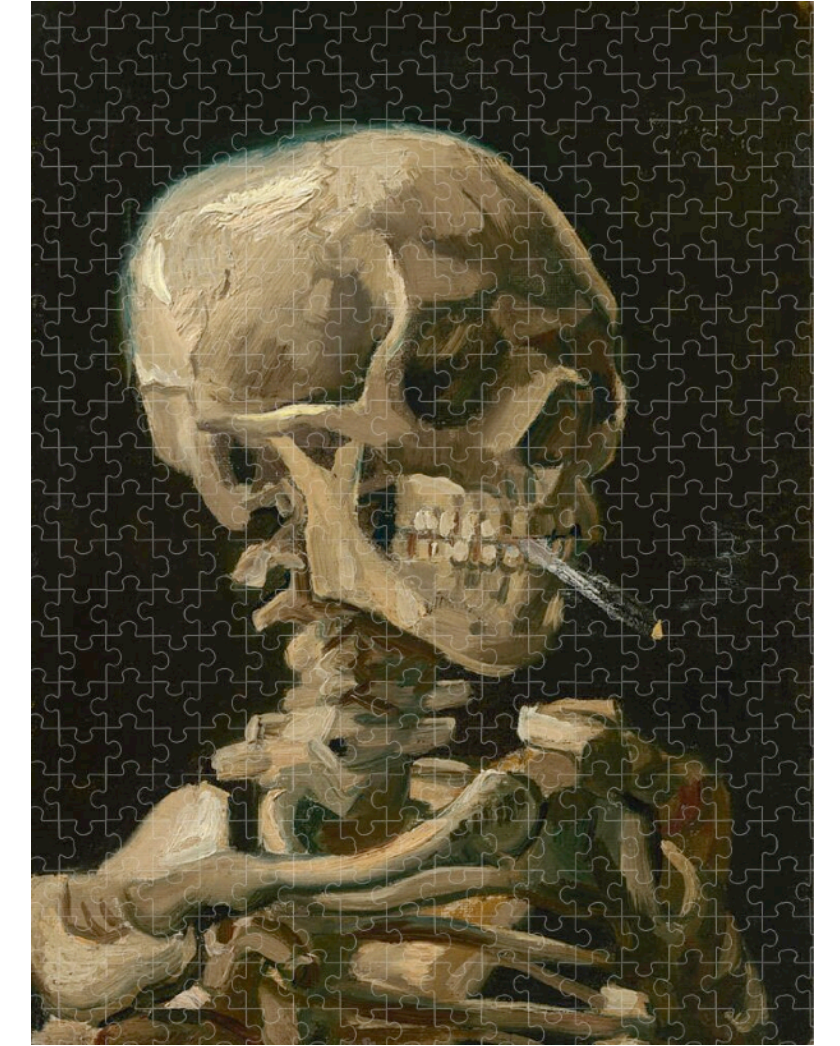
Phi2 shift



Gauge fixing skeletons in the closet

In 1+1D, could change basis to plaquettes plus gauge dofs:

$$\{U_\mu(x)\} \leftrightarrow \{U_p(x), \Omega(x)\}$$



This is **not uniquely possible** in 2+1D and higher!

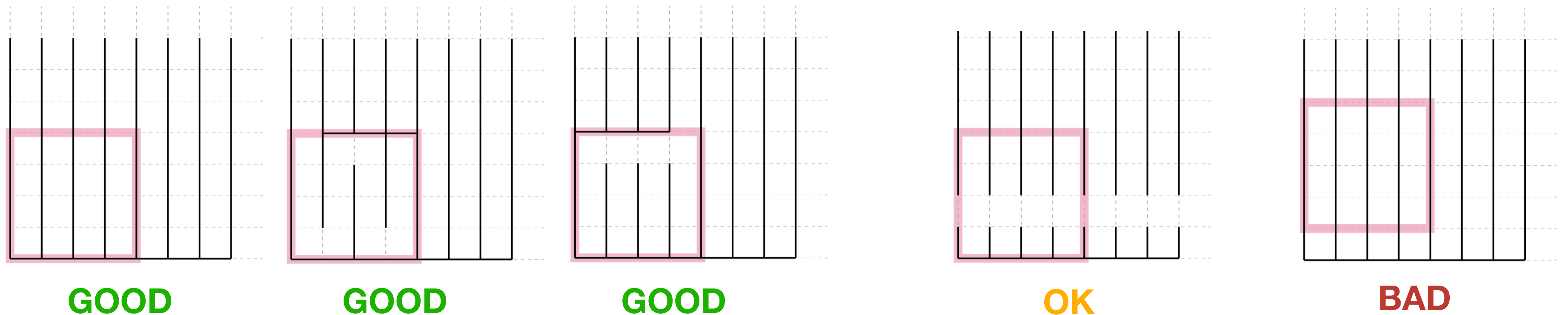
- Either work with subset of plaquettes with a complicated change-of-basis
- Or, simpler, perform maximal-tree gauge fixing and use remaining links as dofs

Have explored both options — No significant benefit to former, more flexibility provided by the latter. **Which maximal tree** is a hyperparameter to optimize.

Maximal tree gauges

Defined by a subset of lattice links containing **no closed loops** to be fixed to **\mathbf{I}**

Interplay between Wilson loop geometry and maximal tree choice

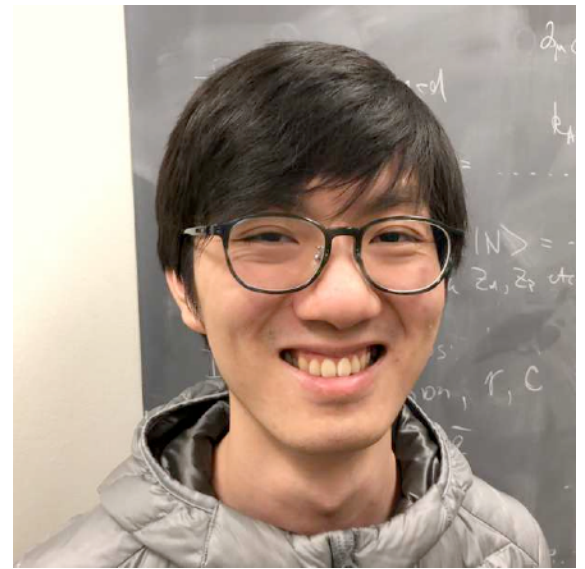


Exponential variance improvement
with Wilson loop area

Worse performance as
opening is moved lower

No variance
improvements

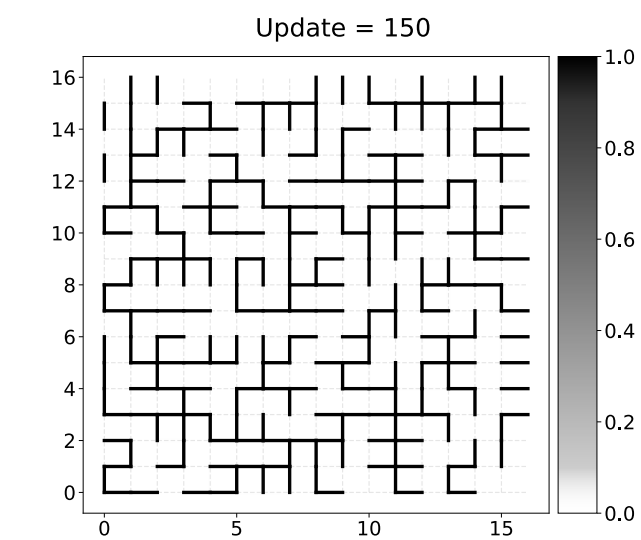
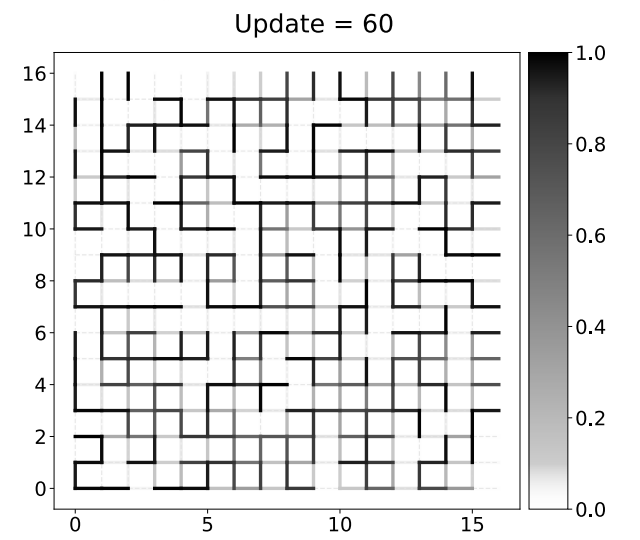
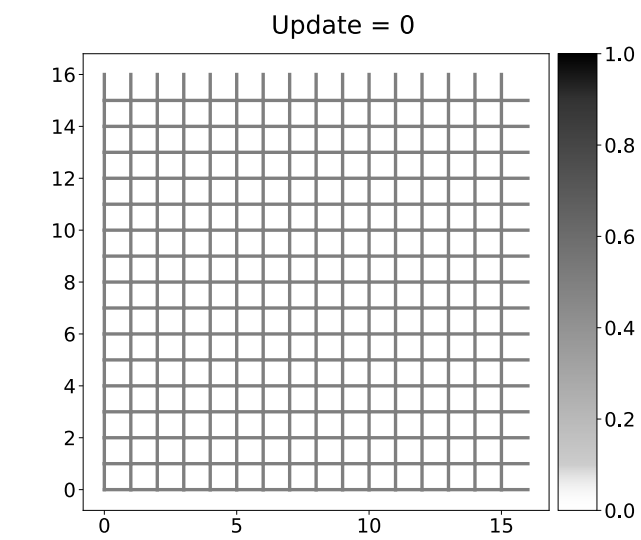
Could we “learn” the maximal tree?



Yin Lin

We define a parameterized gauge fixing functional

- Fix gauge by minimizing over gauge orbit
- Includes Coulomb, Landau, and all maximal tree gauges
- Can be optimized using adjoint state method



General gauge-fixing functional

$$E \propto - \sum_{x,\mu} \text{Tr} (p_\mu(x) U_\mu^g(x))$$

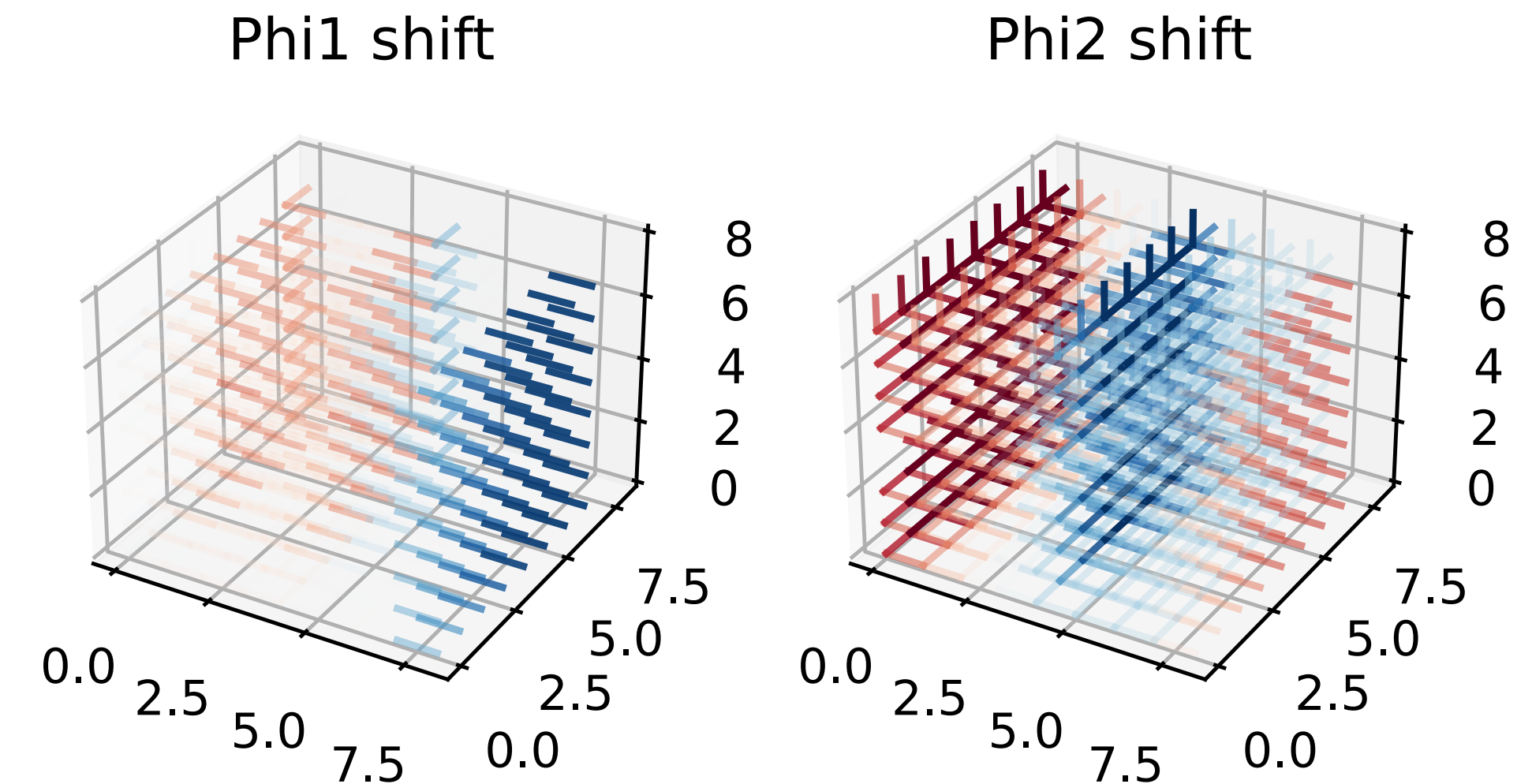
$$\left\{ \begin{array}{l} p_i(x) = 1 \text{ (Coulomb gauge)} \\ p_\mu(x) = 1 \text{ (Landau gauge)} \\ p_\mu(x) = k_\mu(x) \in \{0,1\} \text{ (Max. trees)} \end{array} \right.$$

See poster at Lattice '24!

Summary

Using complex analysis we can ...

- **Deform observables** $\mathcal{O} \rightarrow \mathcal{Q}$, where $\langle \mathcal{O} \rangle = \langle \mathcal{Q} \rangle$ but $\text{Var}[\mathcal{O}] \neq \text{Var}[\mathcal{Q}]$.
I.e., no systematic error!
- **Minimize variance numerically** (using existing MC samples).
- **Achieve far more precise** measurements in proof-of-principle applications to lattice field theories.



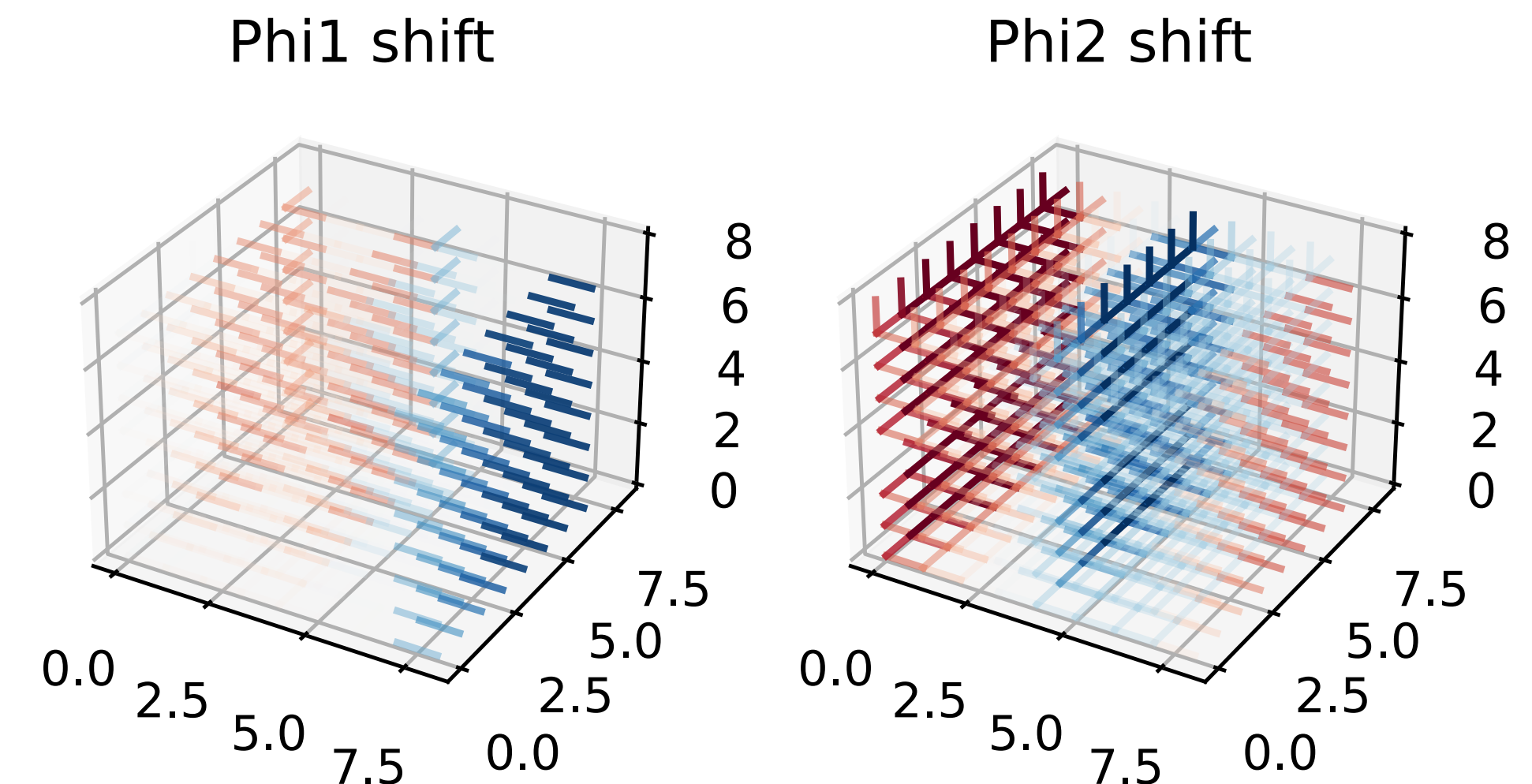
Look out for a paper on 3d and 4d SU(N) results soon! Come chat at the poster session!

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Thanks!
Questions?



Look out for a paper on 3d and 4d SU(N) results soon! Come chat at the poster session!

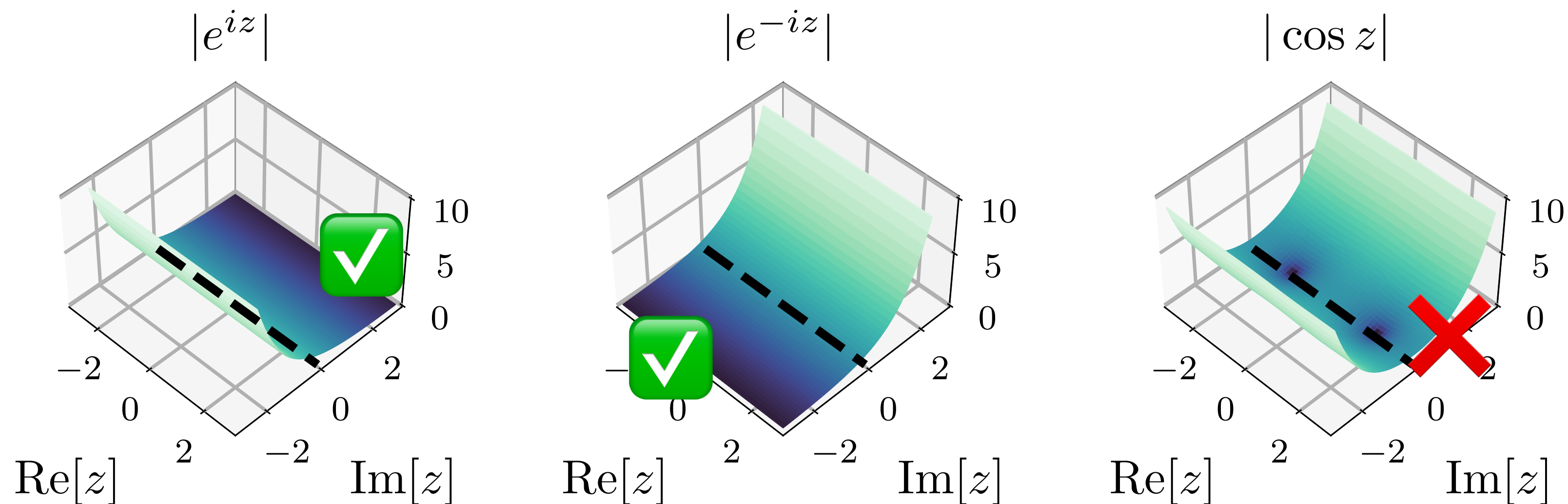
Backup slides

Why select $W_{\mathcal{A}}^{11}$?

Any observable with equivalent expectation value can be taken as the **base observable** for deformation ...

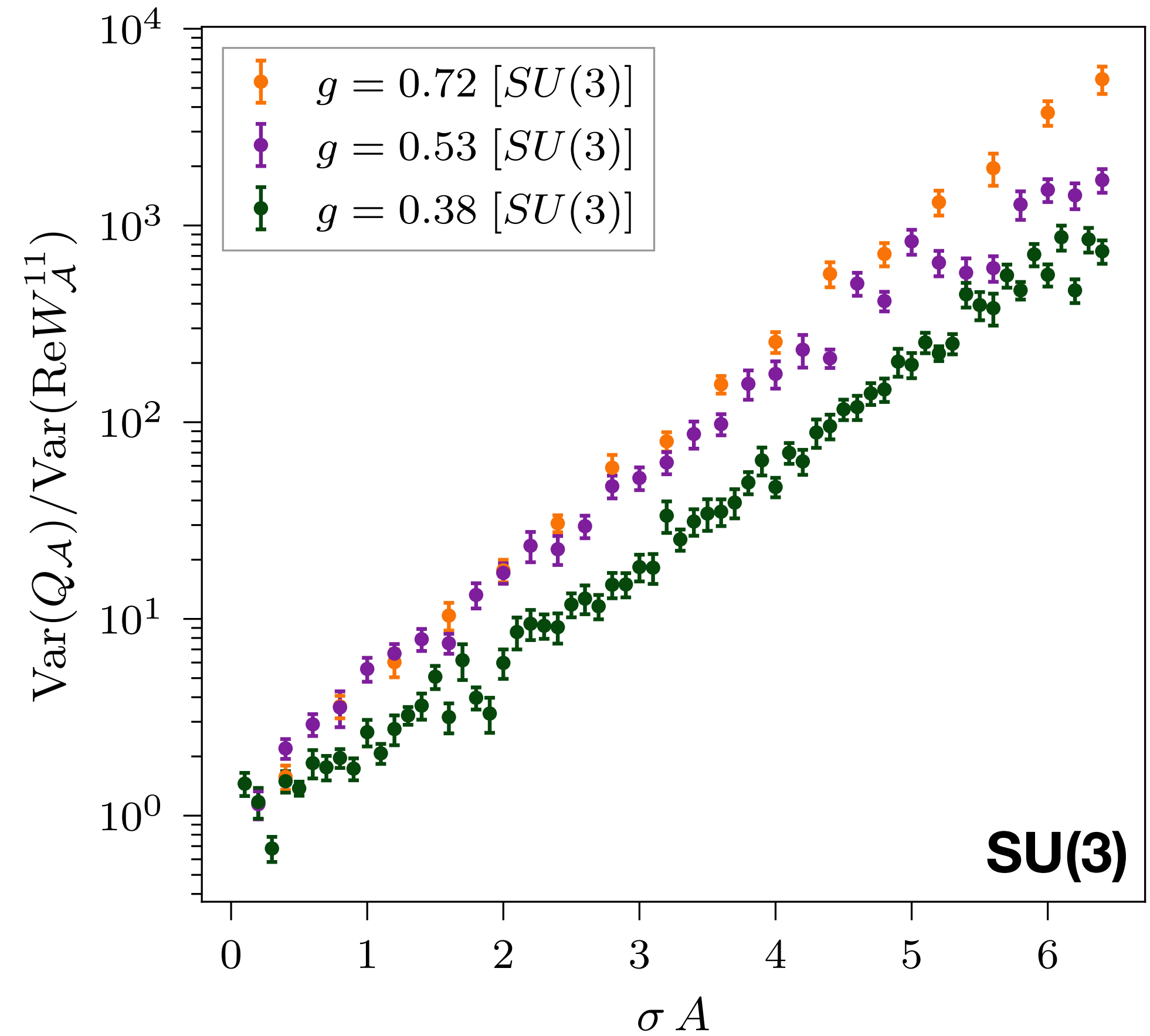
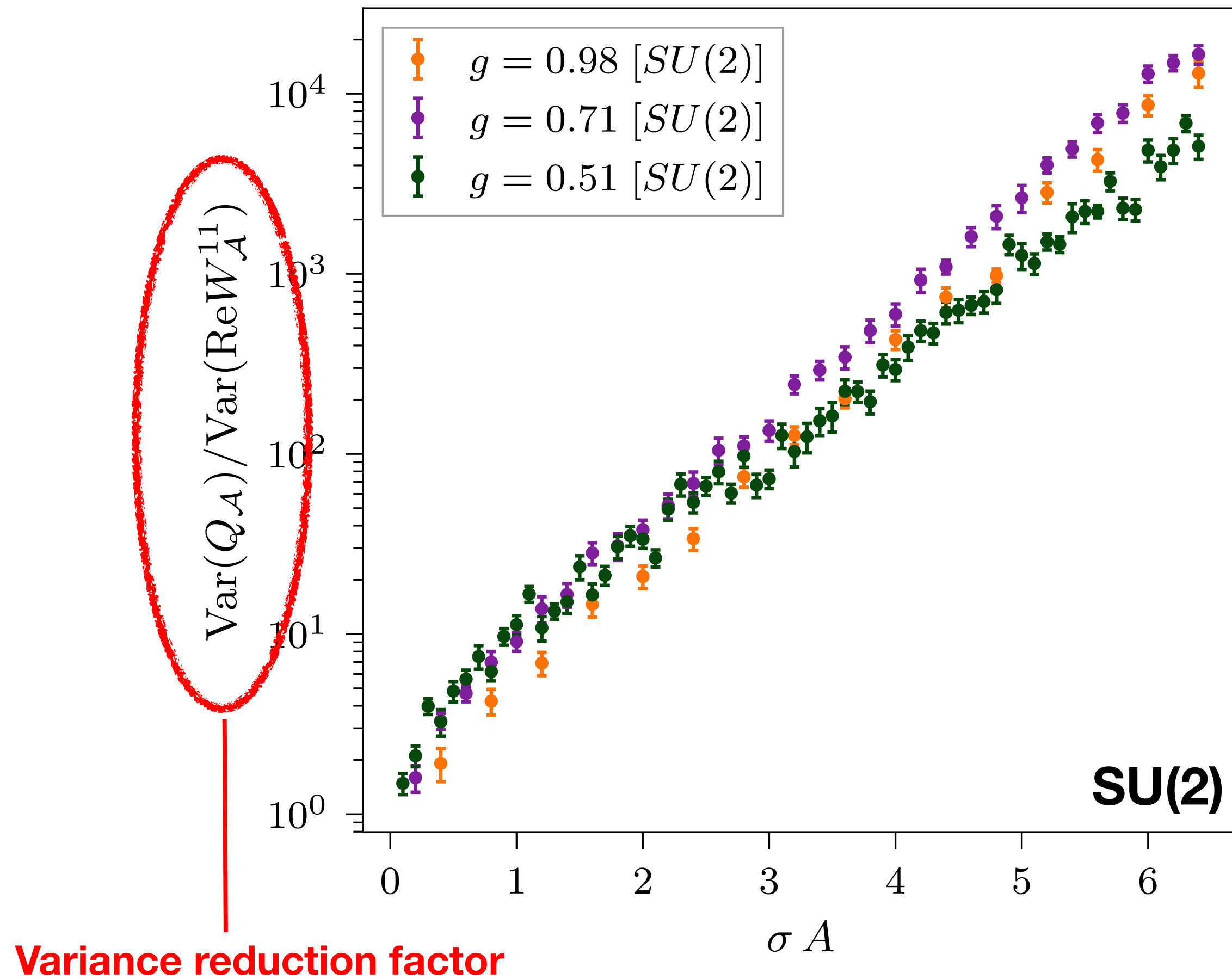
1D example:
$$\int_0^{2\pi} \frac{dz}{2\pi} e^{iz} e^{\beta \cos z} = \int_0^{2\pi} \frac{dz}{2\pi} e^{-iz} e^{\beta \cos z} = \int_0^{2\pi} \frac{dz}{2\pi} \cos z e^{\beta \cos z} = I_1(\beta)$$

... however, some choices are better than others!



SU(N) lattice spacing effects

Similar variance reduction effects across all 3 lattice spacings:



Complex scalar theory

Use phase-magnitude decomposition for variables $\phi_t = R_t e^{i\theta_t}$

Holomorphic:
$$S = -2 \sum_{t=0}^{L-1} R_t R_{t+1} \cos(\theta_{t+1} - \theta_t) + V(R)$$

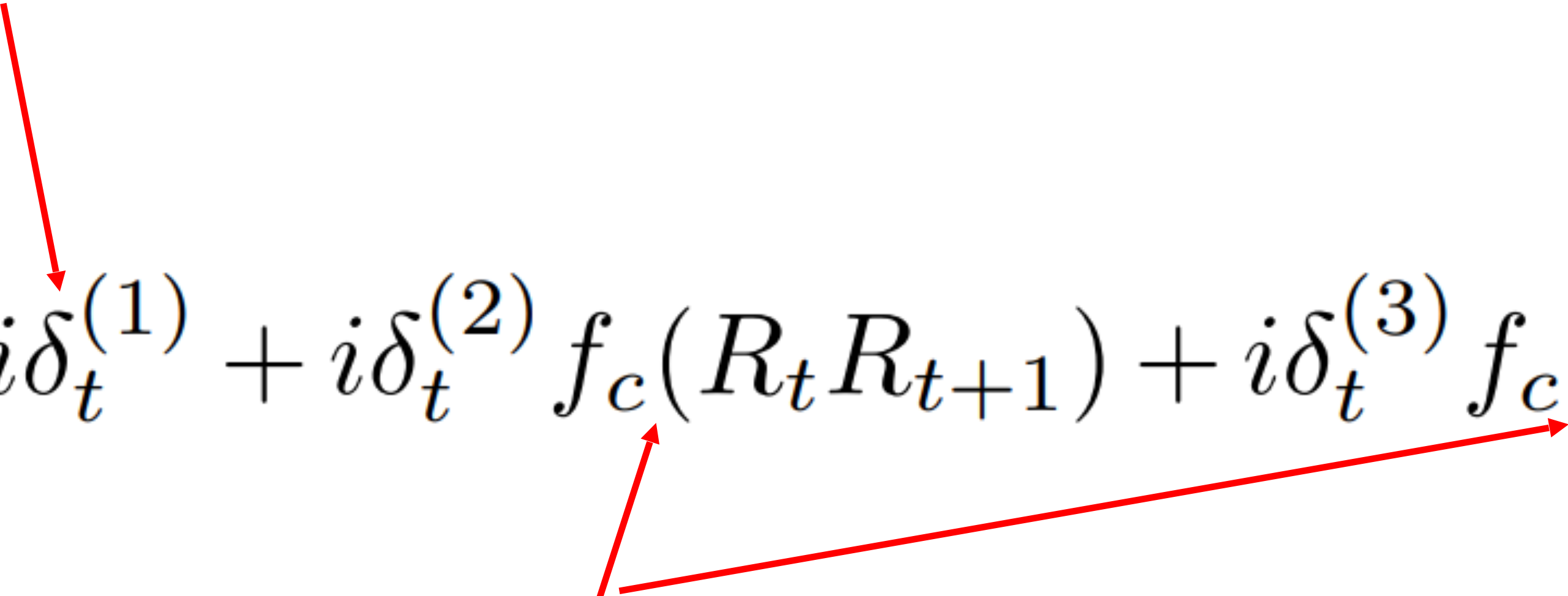
$$V(R) = \sum_t (2 + m^2) R_t^2 + \lambda R_t^4$$

Interested in correlation functions

$$G_t = \langle R_t R_0 e^{i\theta_t - i\theta_0} \rangle \equiv \langle C_t(R, \theta) \rangle$$

Deformation for scalar theory

Intuition: phase differences appear in action similarly to phases of Schwinger, use shifts into imaginary direction

$$\tilde{\theta}_t = \theta_t + i\delta_t^{(1)} + i\delta_t^{(2)} f_c(R_t R_{t+1}) + i\delta_t^{(3)} f_c(R_{t-1} R_t)$$


Extra terms inspired by small phase fluctuation expansion.

Results: 0+1D ϕ^4 correlators

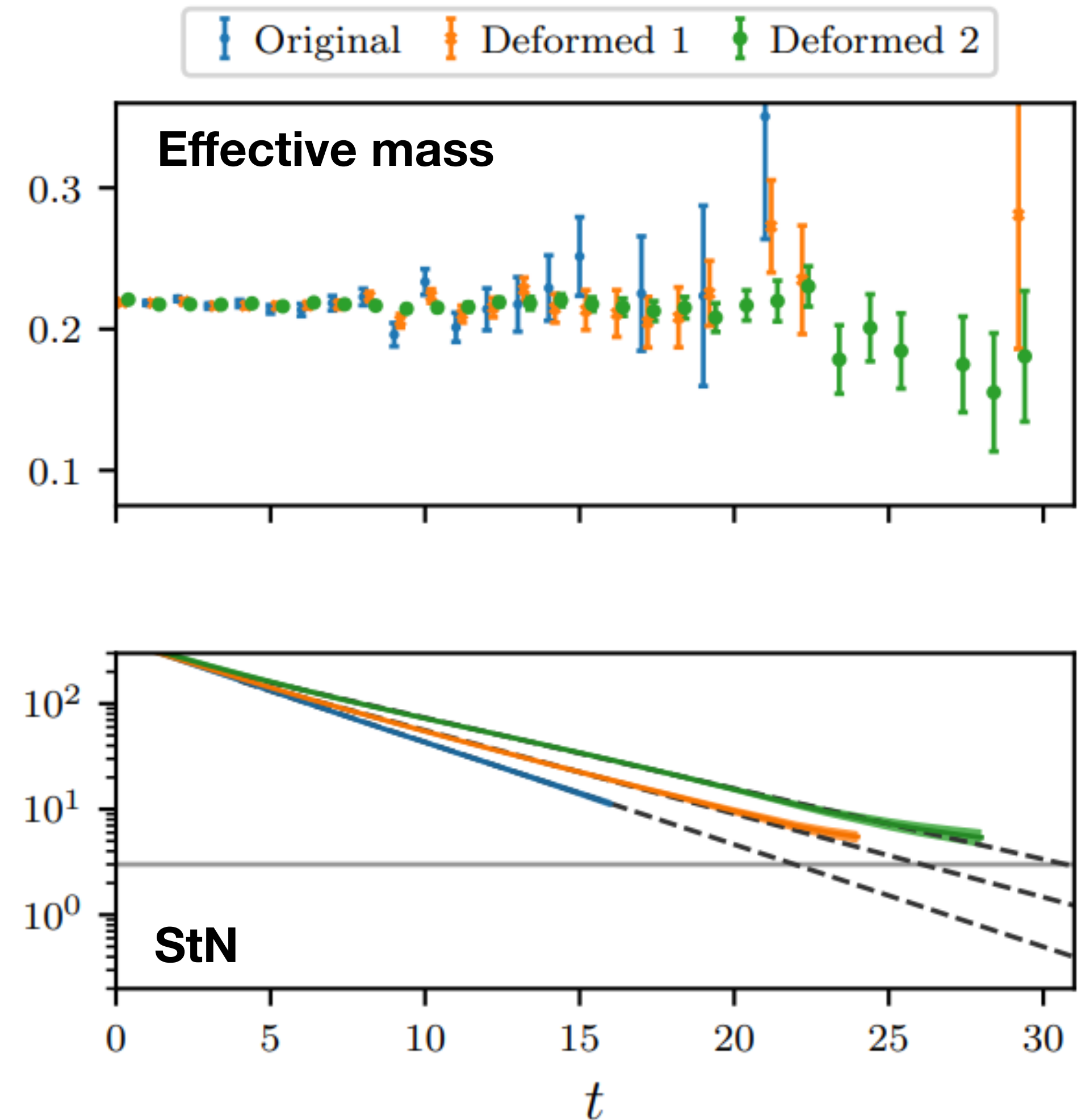
Using polar dofs in the path integral

$$\phi(t) = R(t)e^{i\theta(t)}$$

and the holomorphic action

$$S[R, \theta] = -2 \sum_t R(t) R(t+1) \cos[\theta(t+1) - \theta(t)] \\ + \sum_t V(R(t))$$

$$V(R) \equiv (2 + m^2) R^2 + \lambda R^4$$



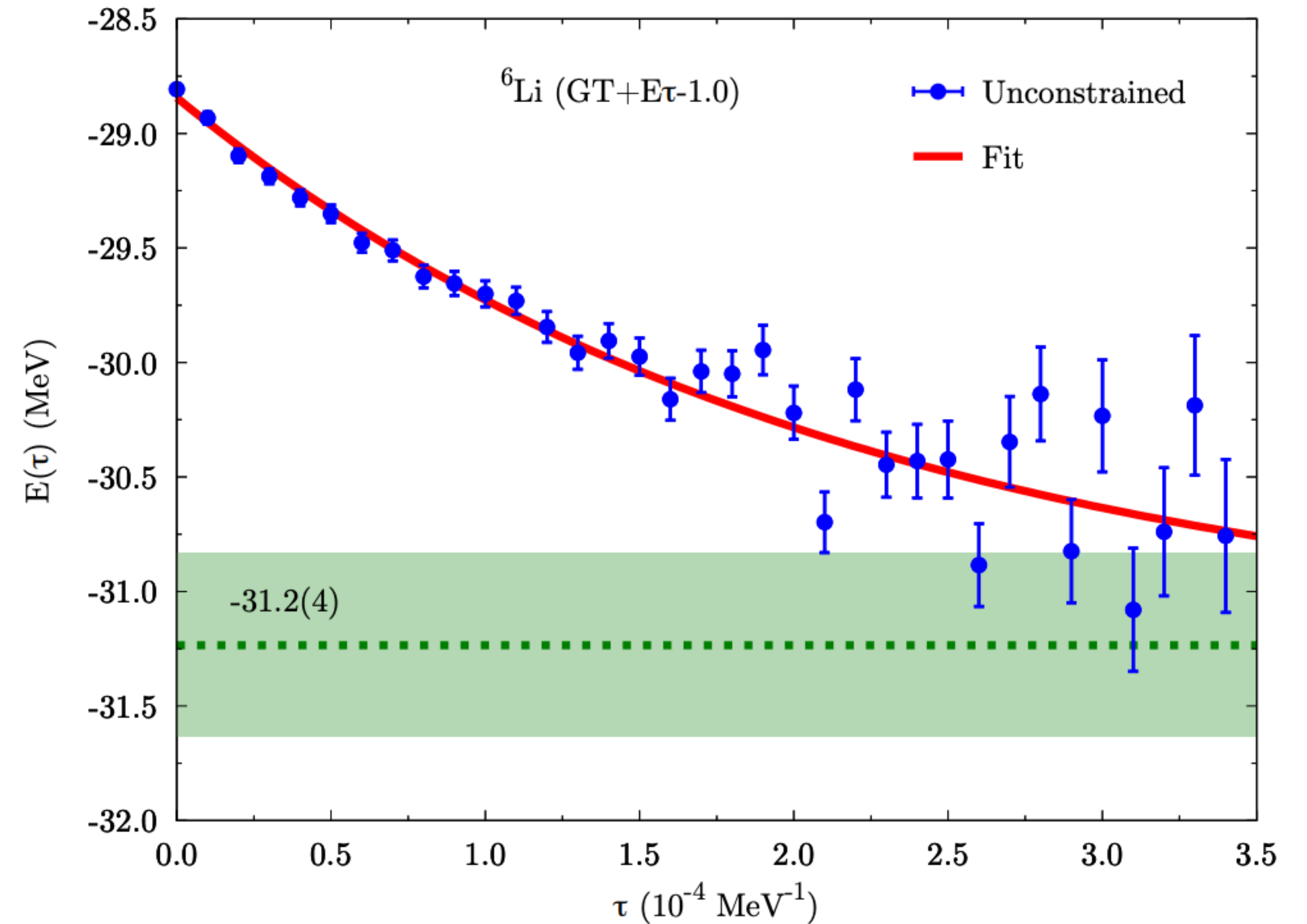
Non-lattice applications: GFMC/AFDMC

[Gandolfi, et al. 2001.01374]

Large t : StN decays exponentially

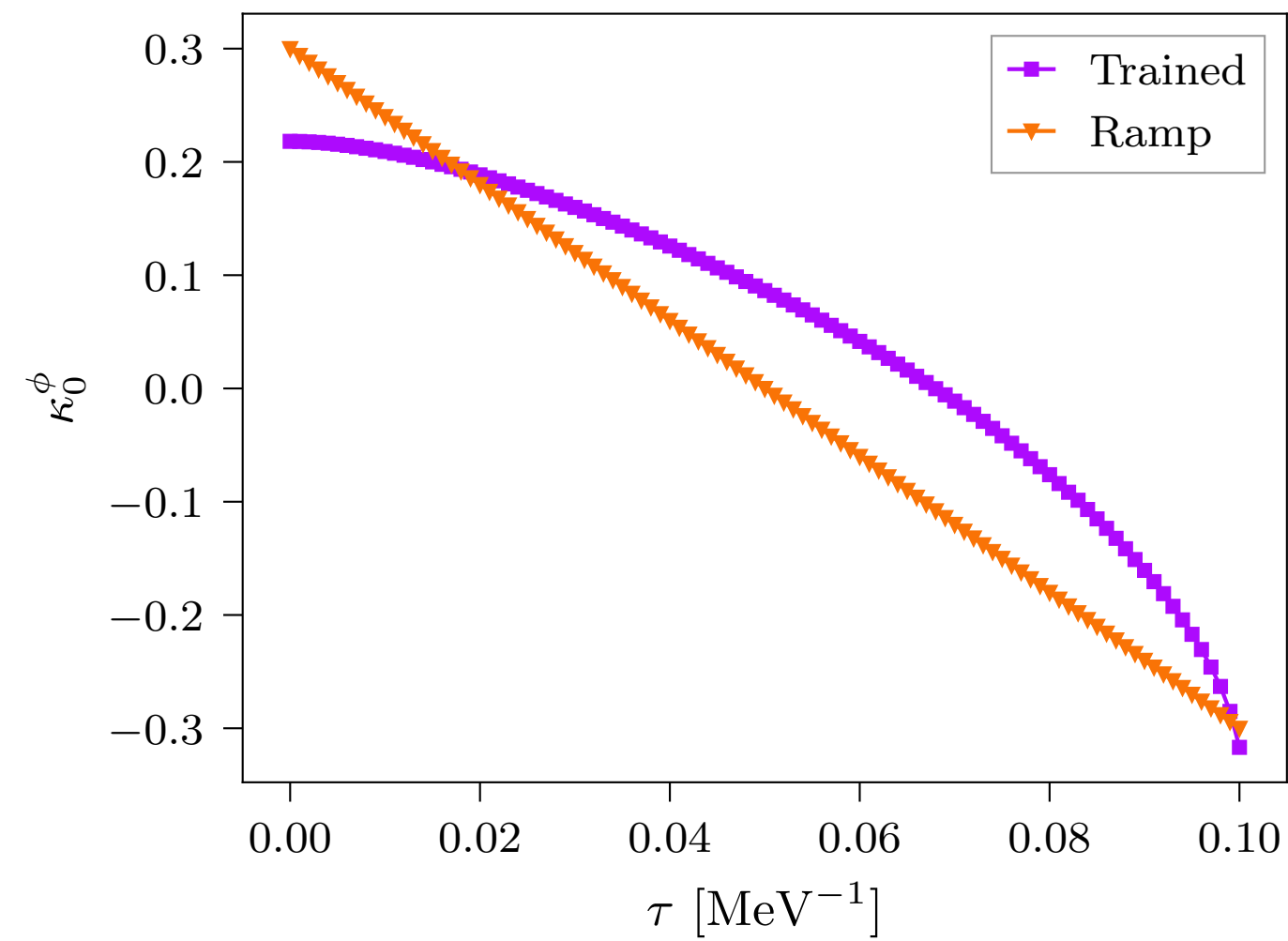
Small t : Excited state effects

To extract physical information,
fit excited state model

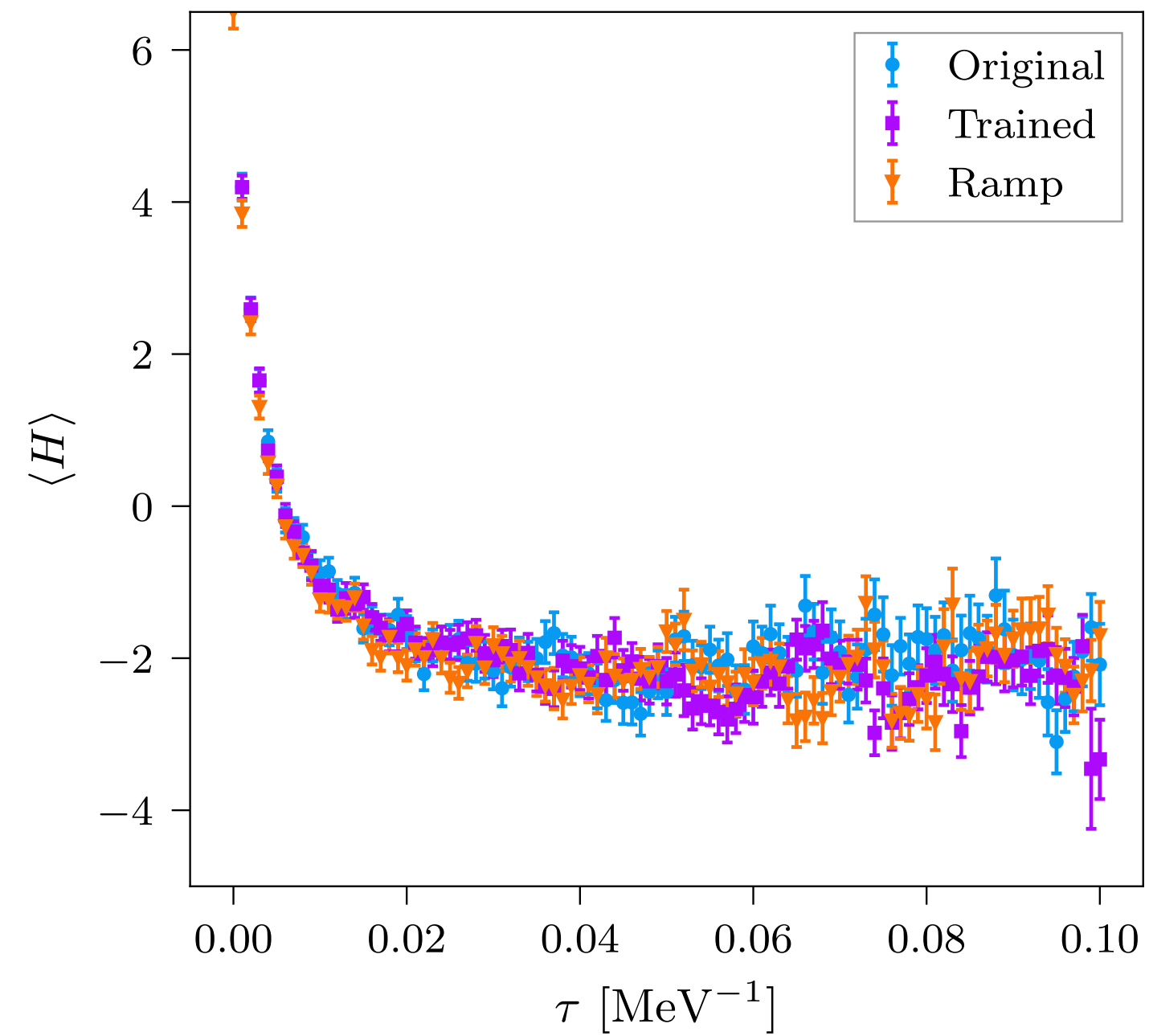


GFMC results

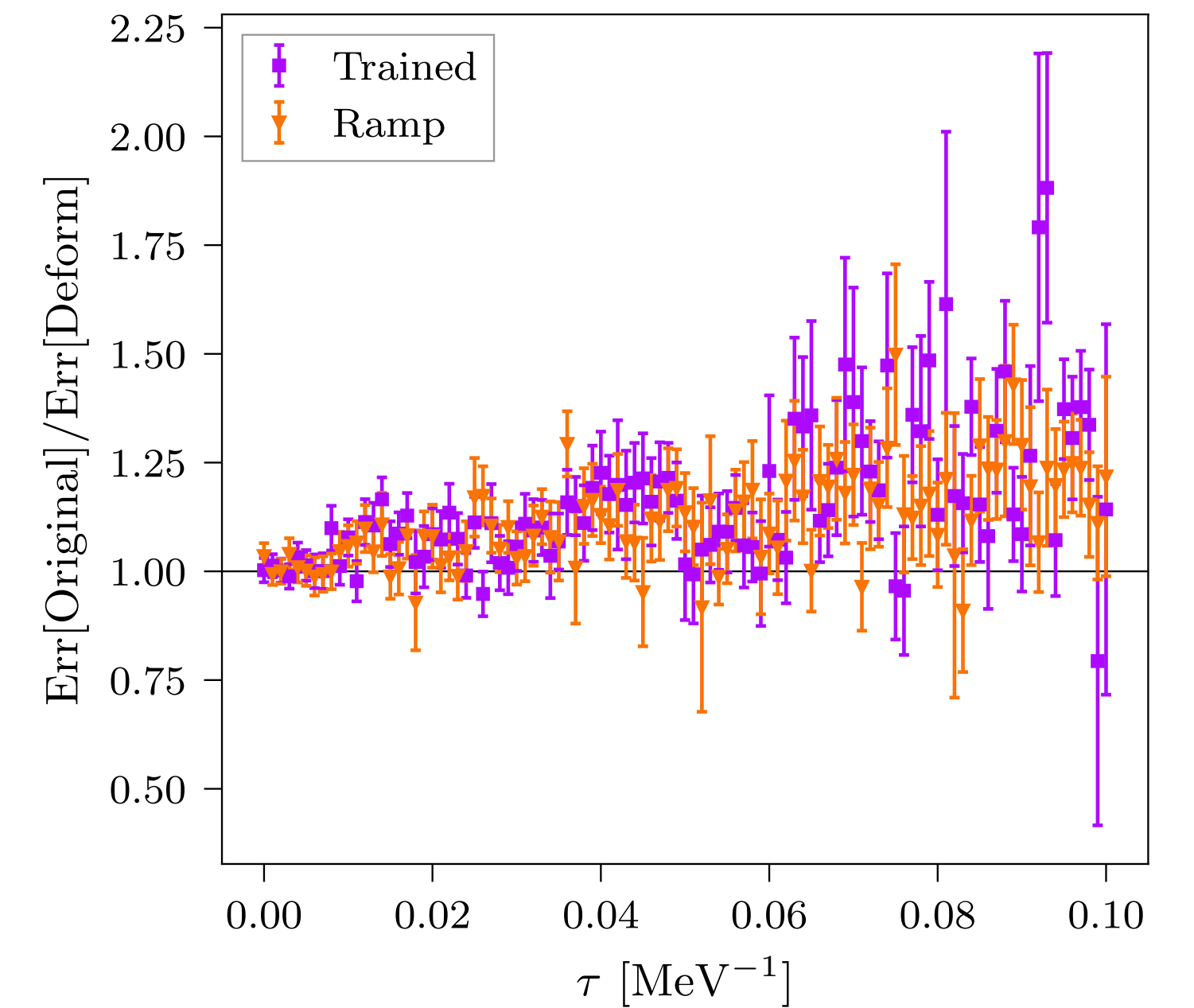
Learned imaginary shift
vs Euclidean time



Measured deuteron
binding energy



Improvement ratio

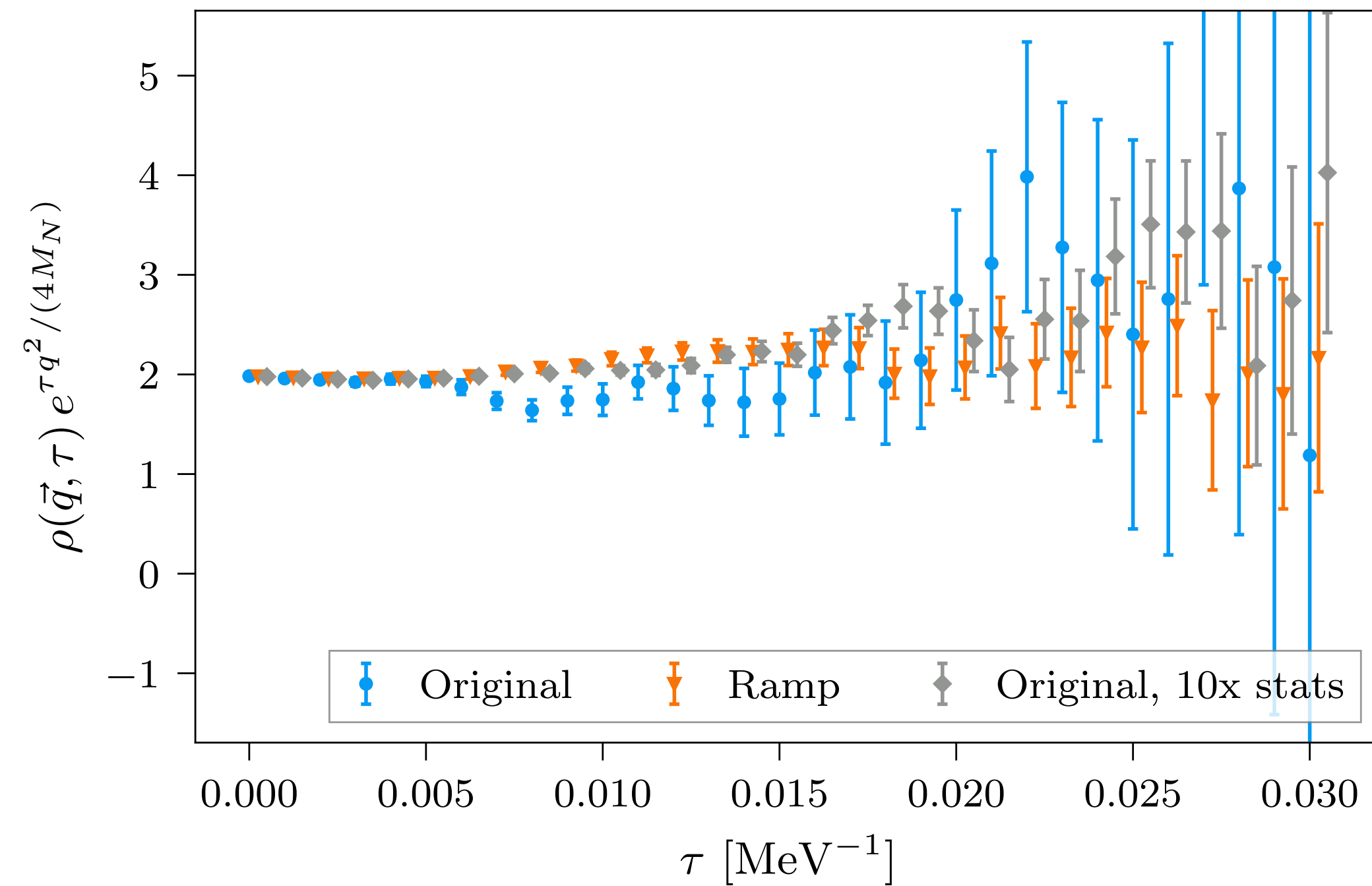


No spectacular results for $\langle H \rangle$, but ...

GFMC results

... deuteron Euclidean density response $\rho(\vec{q})$ significantly improved.

$$\vec{q} = (0, 0, 800) \text{ MeV}$$



$$\vec{q} = (600, 600, 600) \text{ MeV}$$

