

Variance Reduction

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ALPHA
Collaboration



CLS
based



Introduction

Stochastic Estimation

Exact Smeared All-to-all

A Hybrid Approach (Stochastic LapH)

What is Variance Reduction?

Path integral formulation of QFT (finite, discrete, Euclidean space-time lattice): QFT vacuum expectation values \leftrightarrow Ensemble Averages

$$\langle 0 | \hat{\mathcal{O}}_1(x_1) \dots \hat{\mathcal{O}}_N(x_N) | 0 \rangle_{con} = \frac{1}{Z} \int \mathcal{D}\phi \mathcal{O}_1(x_1) \dots \mathcal{O}_N(x_N) e^{-S[\phi]}$$

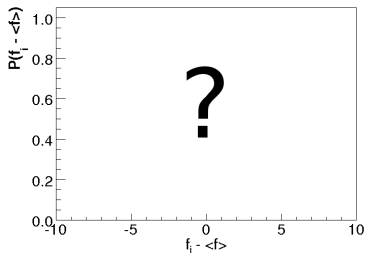
Un-biased estimator of $\langle 0 | \hat{\mathcal{O}}_1(x_1) \dots \hat{\mathcal{O}}_N(x_N) | 0 \rangle_{con}$:

$$\langle \mathcal{O}_1(x_1) \dots \mathcal{O}_N(x_N) \rangle_{N_{conf}} \equiv \frac{1}{N_{conf}} \sum_{i=1}^{N_{conf}} \mathcal{O}_1(x_1) \dots \mathcal{O}_N(x_N)[\phi_i]$$

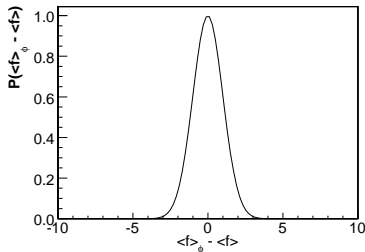
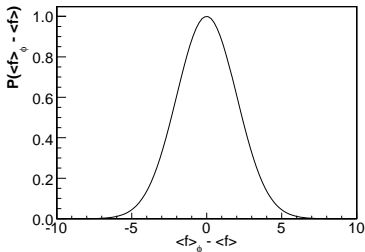
where the ϕ_i are distributed according to $e^{-S[\phi]}/Z$. Treat $f_i = \mathcal{O}_1(x_1) \dots \mathcal{O}_N(x_N)[\phi_i]$ as a random variable. Variance:

$$\begin{aligned} \text{Var}[f] &\equiv \lim_{N_{conf} \rightarrow \infty} \langle f^2 \rangle_{N_{conf}} - \langle f \rangle_{N_{conf}}^2 \\ &= \langle 0 | \hat{f}^2 | 0 \rangle_{con} - \langle 0 | \hat{f} | 0 \rangle_{con}^2 \end{aligned}$$

Central Limit Theorem: As $N_{conf} \rightarrow \infty$, distribution of $\langle f \rangle_{N_{conf}}$ approaches normal with mean $\langle f \rangle = \langle 0 | \hat{f} | 0 \rangle_{con}$ and variance $2\tau_f^{int} \text{Var}[f] / N_{conf}$.



As N_{conf} is increased, the distribution of $\langle f \rangle_{N_{conf}}$ narrows, with width $\sqrt{2\tau_f^{int} \text{Var}[f] / N_{conf}}$.



Two observables \hat{f} and \hat{g} , such that $\langle 0|\hat{f}|0\rangle = \langle 0|\hat{g}|0\rangle$ but $\text{Var}[f] > \text{Var}[g]$. We should measure g , as it has a smaller error.

Simple example (Lüscher '10, arXiv:1002.4232 [hep-lat]):
Translation Invariant observable $\hat{O}(x)$.

$$\mathcal{E} = \langle 0|\hat{O}(0)|0\rangle = \frac{1}{V} \sum_x \langle 0|\hat{O}(x)|0\rangle$$

Variance of the unsummed observable:

$$v_1 = \langle 0|\hat{O}(0)^2|0\rangle - \langle 0|\hat{O}(0)|0\rangle^2$$

Variance of the **summed observable**:

$$\begin{aligned}v_2 &= \frac{1}{V^2} \sum_{x,y} \langle 0 | \hat{O}(x) \hat{O}(y) | 0 \rangle - \frac{1}{V^2} \left(\sum_x \langle 0 | \hat{O}(x) | 0 \rangle \right)^2 \\&= \frac{1}{V^2} \left[\sum_{\substack{x,y \\ ||x-y|| < R}} \langle 0 | \hat{O}(x) \hat{O}(y) | 0 \rangle + \sum_{\substack{x,y \\ ||x-y|| > R}} \langle 0 | \hat{O}(x) \hat{O}(y) | 0 \rangle \right] - \mathcal{E}^2 \\&= \frac{1}{V^2} \left[\sum_{\substack{x,y \\ ||x-y|| < R}} \langle 0 | \hat{O}(x) \hat{O}(y) | 0 \rangle + \sum_{\substack{x,y \\ ||x-y|| > R}} \langle 0 | \hat{O}(x) | 0 \rangle \langle 0 | \hat{O}(y) | 0 \rangle \right] - \mathcal{E}^2 \\&= \frac{1}{V^2} \sum_{\substack{x,y \\ ||x-y|| < R}} \langle 0 | \hat{O}(x) \hat{O}(y) | 0 \rangle + \mathcal{O}\left(\frac{R^2}{V^2}\right) \\&\sim \frac{1}{V}\end{aligned}$$

‘Variance Reduction’: The clever construction of observables \hat{g} which attempt to minimize $Var[g]$.

Variance and the Signal-to-noise problem

Observables are typically extracted from the **temporal fall-off** of correlation functions between hadron interpolating fields.

$$C^{2pt}(t) = \langle \mathcal{O}(t) \bar{\mathcal{O}}(0) \rangle = \sum_n |A_n|^2 e^{-E_n t}$$

$$C^{3pt}(t_1, t_2) = \langle \mathcal{O}(t_1 + t_2) J(t_2) \bar{\mathcal{O}}(0) \rangle = \sum_{mn} A_m A_n^* \mathcal{M}_{mn} e^{-E_m t_1} e^{-E_n t_2}$$

$$\mathcal{M}_{mn} = \langle m | \hat{J} | n \rangle, \quad A_n = \langle 0 | \hat{\mathcal{O}} | n \rangle$$

In order to get the desired asymptotic behaviour (i.e. the 'ground state') **all t 's must be large**.

Nearly all correlation functions have **exponential decay of signal-to-noise with t** .

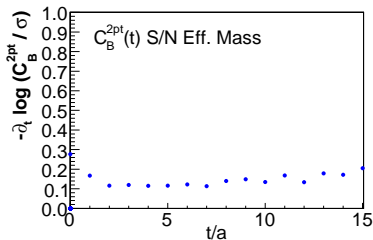
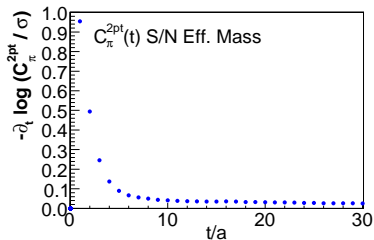
Treat the signal-to-noise ratio as a correlation function

$C(t) = S(t)/N(t)$. Examine the effective 'mass'

$$m_{\text{eff}} = \ln(C(t)/C(t+1))$$

Examine two representative cases: **Pion** and **pseudoscalar static-light** meson (CLS ensemble, $a = 0.07\text{fm}$,

$m_{\pi} = 230, 400\text{MeV}$).



How can I understand this? Lüscher '10 (arXiv:1002.4232 [hep-lat]), Lepage '89

Examine the variance of a **pion** correlation function:

$$\begin{aligned}\mathcal{O}_\pi(t) &= \sum_{\mathbf{x}} \bar{u} \gamma_5 d(\mathbf{x}, t) \\ C_\pi^{2pt}(t) &= \langle \mathcal{O}_\pi(t) \bar{\mathcal{O}}_\pi(0) \rangle \\ \text{Var}[C_\pi^{2pt}(t)] &\sim \langle \mathcal{O}_{\pi\pi}(t) \bar{\mathcal{O}}_{\pi\pi} \rangle - [C_\pi^{2pt}(t)]^2 \\ \frac{C_\pi^{2pt}(t)}{\sqrt{\text{Var}[C_\pi^{2pt}(t)]}} &\sim e^{-(m_\pi - E_{\pi\pi}/2)t} \sim \mathbf{1}\end{aligned}$$

And a **pseudoscalar static-light** meson correlation function:

$$\begin{aligned}C_B^{2pt}(t) &= \langle \mathcal{O}_B(t) \bar{\mathcal{O}}_B(0) \rangle \\ \text{Var}[C_B^{2pt}(t)] &\sim \langle \mathcal{O}_\pi(t) \bar{\mathcal{O}}_\pi \rangle \\ \frac{C_B^{2pt}(t)}{\sqrt{\text{Var}[C_B^{2pt}(t)]}} &\sim e^{-(m_B - m_\pi/2)t}\end{aligned}$$

Things I'm not going to talk about

▶ Autocorrelation:

- ▶ Reduction in autocorrelation \Rightarrow reduction in 'effective' variance.
- ▶ Autocorrelation Refs: Sommer, Schaefer, Virotta '10 (arXiv:1009.5228 [hep-lat]); Wolff '04 (hep-lat/0306017)

▶ 'Multi-level'-type algorithms

- ▶ In bosonic theories, an observable is factored into sub-lattice expectation values.
- ▶ This achieves an exponential error reduction compared to conventional simulations.
- ▶ Refs: Lüscher, Weisz '01 (hep-lat/0108014); Della Morte, Giusti '10 (0806.2601 [hep-lat]).

▶ Interpolator Improvement

- ▶ Hasten onset of asymptotic behavior, Improve signal to noise
- ▶ Quark/Link Smearing: Basak, et al. '06 (hep-lat/0509179)
- ▶ Solutions of a GEVP: Blossier, et al. '09 (arXiv:0902.1265 [hep-lat])
- ▶ Clever operator construction/choice: Basak, et al. '05 (hep-lat/0506029)

What I *will* talk about

Correlators with fermion fields must be wick-contracted down fermion two-point functions (**quark propagators**)

$$\langle \mathcal{O}_\pi(t) \mathcal{O}_\pi(0) \rangle = \sum_{\mathbf{x}, \mathbf{y}} \text{Tr} [Q(\mathbf{x}, t | \mathbf{y}, 0)^2]$$

$$Q(\mathbf{x}, t | \mathbf{y}, 0) = \langle q(\mathbf{x}, t) \bar{q}(\mathbf{y}, 0) \rangle$$

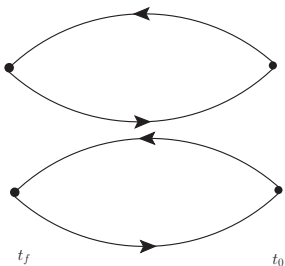
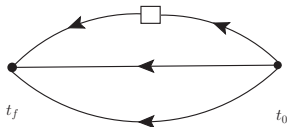
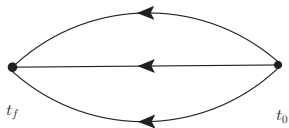
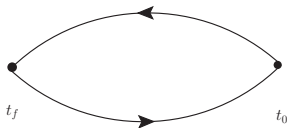
Quark propagators are the inverse of the ($V^3 \times T \times N_{spin} \times N_{color}$ dimensional) fermionic Dirac matrix, and cannot be inverted directly.

(Costly) Algorithms (which iteratively apply the matrix) can obtain the inverse ($Q = M^{-1}$) acting on something.

$$M\psi = \eta \Rightarrow \psi = M^{-1}\eta$$

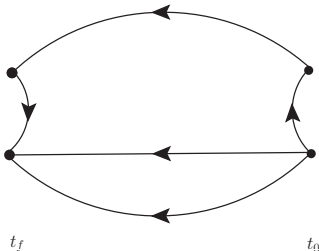
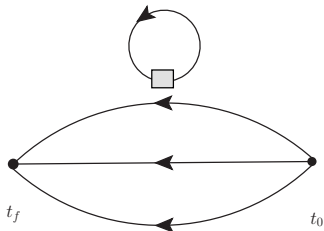
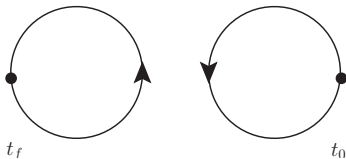
Some correlation functions only require 'point-to-all' propagators, i.e $Q(\mathbf{x}, t|\mathbf{x}_0, t_0) = M^{-1}\delta_{\mathbf{x}\mathbf{x}_0}\delta_{tt_0}$.

- ▶ 'connected' correlation functions (without definite \mathbf{p}_0)
- ▶ Mom. proj. at 'sink' will suffice
- ▶ Some over source sites will increase statistics. (Beware!)



Others require 'all-to-all' propagators

- ▶ Hadrons with definite \mathbf{p}_0
- ▶ Flavor singlet quantities
- ▶ Multi-hadron states



All-to-all propagators cannot be done trivially. Clever estimators can be constructed.

The topic of these lectures will be the **construction and improvement** (i.e. variance reduction) of estimators for all-to-all quark propagators.

Introduction

Stochastic Estimation

Exact Smearred All-to-all

A Hybrid Approach (Stochastic LapH)

Stochastic estimates of inverse matrices

Simple Idea: Choose random numbers η_i from some probability distribution p with the properties

- ▶ $\langle \eta \rangle \equiv \int d\eta \eta p(\eta) = 0$
- ▶ $\langle \eta \eta^* \rangle = \langle |\eta|^2 \rangle = 1$

Then for N independent random numbers we have: ($i, j = 1 \dots N$)

$$\langle \eta_i \eta_j^* \rangle = \begin{cases} \langle |\eta|^2 \rangle = 1, & i = j \\ |\langle \eta \rangle|^2 = 0, & i \neq j \end{cases}$$

$$\text{Var}[\eta_i \eta_j^*] = \begin{cases} \text{Var}[|\eta|^2], & i = j \\ \text{Var}[\eta] \text{Var}[\eta^*], & i \neq j \end{cases}$$

So that by the Central Limit Thm.

$$\langle \eta_i \eta_j^* \rangle_R \equiv \frac{1}{R} \sum_{r=1}^R \eta_i^{(r)} \eta_j^{(r)*}$$

$$\lim_{R \rightarrow \infty} \langle \eta_i \eta_j^* \rangle_R = \langle \eta_i \eta_j^* \rangle + \mathcal{O}(\sqrt{\text{Var}[\eta_i \eta_j^*]/R}) \rightarrow \delta_{ij}$$

We can estimate a (N-dimensional) **matrix inverse**:

$$M\psi^{(r)} = \eta^{(r)} \rightarrow \psi^{(r)} = M^{-1}\eta^{(r)}$$
$$\lim_{R \rightarrow \infty} \langle \psi \eta^* \rangle_R = M^{-1} + \mathcal{O}(\sqrt{\text{Var}[\psi \eta^*]/R})$$

What about the variance?

$$\begin{aligned} \text{Var}[\psi_i \eta_j^*] &= \langle |\psi_i \eta_j^*|^2 \rangle - |\langle \psi_i \eta_j^* \rangle|^2 \\ &= \sum_{k,l} M_{ik}^{-1} [M_{il}^{-1}]^* \langle \eta_k \eta_l^* \eta_j \eta_j^* \rangle - |M_{ij}^{-1}|^2 \\ &= \sum_{k \neq j} |M_{ik}^{-1}|^2 \langle |\eta|^2 \rangle^2 + |M_{ij}^{-1}|^2 (\langle |\eta|^4 \rangle - 1) \\ &= \sum_{k \neq j} |M_{ik}^{-1}|^2 + |M_{ij}^{-1}|^2 (\langle |\eta|^4 \rangle - 1) \end{aligned}$$

What type of noise should I use?

Real Gaussian: η real, $p(\eta) = e^{-\eta^2/2}/\sqrt{2\pi}$

$$\langle \eta \rangle = \int_{-\infty}^{\infty} d\eta \eta p(\eta) = 0$$

$$\langle \eta^2 \rangle = \int_{-\infty}^{\infty} d\eta \eta^2 p(\eta) = 1$$

$$\langle \eta^4 \rangle = \int_{-\infty}^{\infty} d\eta \eta^4 p(\eta) = 3$$

$$\langle \eta_i \eta_j \rangle = \delta_{ij}, \quad \text{Var}[\eta_i \eta_j] = 1 + \delta_{ij}$$

So the **variance** is:

$$\begin{aligned} \text{Var}[\psi_i \eta_j] &= \sum_{k \neq j} |M_{ik}^{-1}|^2 + |M_{ij}^{-1}|^2 (3 - 1) \\ &= \sum_k |M_{ik}^{-1}|^2 + |M_{ij}^{-1}|^2 \end{aligned}$$

Complex Gaussian: η Complex, $p(\eta) = e^{-\eta^*\eta}/\pi$

$$\langle |\eta|^4 \rangle = \int_{\mathcal{C}} d\eta (\eta^*\eta)^2 p(\eta) = 2$$

$$\text{Var}[|\eta|^2] = \int_{\mathcal{C}} d\eta (|\eta|^2 - 1)^2 p(\eta) = 1$$

So the variance is:

$$\text{Var}[\psi_i \eta_j] = \sum_{k \neq j} |M_{ik}^{-1}|^2 + |M_{ij}^{-1}|^2$$

Unimodular noise

$$Z_2: p(1) = p(-1) = 1/2$$

$$\langle \eta^4 \rangle = 1$$

$$\text{Var}[\eta^2] = \langle (\eta^2 - \langle \eta^2 \rangle)^2 \rangle = 0$$

$$Z_4: p(1) = p(-1) = p(i) = p(-i) = 1/4$$

$$(\text{or } Z_2 \times Z_2: p(1+i) = p(1-i) = p(-1+i) = p(-1-i) = 1/4)$$

$$\text{Var}[|\eta|^2] = 0$$

$$\langle |\eta|^4 \rangle = 1$$

$$U(1): \eta = e^{i\theta}, \quad p(\eta) = 1/(2\pi)$$

$$\langle |\eta|^4 \rangle = 1$$

$$\text{Var}[|\eta|^2] = 0$$

Stochastic Estimate of products of M^{-1}

For an unbiased estimator, need independent stochastic estimates for each factor of M^{-1} .

Caveat (the 'sourceless' method): I can obtain $\sum_k M_{ik}^{-1} [M_{jk}^{-1}]^*$ with a single stochastic estimate

$$\begin{aligned}\langle \psi_i \psi_j^* \rangle &= \sum_{k,l} M_{ik}^{-1} [M_{jl}^{-1}]^* \langle \eta_k \eta_l^* \rangle \\ &= \sum_k M_{ik}^{-1} [M_{jk}^{-1}]^*\end{aligned}$$

The variance of products of M^{-1} can be easily calculated:
EXERCISE!!!

Introduce a set of **projectors** $P_{ij}^{[d]}$, $d = 1 \dots N_d$

- ▶ $\sum_d P_{ij}^{[d]} = \delta_{ij}$
- ▶ $\sum_k P_{ik}^{[d]} P_{kj}^{[d']} = \delta_{dd'} P_{ij}^{[d]}$

Now the matrix inverse can be estimated by

$$M_{ij}^{-1} = \sum_d \langle \psi_i^{[d]} \eta_j^{[d]*} \rangle$$
$$\eta_j^{[d]} = P_{jk}^{[d]} \eta_k, \quad \psi_i^{[d]} = M_{ij}^{-1} \eta_j^{[d]}$$

Variance Revisited

$$\begin{aligned}\text{Var}\left[\sum_d \psi_i^{[d]} \eta_j^{[d]*}\right] &= \sum_d \text{Var}[\psi_i^{[d]} \eta_j^{[d]*}] \\ \text{Var}[\psi_i^{[d]} \eta_j^{[d]*}] &= \sum_{k,l} M_{ik}^{-1} [M_{il}^{-1}]^* \langle \eta_j^{[d]} \eta_j^{[d]*} \eta_k^{[d]} \eta_l^{[d]*} \rangle - |\langle \psi_i^{[d]} \eta_j^{[d]*} \rangle|^2\end{aligned}$$

If $j \notin d$, $\langle \psi_i^{[d]} \eta_j^{[d]} \rangle = 0$ and $\text{Var}[\psi_i^{[d]} \eta_j^{[d]}] = 0$. So

$$\begin{aligned}\text{Var}\left[\sum_d \psi_i^{[d]} \eta_j^{[d]}\right] &= \text{Var}[\psi_i^{[d_j]} \eta_j^{[d_j]}] \\ &= \sum_{k \in d_j} |M_{ik}^{-1}|^2 + |M_{ij}^{-1}|^2 (\langle |\eta|^4 \rangle - 1)\end{aligned}$$

There are now *significantly fewer terms* in the sum over k .

The Homeopathic Limit (Maximal Dilution)

Take $N_d = N$. Examine the Variance again:

$$\begin{aligned}\text{Var}\left[\sum_d \psi_i^{[d]} \eta_j^{[d]}\right] &= \sum_{k \in d_j} |M_{ik}^{-1}|^2 + |M_{ij}^{-1}|^2 (\langle |\eta|^4 \rangle - 1) \\ &= |M_{ij}^{-1}|^2 (\langle |\eta|^4 \rangle - 1)\end{aligned}$$

This estimate has zero variance for **unimodular noise**. The Homeopathic limit can be reached with a **single noise vector**.

Start with a single (unimodular) noise vector. I can reduce the error by

- ▶ Adding additional noise sources: decreases like $\sim \sqrt{\text{Var}/R}$
- ▶ Adding additional dilution projectors: reaches the **exact answer with $N_d = N$**

Fermionic Fields

All the above considerations apply to the **Dirac matrix**. Compound indices on η_i : $i = \{a, \alpha, \mathbf{x}, t\}$

Dilution Schemes: (Possible choices of dilution projectors)

- ▶ Time Dilution : $N_d = N_t$, $P^{[d]} = \delta_{td}(\otimes 1_{spin} \otimes 1_{color} \otimes 1_{space})$
- ▶ Spin Dilution : $N_d = N_s$, $P^{[d]} = \delta_{d\alpha}$
- ▶ Color Dilution : $N_d = N_c$, $P^{[d]} = \delta_{da}$
- ▶ Space Even-Odd Dilution : $N_d = 2$. Each projector covers the even or odd sites of the lattice.

Time dilution is a good idea

For connected correlation functions, full **time dilution** is a good idea. Look at the variance:

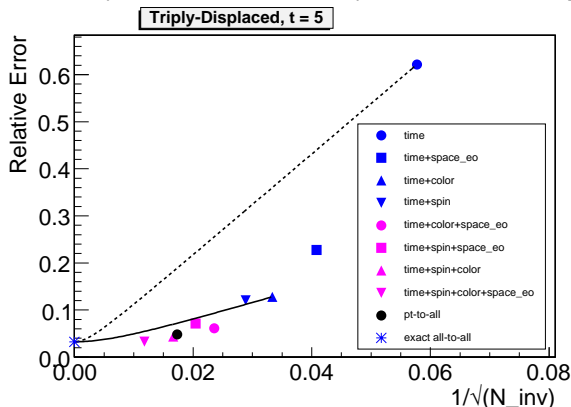
$$\text{Var}\left[\sum_d \psi_{a\alpha}(\mathbf{x}, t)^{[d]} \eta_{a_0\alpha_0}(\mathbf{x}_0, t_0)^{[d]*}\right] = \sum_{(a', \alpha', \mathbf{x}', t') \in d} |M_{(a\alpha|a'\alpha')}(\mathbf{x}, t|\mathbf{x}', t')|^2$$

If time dilution is not used, $\text{Var}[\psi(t)\eta^*(t_0)]$ is independent of $t - t_0$ resulting in a **signal-to-noise problem**.

For fully **disconnected correlators**, variance on the exact all-to-all is independent of time, so full time dilution may be overkill.

Dilution tests

Examine a single observable ([Nucleon correlator](#) at a fixed time separation). Bulava, et al '08 (arXiv:0810.1469 [hep-lat])



Solid/Dashed lines:

$$\sigma_s / \sqrt{N_{inv}} + \sigma_g$$

The rel. error falls off faster than $1/\sqrt{N_d}$

Correlator Construction

With full time dilution:

$$\begin{aligned} Q^{-1}(t|t_0) &= \sum_d \langle \psi^{[d]}(t) \eta^{[d]*}(t_0) \rangle \\ &= \sum_{d \in t_0} \langle \psi^{[d,t_0]}(t) \eta^{[d,t_0]*}(t_0) \rangle \end{aligned}$$

Meson 2pt correlation function:

$$C_{mn}^{2pt}(t - t_0) = \sum_{d,d'} \rho_m^{[d,d';t_0]}(t) \omega_n^{[d,d',t_0]*}(t_0)$$

$$\rho_m^{[d,d';t_0]}(t) = \psi^{[d,t_0]\dagger}(t) \Gamma_m \psi^{[d',t_0]}(t)$$

$$\omega_n^{[d,d';t_0]}(t_0) = \eta^{[d,t_0]\dagger}(t_0) \Gamma_n \eta^{[d',t_0]}(t_0)$$

Baryon 2pt correlation function:

$$C_{mn}^{2pt}(t - t_0) = \sum_{d_1, d_2, d_3} \Delta_m^{[d_1, d_2, d_3; t_0]}(t) \Omega_n^{[d_1, d_2, d_3, t_0]*}(t_0)$$

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A Hybrid Approach (Stochastic LapH)

Quark Smearing

Quark Smearing reduces the level of **excited state contamination**.

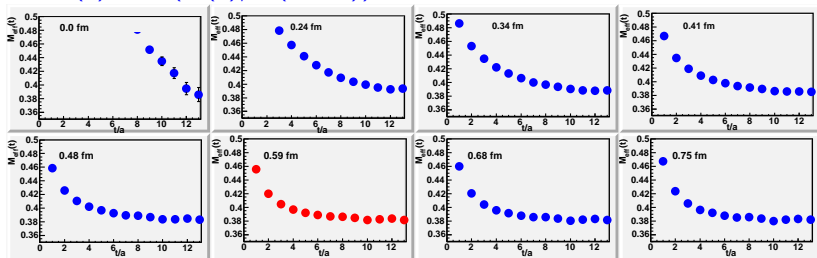
$$Q^{-1} \rightarrow \tilde{Q}^{-1} = S Q^{-1} S^\dagger$$

S is diagonal in time, independent of spin, gauge covariant, and isotropic.

Typical choice: $S = e^{-\frac{\sigma}{2}\Delta}$, σ is a tunable parameter controlling the 'width' of the 'wavefunction'.

To tune, examine the effective mass:

$$am_{\text{eff}}(t) = \ln(C(t)/C(t+1)).$$



A new quark smearing algorithm

Examine the smearing operator in its eigenbasis:

$$S = e^{-\frac{\sigma}{2}\Delta} = \sum_n e^{-\frac{\sigma}{2}\lambda_n} v_n v_n^\dagger$$

Gaussian decreases the weight of high-lying laplacian eigenmodes.

\Rightarrow a new kind of smearing: [truncation in laplacian eigenmodes](#).

(Peardon, et al. '10 arXiv:0905.2160 [hep-lat])

$$S = \Theta(\sigma_s^2 + \Delta) = \sum_{n, \lambda_n < \sigma_s^2} v_n v_n^\dagger \quad (1)$$
$$\approx \sum_{n=1}^{N_v} v_n v_n^\dagger$$

Exact Smeard all-to-all

This 'distillation' allows the **exact smeared quark propagator** (or the smeared-unsmeared quark propagator) to be calculated with $N_{inv} = N_v \times N_t \times N_{spin}$ inversions.

$$\eta^{[n,t_0,\alpha_0]} = v_n[t_0] \otimes \delta_{\alpha\alpha_0} \otimes \delta_{tt_0}$$

$$\psi^{[n,t_0,\alpha_0]} = Q^{-1} \eta^{[n,t_0,\alpha_0]}$$

$$Q_{\alpha\alpha_0}^{-1}(t|t_0)S = \sum_{n=1}^{N_v} \psi_{\alpha}^{[n,t_0,\alpha_0]}(t) \eta^{[n,t_0,\alpha_0]*}$$

Smeared-Smeared Correlator Construction

Correlator construction is particularly simplified with smeared-smeared propagators.

$$\begin{aligned}\tilde{Q}^{-1}(t|t_0) &= \sum_{m,n=1}^{N_v} v_m[t] v_m^\dagger[t] Q^{-1}(t|t_0) v_n[t_0] v_n^\dagger[t_0] \\ &= \sum_{m,n=1}^{N_v} v_m[t] K_{mn}(t|t_0) v_n^\dagger[t_0]\end{aligned}$$

Meson correlation functions:

$$\begin{aligned}C_{mn}^{2pt}(t - t_0) &= \text{Tr} \left[M^m[t] K(t|t_0) M^{n\dagger}[t_0] K(t_0|t) \right] \\ M_{ij}^m[t] &= \sum_{\mathbf{x}, \mathbf{x}'} v_i^\dagger[t](\mathbf{x}) D_m(\mathbf{x}|\mathbf{x}') v_j[t](\mathbf{x}')\end{aligned}$$

Baryon correlation functions:

$$C_{mn}^{2pt}(t - t_0) = \sum_{\substack{ijk \\ i'j'k'}} \Omega_{ijk}^m[t] K_{ii'}(t|t_0) K_{jj'}(t|t_0) K_{kk'}(t|t_0) \Omega_{i'j'k'}^{n\dagger}[t_0]$$

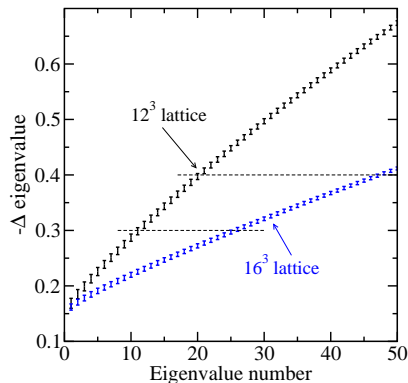
Current Insertions:

$$K_{ij}(t_f|t_0) \rightarrow G_{ij}^J(t_f|t|t_0)$$
$$G_{ij}^J(t_f|t|t_0) = \psi^{[i,t_f,\alpha_f]\dagger}(t) J \psi^{[j,t_0,\alpha_0]}(t)$$

Multi-hadron correlation functions are more complicated products of M, K, Ω, G

The V^2 Problem

To maintain fixed σ_s , N_V (and N_{inv}) **increases like $\sim V$** .



Each inversion cost **increases like $V \rightarrow V^2$** total cost increase.

Error only **goes down like $\sim V$** .

Preliminary Distillation Results (Small Volumes)

- ▶ **Nucleon, Δ , and Ω baryon** excitation spectra: Bulava, et al. '10 (arXiv:1004.5072 [hep-lat])
- ▶ **Isovector meson** excitation spectra: Morningstar, et al. '10 (arXiv:1011.6573 [hep-lat])
- ▶ **Isoscalar meson** excitation spectra: Dudek et al. '11 (arXiv:1102.4299 [hep-lat])
- ▶ **$I = 2 \pi - \pi$** scattering phase shifts: Dudek et al. '10 (arXiv:1011.6352 [hep-lat])

Introduction

Stochastic Estimation

Exact Smeared All-to-all

A Hybrid Approach (Stochastic LapH)

A new way to introduce noise

Ordinary Lattice Noise : $\eta_{a\alpha}^{(r)}(\mathbf{x}, t) \in Z_4$

LapH(Laplacian-Heaviside) Noise :

$$\eta_{a\alpha}^{(r)}(\mathbf{x}, t) = \sum_n d_{n\alpha t}^{(r)} v_{na}[t](\mathbf{x}), \quad d_{n\alpha t}^{(r)} \in Z_4$$

Two distinct features of LapH Noise :

- ▶ *Dramatically fewer random numbers* than standard noise
- ▶ η is now *gauge-covariant*

Correlator construction is the same as discussed previously, with source smearing.

Types of LapH dilution schemes

Dilution can now occur in $time \otimes spin \otimes eigenmode$ space.

For each type of dilution, one has three simple options:

- ▶ **Full (F)**: full dilution
- ▶ **Interlace- N (IN)**: Each projector has support separated by N
- ▶ **Block- N (BN)**: Each projector has support on N adjacent

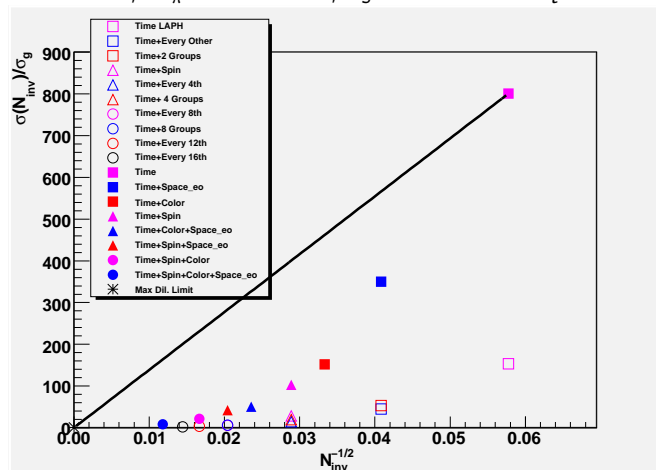
Examples:

- ▶ $[F, F, F]$: The maximal dilution limit.
- ▶ $[F, F, I8]$: Full time and spin dilution, support on every 8th eigenvector
- ▶ $[I16, F, I8]$: Support on every 16th timeslice, full spin dilution, support on every 8th eigenvector.

Numerical Tests/Demonstrations

Laph Noise vs. Lattice Noise (J. Bulava '09, P.H.D Thesis)

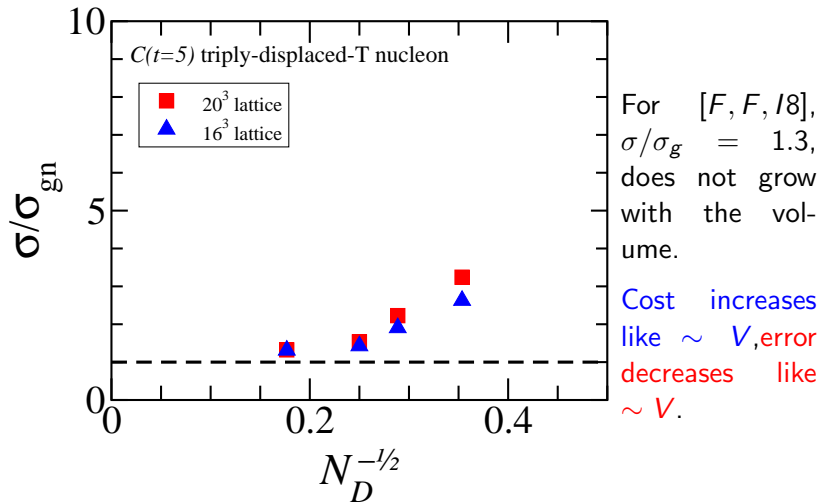
$20^3 \times 128$, $m_\pi \sim 400\text{MeV}$, $a_s = .12\text{fm} = 3a_t$.



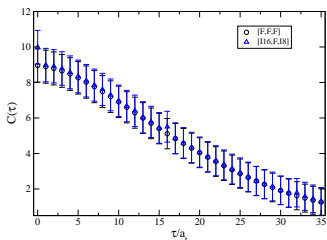
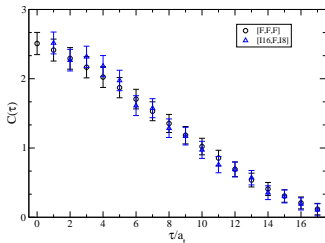
Volume dependence of higher LapH dilution schemes (Morningstar, et al. *to appear*)

Dilution schemes, from left to right:

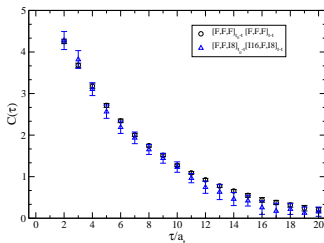
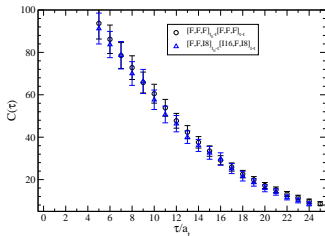
$[F, F, 18]$, $[F, N, 16]$, $[F, N, 12]$, $[F, N, 18]$.



- ▶ Isoscalar(σ and η') mesons: Foley, et al. '10 (arXiv:1011.6573 [hep-lat])
- ▶ Examine **disconnected contributions** in Distillation vs. Stochastic LapH.



- ▶ Two-pion systems: Foley, et al. '10 (arXiv:1011.6573 [hep-lat])
- ▶ Disconnected diagrams from $\pi\pi - \pi\pi$ (box diagram) and $\rho - \pi\pi$ (triangle diagram).



Conclusions

- ▶ All-to-all propagators are required to evaluate many correlation functions, and (probably) have to be estimated stochastically.
- ▶ If you don't have to do all-to-all, a careful cost-benefit analysis should be performed, analyzing the variance, etc.
- ▶ Introducing additional dilution projectors reduces the variance faster than $N_{inv}^{-1/2}$, to a point.
- ▶ LapH noise seems to perform better (factor of ~ 8) than lattice noise, although the construction of the eigenvectors is a non-trivial (but relatively small) cost.
- ▶ Additionally, highly diluted Laph noise has a volume independent effectiveness. The $[F, F, 18]$ scheme has an error only 30% larger than the exact answer, for a factor $\sim 4 - 32$ less inversions (on a 2fm) lattice.

Additional all-to-all techniques

- ▶ Low-mode preconditioning: Neff, et al. '01 (hep-lat/0106016)
- ▶ Even-odd preconditioning: Blossier, et al. '10 (arXiv:1004.2661 [hep-lat])
- ▶ Eigenspectrum noise subtraction: Guerrero, et al. '09 (arXiv:1001.4366 [hep-lat])
- ▶ Hopping parameter acceleration: Bali et al. '10 (arXiv:0910.3970 [hep-lat])
- ▶ Truncated solver method: Bali et al. '10 (arXiv:0910.3970 [hep-lat])
- ▶ Domain decomposition improvement: Burch and Hagen '06 (arXiv:hep-lat/0609011)