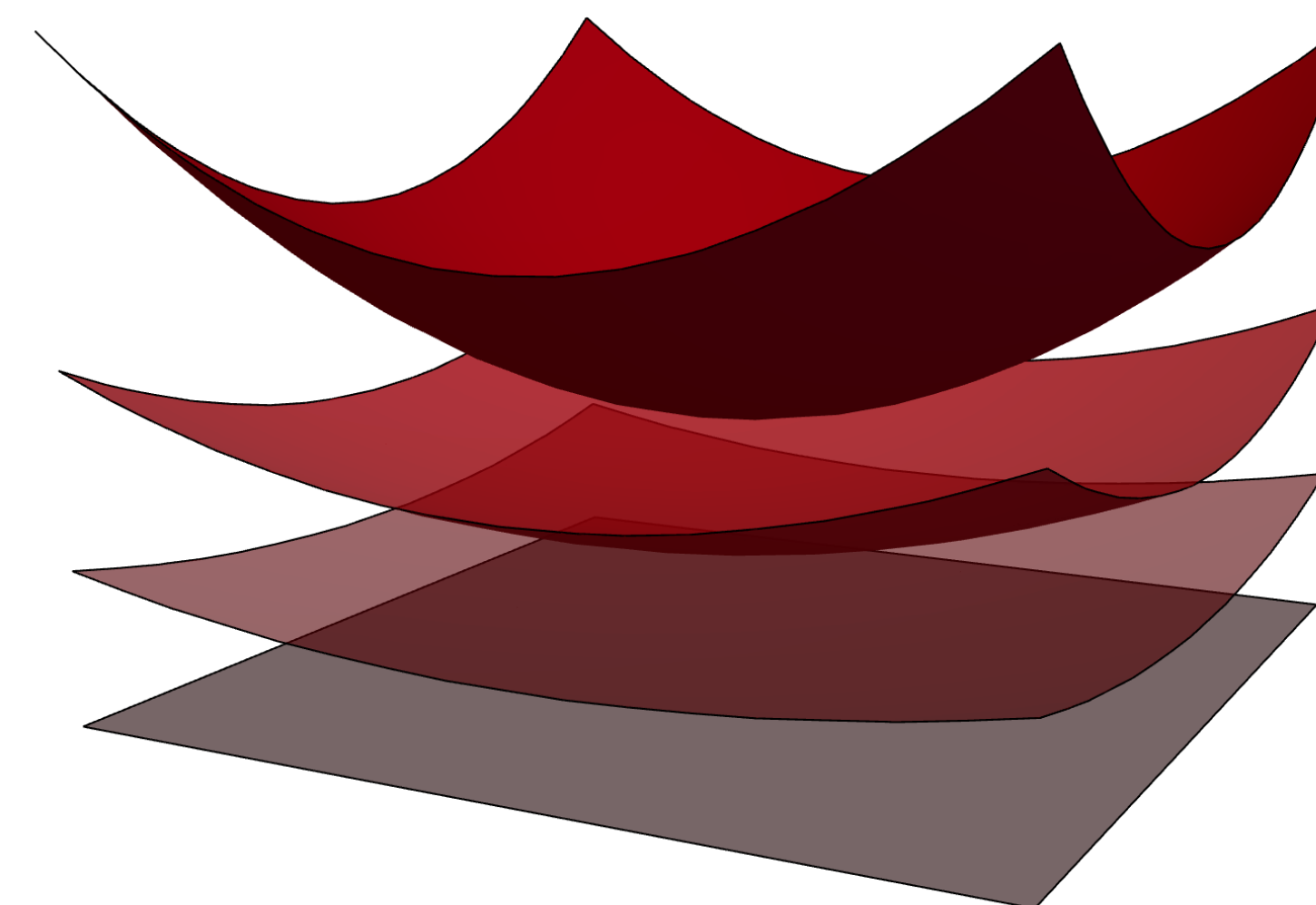
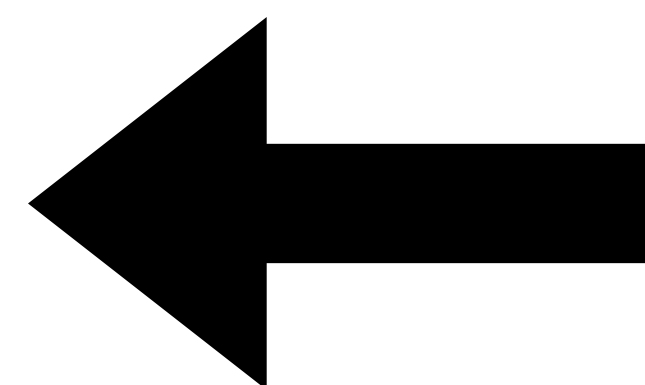
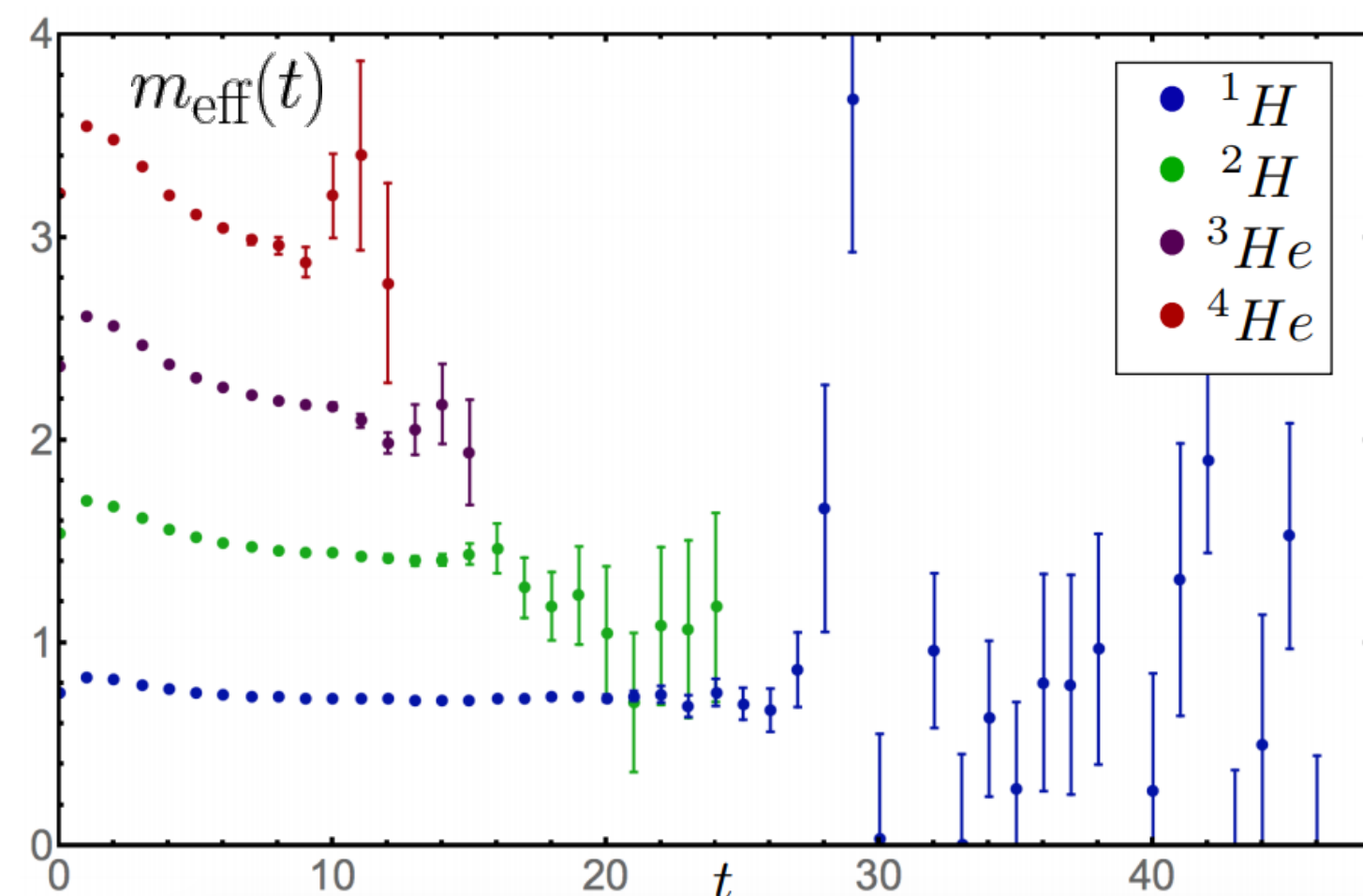


# Mitigating signal-to-noise problems using learned contour deformations



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University of Bern

<sup>b</sup>  
UNIVERSITÄT  
BERN

+ W. Detmold / H. Lamm / Y. Lin / P. Shanahan  
M. Wagman / N. Warrington

# A “Signal-to-Noise Problem”

Can define the Signal-to-Noise (StN) ratio of a Monte Carlo estimator for  $\langle \mathcal{O} \rangle$

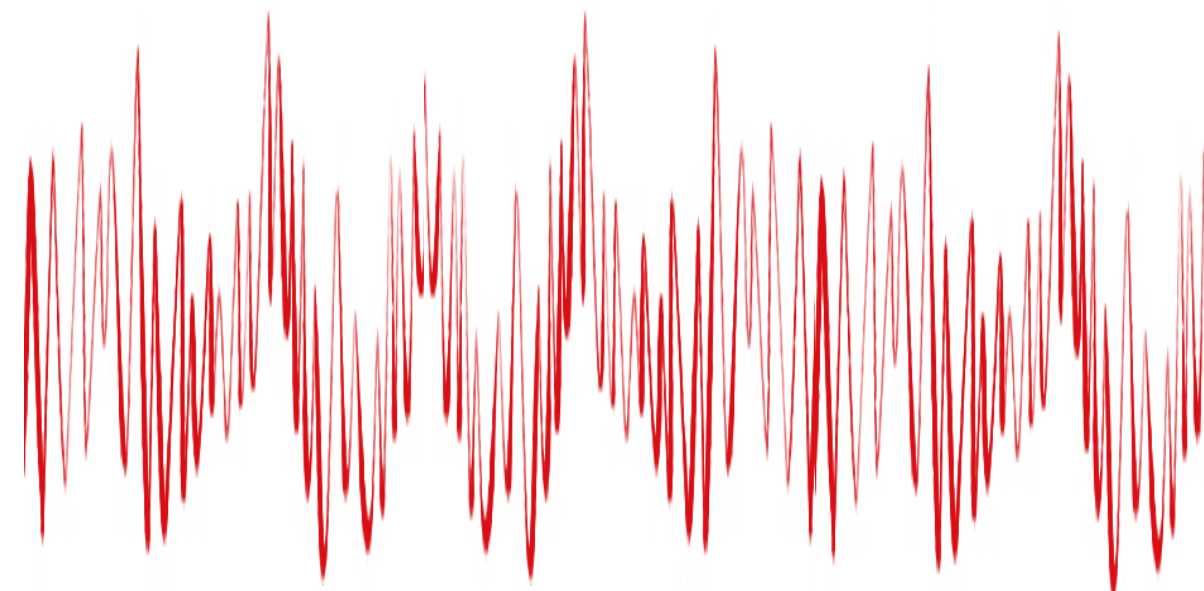
$$\text{StN} \equiv |\langle \mathcal{O} \rangle| / \text{Err}[\mathcal{O}]$$

We have a **StN problem** if fluctuations dominate the mean ...

$$\text{Err}[\mathcal{O}] = \sqrt{\text{Var}[\mathcal{O}]/n} \gg |\langle \mathcal{O} \rangle|$$

**NOISE**  **SIGNAL**

... leading to  $\text{StN} \ll 1$



**SIGNAL?**  
**NOISE?**

# Correlation functions

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

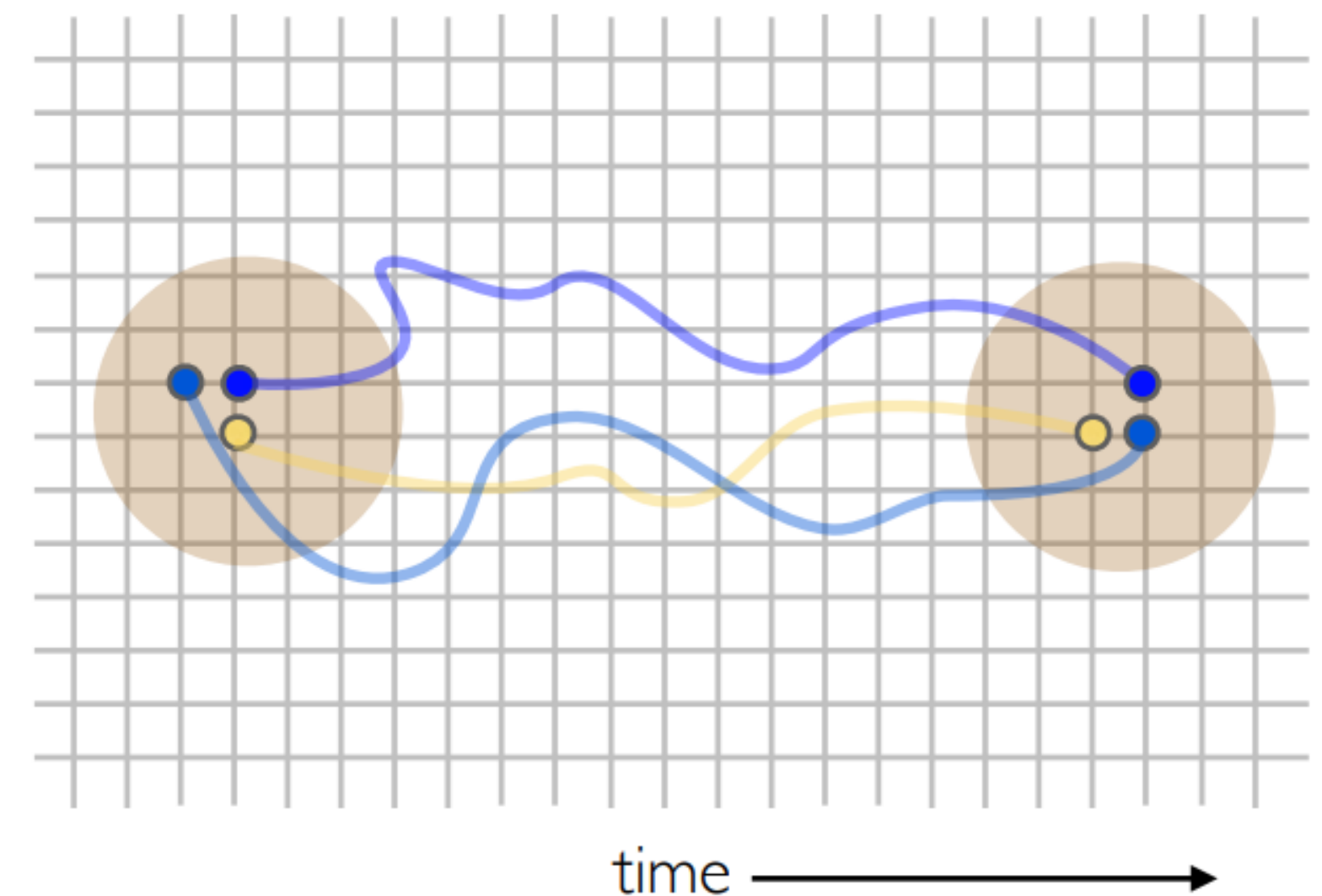
$$\langle \mathcal{A}(t) \mathcal{A}^\dagger(0) \rangle = \sum_n Z_n e^{-E_n t} \xrightarrow{t \gg (\Delta E)^{-1}} Z_0 e^{-E_0 t}$$

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.

[Detmold, INT-14-57W]



E.g. for the nucleon in lattice QCD

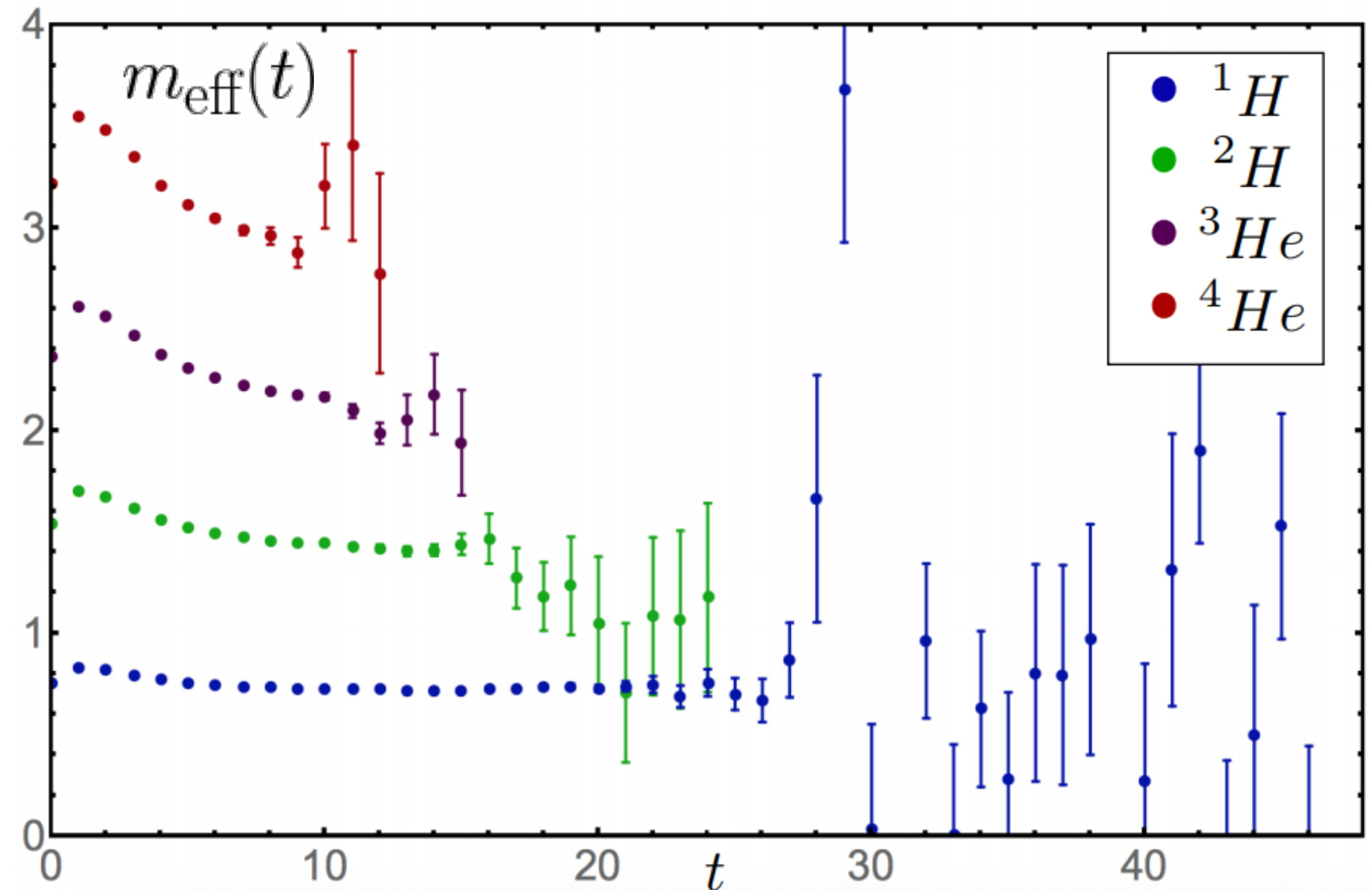
# StN problems in correlation functions

[Wagman, Lattice 2018]

Must find **plateau**\*

- Small  $t$ : Excited-state contamination
- Large  $t$ : **Noise problem!**

**THIS TALK**



\* Or regime dominated by only a few states



# Variance analysis

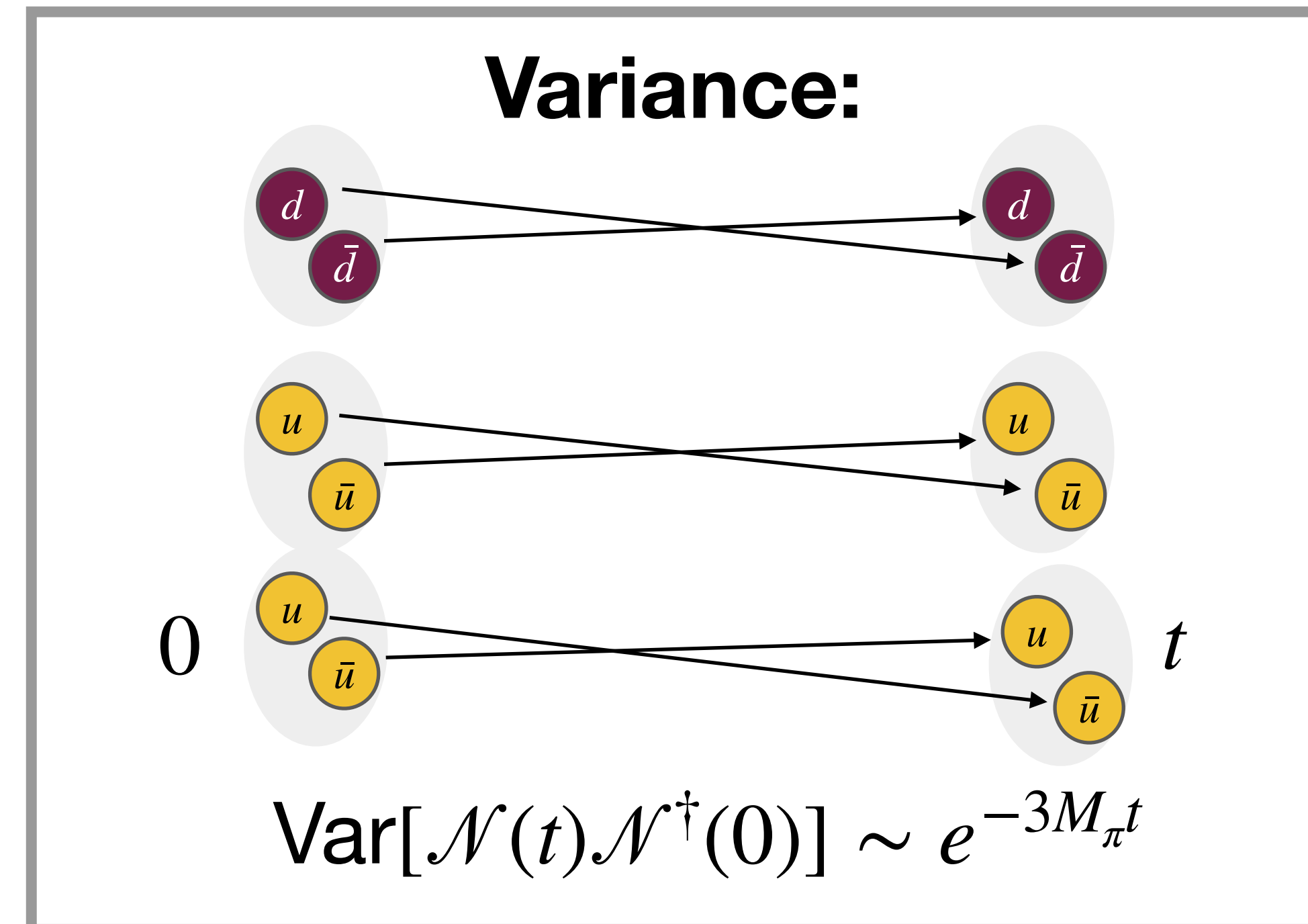
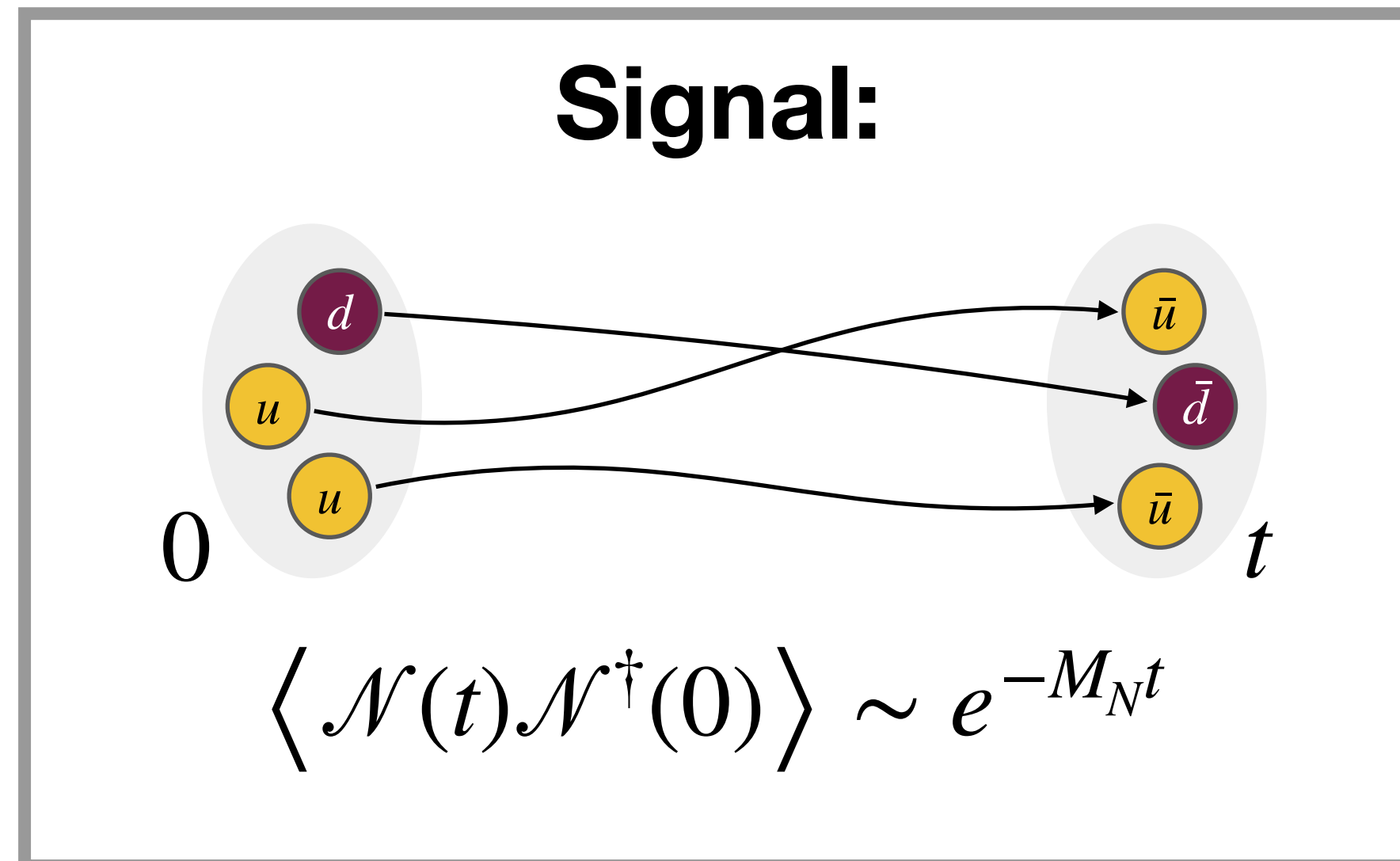
[Parisi Phys. Rept. 103 (1984) 203]  
 [Lepage TASI Proceedings (1989) 97]

Variance scaling related to physical states

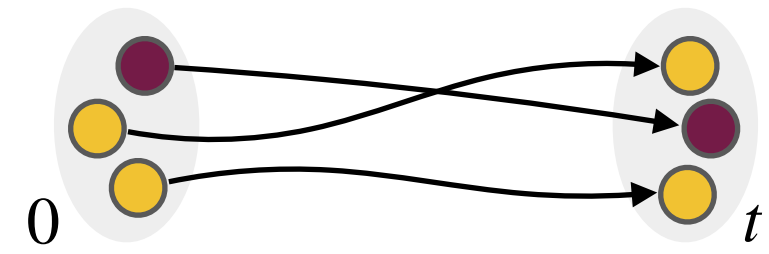
$$\text{Var}[\mathcal{A}(t)\mathcal{A}^\dagger(0)] \sim \left\langle \underbrace{\mathcal{A}(t)\mathcal{A}^\dagger(t)}_{\text{annihilating}} \underbrace{\mathcal{A}^\dagger(0)\mathcal{A}(0)}_{\text{creating}} \right\rangle \sim e^{-M_{A^\dagger A} t}$$

Interpret as annihilating / creating a physical state

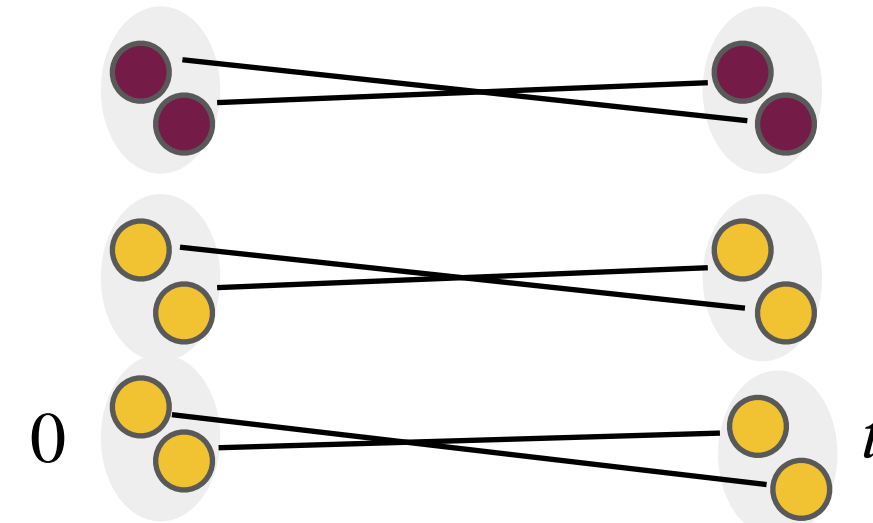
**Example:** StN of the nucleon given by



# Variance analysis: nucleon



$$\langle \mathcal{N}(t) \mathcal{N}^\dagger(0) \rangle \sim e^{-M_N t}$$



$$\text{Var}[\mathcal{N}(t) \mathcal{N}^\dagger(0)] \sim e^{-3M_\pi t}$$



$$\text{StN}[\mathcal{N}(t) \mathcal{N}^\dagger(0)] = \frac{\langle \mathcal{N}(t) \mathcal{N}^\dagger(0) \rangle}{\sqrt{\text{Var}[\mathcal{N}(t) \mathcal{N}^\dagger(0)]}} \sim e^{-(M_N - 3M_\pi/2)t}$$

**Exponentially bad!**

**Similar problems affect 3-point and higher correlation functions!**

# Noise problem = sign problem

Intuitively a noise problem implies ...

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

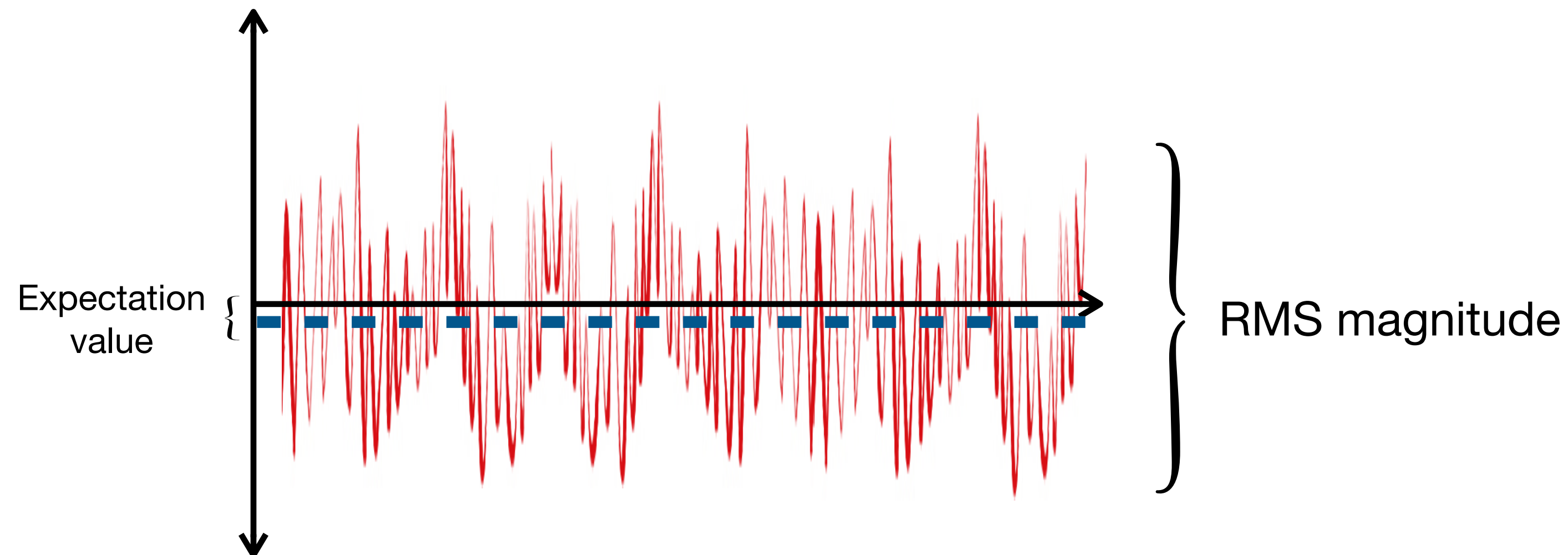
The RMS magnitude ...

$\gg$

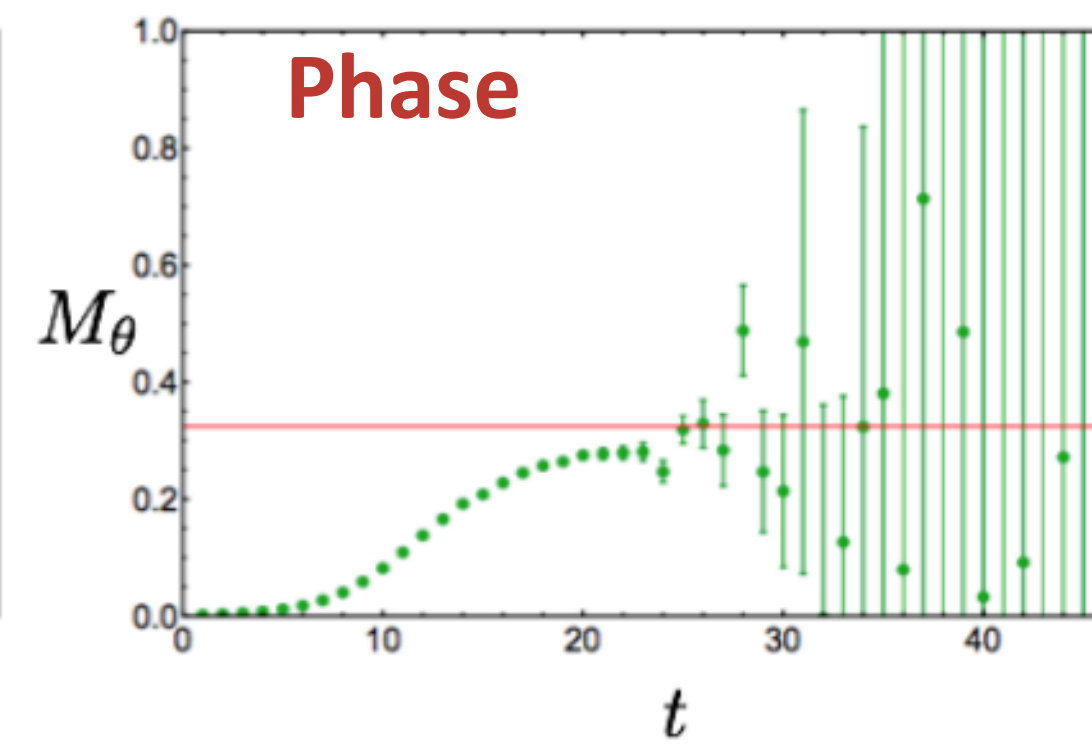
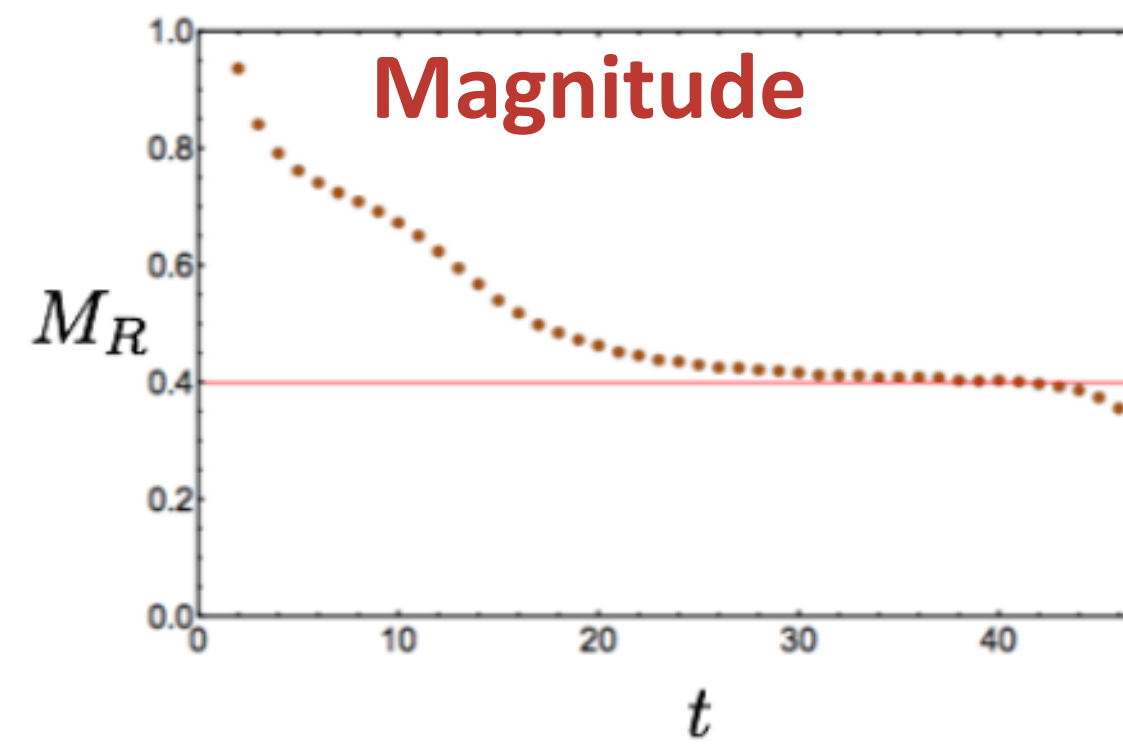
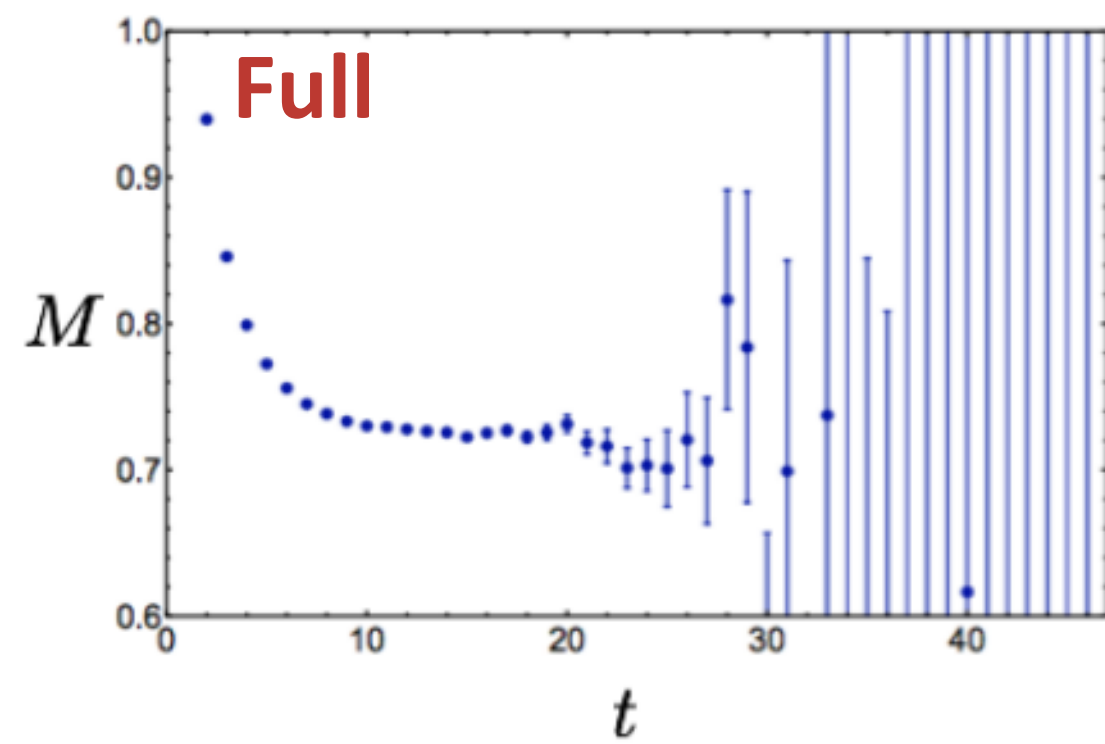
$$|\langle C(t) \rangle|$$

... is exponentially larger than the expectation value.

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



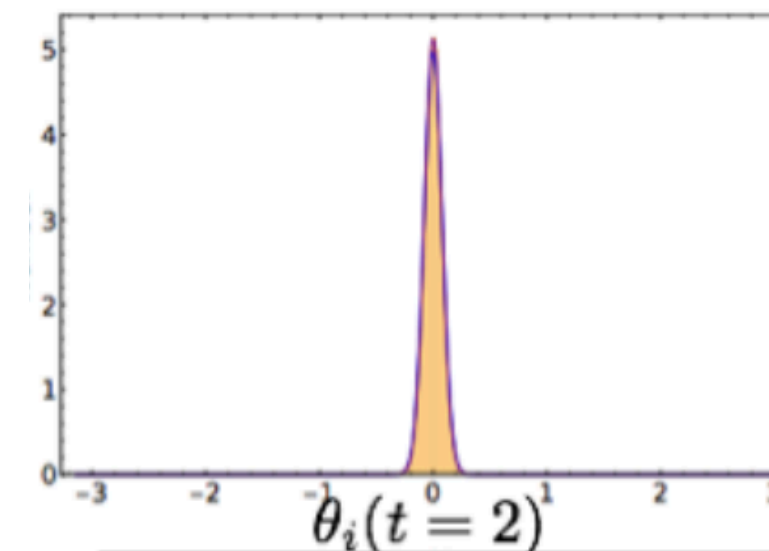
# Noise problem = sign problem



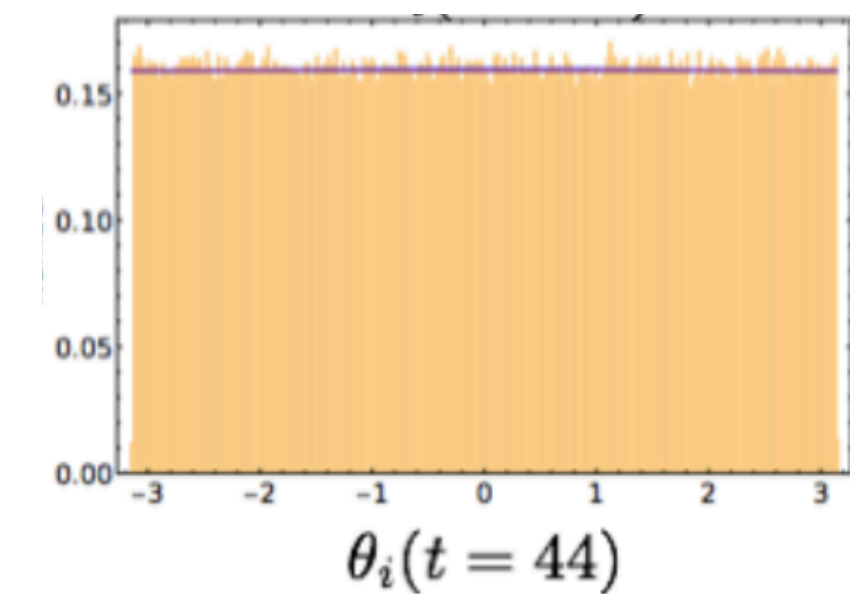
Noise in nucleon effective mass carried by complex phase

[Wagman & Savage, PRD96 (2017) 114508]

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



vs.

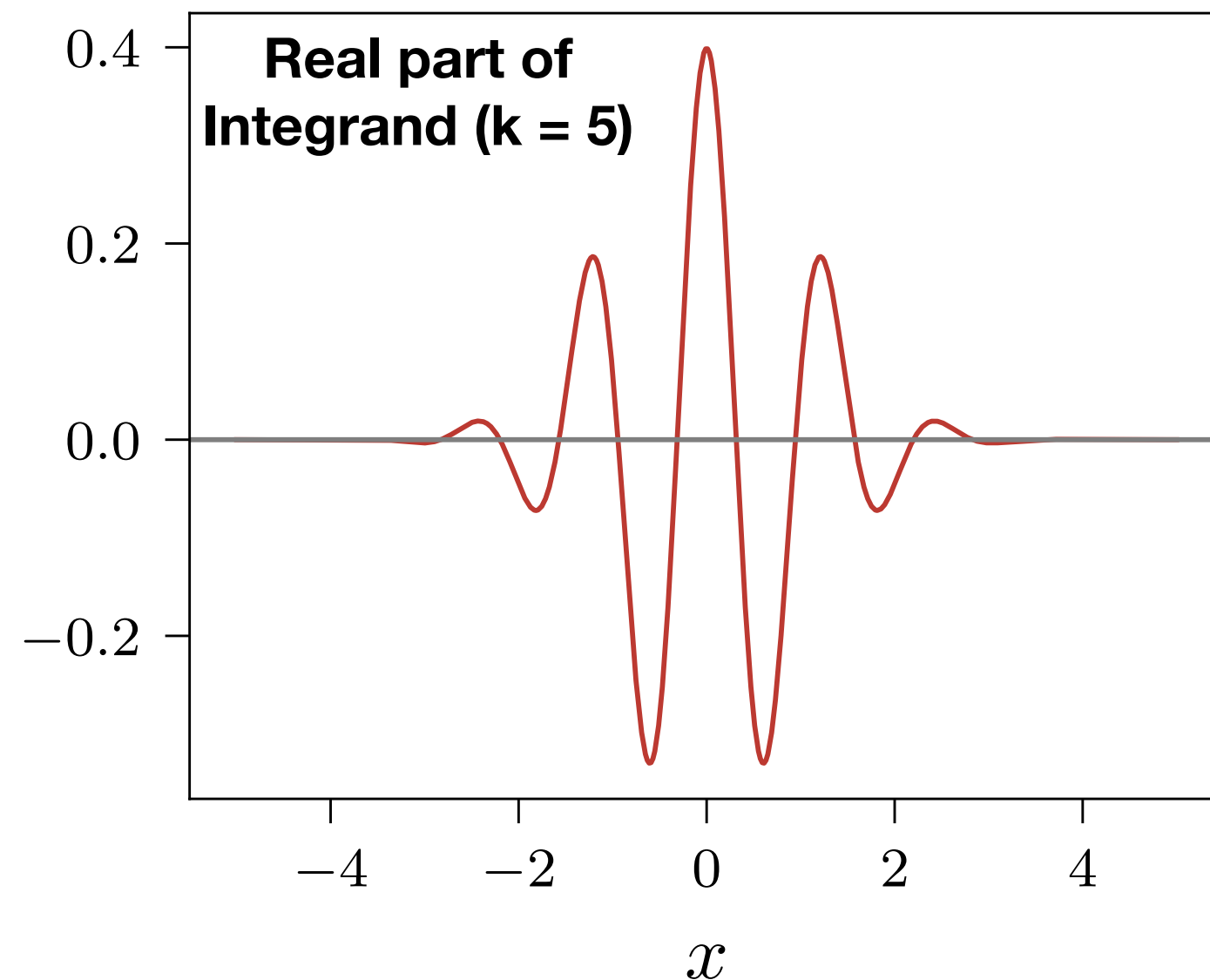


# Case study: Gaussian StN problem

**Toy example:** Simple “observable” in a Gaussian “theory”

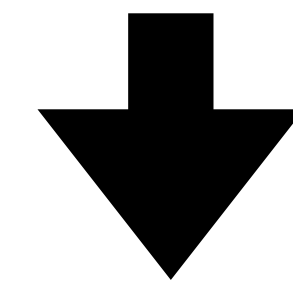
$$\langle e^{ikx} \rangle = \frac{1}{Z} \int_{-\infty}^{\infty} dx [e^{ikx}] e^{-x^2/2}$$

**Monte Carlo approach:** Sample from distribution  $p(x) = e^{-x^2/2}/Z$



**Signal:**  $\langle e^{ikx} \rangle \sim \text{exp. small in } k^2$

**Variance:**  $\text{Var}[e^{ikx}] \sim \langle |e^{ikx}|^2 \rangle = 1$



**StN and sign problem!**

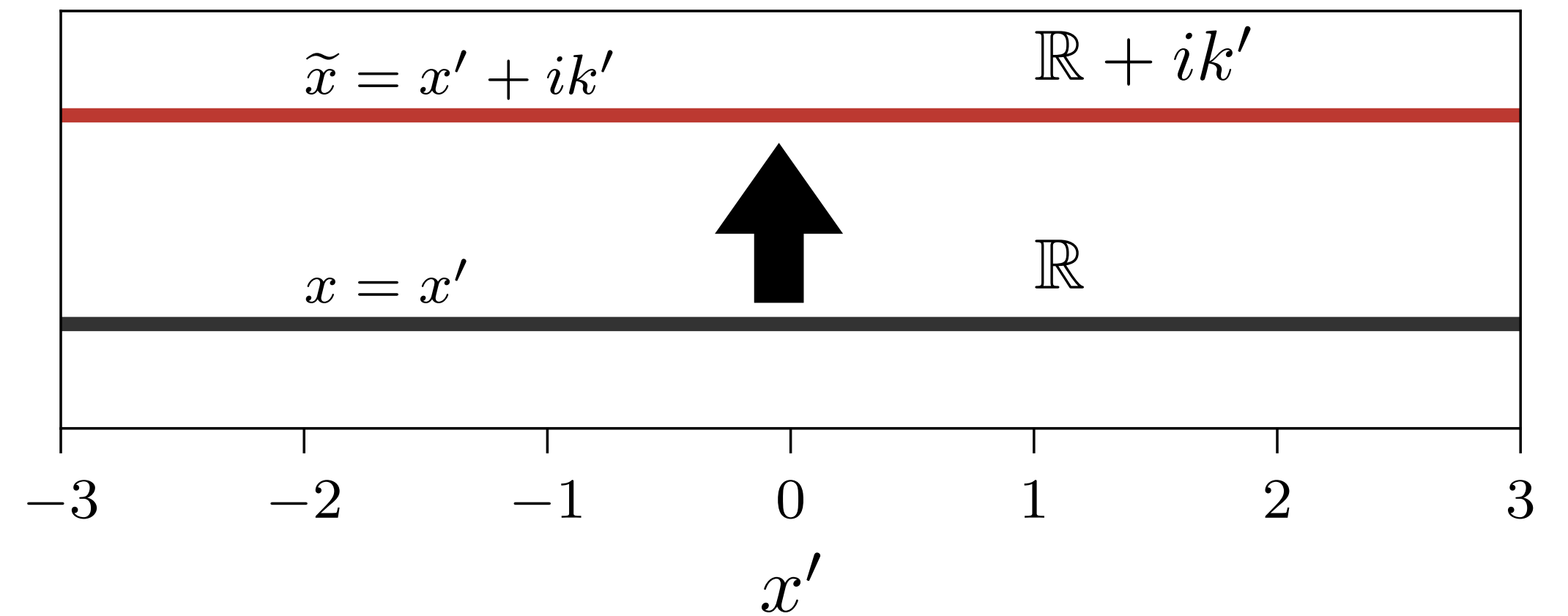


# Case study: Gaussian StN problem

## Deformation approach:

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

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

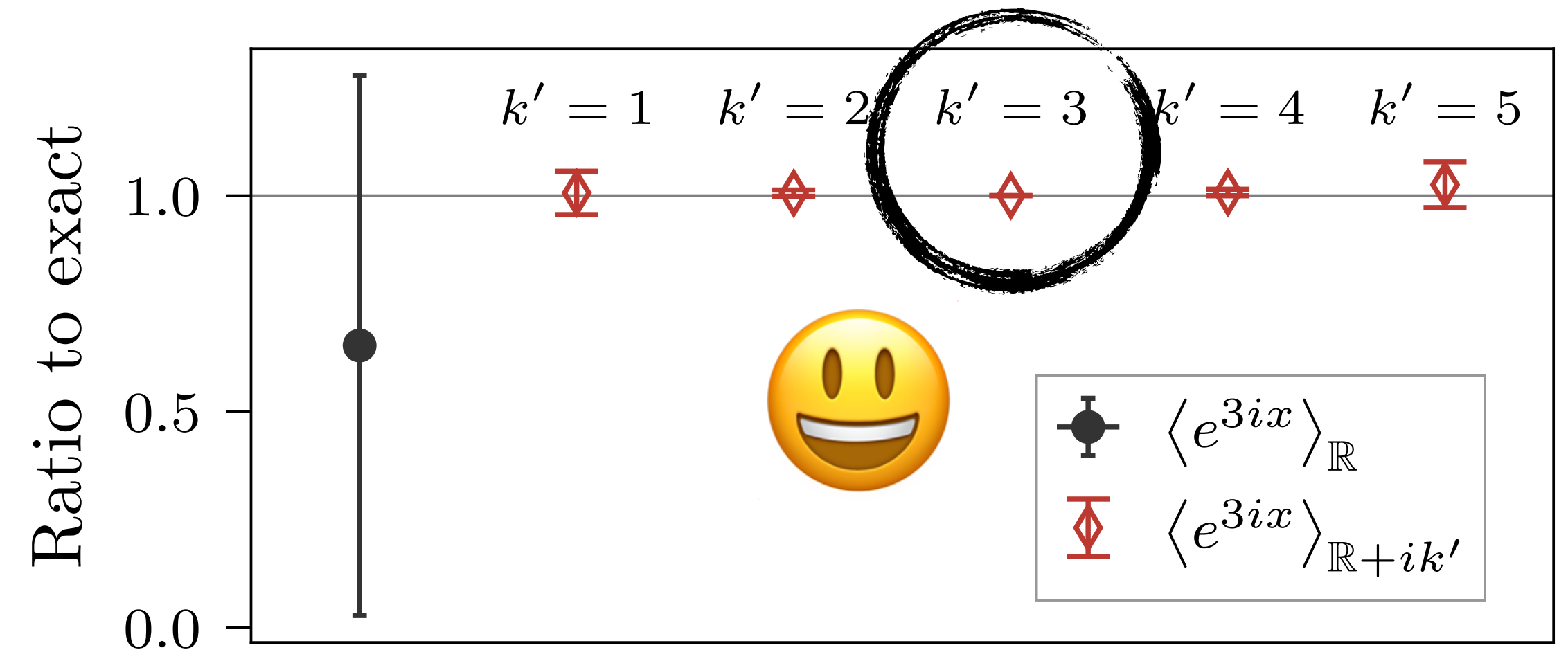


# Case study: Gaussian StN problem

**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  $e^{-\frac{1}{2}x'^2 - ix'k' + \frac{1}{2}k'^2}$ .

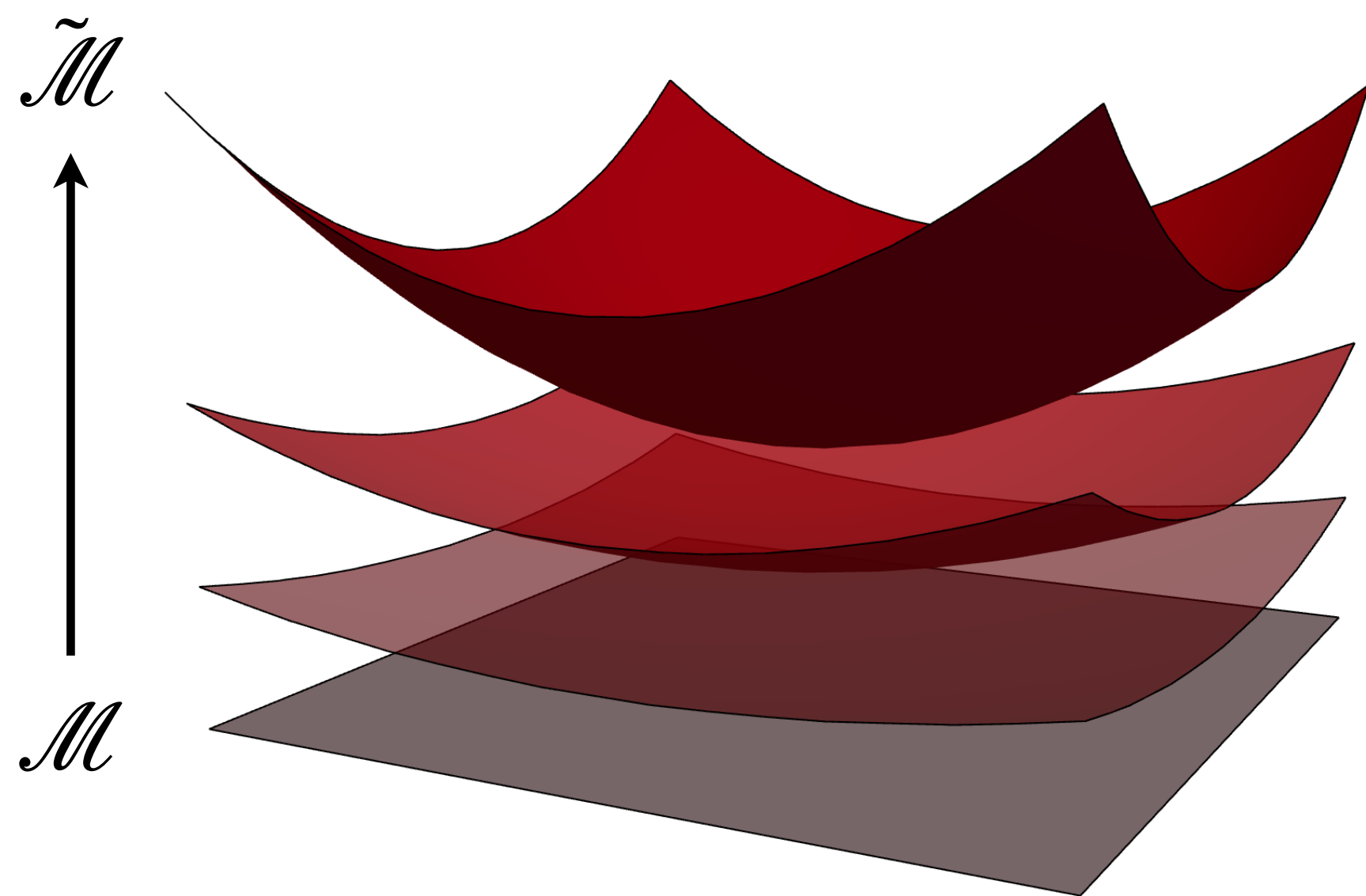
$$\text{Result: } \langle e^{ikx} \rangle = \left\langle e^{ix'(k-k')} e^{\frac{1}{2}k'^2 - kk'} \right\rangle$$



# Deforming the path integral

Just high-dimensional contour deformation...

$$\langle \mathcal{O} \rangle = \int_{\mathcal{M}} \mathcal{D}U e^{-S(U)} \mathcal{O}(U) = \int_{\tilde{\mathcal{M}}} \mathcal{D}\tilde{U} e^{-S(\tilde{U})} \mathcal{O}(\tilde{U})$$



Many works applying this to “extensive” sign problems (e.g. non-zero density, real time)

**Cristoforetti+** PRD86(074506), PRD88(051501), PRD89(114505)

**Aarts** PRD88(094501)

**Alexandru+** PRD93(014504), JHEP05(053), PRD96(094505), PRD98(054514), PRD98(034506), PRD97(094510), PRL121(191602)

**Fujii+** JHEP12(125)

**Tanizaki+** NJP18(033002)

**Mori+** PTEP2018(023B04), PRD99(014033)

**Alexandru+** PRL117(081602), PRD95(114501)

**Mou+** JHEP11(135)

...

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

# Path integral deformations for observables

- Deform the path integral **numerator** only:

$$\langle \mathcal{O} \rangle = \frac{1}{Z} \int_{\tilde{\mathcal{M}}} e^{-S(\tilde{U})} \mathcal{O}(\tilde{U}) \xrightarrow{\text{manifold coordinates}} \frac{1}{Z} \int_{\mathcal{M}} J(U) e^{-S(\tilde{U}(U))} \mathcal{O}(\tilde{U}(U))$$

- Define a new observable  $\mathcal{Q}(U)$

$$\mathcal{Q}(U) \equiv e^{-[S_{\text{eff}}(U) - S(U)]} \mathcal{O}(\tilde{U}(U))$$

where  $S_{\text{eff}}(U) \equiv S(\tilde{U}(U)) - \log J(U)$

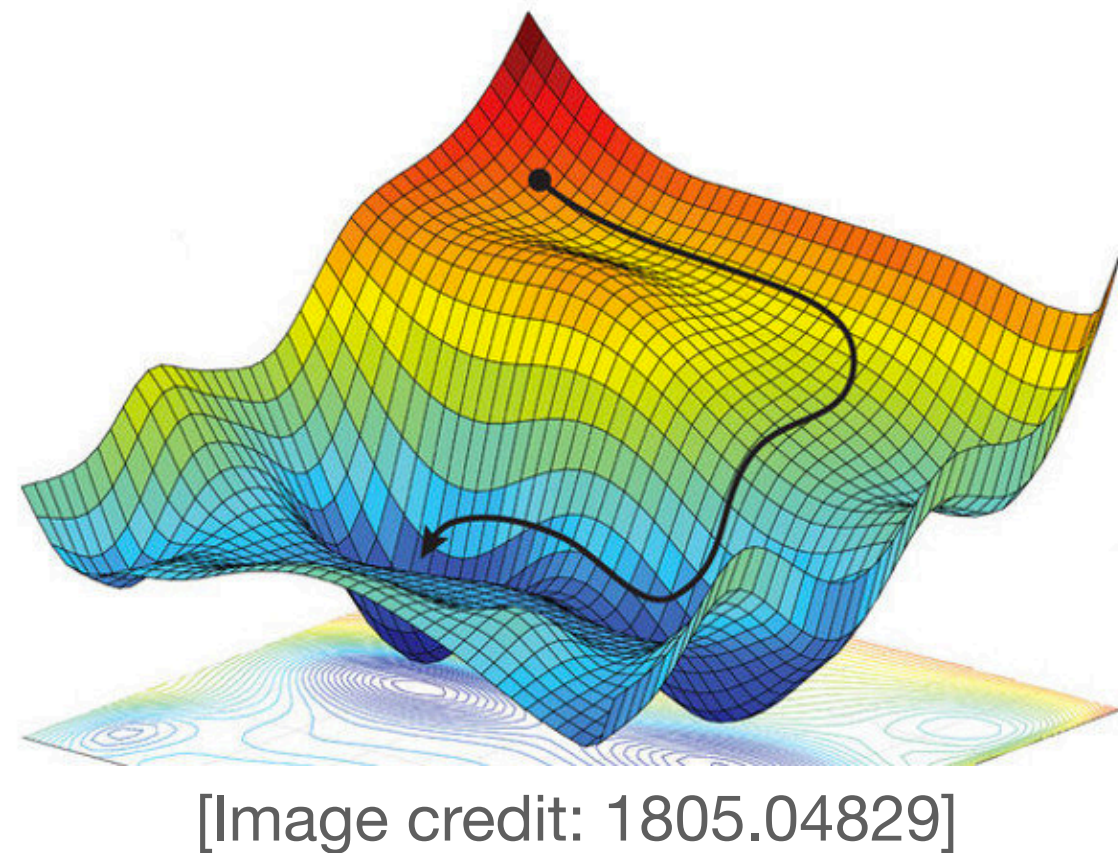
- $\mathcal{Q}(U)$  has identical expectation value, but different variance:

$$\langle \mathcal{Q}(U) \rangle = \langle \mathcal{O}(U) \rangle$$
$$\text{Var}[\mathcal{Q}(U)] \neq \text{Var}[\mathcal{O}(U)]$$



# Optimizing the variance

Choice of  $\tilde{U}(U)$  defines  $\mathcal{Q}(U)$  and determines the variance.



Parameterize  $\tilde{U}(U)$  using ML techniques, then **numerically minimize variance**.

Gradients of variance w.r.t. params defined using **original Monte Carlo ensemble**:

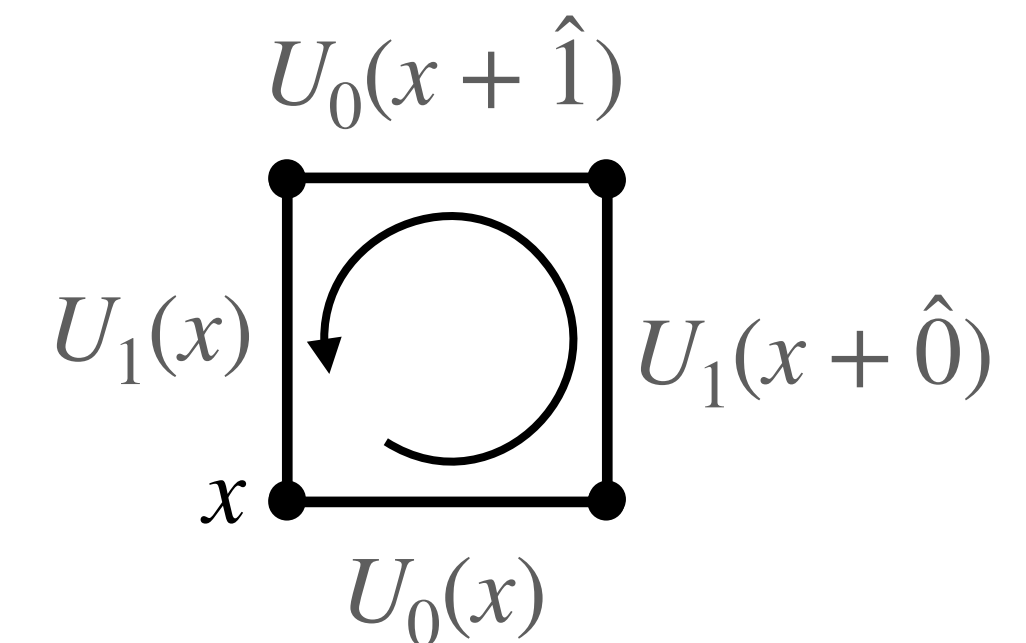
$$\begin{aligned}\nabla_{\vec{\omega}} \text{Var}[\text{Re } \mathcal{Q}] &= \langle \nabla_{\vec{\omega}} (\text{Re } \mathcal{Q})^2 \rangle = 2 \langle \text{Re } \mathcal{Q} \text{ Re } \nabla_{\vec{\omega}} \mathcal{Q} \rangle \\ &= 2 \left\langle (\text{Re } \mathcal{Q}) \text{Re} \left( \mathcal{Q}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

# $SU(N)$ pure-gauge theory

Simple action for the theory given in terms of **plaquettes**

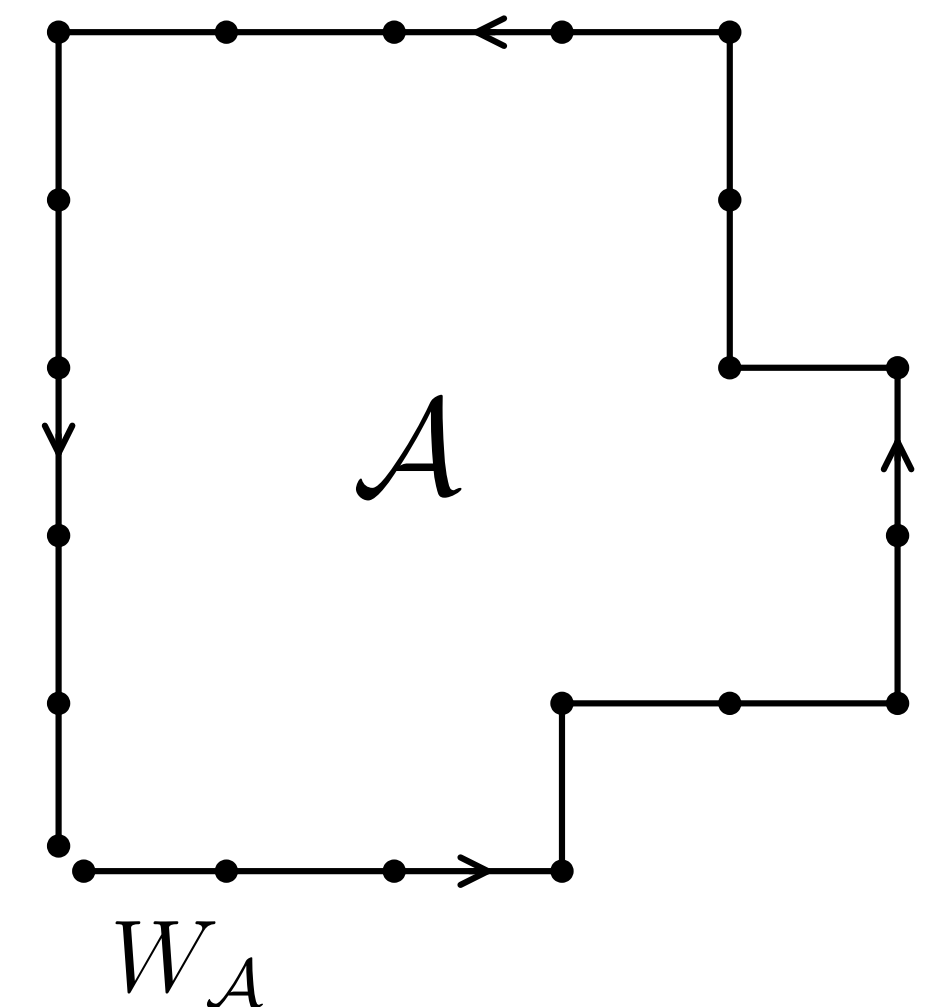
$$S[P] = -\frac{1}{g^2} \sum_x \text{tr}(P_x + P_x^{-1}) \xrightarrow{\text{ctm.}} \frac{1}{2g^2} \text{tr}(F^2)$$

$$P_x = U_0(x)U_1(x + \hat{0})U_0^\dagger(x + \hat{1})U_0^\dagger(x)$$

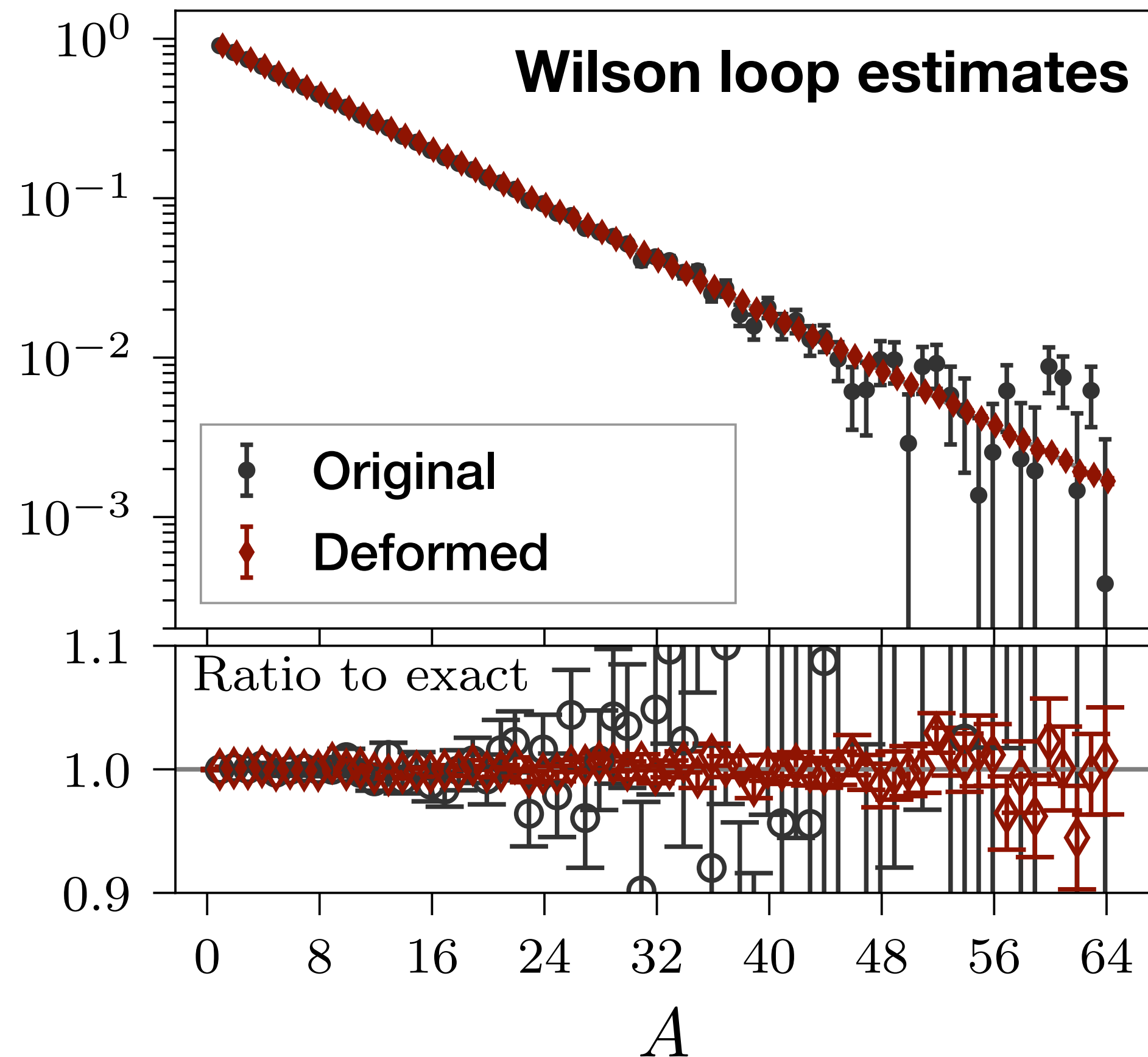


**Wilson loop** observables  $W_{\mathcal{A}}$ :

- Generalize plaquettes to arbitrary loops
- Probe of confinement (area vs perimeter law scaling)



# 1+1D SU(3) Wilson loops

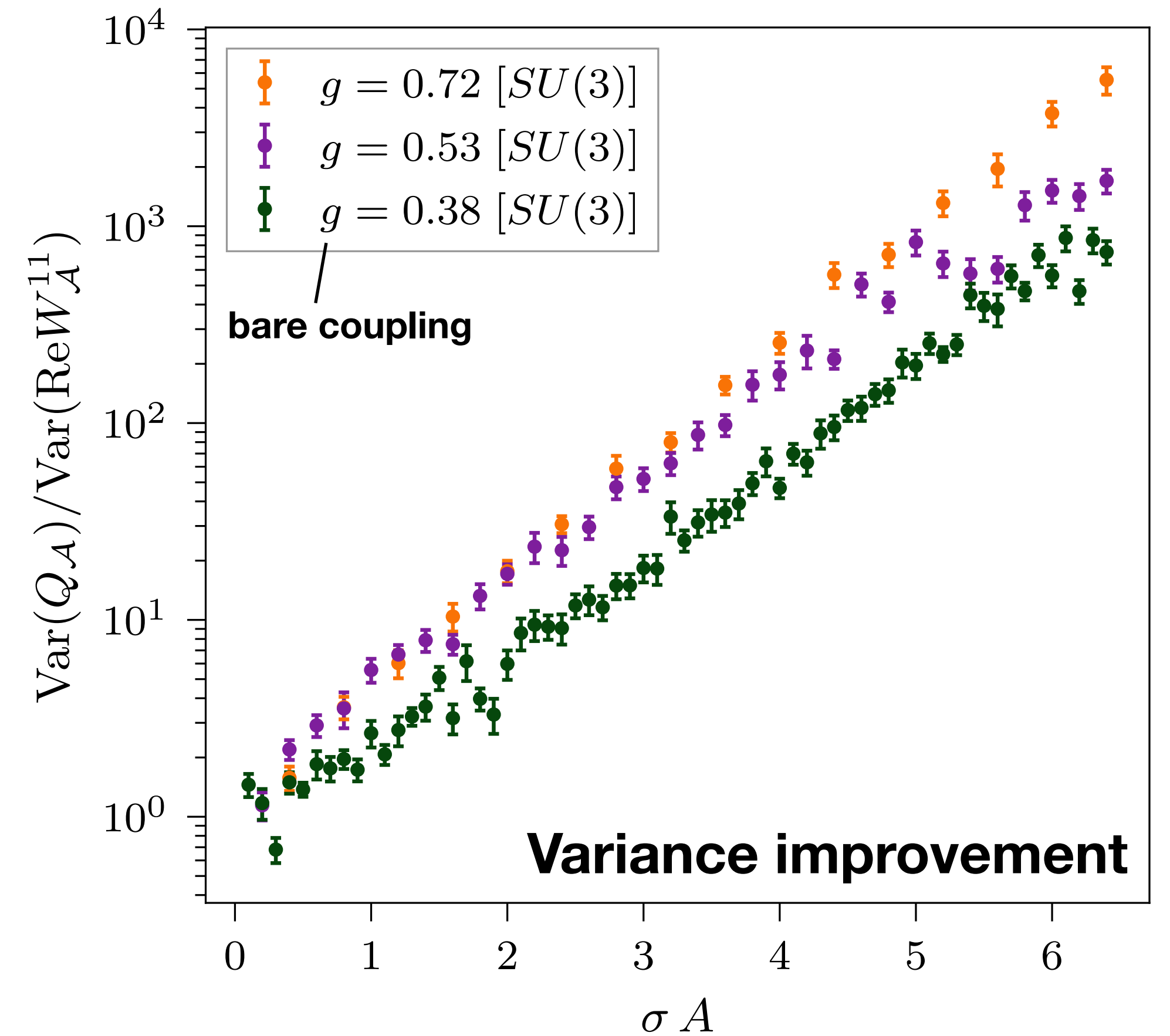
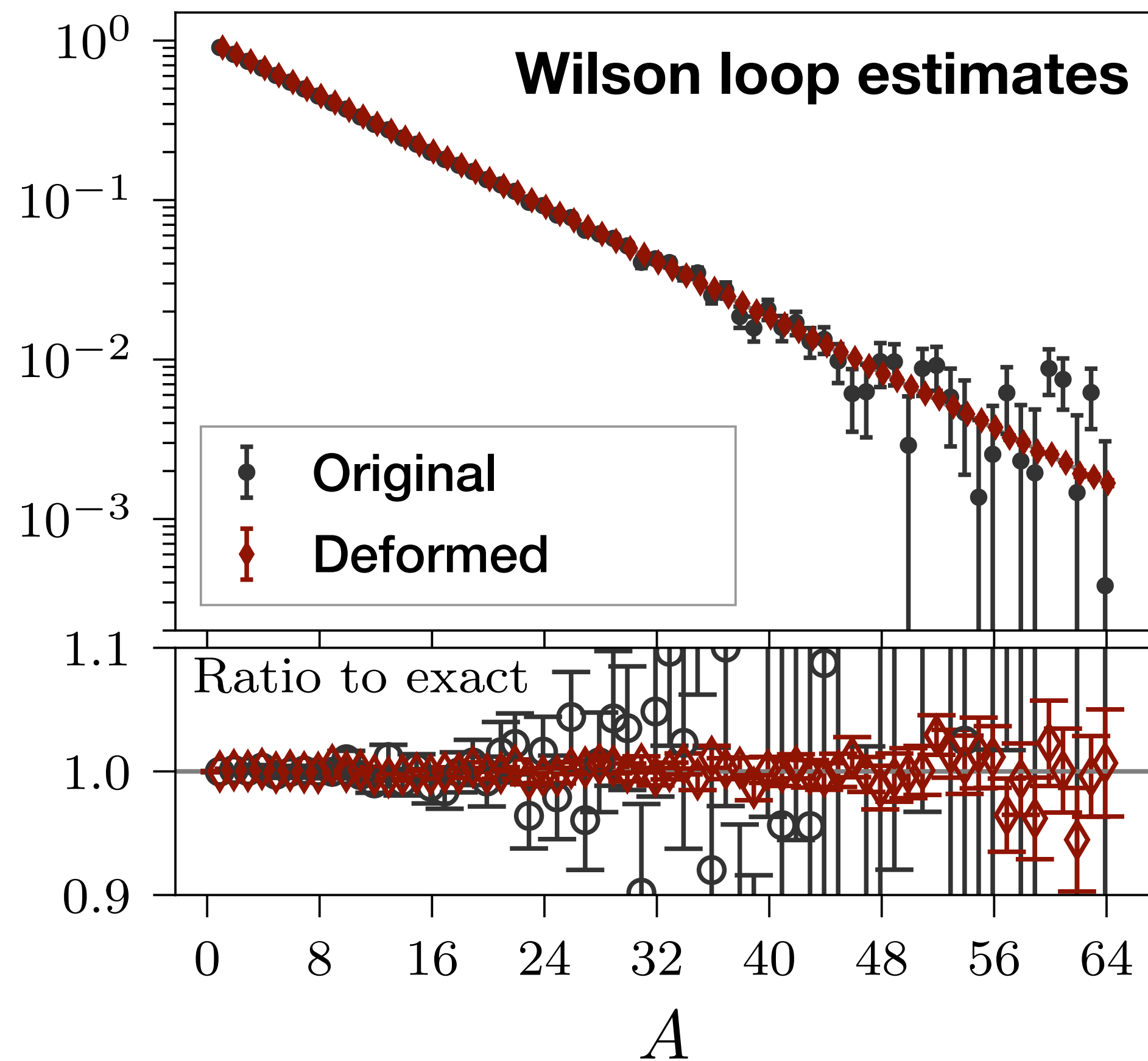


## “Constant shift” deformations:

- 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

# 1+1D SU(3) Wilson loops

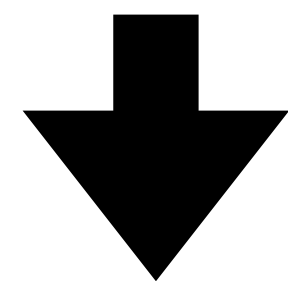
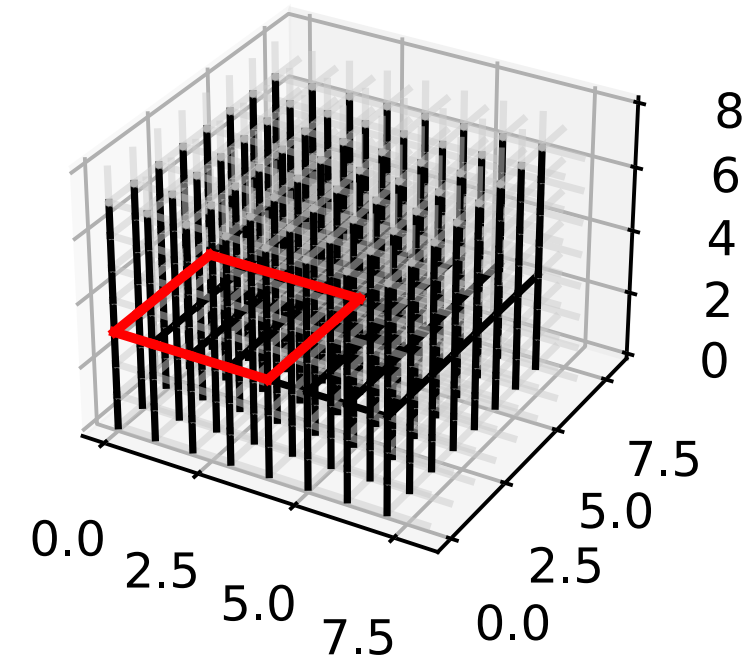
Significant variance reduction at large  $A$  for all couplings, no bias.





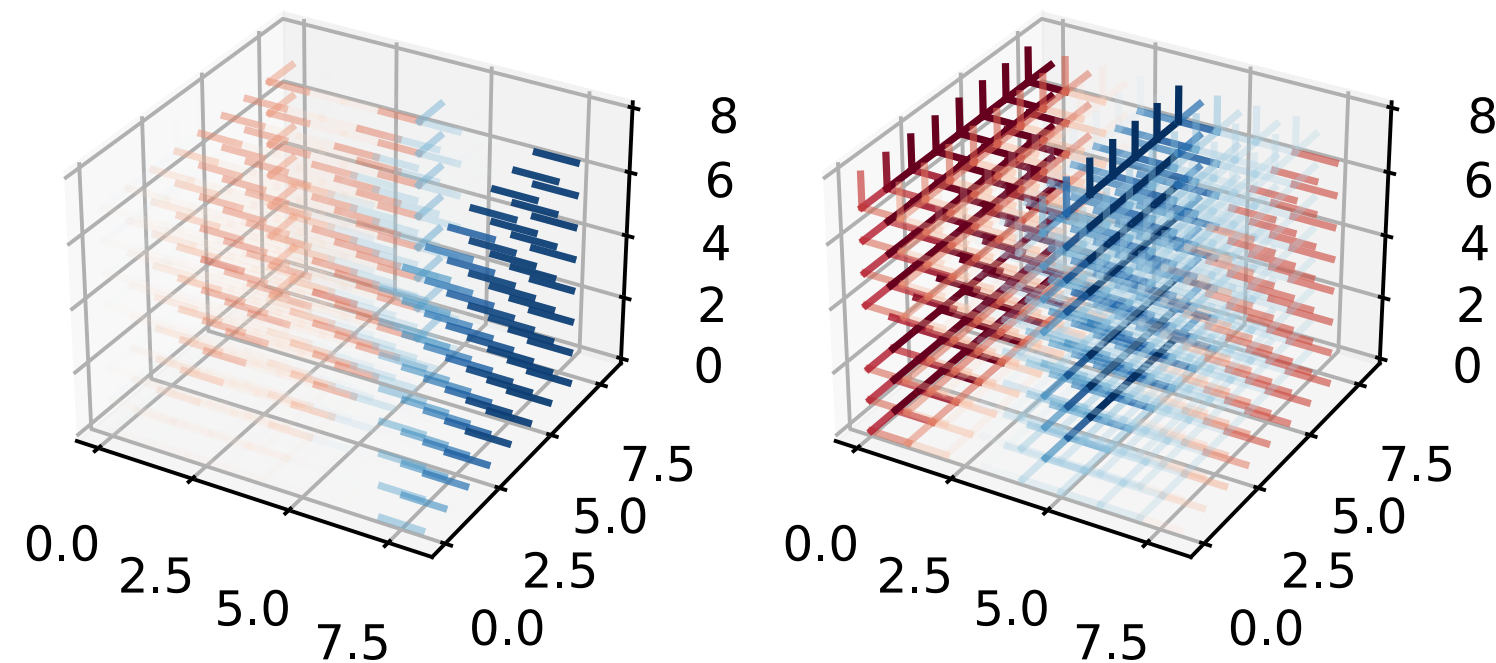
# 2+1D SU(2) Wilson loops

Wilson loop and gauge fixing scheme

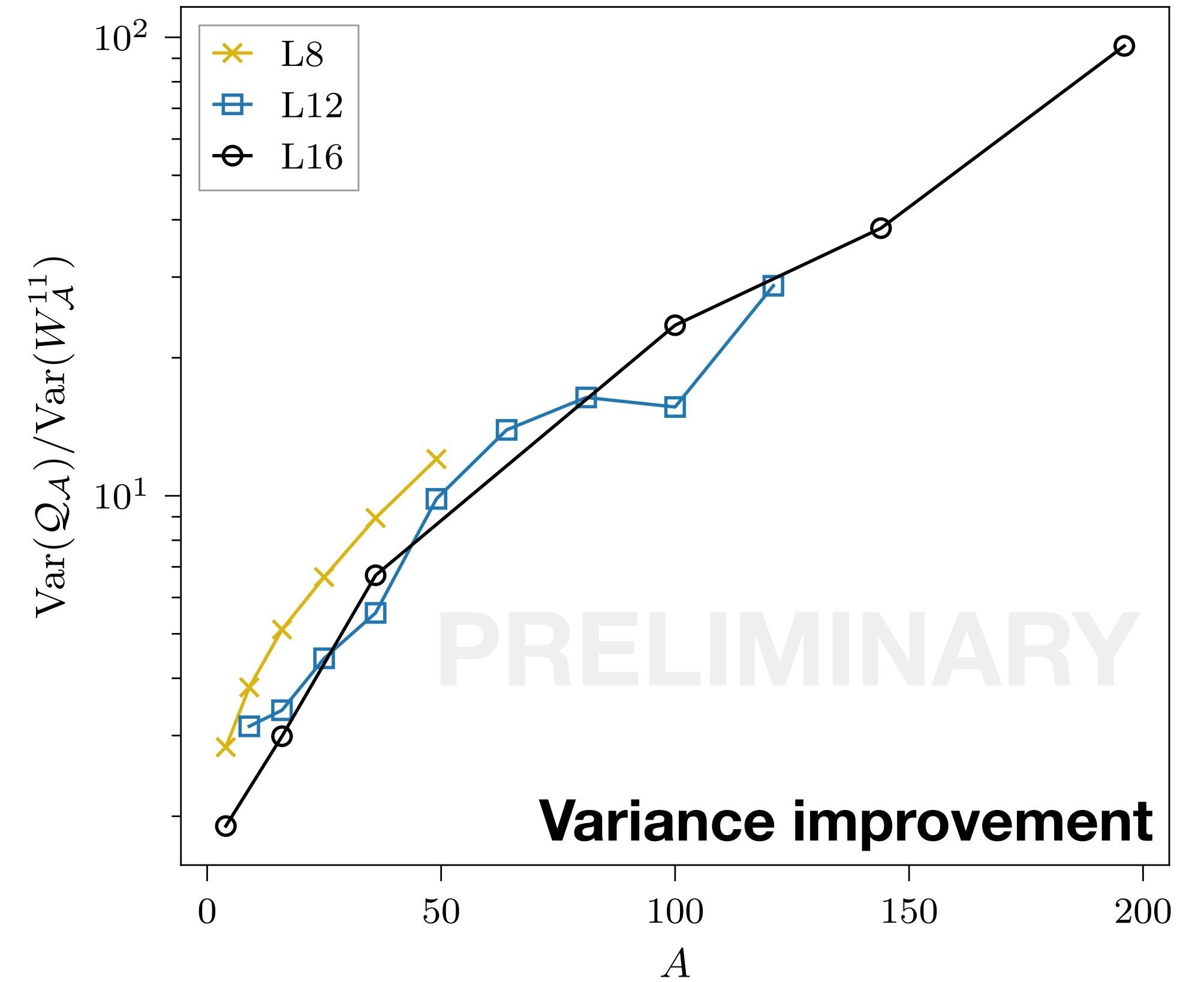


Phi1 shift

Phi2 shift



Learned shifts

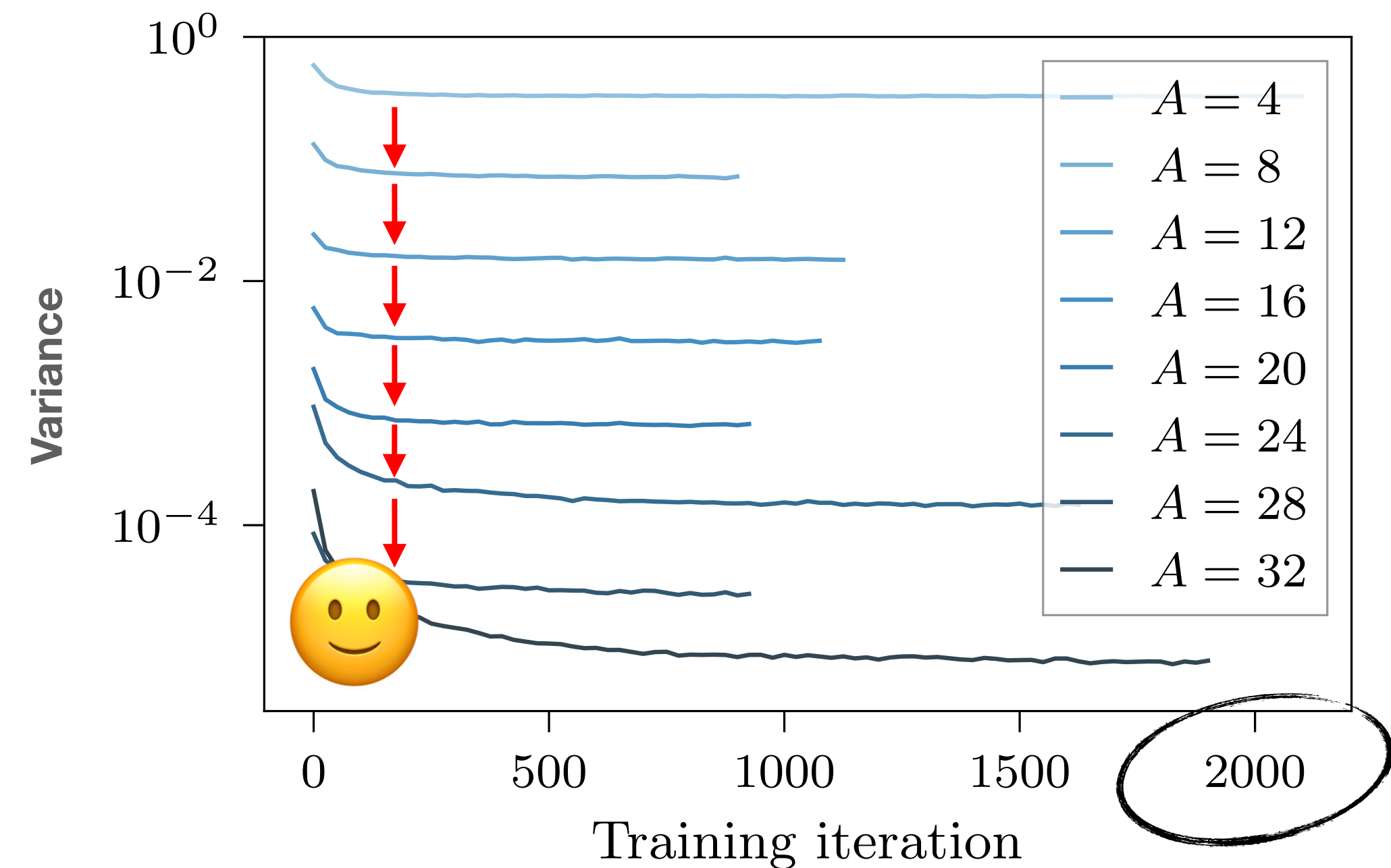
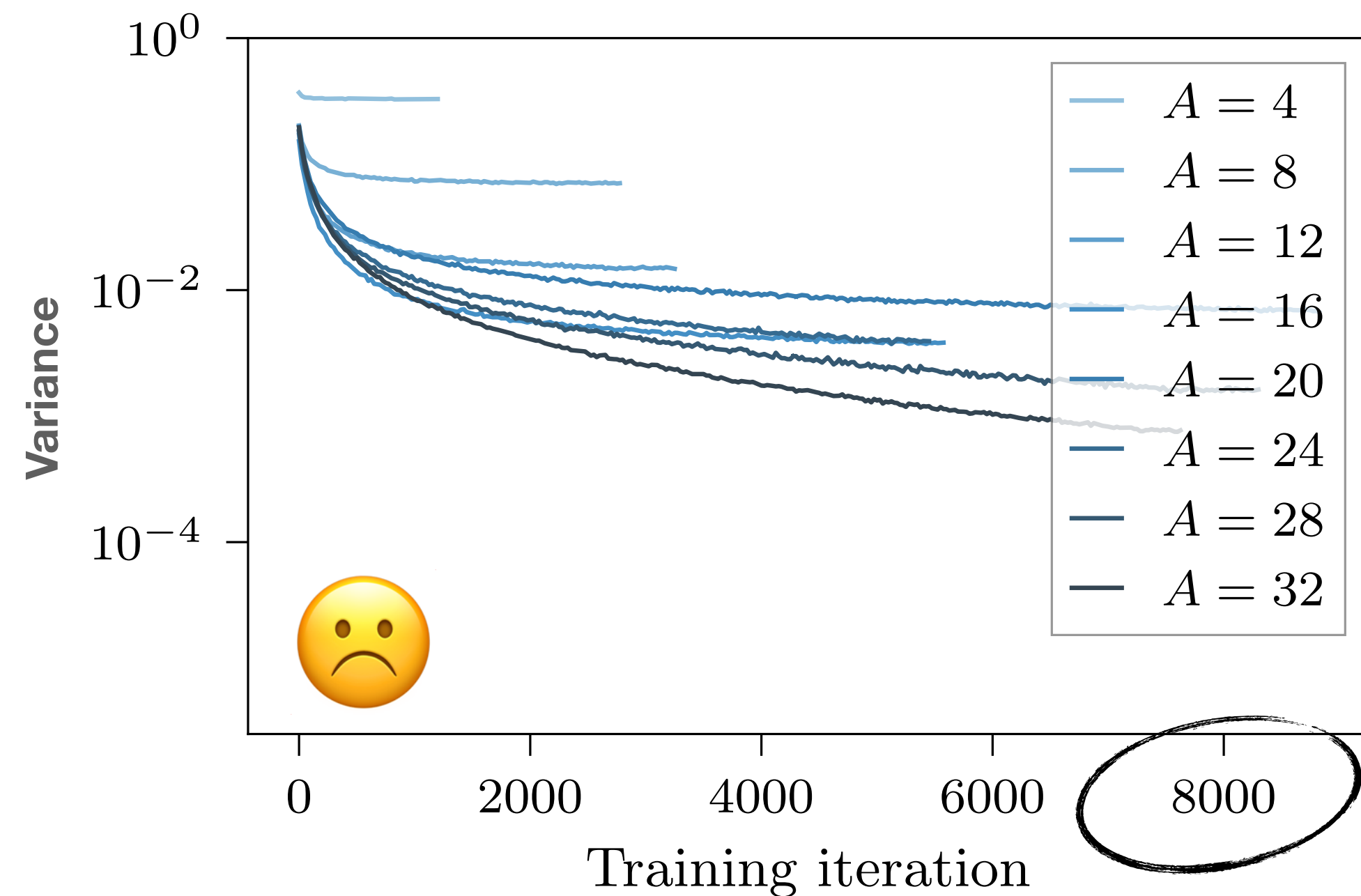




# Takeaway: Transfer learning works well

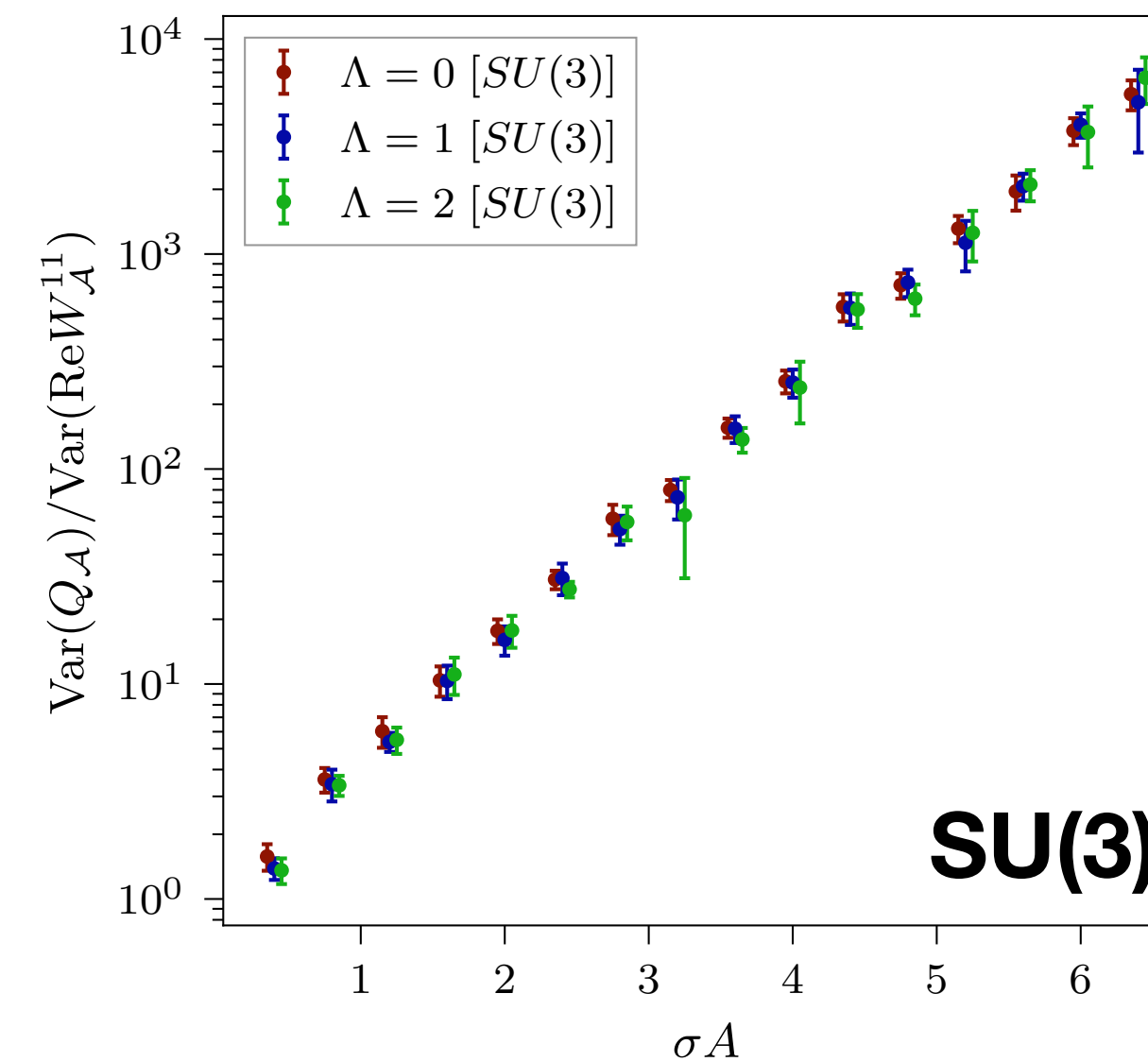
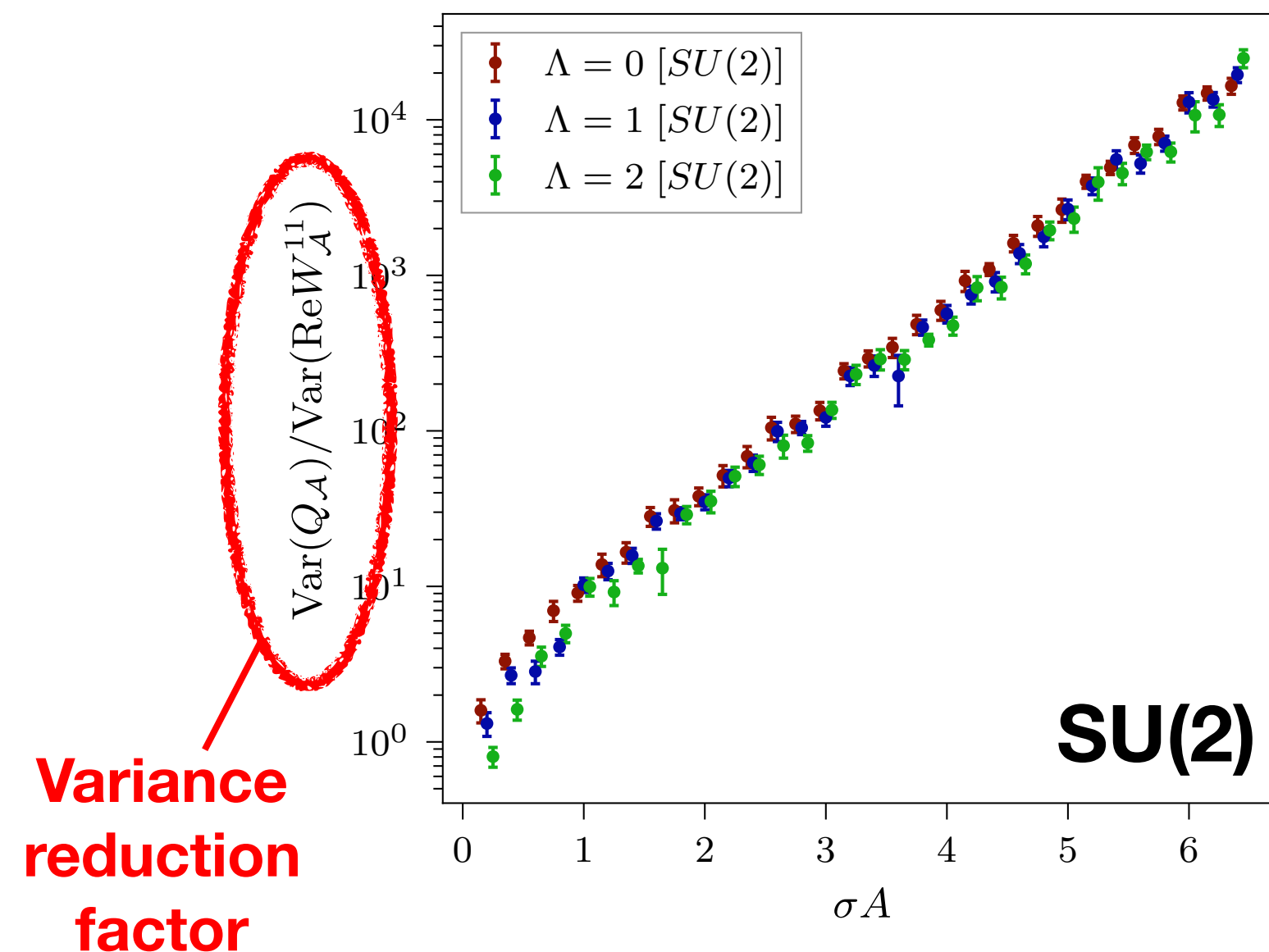
**Challenge:** Different optimal manifold per observable (e.g. per Wilson loop  $A$ )

**Solution:** “Transfer” optimized manifold to initialize next search



# Takeaway: (so far) simple manifolds work

- Contour deformations based on field-independent transformations only
  - Using a Fourier expansion to introduce field-dependence not useful



- Deformations are related to normalizing flows...  
can we use NNs & other methods from this domain?

See e.g. [Rodekamp+ PRB106 (2022) 125139]

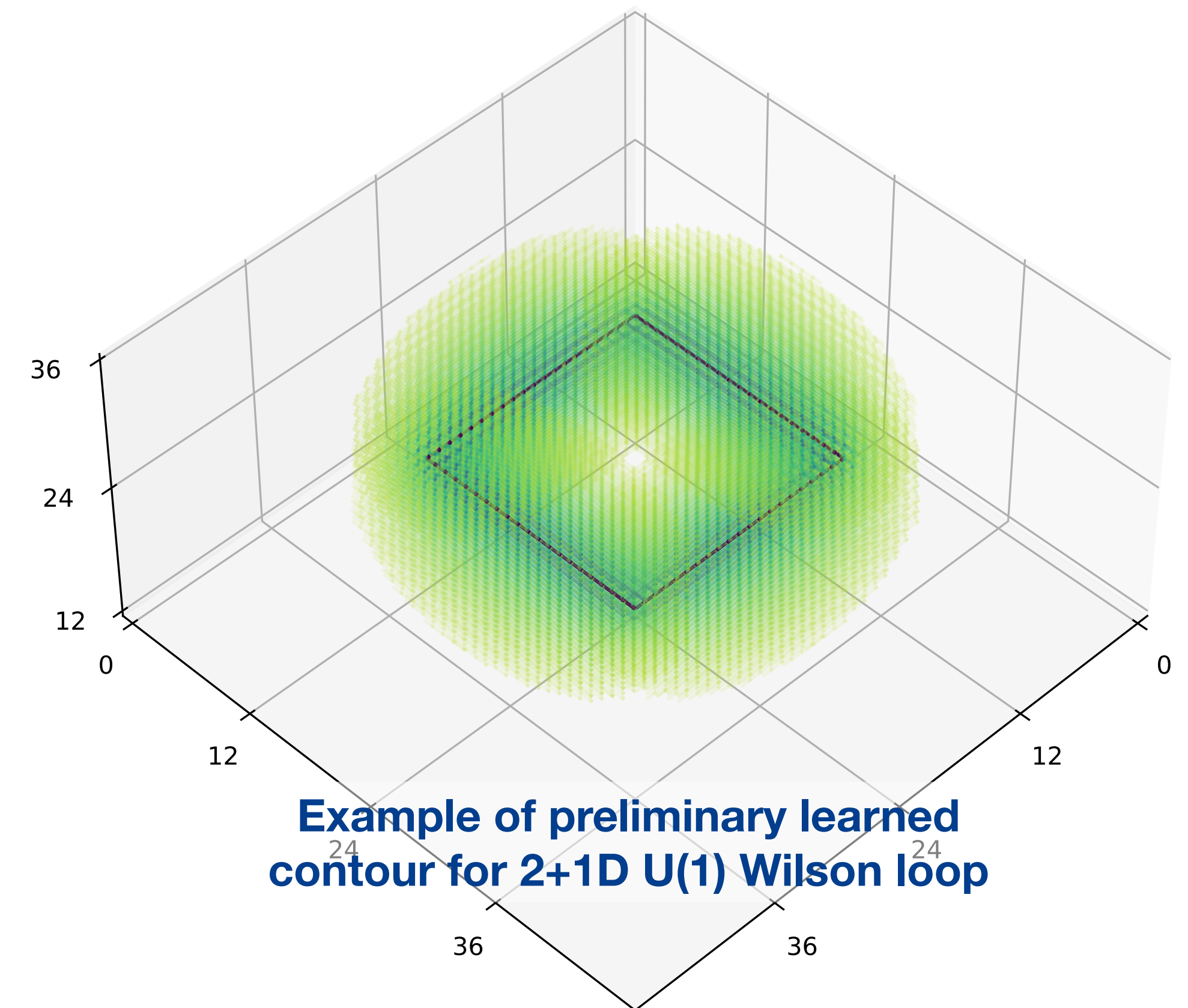
# Summary

Thanks!  
Questions?

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.

**Future work:** 3D & 4D, fermions, more advanced ML techniques?



# Backup Slides

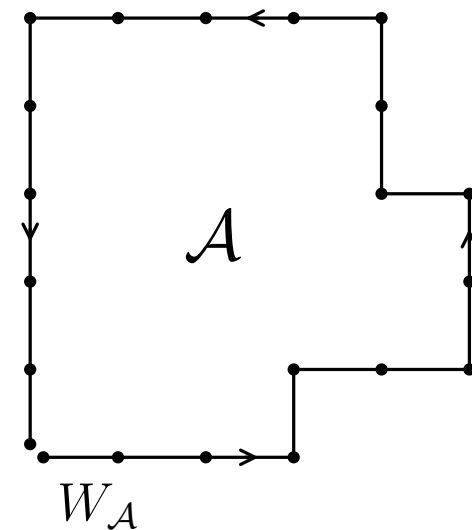
# Application to 1+1D $U(1)$ gauge theory

Rewrite path integral in terms of **plaquette angles**

$$\langle \mathcal{O} \rangle = \frac{1}{Z} \int \prod_x dP_x \mathcal{O}[P] e^{-S[\arg(P)]}$$

$$S[\phi] = -\beta \sum_x \cos(\phi_x)$$

**Wilson loops** are a sign-afflicted observable



$$W_{\mathcal{A}} = \left[ \prod_{x \in \mathcal{A}} P_x \right] = e^{i \sum_{x \in \mathcal{A}} \theta_x}$$

Four lattice spacings

$\sigma$	$V$	$g$	$\beta$
0.4	64	0.52	1.843
0.3	64	0.47	2.296
0.2	64	0.40	3.124
0.1	64	0.30	5.555

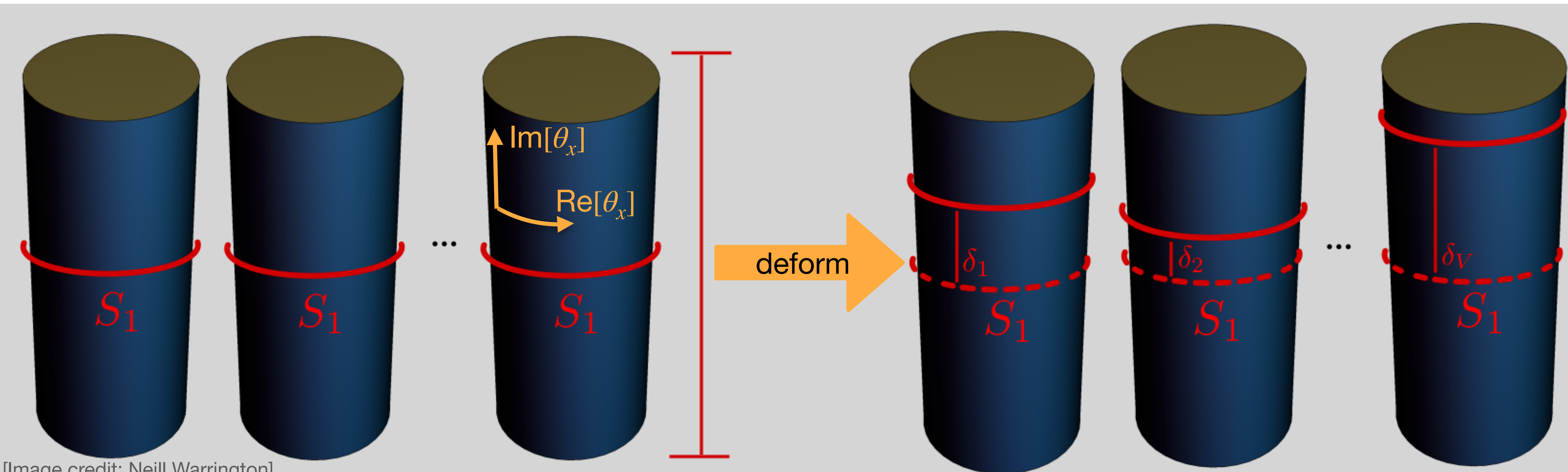


# U(1) deformation

Shift plaquette integral in complex direction

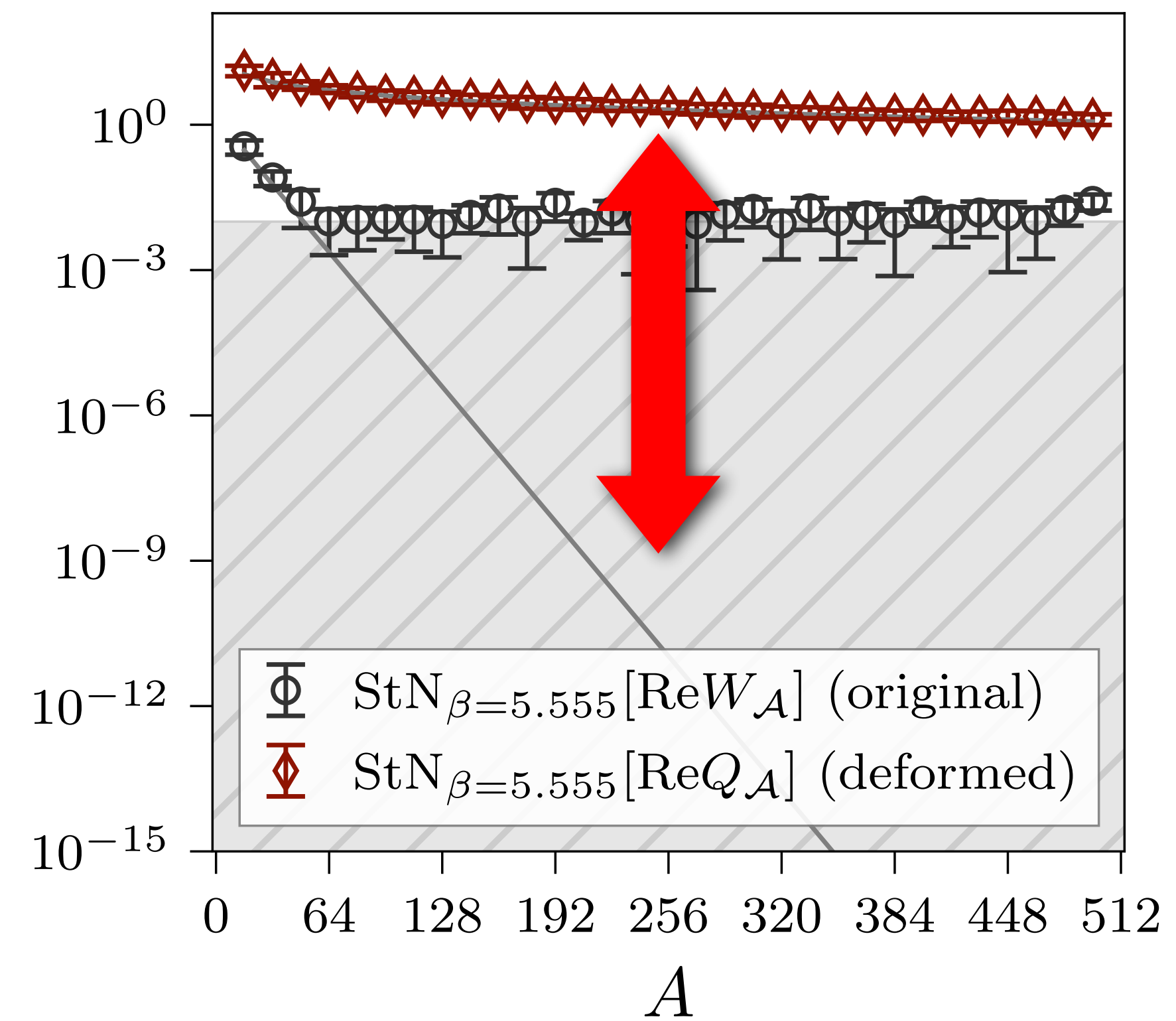
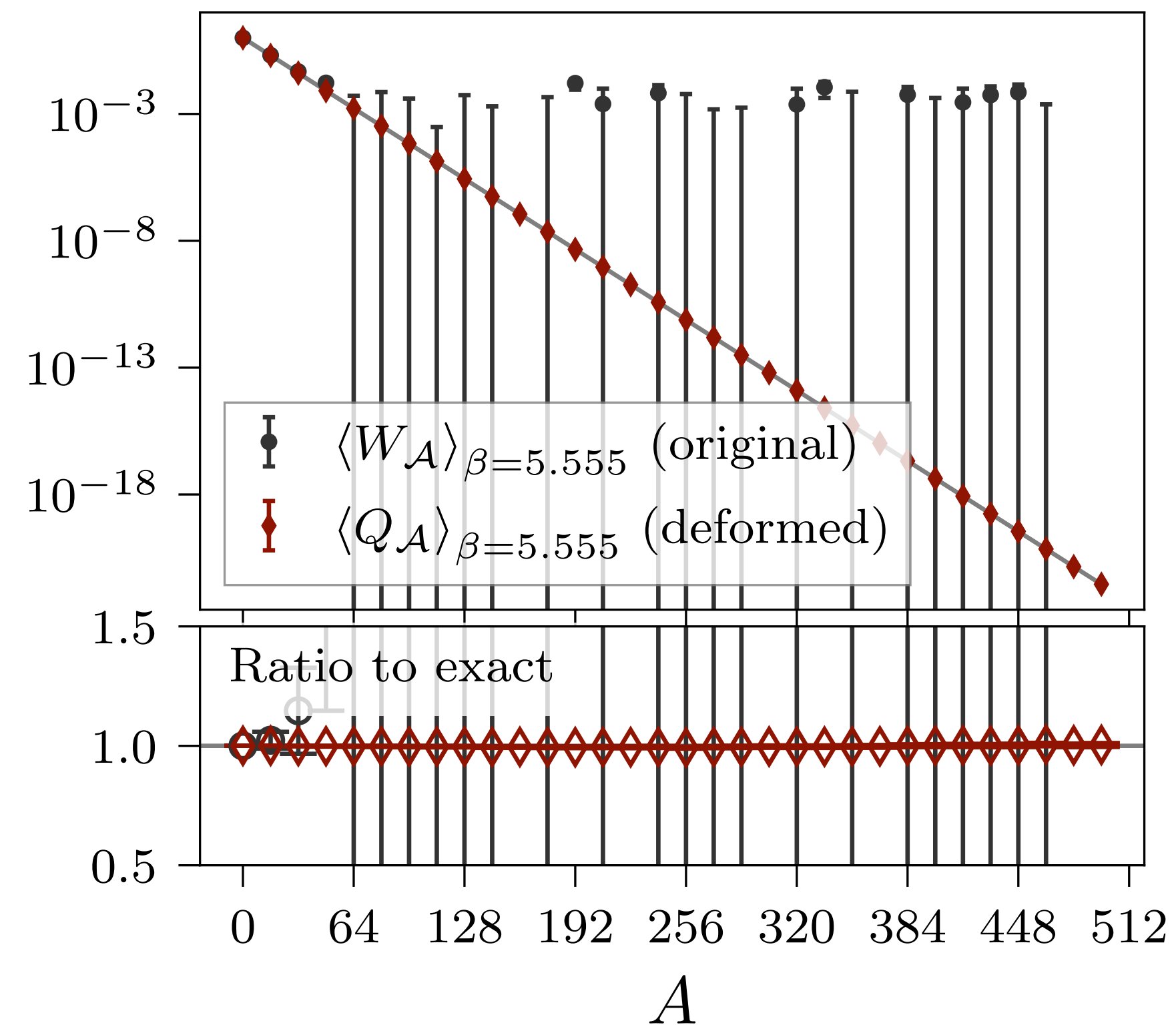
$$\tilde{\phi}_x = \phi_x + i\delta_x$$

Trivial Jacobian!



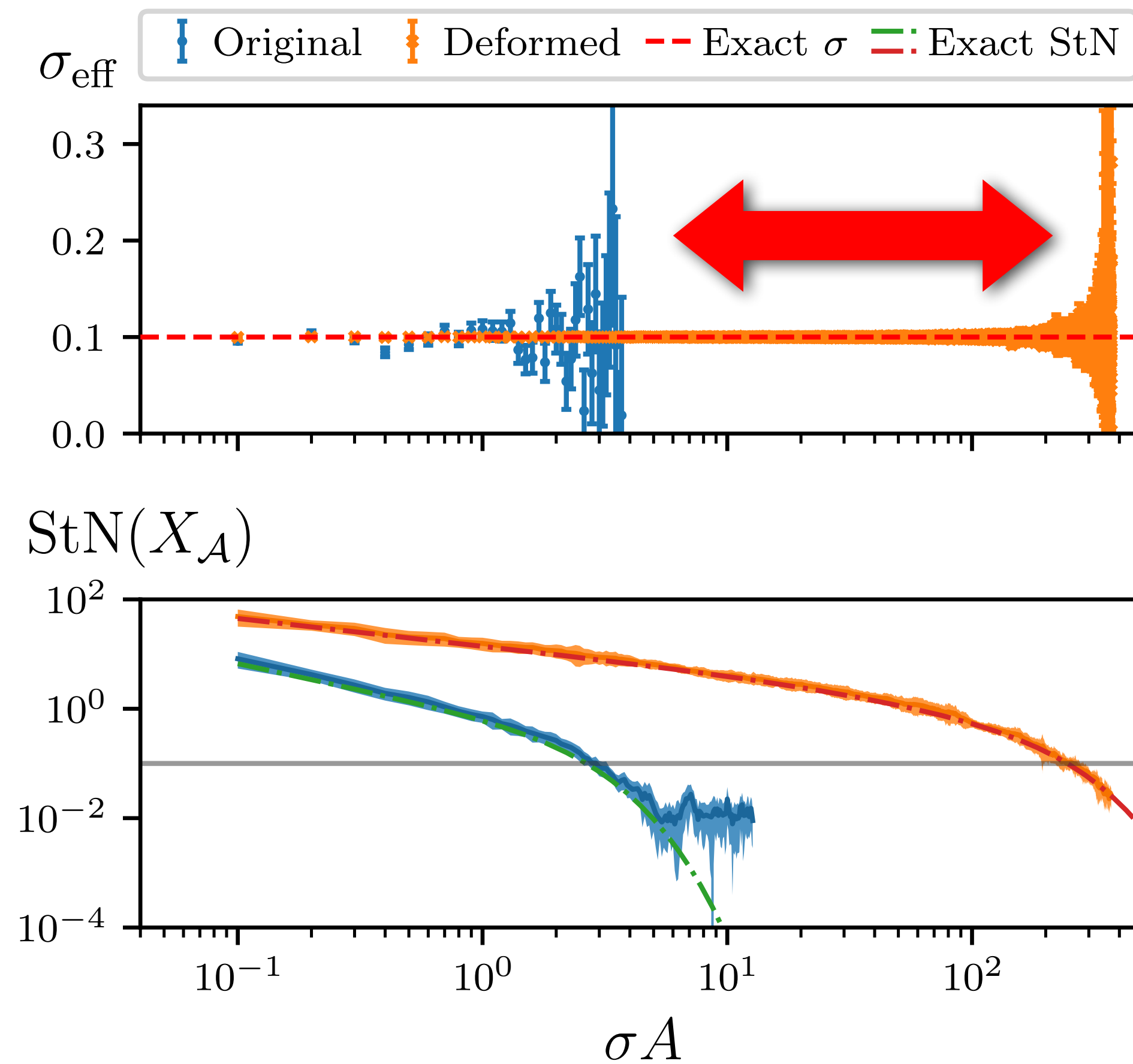
# $U(1)$ results

Exponential improvement in variance:



# $U(1)$ results

Much longer plateaus in effective string tension  $\sigma_{\text{eff}}$

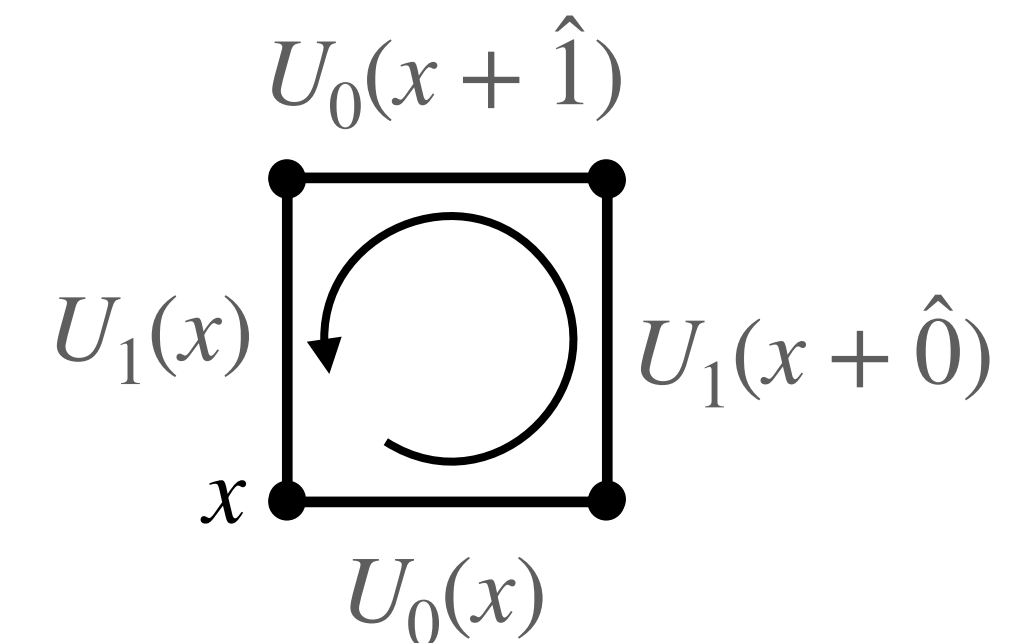


# 1+1D $SU(N)$ gauge theory

Simple action for the theory given in terms of **plaquettes**

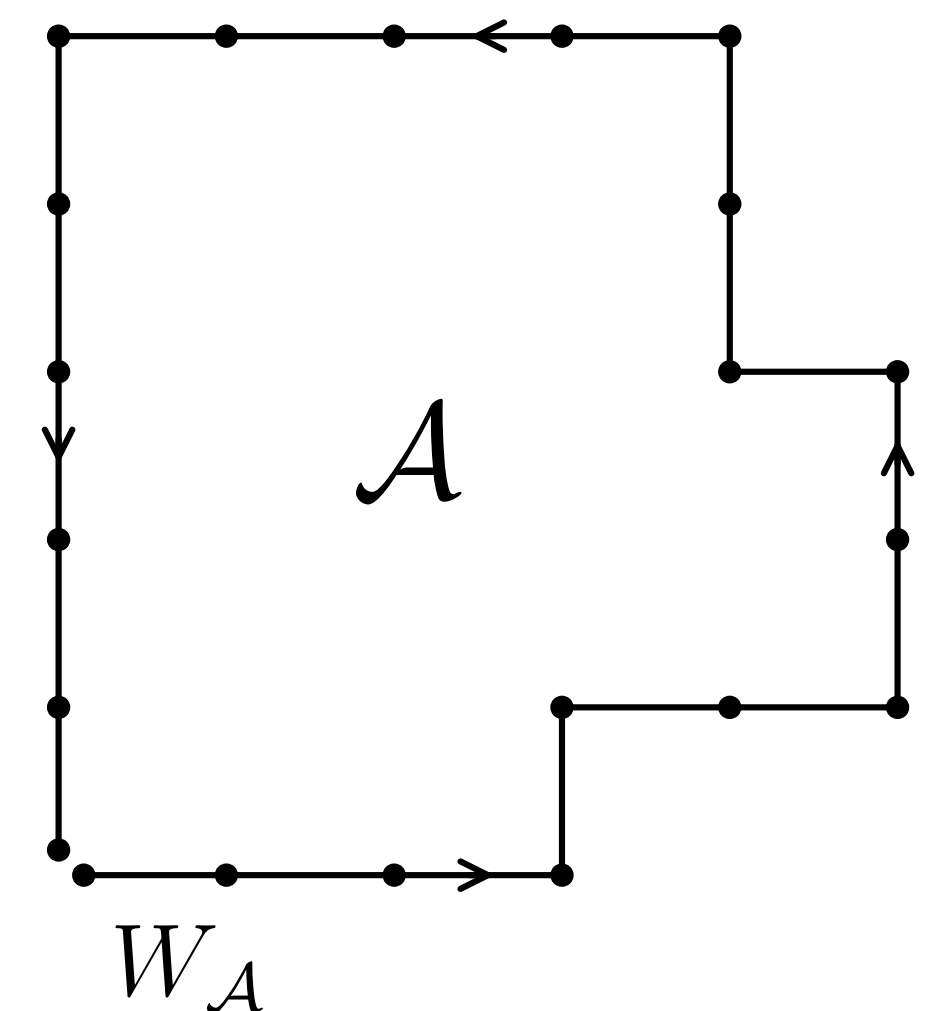
$$S[P] = -\frac{1}{g^2} \sum_x \text{tr}(P_x + P_x^{-1}) \xrightarrow{\text{ctm.}} \frac{1}{2g^2} \text{tr}(F^2)$$

$$P_x = U_0(x)U_1(x + \hat{0})U_0^\dagger(x + \hat{1})U_0^\dagger(x)$$



**Wilson loop** observables:

- Generalize plaquettes to arbitrary loops
- Probe of confinement (area vs perimeter law scaling)

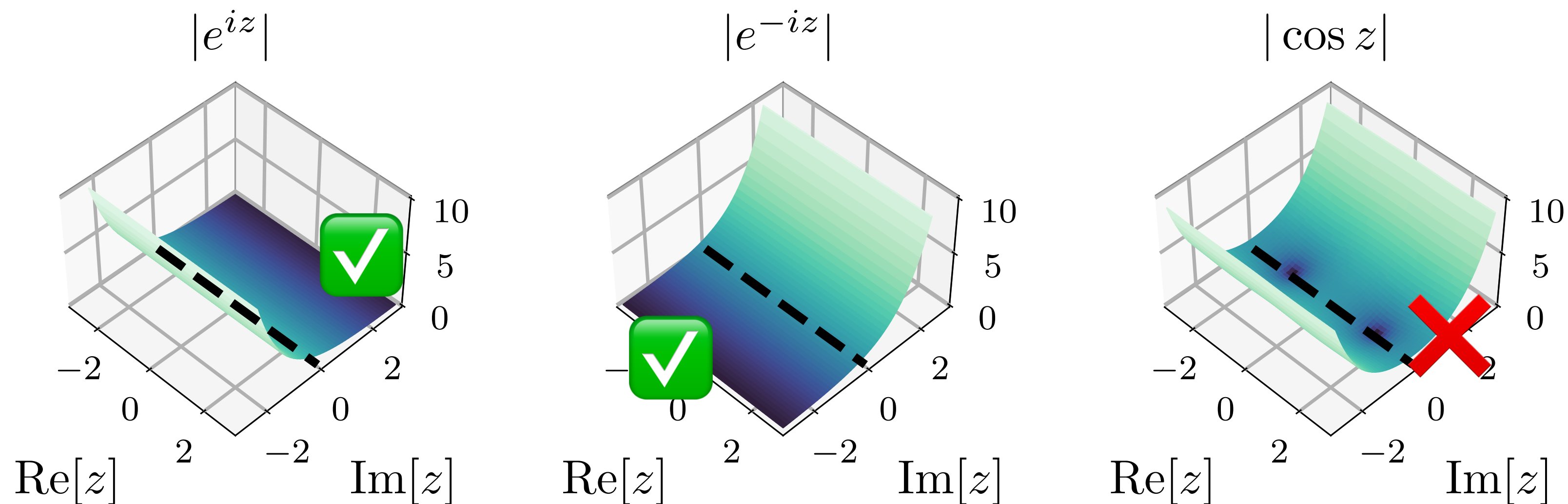


# Why select $W_{\mathcal{A}}^{11}$ ?

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!





# 1+1D $SU(N)$ gauge theory

Path integral can be written over plaquettes directly

$$\langle \mathcal{O} \rangle = \frac{1}{Z} \int \prod_x dP_x \mathcal{O}[P] e^{-S[P]} \quad S[P] = -\frac{1}{g^2} \sum_x \text{tr}(P_x + P_x^{-1})$$

Analytic continuation straightforward with the complexification  $SU(N) \rightarrow SL(N, \mathbb{C})$ .

Singularities of  $P_x^{-1}$  easily avoided by contour parameterization.

We study three lattice spacings and both  $SU(2)$  and  $SU(3)$ :

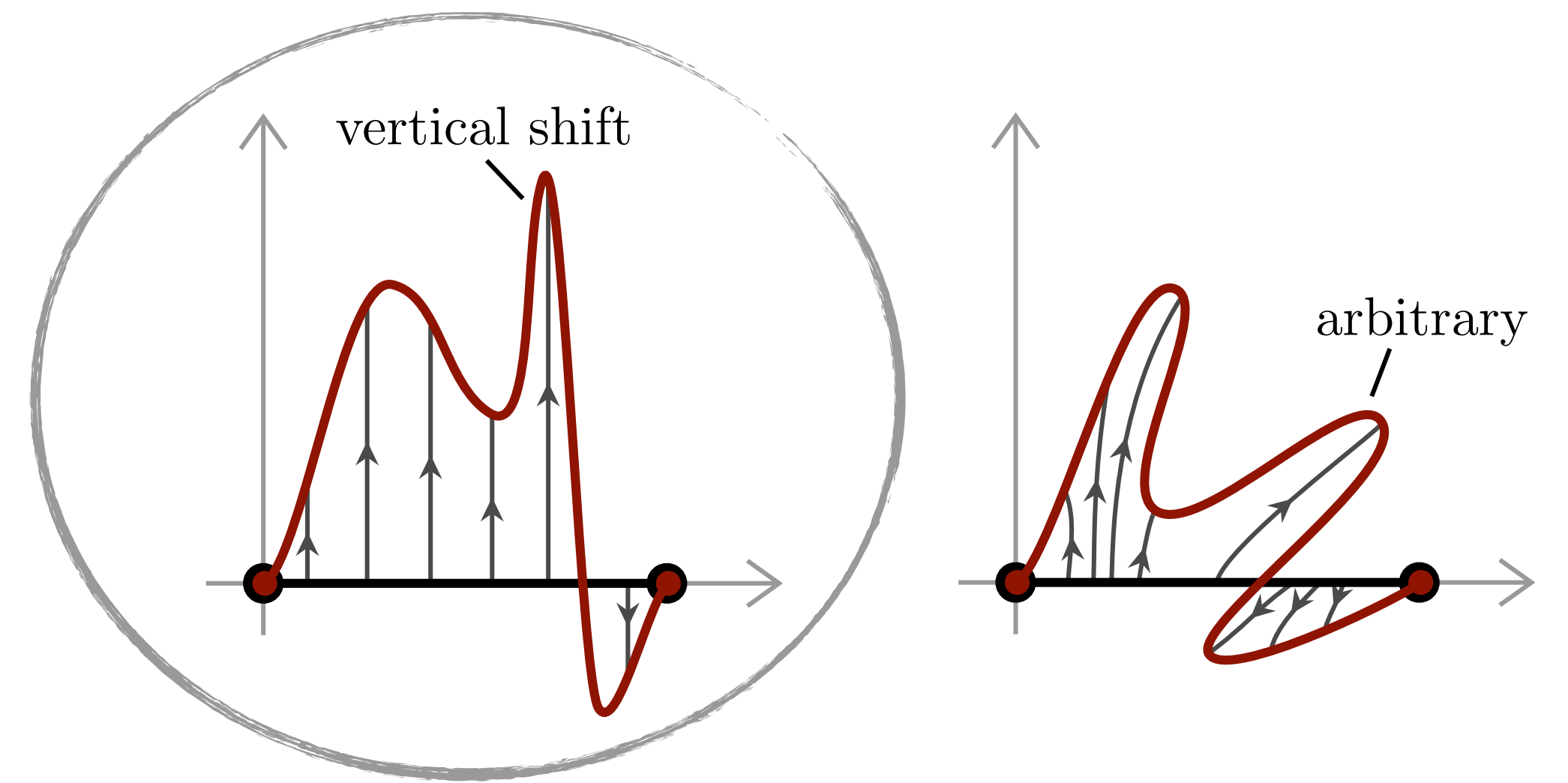
$\sigma$	$V$	$SU(2)$		$SU(3)$	
		$g$	$\beta$	$g$	$\beta$
0.4	16	0.98	4.2	0.72	11.7
0.2	32	0.71	8.0	0.53	21.7
0.1	64	0.51	15.5	0.38	41.8

(Parameters chosen to fix  $\sigma V$ )

# Deformation

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

[Alexandru et al. PRD97 (2018) 094510]



**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

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

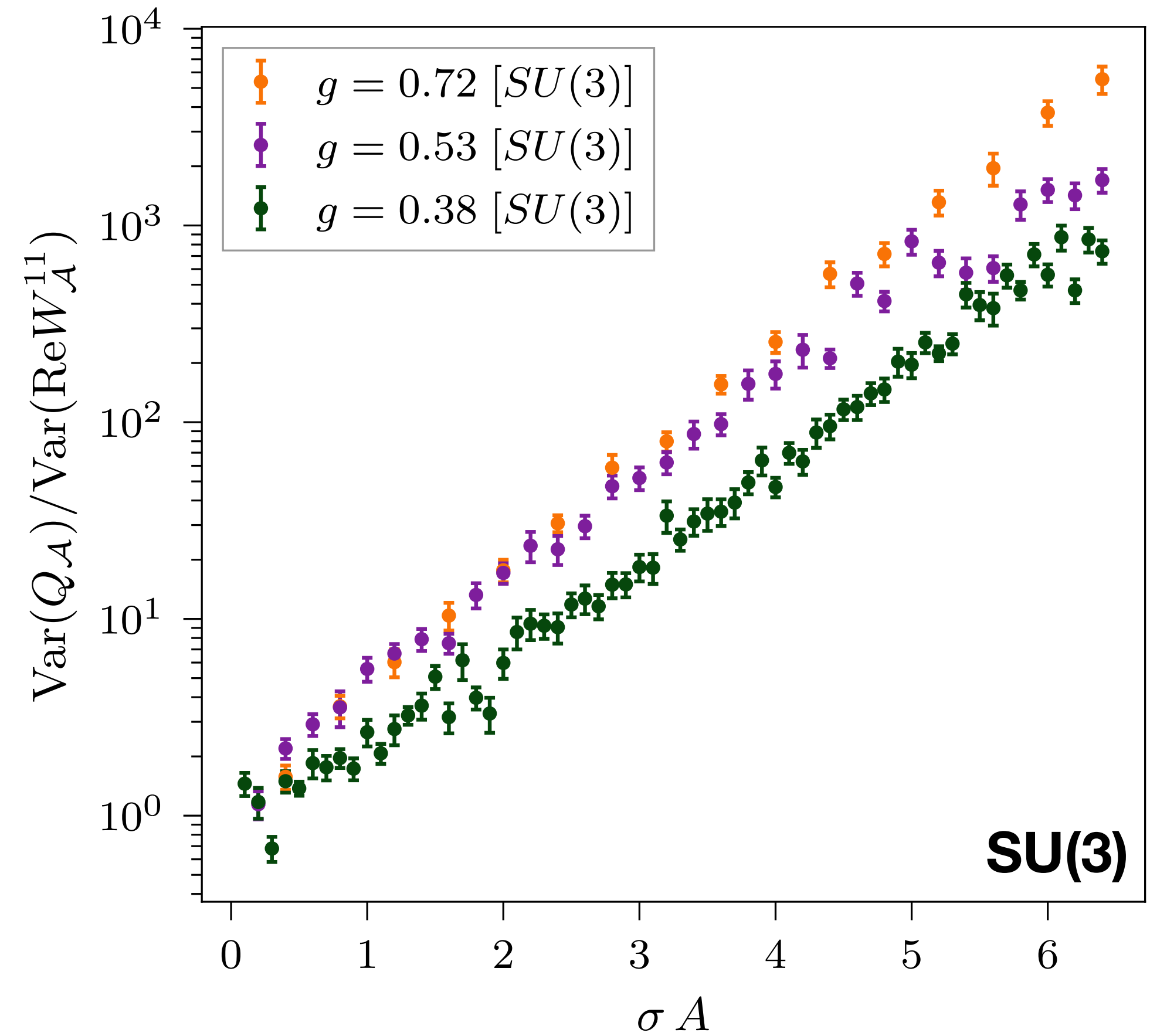
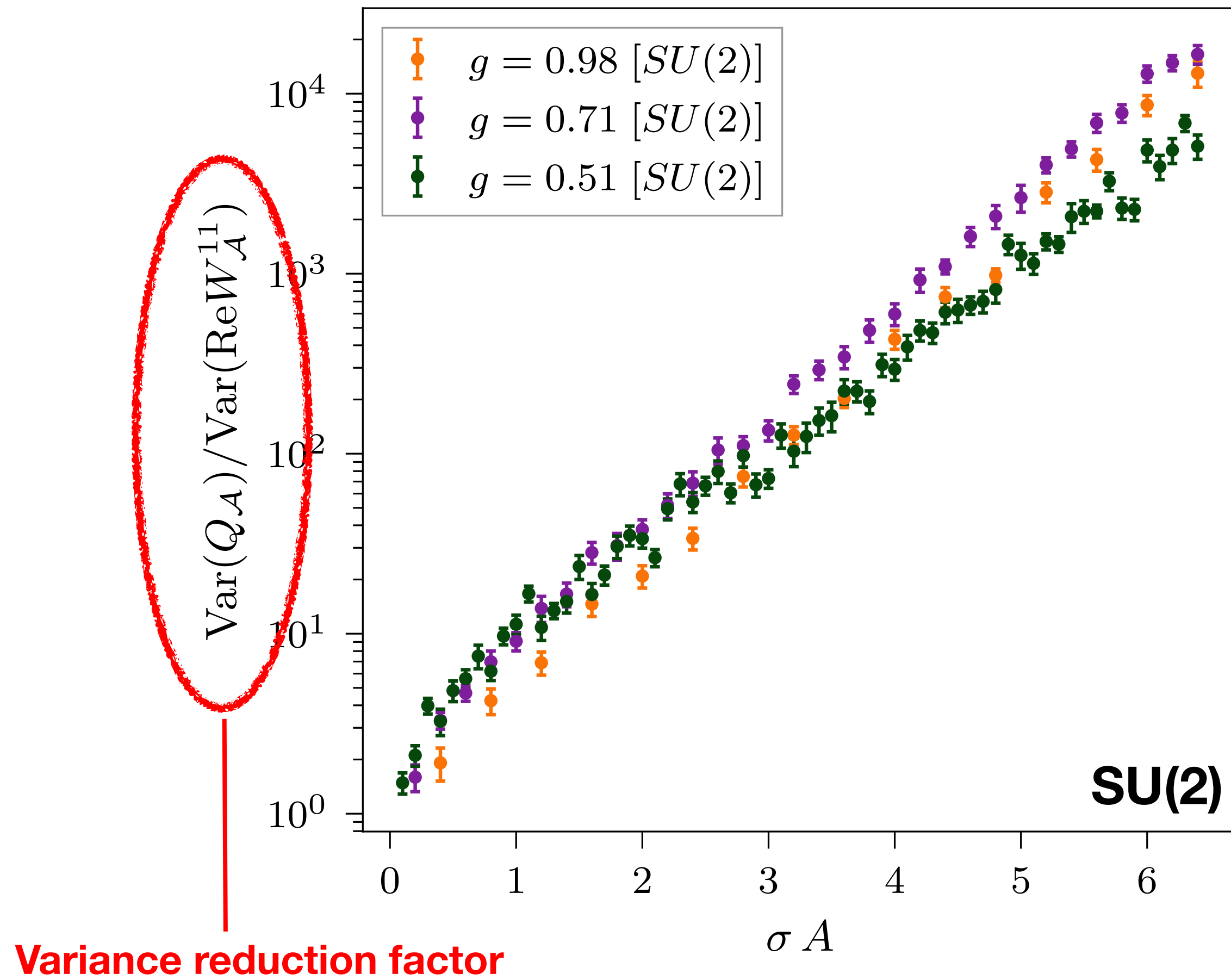
$$f_{\phi^a} = \sum_{m=1}^{\Lambda} \kappa_m^{xy;a} \sin(m\phi_y^a + \zeta_m^{xy;a}) \left\{ 1 + \sum_{n=1}^{\Lambda} \left[ \sum_{r=1}^3 \lambda_{mn}^{xy;ar} \sin(2n\theta_y^r) + \sum_{\substack{s \neq a \\ s=1}}^5 \eta_{mn}^{xy;as} \sin(n\phi_y^s + \chi_{mn}^{xy;as}) \right] \right\}.$$

**Triangular Jacobian:**  $f(\Omega)$  only allowed to depend on  $y \leq x$ . Jacobian determinant calculable from diagonal elements.

← This is key for scalability

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

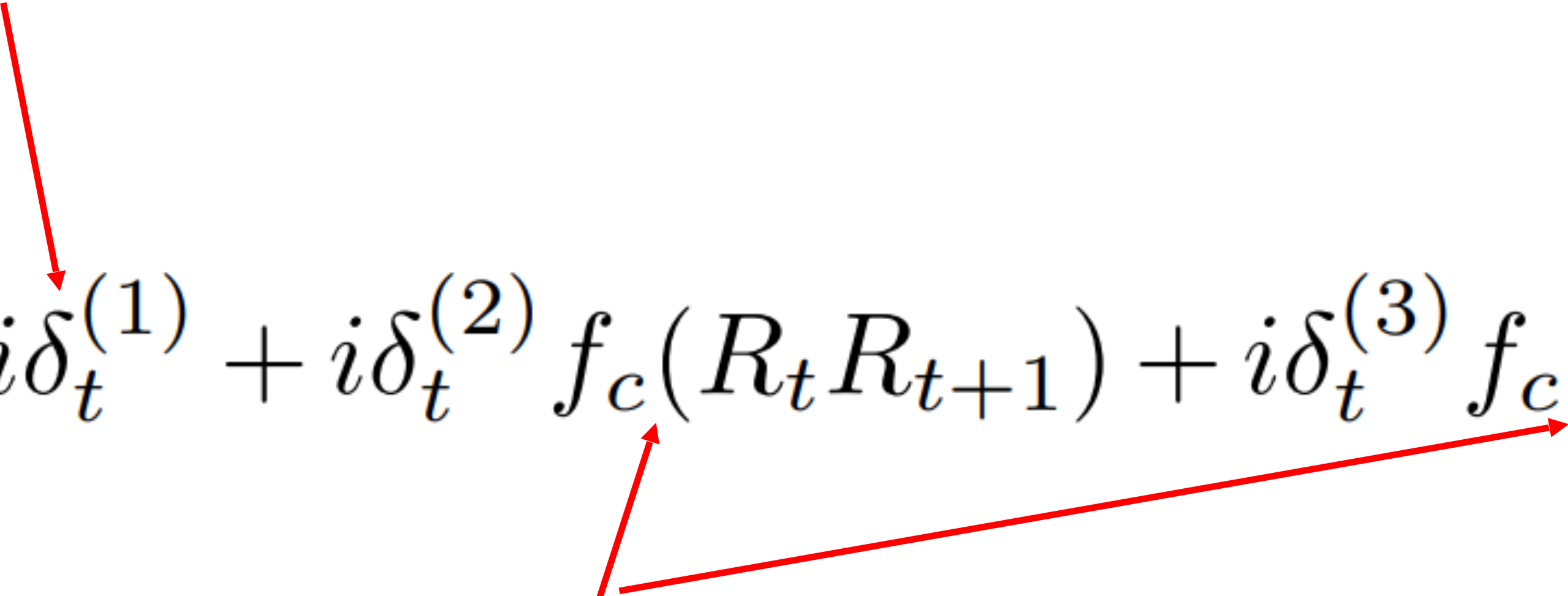
$$V(R) = \sum_t (2 + m^2) R_t^2 + \lambda R_t^4$$

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 $\phi^4$ correlators

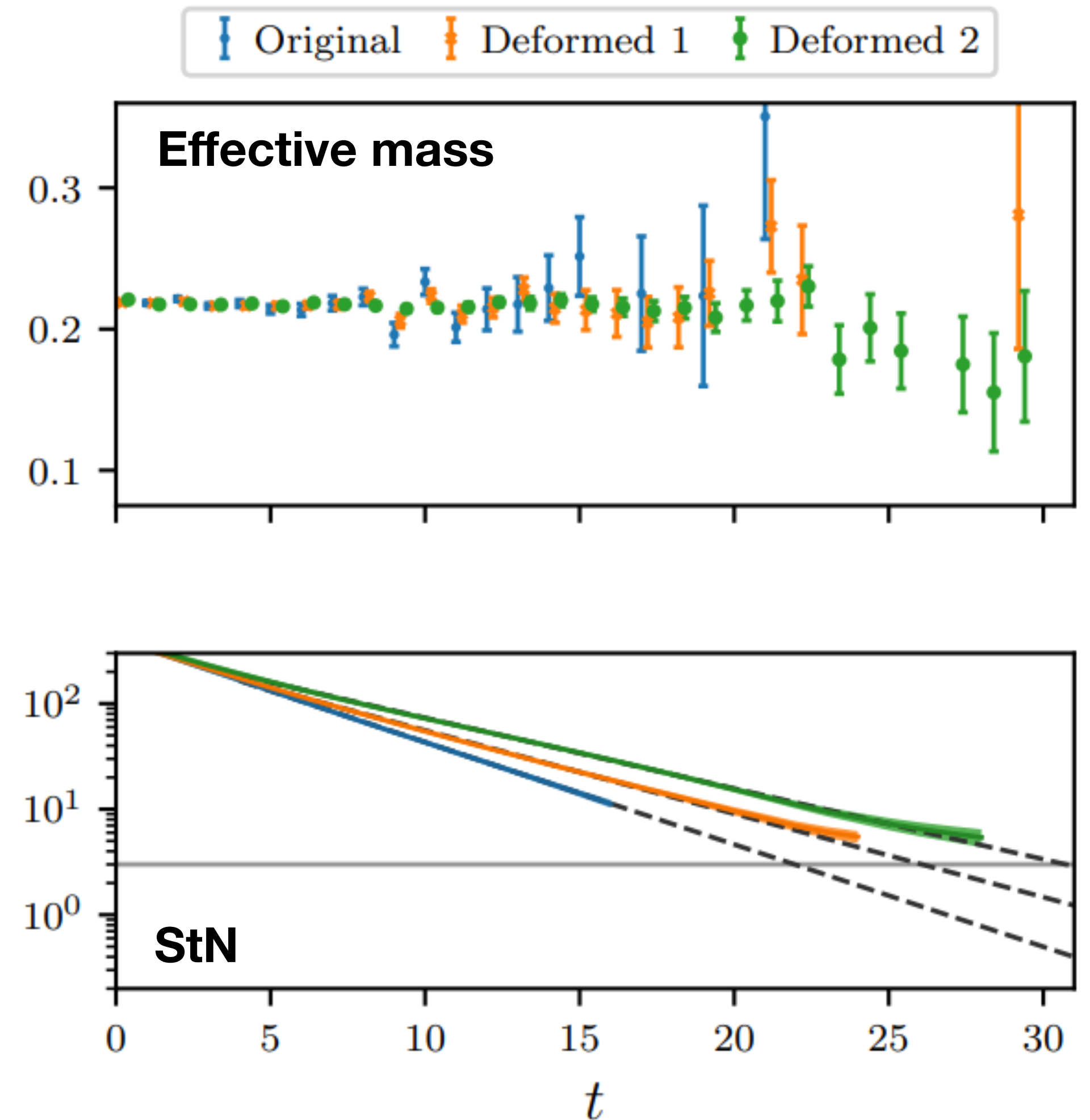
Using polar dofs in the path integral

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

and the holomorphic action

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

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

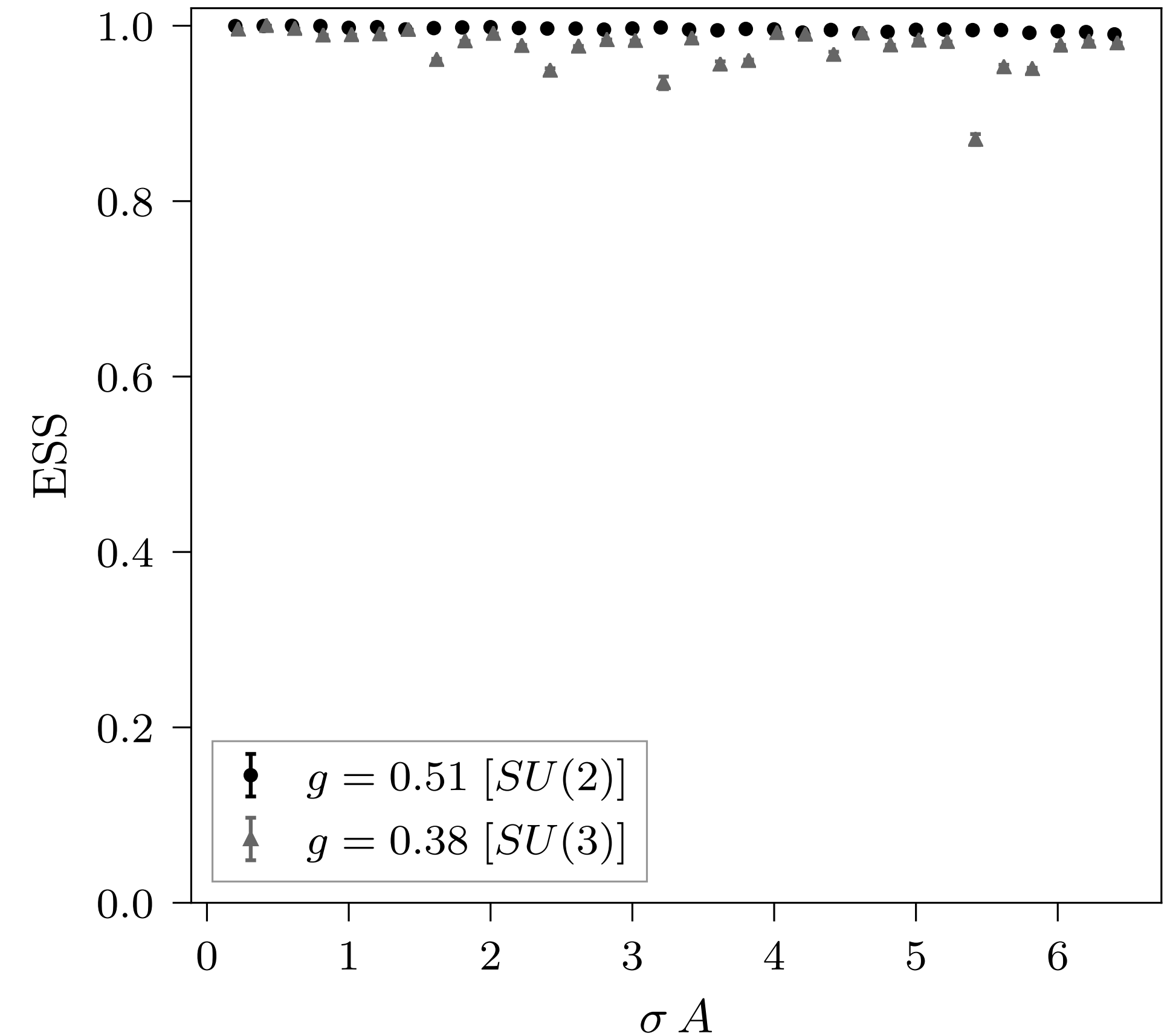


# Overlap

“Effective sample size” (ESS) measures degree of overlap between MC distribution and observable path integral

- ESS = 1 implies observable magnitude is always 1, sample-by-sample, i.e. best possible importance sampling

We observe ESS consistently  $\gtrsim 80\%$



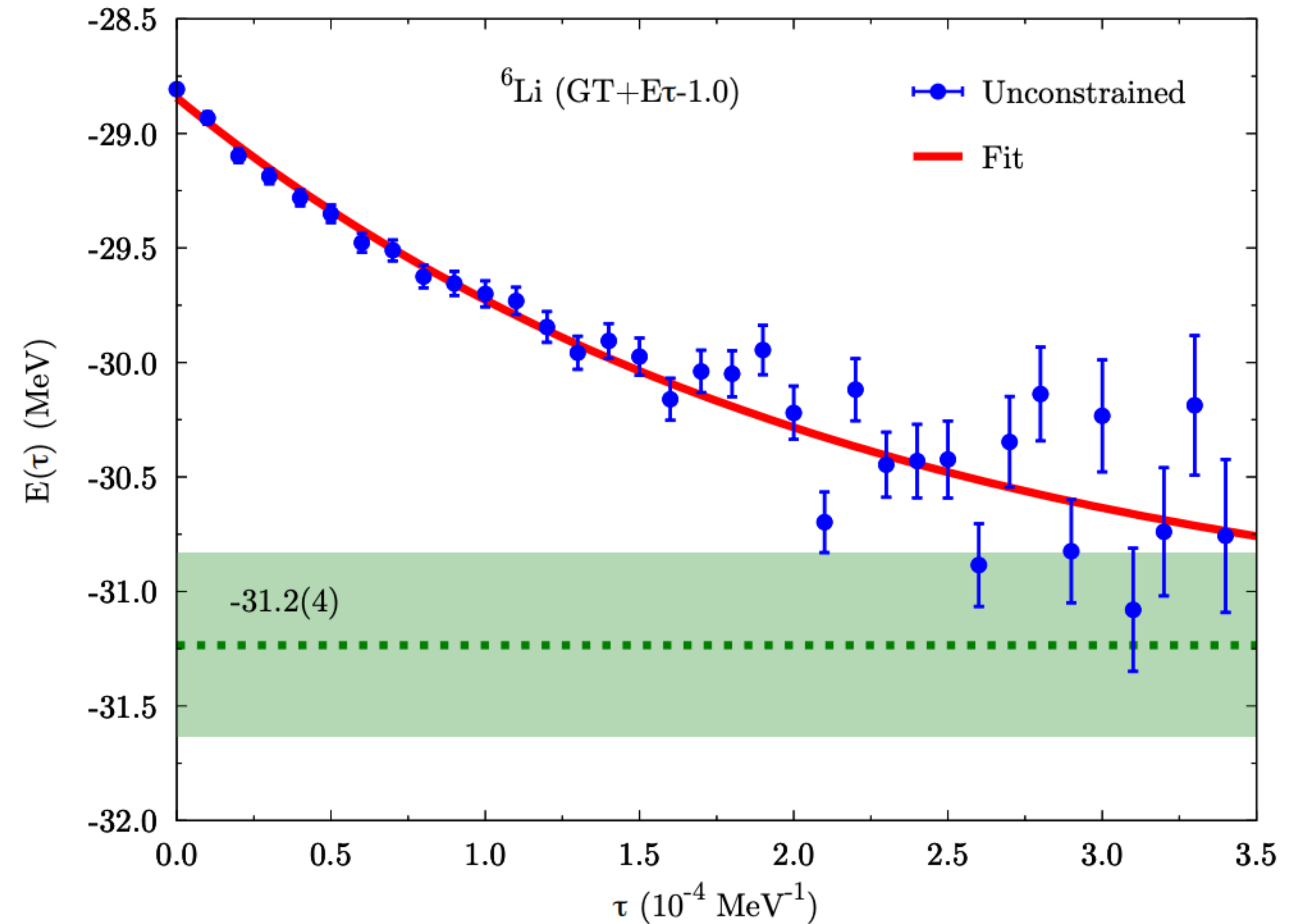
# Non-lattice applications: GFMC/AFDMC

[Gandolfi, et al. 2001.01374]

**Large  $t$ :** StN decays exponentially

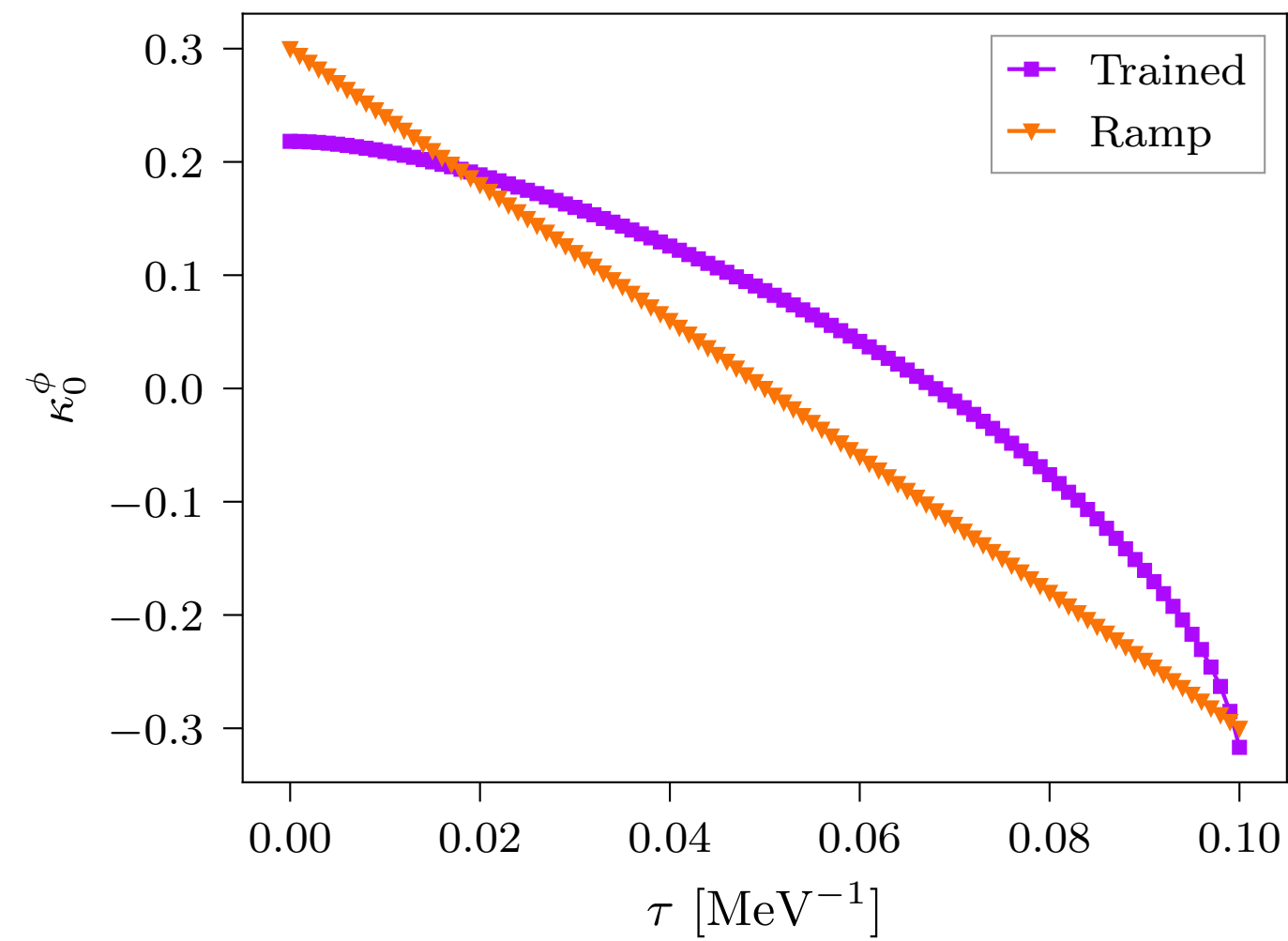
**Small  $t$ :** Excited state effects

To extract physical information,  
fit excited state model

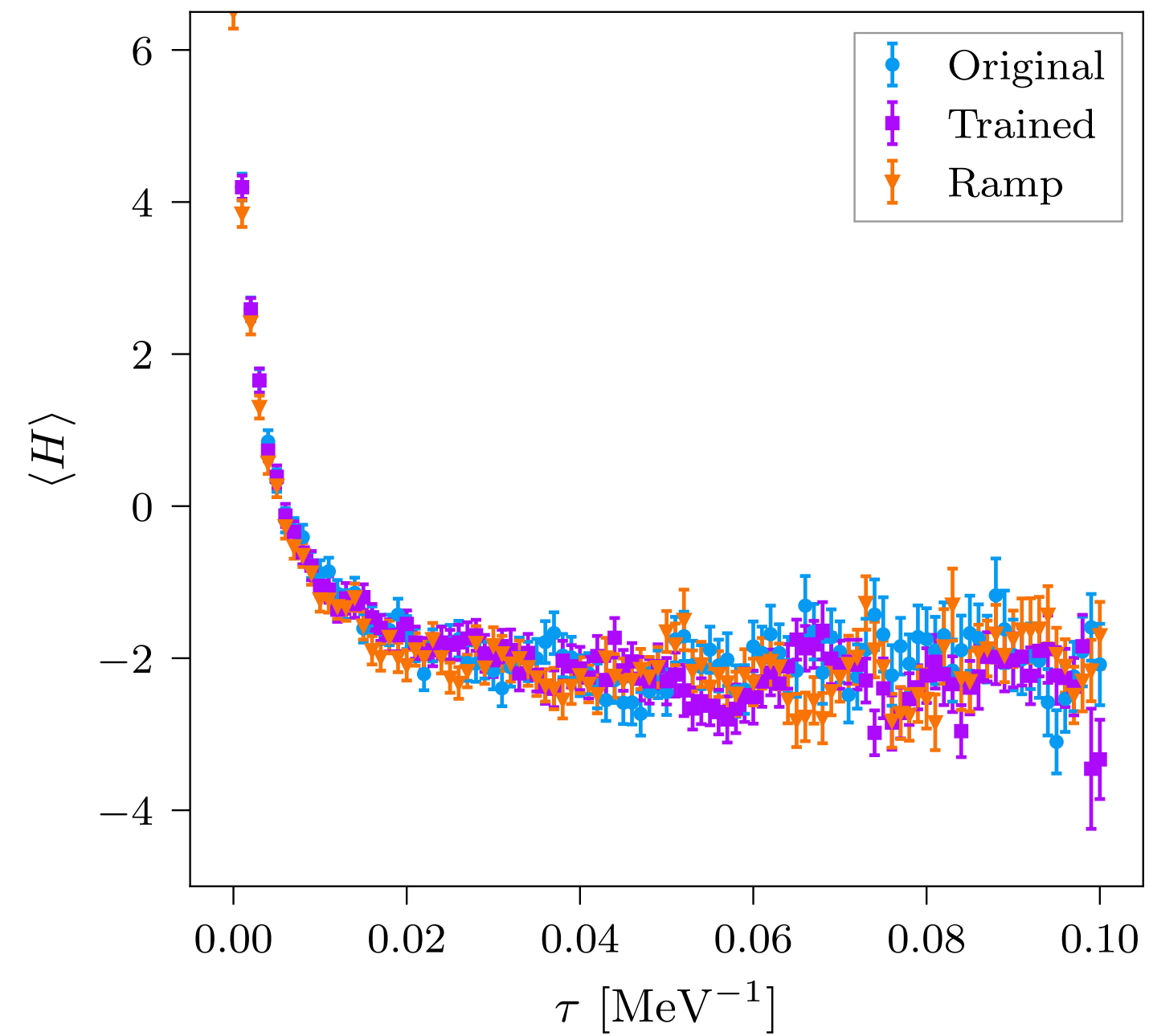


# GFMC results

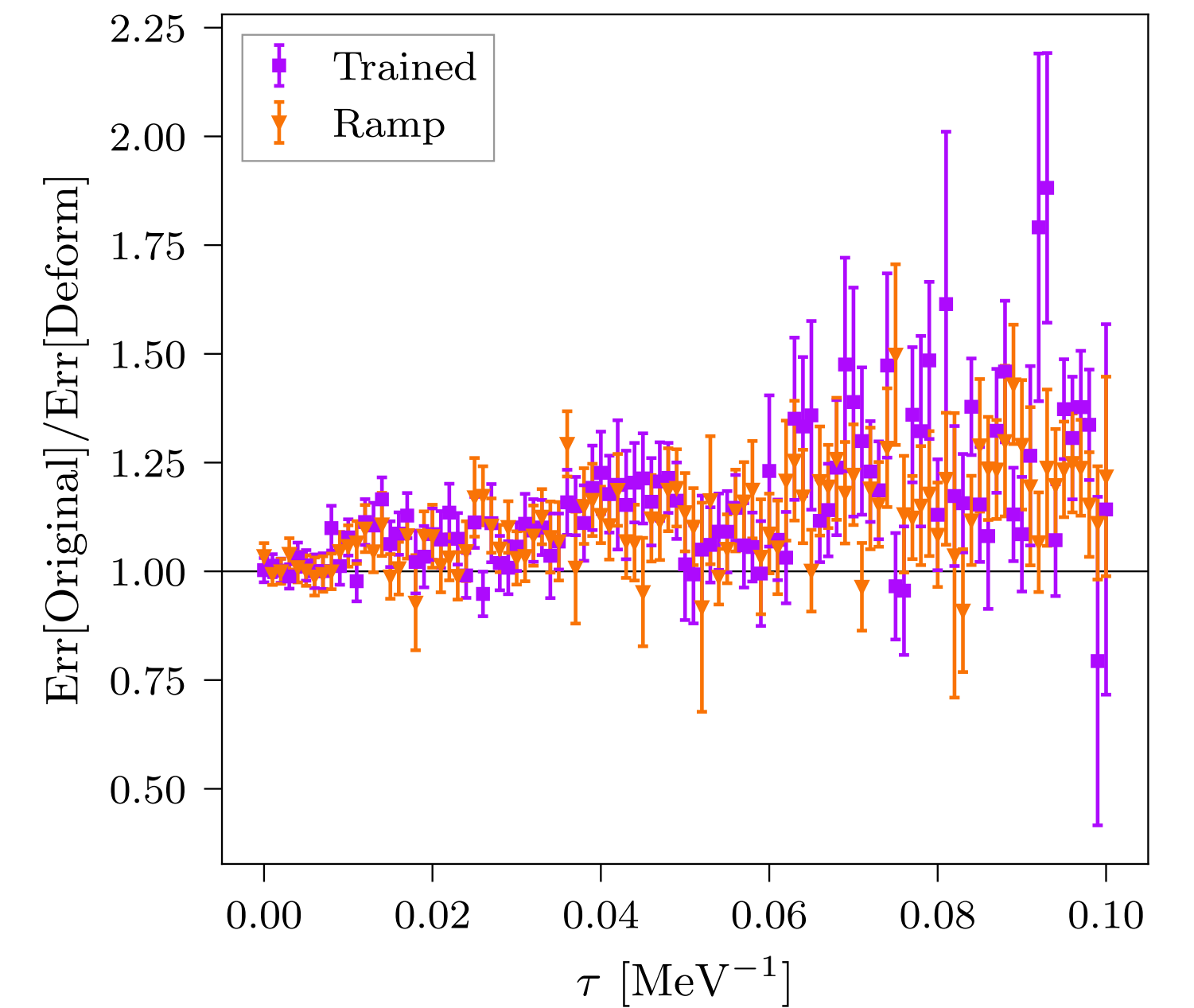
Learned imaginary shift  
vs Euclidean time



Measured deuteron  
binding energy



Improvement ratio

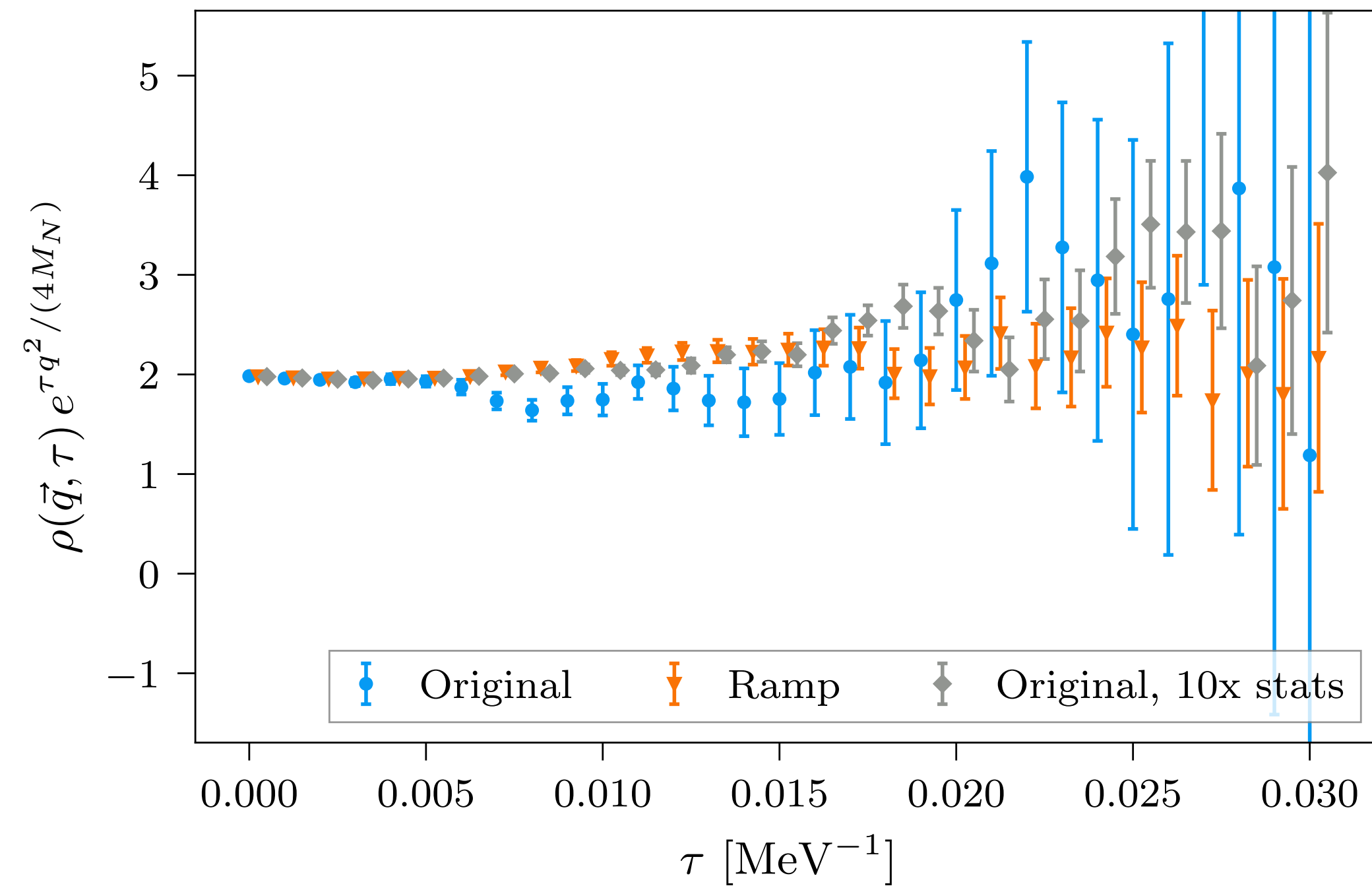


No spectacular results for  $\langle H \rangle$ , but ...

# GFMC results

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

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



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

