Multi-level Monte Carlo integration: HVP

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Dalla Brida, LG, Harris, Pepe, PLB 816 (2021) 136191 [arXiv:2007.02973] LG, Harris, Nada, Schaefer, EPJC 79 (2019) 586 [arXiv:1903.10447]

Muon g - 2 theory initiative workshop - Virtual Meeting - June 30th 2021

The bottleneck: signal/noise ratio for HVP (HLbL,...)

• The HVP contribution to $a_{\mu} = (g-2)_{\mu}/2$ reads

$$a_{\mu}^{\mathrm{HVP}} = \left(\frac{\alpha}{\pi}\right)^2 \int_0^{\infty} dx_0 K(x_0, m_{\mu}) G(x_0)$$

where

$$G(x_0) = \int d^3x \langle J_k^{\mathrm{em}}(x) J_k^{\mathrm{em}}(0) \rangle$$

with $K(x_0, m_\mu)$ being a known function

 For the light-connected contribution (by far the largest)

$$\frac{\sigma^2_{G^{\rm conn}_{u,d}}(x_0)}{[G^{\rm conn}_{u,d}(x_0)]^2} \propto \frac{1}{n_{cnfg}} e^{2(M_{\rho} - M_{\pi})|x_0|}$$

where M_{ρ} is the lightest state in that channel. Signal lost after 1.5-2.5 fm (depending on $m_{u,d}$) due to exp. increase of statistical error



$$n_{cnfg} = n_0 = 25 \,, \ n_{tot} = n_0 \cdot n_1$$



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- Sharp rise of σ² with x₀ when computed by a standard 1-level integration (red points) is automatically flattened out by the 2-level integration (blue-points)
- Accurate computations can be obtained at large distances: no need for any modeling of the long-distance behaviour of G_{u,d}





Why/How does it work ?



where

$$\langle\!\langle O_0[U_{\Omega_0}]\rangle\!\rangle_{\Lambda_0} = \frac{1}{Z_{\Lambda_0}} \int \mathcal{D}U_{\Lambda_0} e^{-S_0[U_{\Omega_0}]} O_0[U_{\Omega_0}]$$

Two-level integration:

- n_0 configurations U_{Λ_1}

- n_1 configurations U_{Λ_0} and U_{Λ_2} for each U_{Λ_1}
- ▶ If $\langle\!\langle \cdot \rangle\!\rangle_{\Lambda_i}$ can be computed efficiently with a statistical error comparable to its central value, then the prefactor in the signal/noise ratio changes as (until S/N problem solved) $n_0 \rightarrow n_0 n_1^2$

at the cost of generating approximatively $n_0 n_1$ level-0 configurations

Split-even estimator of disconnected contribution

- Advantage of multi-level sets in when variances are due to fluctuations of gauge field
- The disconnected Wick contraction reads

$$t(x) = \operatorname{Tr} \left[\gamma_k \{ D_{m_u}^{-1}(x, x) - D_{m_s}^{-1}(x, x) \} \right]$$
$$= (m_s - m_u) \operatorname{Tr} \left[\gamma_k D_{m_u}^{-1} D_{m_s}^{-1}(x, x) \right]$$



$$\theta(x) = \frac{(m_s - m_u)}{N_s} \sum_{i=1}^{N_s} \operatorname{Im} \left[\eta_i^{\dagger}(x) \gamma_k \{ D_{m_u}^{-1} D_{m_s}^{-1} \eta_i \}(x) \right]$$

is expensive. It requires $O(10^4)$ random fields η for its σ^2 to be dominated by gauge fluctuations

Why random noise much larger than gauge one? Computable and understandable in QFT





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Split-even stochastic estimator $[\langle \eta(x)\eta^{\dagger}(y)\rangle = \delta_{xy}]$

$$\tau(x) = \frac{(m_s - m_u)}{N_s} \sum_{i=1}^{N_s} \operatorname{Im} \left[\{\eta_i^{\dagger} D_{m_u}^{-1}\}(x) \, \gamma_k \, \{D_{m_s}^{-1} \eta_i\}(x) \right]$$

requires $O(10^2)$ random fields η to hit gauge noise. Gain: 2 orders of magnitude. Definition suggested by the QFT analysis of the variance.

Used in the past for pseudoscalar density in TMQCD (one-end trick)



Split-even estimator of disconnected contribution

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$$f(x) = \operatorname{Tr} \left[\gamma_k \{ D_{m_u}^{-1}(x, x) - D_{m_s}^{-1}(x, x) \} \right]$$
$$= (m_s - m_u) \operatorname{Tr} \left[\gamma_k D_{m_u}^{-1} D_{m_s}^{-1}(x, x) \right]$$



Split-even stochastic estimator [⟨η(x)η[†](y)⟩ = δ_{xy}]

$$\tau(x) = \frac{(m_s - m_u)}{N_s} \sum_{i=1}^{N_s} \operatorname{Im} \left[\{\eta_i^{\dagger} D_{m_u}^{-1}\}(x) \gamma_k \{D_{m_s}^{-1} \eta_i\}(x) \right]$$

combined with multi-level integration is a solution for a precise computation of the disconnected contribution

It is already being applied in production phase for HVP by CLS (Mainz)



Conclusions & Outlook

- Permille precision and accuracy on HVP is the challenge for lattice QCD
- Our strategy: new integration and estimators (better "machine" and "experiment")



- Multi-level integration reduces the variance exponentially:
 - with the time-distance of the currents
 - when pion mass gets lighter (physical point)
- ▶ Next step: R&D ⇒ production. Significant human and numerical resources needed
- Analogous variance-reduction pattern expected to work out also for lattice calibration, electromagnetic corrections, HLbL, ...