

Algorithmic advances in NSPT

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Based on work done in collaboration with
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Motivation

Why perturbation theory?

Renormalization

$$\text{Ex.: } O(q) \stackrel{q \rightarrow \infty}{\approx} \alpha_{\overline{\text{MS}}}(q) + k_1 \alpha_{\overline{\text{MS}}}(q)^2 + k_2 \alpha_{\overline{\text{MS}}}(q)^3 + \dots$$

$\approx 10\% \qquad \qquad \approx 1\%$

but also

- ▶ quark-masses
- ▶ effective electro-weak Hamiltonians
- ▶ small-flow time expansions
- ▶ quasi-PDF
- ▶ ...

Improvement

$$\text{Ex.: } \tilde{g}_0^2 = g_0^2 [1 + b_g(g_0) a m_q], \quad b_g = O(g_0^2)$$

and in general

$$O^{\text{lat}}(q, a) - O(q) \stackrel{a \rightarrow 0}{\approx} g_0 \delta_1(a, q) + g_0^2 \delta_2(a, q) + \dots$$

Motivation

Need for automation

Perturbative **lattice** calculations are particularly **difficult** and human **time consuming** ...

Why?

- ▶ Feynman rules can be very complicated
- ▶ In gauge-theories new vertices appear at each order
⇒ the number of diagrams grows very rapidly
- ▶ Requires numerical evaluation even for simple diagrams
⇒ naive computational cost $\propto V_{4d}^N$ @ N loops

A possible **solution** ...

Numerical stochastic perturbation theory (NSPT)

- ✓ Fully automated
- ✓ Easy to set-up and very flexible tool
- ✓ Allows for high-order computations
- ✗ Systematic and statistical errors

Introduction

The Parisi and Wu's way

(Parisi, Wu '81; Zwanziger '81; Batrouni et. al. '85; Zinn-Justin '86; Damgaard, Hüffel '87; Zinn-Justin, Zwanziger '88)

Ex.: lattice φ^4 -theory

$$S(\varphi) = \sum_x \left\{ \frac{1}{2} \varphi(x) \Delta \varphi(x) + \frac{g}{4!} \varphi(x)^4 \right\}, \quad \Delta = (-\partial_\mu^* \partial_\mu + m^2) > 0$$

$$\langle \mathcal{O} \rangle = \frac{1}{Z} \int d\varphi \mathcal{O}(\varphi) e^{-S(\varphi)}$$

Langevin equation

$$\partial_t \phi(t, x) = - \frac{\delta S(\phi)}{\delta \phi(t, x)} + \eta(t, x)$$

$$\frac{\delta S(\phi)}{\delta \phi(t, x)} = \Delta \phi(t, x) + \frac{g}{3!} \phi(t, x)^3, \quad \langle \eta(t, x) \eta(s, y) \rangle_\eta = 2 \delta_{xy} \delta(t - s)$$

Stochastic quantization

$$\lim_{t \rightarrow \infty} \langle \phi(t, x_1) \dots \phi(t, x_n) \rangle_\eta = \langle \varphi(x_1) \dots \varphi(x_n) \rangle$$

$$\lim_{t \rightarrow \infty} P(t, \phi) = P_{\text{eq}}(\phi) \propto e^{-S(\phi)}$$

Introduction

Stochastic perturbation theory

(Parisi, Wu '81; Damgaard, Hüffel '87)

Perturbative field

$$\phi = \phi_0 + g \phi_1 + g^2 \phi_2 + \dots$$

Forces

$$\frac{\delta S(\phi)}{\delta \phi} = F_0(\phi_0) + g F_1(\phi_0, \phi_1) + \dots$$

Dynamics

$$\partial_t \phi_0 = -\Delta \phi_0 + \eta$$

$$\partial_t \phi_1 = -\Delta \phi_1 - \frac{1}{3!} \phi_0^3$$

...

$$\partial_t \phi_r = -F_r(\phi_0, \dots, \phi_r) + \delta_{r0} \eta \quad [F_r(\phi) = \Delta \phi_r + V_r(\phi_0, \dots, \phi_{r-1})]$$

Observables

$$\mathcal{O}(\phi) = \mathcal{O}_0(\phi_0) + g \mathcal{O}_1(\phi_0, \phi_1) + \dots \quad \text{Ex.: } \phi^2 = \phi_0^2 + g 2\phi_0 \phi_1 + \dots$$

$$\lim_{t \rightarrow \infty} \langle \mathcal{O}_r(\phi_0, \dots, \phi_r) \rangle_\eta = k_r^{\mathcal{O}} \quad \text{where} \quad \langle \mathcal{O} \rangle = k_0^{\mathcal{O}} + g k_1^{\mathcal{O}} + \dots$$

Numerical stochastic perturbation theory

Go numerical!

(Di Renzo et. al. '94; Di Renzo, Scorzato '04)

Computerize

$$t \rightarrow t_n = n\varepsilon, \quad n \in \mathbb{N} \quad \text{and} \quad r = 0, 1, \dots, N$$

Discrete dynamics

$$\text{Ex.: } \phi_0(t_{n+1}) = -\varepsilon \Delta \phi_0(t_n) + \sqrt{\varepsilon} \eta(t_n)$$

$$\phi_1(t_{n+1}) = -\varepsilon \Delta \phi_1(t_n) - \frac{\varepsilon}{3!} \phi_0(t_n)^3$$

...

$$\phi_r(t_{n+1}) = -\varepsilon F_r(\phi_0(t_n), \dots, \phi_r(t_n)) + \delta_{r0} \sqrt{\varepsilon} \eta(t_n)$$

Order-by-order ops.

$$\phi = \{\phi_0, \dots, \phi_N\} \quad \text{Ex.: } \phi + \chi \rightarrow \phi_r + \chi_r, \quad \phi^2 \rightarrow \phi_r^2 = \sum_{s=0}^r \phi_{r-s} \phi_s$$

Stochastic estimates

$$\bar{\mathcal{O}}_r = \frac{1}{N_{\text{cnfg}}} \sum_{n=0}^{N_{\text{cnfg}}} \mathcal{O}_r(\phi_0(t_n), \dots, \phi_r(t_n)) \quad \text{Ex.: } \phi^2 = \{\phi_0^2, 2\phi_0\phi_1, \dots\}$$

Langevin based NSPT

Some known (and proven) facts

[Statistical errors]

$$\sigma(\overline{\mathcal{O}}_r)^2 = N_{\text{cnfg}}^{-1} \times \tau_{\text{int}}(\mathcal{O}_r) \times \text{var}(\mathcal{O}_r) \quad [\mathcal{O} \equiv \text{multiplicatively ren.}]$$

▶ Autocorrelations

(Zinn-Justin '86; Zinn-Justin, Zwanziger '88)

$$\tau_{\text{int}}(\mathcal{O}_r) \stackrel{a \rightarrow 0}{\propto} a^{-2}$$

▶ Variances

(Lüscher '15)

$$\text{var}(\mathcal{O}_r) = \overline{\mathcal{O}_r^2} - (\overline{\mathcal{O}_r})^2 \stackrel{a \rightarrow 0}{\propto} \ln(a)$$

IMPORTANT: $\text{var}(\mathcal{O}_r) \neq 2r$ -order of $\langle \mathcal{O}^2 \rangle - \langle \mathcal{O} \rangle^2$ [$r \neq 0$]

\Rightarrow they are **NOT** given by the theory **ALONE!**

[Systematic errors]

(Batrouni et. al. '85; Kronfeld '93)

$$\lim_{N_{\text{cnfg}} \rightarrow \infty} \overline{\mathcal{O}}_r = k_r^{\mathcal{O}} + \mathcal{O}(\varepsilon^p) \quad [p \equiv \text{order integration scheme}]$$

IMPORTANT: NO accept-reject step: **NOT** analytic in $g!$

\Rightarrow inexact algorithm and step-size errors

Stochastic molecular dynamics algorithm

The Horowitz's way

(Horowitz '85, '87, '91; Jansen, Liu '95; Lüscher, Schaefer '11)

Stochastic molecular dynamics (SMD)

$$\partial_t \phi(t, x) = \pi(t, x)$$

$$\partial_t \pi(t, x) = -\frac{\delta S(\phi)}{\delta \phi(t, x)} - \gamma \pi(t, x) + \eta(t, x)$$

$$\langle \eta(t, x) \eta(s, y) \rangle_\eta = 2\gamma \delta_{xy} \delta(t - s)$$

- ▶ Adjustable “friction” parameter $\gamma > 0$
- ▶ Coincides with Langevin equation for $\gamma \rightarrow \infty$
- ▶ Once perturbatively expanded, if $\gamma = \text{const.}$:

- ▶ **Autocorrelations:** $\tau_{\text{int}}(\mathcal{O}_r) \stackrel{a \rightarrow 0}{\approx} c_{\mathcal{O}_r}(\gamma) a^{-2}$

- ▶ **Variances:** $\text{var}(\mathcal{O}_r) \stackrel{a \rightarrow 0}{\approx} c'_{\mathcal{O}_r}(\gamma) \ln(a)$

(MDB, Lüscher '17; MDB, Garofalo, Kennedy '17)

Phase-space stochastic quantization

$$\lim_{t \rightarrow \infty} \langle \phi(t, x_1) \dots \phi(t, x_n) \rangle_\eta = \langle \varphi(x_1) \dots \varphi(x_n) \rangle$$

$$P_{\text{eq}}(\pi, \phi) \propto e^{-H(\pi, \phi)}, \quad H(\pi, \phi) = \frac{1}{2} \sum_x \pi(x)^2 + S(\phi)$$

SMD based NSPT

The algorithm in a nutshell

(MDB, Garofalo, Kennedy '15, '17; MDB, Lüscher '16, '17)

Start:

$$t \rightarrow t_n = n\epsilon, \quad n \in \mathbb{N}, \quad \phi = \sum_{r=0}^N g^r \phi_r, \quad \pi = \sum_{r=0}^N g^r \pi_r$$

Step 1: Momentum rotation

$$\pi \rightarrow c_1 \pi + c_2 v$$

$$c_1 = e^{-\gamma\epsilon}, \quad c_2 = (1 - c_1^2)^{1/2}, \quad \langle v(x)v(y) \rangle_v = \delta_{xy} \quad (v = v_0)$$

Step 2: Molecular dynamics step

$$\left. \begin{aligned} I_{\pi,h} : \pi &\rightarrow \pi - hF(\phi) \\ I_{\phi,h} : \phi &\rightarrow \phi + h\pi \end{aligned} \right\} \begin{array}{l} \text{Symplectic reversible int.} \\ t_n \rightarrow t_{n+1} \end{array}$$

Ex. LPF: $I_\epsilon = I_{\pi,\epsilon/2} I_{\phi,\epsilon} I_{\pi,\epsilon/2}$

Step 1 \rightarrow **Step 2** \rightarrow **Step 1** $\rightarrow \dots \rightarrow$ **Step 2**

Step 3: Measure $\quad \bar{\mathcal{O}}_r = \frac{1}{N_{\text{cnfg}}} \sum_{n=0}^{N_{\text{cnfg}}} \mathcal{O}_r(\phi_0(t_n), \dots, \phi_r(t_n))$

SMD based NSPT

Convergence to a stationary distribution

(MDB, Lüscher '17)

The discrete stochastic processes $\{\phi_r(t_n), \pi_r(t_n)\}$, $r = 0, 1, \dots$ all converge for $t_n \rightarrow \infty$ to a **unique** and **stationary** distribution iff:

1. $\Delta > 0$

2. $\epsilon^2 \|\Delta\| < \kappa$

$$S_0 = \frac{1}{2} \sum_x \phi_0(x) \Delta \phi_0(x)$$

$$\|\Delta\| \equiv \lambda_{\max} \approx 16$$

Leading-order distros [$r = 0$]

$$P_{\text{eq}}(\pi_0, \phi_0) \propto e^{-\hat{H}_0(\pi_0, \phi_0)}$$

$$\hat{H}_0(\pi_0, \phi_0) = \frac{1}{2} \sum_x \pi_0(x)^2 + \hat{S}_0(\phi_0), \quad \hat{S}_0 = \frac{1}{2} \sum_x \phi_0(x) \hat{\Delta} \phi_0(x)$$

► LPF: $\hat{\Delta} = \Delta(1 - \frac{1}{4}\epsilon^2\Delta)$ $\hat{\Delta} > 0 \Rightarrow \kappa = 4$

► OMF2: $\hat{\Delta} = \Delta(1 + a_1 \epsilon^2 \Delta + O(\epsilon^4))$ $a_1 \approx -2.46 \times 10^{-3}$, $\kappa = 6.51$

► OMF4: $\hat{\Delta} = \Delta(1 + a_2 \epsilon^4 \Delta^2 + O(\epsilon^6))$ $a_2 \approx -2.58 \times 10^{-5}$, $\kappa = 9.87$

(Omelyan, Mryglod, Folk '03)

N.B.: $\|\hat{\Delta}_{\text{LPF}}/\Delta\| \approx 25 \|\hat{\Delta}_{\text{OMF2}}/\Delta\| \approx 300 \|\hat{\Delta}_{\text{OMF4}}/\Delta\|$ [@ fixed cost per MDU]

NSPT in lattice QCD

SU(3) Yang-Mills theory

(MDB, Lüscher '17)

SMD equations

$$\partial_t U_t(x, \mu) = g_0 \pi_t(x, \mu) U_t(x, \mu)$$

$$\partial_t \pi_t(x, \mu) = -g_0 (\partial_{x, \mu}^a S_G)(U_t) T^a - \gamma \pi_t(x, \mu) + \eta_t(x, \mu)$$

$$\langle \eta_t^a(x, \mu) \eta_s^b(y, \nu) \rangle = 2\gamma \delta^{ab} \delta_{\mu\nu} \delta_{xy} \delta(t - s)$$

Perturbative fields

$$U(x, \mu) = e^{g_0 A_\mu(x)} = 1 + \sum_{r=0}^N g_0^{r+1} U_r(x, \mu), \quad \pi(x, \mu) = \sum_{r=0}^N g_0^r \pi_r(x, \mu)$$

Step 1. Momentum rotation (very much as for ϕ^4)

Step 2. Molecular dynamics step

$$\left. \begin{aligned} I_{\pi, h} : \pi &\rightarrow \pi - h g_0 \partial S_G(U) \\ I_{U, h} : U &\rightarrow e^{h g_0 \pi U} \end{aligned} \right\} \begin{array}{l} \text{Symplectic reversible int.} \\ t_n \rightarrow t_{n+1} \end{array}$$

WARNING: Convergence to a unique and stationary distro. requires special care due to gauge symmetry, but it can be guaranteed! [@ BACKUP]

The gradient flow coupling

A highly non-trivial case study

Gradient flow

$$\partial_t B_\mu(t, x) = D_\nu G_{\nu\mu}(t, x) \quad [t \equiv \text{flow time}]$$

$$G_{\mu\nu} = \partial_\mu B_\nu - \partial_\nu B_\mu + [B_\mu, B_\nu], \quad B_\mu(0, x) = A_\mu(x)$$

Schrödinger functional (Lüscher et. al. '92)

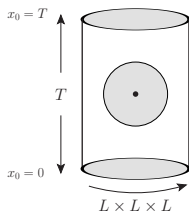
$$A_k(x)|_{x_0=0} = A_k(x)|_{x_0=T} = 0$$

$$A_\mu(x + \hat{k}L) = A_\mu(x)$$

Basic quantity (Lüscher '10; Fodor et. al. '12; Fritzsche, Ramos '13)

$$t^2 \langle E(t, x) \rangle, \quad E(t, x) = \frac{1}{4} G_{\mu\nu}^a(t, x) G_{\mu\nu}^a(t, x)$$

$$T = L, \quad x_0 = T/2, \quad \sqrt{8t} = 0.3 \times L$$



NSPT observables

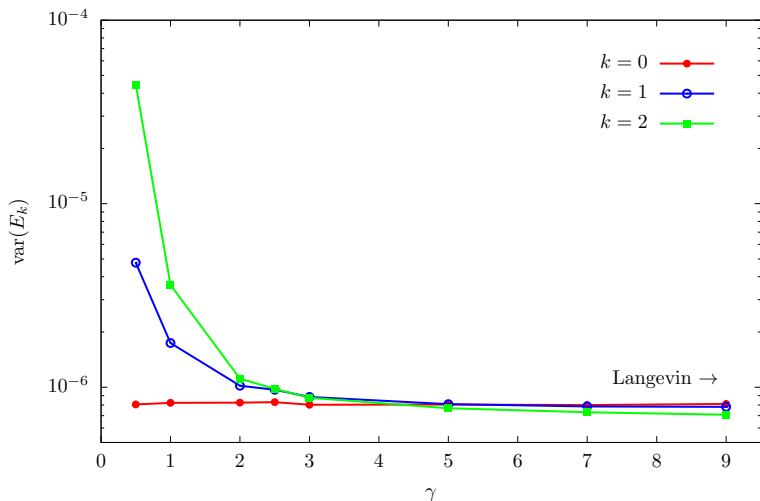
[using $\alpha_{\overline{\text{MS}}} \leftrightarrow g_0$ (Lüscher, Weisz '95)]

$$t^2 \langle E(t, x) \rangle = g_0^2 E_0 + g_0^4 E_1 + g_0^6 E_2 + \dots$$

$$\stackrel{a/L \rightarrow 0}{=} k_0 \{ \alpha_{\overline{\text{MS}}}(q) + k_1 \alpha_{\overline{\text{MS}}}(q)^2 + k_2 \alpha_{\overline{\text{MS}}}(q)^3 + \dots \}, \quad q^{-1} = \sqrt{8t}$$

Algorithm dependence of the variances

Variances $\text{var}(E_k)$ as a function of γ for $L/a = 16$

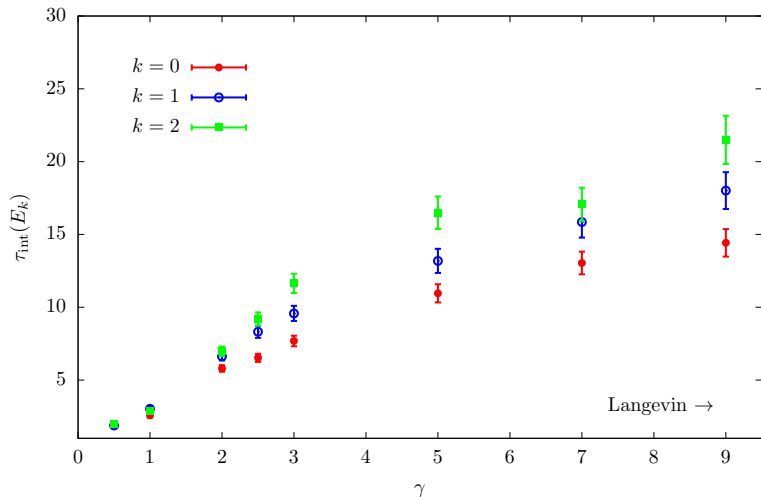


IMPORTANT:

$\text{var}(E_k) \neq [\langle E^2 \rangle - \langle E \rangle^2]_{g_0^{4(k+1)}}$ if $k \neq 0 \Rightarrow \text{var}(E_{1,2})$ are **algorithm dependent!**

Autocorrelations

Integrated autocorrelations $\tau_{\text{int}}(E_k)$ (in MDU) as a function of γ for $L/a = 16$

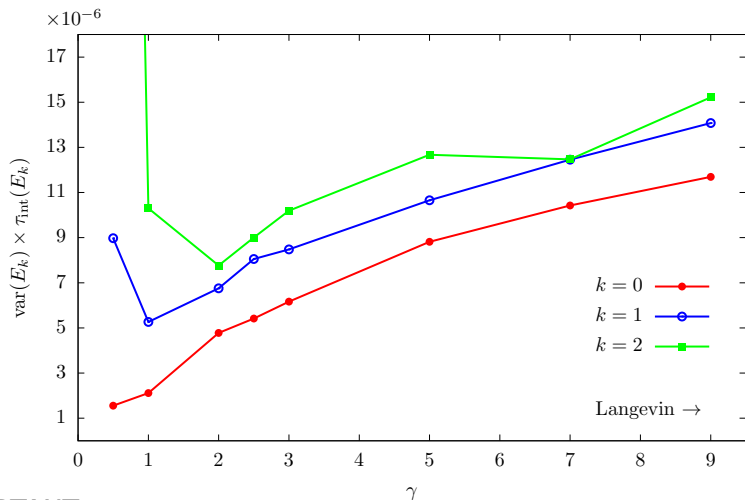


REMARK:

The autocorrelations of the different orders have a similar γ -dependence

Statistical errors

Products $\text{var}(E_k) \times \tau_{\text{int}}(E_k)$ as a function of γ for $L/a = 16$

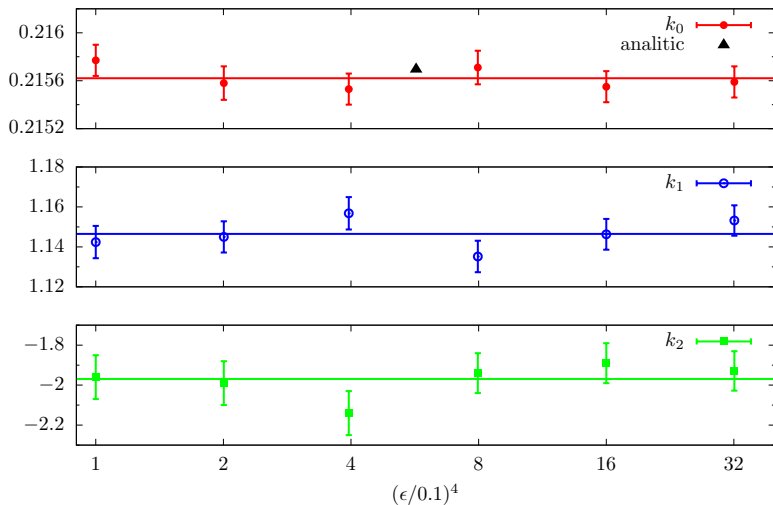


IMPORTANT:

Algorithms tuned for **small** autocorrelations tend to have **large** variances and viceversa: a **compromise** needs to be found!

Systematic errors

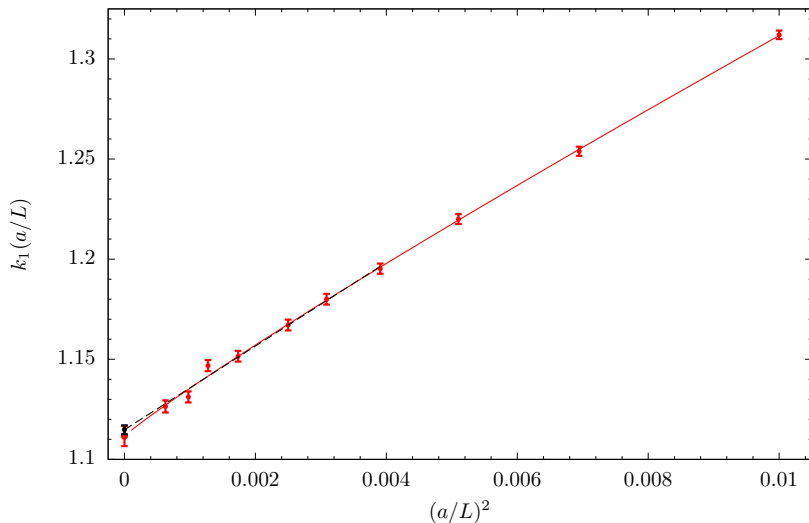
Coefficients $k_i(a/L)$, $i = 0, 1, 2$, for $L/a = 24$, 10^4 measurements per ϵ , OMF4 integrator



NOTE:

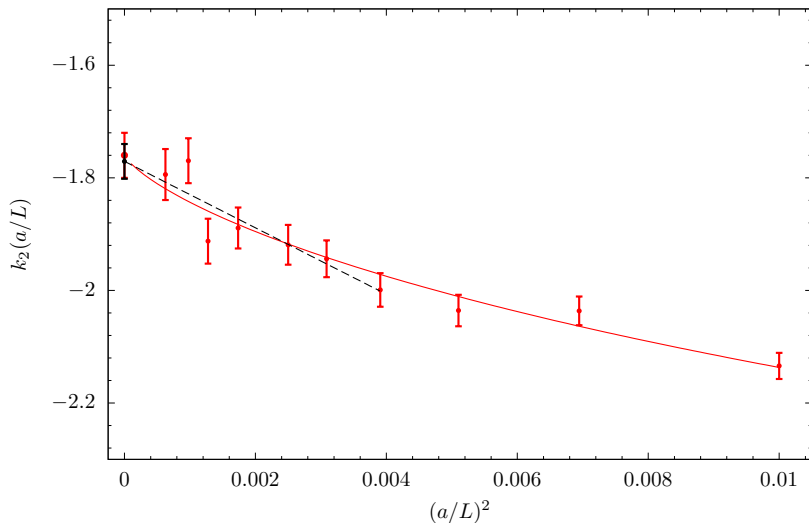
The energy violations per MD step are $\Delta H_i \propto \epsilon^5$, where $H = H_0 + g_0 H_1 + \dots$

Continuum limit of k_1



Symanzik theory: $k_1 \stackrel{a \rightarrow 0}{\approx} a_0 + \{a_1 + b_1 \ln(L/a)\}(a/L)^2 + O(a^3)$

Continuum limit of k_2



Symanzik theory: $k_2 \stackrel{a \rightarrow 0}{=} a_0 + \{a_1 + b_1 \ln(L/a) + c_1 [\ln(L/a)]^2\} (a/L)^2 + O(a^3)$

Outlook & conclusions

Conclusions:

- ▶ NSPT is a **powerful** tool for automatizing LPT calculations
- ▶ Complicated lattice set-ups and observables give rise to hardly any difficulty
- ▶ The recent algorithmic advances opened the way for **precise** and **accurate** results
- ▶ A full implementation for the SU(3) gauge theory can be downloaded at:

luscher.web.cern.ch/luscher/NSPT

Outlook:

- ▶ The inclusion of fermions is in principle straightforward and does not pose any technical difficulty
- ▶ Fermions are not expected to slow down the simulations by a big factor

(MDB, Lüscher '17)

(Di Renzo, Scorzato '04)



BACKUP

NSPT in lattice QCD

Convergence to a stationary distribution

Leading order dynamics

$$S_G = S_0 + g_0 S_1 + g_0^2 S_2 + \dots$$

$$S_0 = \frac{1}{2} \sum_{x,y} A_\mu(x) \Delta_{\mu\nu}(x,y) A_\nu(y), \quad A_\mu(x) = U_0(x, \mu)$$

PROBLEM: $\Delta \geq 0$

1. Gauge modes $A_\mu(x) \sim 0$
2. In a finite volume w/ periodic bc. $\exists A_\mu(x) \not\sim 0$ s.t. $\Delta A = 0$

N.B.: Issues of the perturbative expansion not (N)SPT!

(Gonzalez-Arroyo, Jurkiewicz, Korhals-Altes '83)

SOLUTION:

1. Gauge damping a.k.a. stochastic gauge fixing
2. Choose proper boundary conditions for the fields
e.g. $A_k(x)|_{x_0=0} = A_k(x)|_{x_0=T} = 0$

(Zwanziger '81)

RESULT: Separate convergence criteria for the gauge modes

(MDB, Lüscher '17)

Gauge damped SMD equations

Time-dependent gauge transf.

$$\pi_t(x, \mu) \rightarrow \Lambda_t(x)\pi_t(x, \mu)\Lambda_t(x)^{-1}$$

$$U_t(x, \mu) \rightarrow \Lambda_t(x)U_t(x, \mu)\Lambda_t(x + \hat{\mu})^{-1}$$

Modified SMD equations

$$\partial_t U_t(x, \mu) = g_0 \{ \pi_t(x, \mu) - \nabla_\mu \omega_t(x) \} U_t(x, \mu)$$

$$\partial_t \pi_t(x, \mu) = -g_0 (\partial_{x,\mu}^a S_G)(U_t) T^a - \gamma \pi_t(x, \mu) + \eta_t(x, \mu) + g_0 [\omega_t(x), \pi_t(x, \mu)]$$

$$\langle \eta_t^a(x, \mu) \eta_s^b(y, \nu) \rangle = 2\gamma \delta^{ab} \delta_{\mu\nu} \delta_{xy} \delta(t - s)$$

with

$$\omega_t(x) = g_0^{-1} \partial_t \Lambda_t(x) \Lambda_t(x)^{-1}$$

$$\nabla_\mu \omega_t(x) = U_t(x, \mu) \omega_t(x + \hat{\mu}) U_t(x, \mu)^{-1} - \omega_t(x)$$

Gauge damping

$$\partial_t \omega_t(x) = -\gamma \omega_t(x) + \lambda (d^* C_t)(x), \quad \lambda > 0$$

$$C_t(x, \mu) = \frac{1}{2g_0} \left\{ U_t(x, \mu) - U_t(x, \mu)^{-1} - \frac{1}{3} \text{tr} \left[U_t(x, \mu) - U_t(x, \mu)^{-1} \right] \right\}$$

NSPT in lattice QCD

Gauge damping

Gauge transformation

$$\pi(x, \mu) \rightarrow \Lambda(x)\pi(x, \mu)\Lambda(x)^{-1}, \quad U(x, \mu) \rightarrow \Lambda(x)U(x, \mu)\Lambda(x + \hat{\mu})^{-1}$$

Gauge damping field

$$\Lambda(x) = e^{\epsilon g_0 \omega(x)}, \quad \omega(x) \in \mathfrak{su}(3), \quad \omega(x) = \sum_{r=0}^N g_0^r \omega_r(x)$$

Gauge-damping dynamics (leading-order)

1. $\omega_0(x) \rightarrow c_1 \omega_0(x)$ $c_1 = e^{-\epsilon \gamma}$
2. $\omega_0(x) \rightarrow \omega_0(x) + \epsilon \lambda (d^* A)(x)$ $(d^* A)(x) = -\partial_\mu^* A_\mu(x)$
3. $A_\mu(x) \rightarrow A_\mu(x) - \epsilon (d\omega_0)_\mu(x)$ $(d\omega_0)_\mu(x) = \partial_\mu \omega_0(x)$

N.B.: $\omega(x)$ assures a restoring force on the longitudinal modes!

Convergence criteria

1. $\Delta|_{A^T} > 0$
2. $\epsilon^2 \|\Delta\| < \kappa$
3. $\epsilon^2 \lambda \|\ddot{d}d^*\| < 2(1 + c_1)$ **(N.B.:** $\|\ddot{d}d^*\| \leq 16$)

The Yang-Mills gradient flow in perturbation theory

In PTh, GF obs. are quite non-trivial objects,

$$B_\mu = \sum_{k=1}^{\infty} g_0^k B_{\mu,k} \quad \xrightarrow{\text{GF eq.}} \quad \partial_t B_{\mu,k} - \partial_\nu \partial_\nu B_{\mu,k} = R_{\mu,k}$$
$$B_{\mu,k}|_{t=0} = \delta_{k1} A_\mu$$

where e.g.

$$R_{\mu,1} = 0,$$

$$R_{\mu,2} = 2[B_{\nu,1}, \partial_\nu B_{\mu,1}] - [B_{\nu,1}, \partial_\mu B_{\nu,1}],$$

$R_{\mu,3}$ = complicated and not very illuminating expression
almost as long as this whole sentence ...

Simple at lowest order

$$B_{\mu,1}(t, x) = \int d^4y K_t(x-y) A_\mu(y), \quad K_t(z) = \frac{e^{-z^2/4t}}{(4\pi t)^2}$$

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Simple at lowest order ... but quickly becomes **involved**,

$$B_{\mu,k}(t, x) = \int_0^t ds \int d^4y K_{t-s}(x-y) R_{\mu,k}(s, y), \quad k = 2, 3, \dots$$

The gradient flow coupling

Lattice formulation

Gradient flow \rightarrow Wilson flow

$$\partial_t V_t(x, \mu) V_t(x, \mu)^{-1} = -g_0^2 (\partial_{x, \mu}^a S_G) (V_t) T^a, \quad V_{t=0}(x, \mu) = U(x, \mu)$$

N.B.: Solved using a Runge-Kutta scheme

Schrödinger functional

$$U(x, k)|_{x_0=0} = 1 = U(x, k)|_{x_0=T} \quad \Rightarrow \quad U_r(x, k)|_{x_0=0, T} = 0$$

Perturbative solution

(MDB, Hesse '13)

$$V_t(x, \mu) = e^{g_0 B_\mu(t, x)} = 1 + \sum_{r=0}^N g_0^{r+1} V_{t,r}(x, \mu), \quad V_{t,r}(x, \mu) = U_r(x, \mu)$$

Basic quantity

$$E(t, x) = g_0^2 E_0 + g_0^4 E_1 + g_0^6 E_2 + \dots$$

Coupling

[using $\alpha_{\overline{\text{MS}}} \leftrightarrow g_0$ (Lüscher, Weisz '95)]

$$t^2 \langle E(t, x) \rangle = k_0(a/L) \{ \alpha_{\overline{\text{MS}}}(q) + k_1(a/L) \alpha_{\overline{\text{MS}}}(q)^2 + k_2(a/L) \alpha_{\overline{\text{MS}}}(q)^3 + \dots \}$$

$$\lim_{a/L \rightarrow 0} k_i(a/L) = k_i, \quad i = 1, 2, 3$$