ERROR ANALYSIS FROM STOCHASTIC LOCALITY AND MASTER FIELDS

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STOCHASTIC LOCALITY

[Lüscher '17]

Observable $\mathcal{O}(s,x)$ at Monte Carlo time s at position x_{μ} define true expectation value $\langle \mathcal{O} \rangle$ define $\langle\!\langle \overline{\mathcal{O}} \rangle\!\rangle = \frac{1}{VN} \sum_{s,x} \mathcal{O}(s,x)$

Stochastic locality implies $\lim_{V \to \infty} \langle\!\langle \overline{\mathcal{O}} \rangle\!\rangle = \langle \mathcal{O} \rangle$ at fixed N parallelism to infinite statistics, $N \to \infty$

$$\begin{split} \langle\!\langle \overline{\mathcal{O}} \rangle\!\rangle &\text{ is our best estimator for } \langle \mathcal{O} \rangle \\ &\text{ its error from } \langle \left[\langle\!\langle \overline{\mathcal{O}} \rangle\!\rangle - \langle \mathcal{O} \rangle \right]^2 \rangle = \frac{1}{VN} \sum_{s,x} \Gamma(s,x) \\ &\Gamma(s,x) = \langle \left[\mathcal{O}(s,x) - \langle \mathcal{O} \rangle \right] \left[\mathcal{O}(0) - \langle \mathcal{O} \rangle \right] \rangle \end{split}$$



MASTER FIELDS

Translation invariance for error estimates

 \rightarrow unlocks the possibility to simulate very large lattices $m_\pi L \geq 12$ very expensive, likely small number of configs/fields

Master-field physics program

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at fine lattice spacing suppression of topology freezing effects at intermediate lattice spacing physics suffering from volume effects
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i. magnetic background fields, fine momentum resolution

ii. intermediate windows HVP, finite vol. QED effects

iii. spectral reconstructions [talk by Hansen]

solutions to inverse laplace overcome 4-particle limit possible w/ smeared kernel, e.g. $\frac{\epsilon}{\omega^2+\epsilon^2}$

ordered double limit $V \to \infty, \epsilon \to 0$

iv. ...



Master-field errors - I

In practice our best estimator for Γ is given by

$$\begin{split} \overline{\Gamma}(s,x) &= \tfrac{1}{V} \tfrac{1}{N-s} \sum_{s',x'} \delta \mathcal{O}(s+s',x+x') \delta \mathcal{O}(s',x') \\ & \text{w}/\ \delta \mathcal{O}(s,x) = \mathcal{O}(s,x) - \langle\!\langle \overline{\mathcal{O}} \rangle\!\rangle \end{split}$$

For large volumes, likely few fields, i.e. N small pre-average all fields, $\overline{\mathcal{O}}(x) = \frac{1}{N} \sum_{s} \mathcal{O}(s, x)$

equivalent to $\overline{\Gamma}(x) = \frac{1}{N} \sum_{s} \overline{\Gamma}(s, x)$

truncated sum over space-time approximates the error

$$\sigma^{2} = \frac{1}{V} \left[\sum_{|x| \le r} \overline{\Gamma}(x) + O(e^{-mr}) + O(V^{-1/2}) \right]$$

define
$$\overline{C}(r) = \sum_{|x| \le r} \overline{\Gamma}(x)$$

[Lüscher '17]



Master-field errors - II

In traditional simulations normally we perform volume average

$$\begin{array}{l} \text{take } \langle\!\langle O(s)\rangle\!\rangle = \frac{1}{V} \sum_x \mathcal{O}(s,x) \\ \text{its error given by } \Gamma(s) = \frac{1}{V} \sum_x \Gamma(s,x) \\ \text{truncate sum over } s \text{ to maximal window } W \\ \text{and the error of error is } O(\sqrt{W/N}) \end{array} \tag{Madras, Sokal '88}$$

Extend Γ -method for derived observables [e.g. Hernandez et al. '21] linear error propagation $(\delta F)^2 = \sum_{\alpha\beta} \frac{\partial f}{\partial \mathcal{O}_{\alpha}} \Gamma_{\alpha\beta} \frac{\partial f}{\partial \mathcal{O}_{\beta}}$

We extended Wolff's derivation to the case of master fields

- 1. define an automatic summation radius R
- 2. quantitative grasp over error of error $O(\sqrt{R^D/V})$

Note that $\int d^Dx o \int d\Omega_D d|x||x|^{D-1}$ volumetric factor $|x|^{D-1}$ pushes optimal radius to large distance



Our Laboratory

A pre-master-field setup to test methods and ideas

Lattices w/
$$N_f = 2 + 1$$
 $m_{\pi} = 293 \text{ MeV}, \ a \approx 0.094 \text{ fm}$ $L/a = 32, 48, 64 \rightarrow m_{\pi}L = 4.5, 6.7, 9$ approx. 100 configs for each ensemble

Observables:

- 1. energy density at positive flow time E_t 1 point function: cheap, precise, known for all points
- 2. fermionic two-point correlators $\langle \operatorname{Tr} \left[\Gamma D^{-1}(x,y) \Gamma D^{-1}(y,x) \right] \rangle$ $\Gamma = \gamma_5, \gamma_x, \gamma_y, \gamma_z$ zero momentum, depends on source-sink sep $|x_0 y_0|$ position space, depends on radius $r^2 = \sum_{\mu} (x_{\mu} y_{\mu})^2$



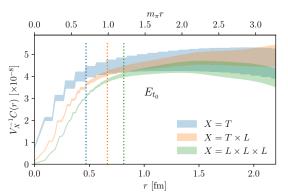


Master-field analysis

We study variance
$$C(r) = \sum_{|x| < r} \langle \left[\mathcal{O}(x) - \langle \mathcal{O} \rangle \right] \left[\mathcal{O}(0) - \langle \mathcal{O} \rangle \right] \rangle$$

$$\lim_{N \to \infty} \frac{1}{N} \sum_{s} \sum_{|x| < r} \overline{\Gamma}(s, x) = C(r)$$

Monte-Carlo sampling of $C \to \mathrm{stat.}$ error of C as error of error master-field analysis along T, TL, L^3 (average other dimensions)



master-field error possible along any direction

$$m_{\pi}L \simeq 6.7$$

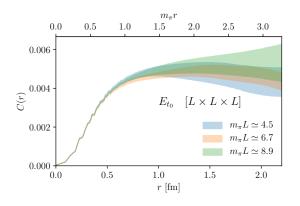
volumetric factor pushes saturation to

right



VOLUME SCALING

Traditional MC: Markov chain long enough to estimate autocorrelations? Master-field: is volume large enough to estimate spatial correlations?



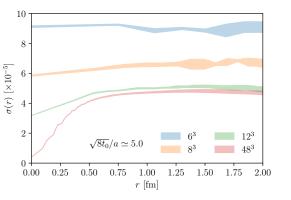
good agreement $m_{\pi}L=4.5, 6.7, 9 \text{ for } E_{t_0}$

Footprint (=r such that error staturates) is observable dependent



Gauge Noise Limit

 \mathcal{O} known on subset of points (common for fermionic observables) e.g. on a grid of points w/ equal distance



error of E_{t_0}

 $m_{\pi}L \simeq 6.7$

less points smaller accuracy

 $\begin{array}{l} {\rm master\mbox{-}field\mbox{\ }analysis} \\ {\rm in\mbox{\ }space\mbox{\ }} L^3 \end{array}$

checking saturation point useful to stop fermionic measurements similar to check scaling with number random sources





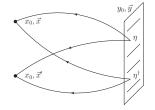


FERMIONIC CORRELATORS

Take zero mom. projection: how to study error a la master-field?

$$G(x,|x_0-y_0|) = \sum_{\boldsymbol{y}} \operatorname{Tr} \left[\Gamma D^{-1}(x,y) \Gamma D^{-1}(y,x) \right]$$

use point-sources and perform master-field error over sampled points use random-noise and perform master-field error



in time by changing position of wall source y_0

in 3D space at the sink \boldsymbol{x} , for fixed wall source

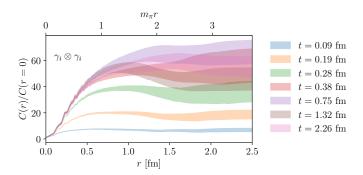
We take average over time (and hits), study MF error in 3D wall





FOOTPRINT

We study ud isovector two-point correlator (data from $m_\pi L \simeq 9$) source-sink separation t related to footprint?



Footprint grows with source-sink sep, but then saturates e.g. in master-field analysis fix summation radius to $1~\mathrm{fm}$ traditional Markov chain: "higher plateaus" \leftrightarrow larger window W

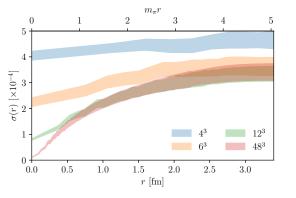






PIONS

Pions play special role, due to absence of signal-to-noise problem saturation at long distance \rightarrow large volumes mandatory \rightarrow several configurations



w/ few points reach gauge noise limit

master-field analysis less relevant in traditional simulations

 $m_{\pi}L \simeq 6.7$

Footprint depends on observable partially quenched theory to get asympt. behavior





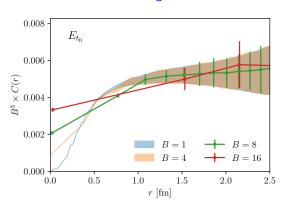
Towards master fields - I

New challenges appear in several aspects of calculation

- Stochastic locality naturally associated to coordinates correlators in position space more natural [e.g. x-space HVP by H. Meyer] truncated sum for mom. projections
- Numerical methods for fermionic correlators w/ good volume scaling stochastic sampling of point sources [e.g. HLbL by RBC/UKQCD] (stoch.) grid of points [e.g. discon. HVP by RBC/UKQCD] factorization of observables [Cé et al.]
- 3. Production of master-fields [Friztsch Lattice '21] algorithmic considerations: SMD, infinite norm [Francis et al. '19] Wilson specific feature: exponentiated clover action [Francis et al. '19]
 - \rightarrow successful generation of L^4 fields w/ $L\simeq 9,18$ fm $a\simeq 0.094$ fm, $m_\pi\simeq 270$ MeV, $m_\pi L=12,25$ on-going efforts at $a\simeq 0.065$ fm

BLOCKING

Analysis & storage costs grow w/ volume, e.g. E_t known on all points perform blocking of points from averages of sub-volumes similar to binning in traditional simulations



If blocks large enough → jackknife, bootstrap

$$m_{\pi}L \simeq 9$$

 $\begin{array}{l} {\rm master\mbox{-}field\mbox{\ analysis}} \\ {\rm in\ space\mbox{\ }} L^3 \end{array}$

blocking
$$L \to L/B$$





Towards master fields - II

[Cé Lattice 2021] 1. Position space $G_{\pi}(r) = \sum_{|x|=r} \langle \operatorname{Tr} \left[\gamma_5 D^{-1}(x,0) \gamma_5 D^{-1}(0,x) \right] \rangle$ in the continuum asympt. behavior $G_{\pi}(r) \propto \frac{m_{\pi}}{r} K_{1}(m_{\pi}r)$ not an exact representation, understand cutoff effects large finite volume contaminations for G_{π} mirror images \rightarrow easy to correct like the cosh effect along temporal direction for nucleon $N\pi$ intermediate states average together points of same size → reduce noise w.r.t. zero momentum proj. encouraging results for mass extraction of nucleon

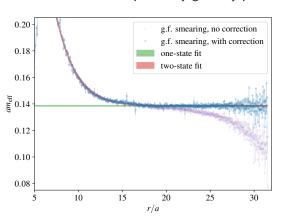
2. Grid of points w/ separation dictated by our study initial TMR study w/ wall sources relevant to get "saturation scales"





TOWARDS MASTER FIELDS - III

Hadronic masses: one possibility given by position space correlators



$$m_{\pi}L \simeq 9$$

correction from mirror images working

excited states modeled w/ secondary K_1

discretization errors under investigation



Conclusions - I

Errors a la master-field

may play interesting role w/ traditional simulations e.g. DWF simulations expensive, fewer configurations improved error estimators if consider D+1 theory? blocking, reduces analysis and storage costs standard analysis techniques, like jackknife preliminary studies on intermediate volumes identify relevant scales \rightarrow planning of new simulation

Master-field observables: re-think definition and numerical methods first tests w/ position space correlators and common TMR sampling strategies (grids, point sources) under study investigations of spectral reconstructions under way





Conclusions - II

Master-field simulations

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important product, relevant beyond master-field program m_\pi L=12 and 24 generated at coarse a for (eClover)-Wilson \to mesonic masses and decay const., spectral reconstruction interplay w/ domain decomposition will be crucial [talk by Boyle, Giusti] \to ratio network/computing power of new machines
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exponentiated Clover formulation of Wilson fermions

to really overcome topology freezing must go finer
thermalization challenge still open!
but many good ideas out there [e.g. talk by Matsumoto, Jung ...]

Thanks for your attention!!



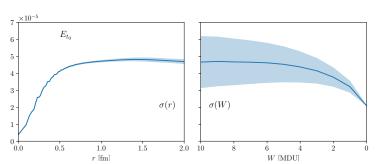


IMPROVED ERROR ESTIMATORS

Master-field analysis

 $V \simeq 48^3 \; \mathrm{points}$

Traditional Monte-Carlo analysis $N \simeq 10^2 \ {\rm confs}$



Exactly same data both plots

If autocorrelations present <code>[in prep.]</code> bin MC data w/ bin size $au_{\rm int}$ before master-field analysis or statistical analysis in full D+1 theory



