Numerical Stochastic Perturbation Theory in the Schrödinger Functional: **Opportunities & Challenges**

Dirk Hesse in collaboration with Michele Brambilla & Francesco Di Renzo

STRONGnet Workshop Madrid 2012

Overview

- What is NSPT and what is it useful for?
- Teaming up NSPT and the Schrödinger Functional.

useful for? hrödinger Functional.

The Schrödinge						
	Fermior					
Dirichlet boundary conditions in time.	$\zeta(\mathbf{x}) = \sum_{y} \tilde{K}(\mathbf{x}, y)$ $\overline{\zeta}(\mathbf{x}) = \sum_{y} \overline{\psi}(y) K$ $P_{+} \psi(x) _{x_{0}=0} = \overline{\psi}(x) F$					
Periodic boundary conditions in space .	$\psi(x + \hat{k}L) = e^{i\theta}$					

M. Lüscher, P.Weisz, R. Narayanan, U.Wolff, 1992. S. Sint, 1996, M. Lüscher 2006.

er Functional



The Schrödinge						
	Fermior					
Dirichlet boundary conditions in time.	$\zeta'(\mathbf{x}) = \sum_{y} \tilde{K}'(\mathbf{x}, y)$ $\overline{\zeta}'(\mathbf{x}) = \sum_{y} \overline{\psi}(y) K$ $P_{-}\psi(x) _{x_{0}=T} = \overline{\psi}(x) L$					
Periodic boundary conditions in space .	$\psi(x + \hat{k}L) = e^{i\theta}$					

M. Lüscher, P.Weisz, R. Narayanan, U.Wolff, 1992. S. Sint, 1996, M. Lüscher 2006.

er Functional



The Schrödinger Functional



M. Lüscher, P.Weisz, R. Narayanan, U.Wolff, 1992. S. Sint, 1996, M. Lüscher 2006.



Perturbation Theory

In **perturbation theory**, we want to obtain **expansions** like this one:

 $\langle O[\phi] \rangle = \frac{1}{Z} \int \mathcal{D}\phi \ O[\phi] \exp \{ S_0[\phi] + \alpha S_1[\phi] + \ldots \} = O^{(0)} + \alpha O^{(1)} + \ldots$

Usually, one calculates $O^{(i)}$ using Feynman diagrams and rules extracted from

- $S[\phi] = S_0[\phi] + \alpha S_1[\phi] + \dots$

PT in the SF

In the Schrödinger Functional one would like to **avoid** PT! t = 0t = T

- Big number of diagrams even at low orders
- Background field makes Feynman rules complicated

Numerical Stochastic PT avoids both!



Stochastic Quantization I

- We want to calculate an **expectation value** $\langle O[\phi] \rangle = \frac{1}{Z} \int \mathcal{D}\phi \ O[\phi] \ e^{-S[\phi]}$
- Introduce a new d.o.f., the stochastic time t.
- The evolution in stochastic time is given by the Langevin Equation,

$$\dot{\phi}_{\eta}(x;t) = -\partial_{\phi_{\eta}(x;t)}$$

With **Gaussian noise** η .

$_{t} S[\phi] + \eta(x;t)$

Parisi, Wu, 1981

Stochastic Quantization 2

Defining the **'noise average'**

$$\langle O \rangle_{\eta} = \frac{1}{Z'} \int \mathcal{D}\eta \ O e$$

One asserts that the **functional integral** can be calculated using

 $\langle O[\phi_{\eta}(x_1;t),\ldots,\phi_{\eta}(x_n;t)]\rangle_{\eta} \xrightarrow{t\to\infty} \langle O[\phi(x_1),\ldots,\phi(x_n)]\rangle$

 $\sum_{z} -\frac{1}{4} \int \mathrm{d}[z,\tau] \ \eta^2(z,\tau)$

Stochastic Perturbation Theory

Split up the action into free and interacting parts $S[\phi] = S_0[\phi] + gS_1[\phi] + \dots$

and formally write $\phi(x;t) = \sum_{i=0}^{\infty} g^i \phi^{(i)}(x;t)$ Using the Langevin equation one may now obtain $\langle O \rangle_{\eta} = \sum_{i=0}^{\infty} g^i \langle O \rangle_{\eta}^{(i)}$ i=0By defining $(\phi + \varphi)^{(r)} = \phi^{(r)} + \varphi^{(r)}$, $(\phi \varphi)^{(r)} = \sum_{r}^{r} \phi^{(i)} \varphi^{(r-i)}$

Numerical Stochastic Perturbation Theory

- In NSPT, one integrates the perturbative Langevin Equation numerically.
- This is similar to hybrid MC methods c.f. Stefan's talk.
- However, there is no perturbative expression for an accept/reject step.
- Hence, one is stuck with a finite integration time τ and has to extrapolate $\tau \to 0$.

Di Renzo, 2004

Stochastic Gauge Fixing

In principle no GF is need. However, if one looks at the Langevin eqn.,

$$\frac{\partial}{\partial t}A^a_{\mu}(\eta, x; t) = D^{ab}_{\nu}F^b_{\nu\mu}(\eta, x; t) + \eta^a_{\mu}(x; t)$$

One finds that for a solution in Fourier space

$$A^{(n)a}_{\mu}(k;t) = T^{ab}_{\mu\nu} \int_0^t ds \, e^{-k^2(t-s)} f^{(n)b}_{\nu}(k,s) + L^{ab}_{\mu\nu} \int_0^t ds \, f^{(n)b}_{\nu}(k,s)$$

The longitudinal component will diverge like a random walk. GF introduces a damping factor and stabilizes the simulation. A similar statement holds for the gauge zero modes.

Zwanziger, 1981

Gauge Fixing Pitfalls

The gauge fixing function in the SF at the boundary reads

$$[d^*q(x)]_{ij} = \begin{cases} (a^2/L^3) \sum_{\mathbf{y}} [a^2/L^3] \\ 0 \end{cases}$$

It acts on the fluctuation field, writing

$$U_{\mu}(x) = \exp\{g_0\}$$

This amounts to suppressing **spatial zero modes** at the boundary.

 $[q_0(0,\mathbf{y})]_{ij}$ if $x_0 = 0, i = j$ else

 $V_{\mu}(x) V_{\mu}(x)$

Lüscher, Narayanan, Weisz, Wolff, 1992



Incorrect gauge fixing leads to a slow increase of noise (which can be confusing).



Correct gauge fixing leads to a drastic decrease of the noise.

Advantages of NSPT

- No Feynman Rules, i.e. easier to implement various actions.
- No Feynman Diagrams, i.e. higher order are only a matter of CPU time.
- NSPT codes can benefit from non-perturbative ones ...
- ... and vice versa!

No Free Lunch

- Way more numerical effort than common PT. Stochastic noise makes extraction of logs difficult.
 - But NSPT has many applications, e.g. improvement:

$$\mathcal{O}_{\mathrm{I}}(a/L) = \frac{1}{1}$$

$$\delta(a/L) = \frac{\mathcal{O}(a/L) - \mathcal{O}(0)}{\mathcal{O}(0)} = \delta^{(0)}$$

$$\frac{\mathcal{O}(a/L)}{+\,\delta(a/L)}$$

 $^{(0)}(a/L) + g_0^2 \delta^{(1)}(a/L) + \dots$

de Divitiis et al. Nucl. Phys. B437

The SF coupling

The SF introduces an external scale T = cL (usually c = I). This enables us to give a precise definition for a coupling $\overline{q}(\mu = 1/L)$

Furthermore one can calculate a discrete beta-function

$$u = \overline{g}^2(L), \quad u' = \overline{g}^2(sL), \quad \sigma(s, u) = u'$$

On the lattice: $\Sigma(s, u, a/L) = u'$

Lüscher, Narayanan, Weisz, Wolff, 1992

Defining the SF coupling

The usual choice for the background field is 'defined' by setting $C = i/L \operatorname{diag}(\phi_1, \phi_2, \phi_3), \quad C' = i/L \operatorname{diag}(\phi'_1, \phi'_2, \phi'_3)$ Where ϕ_i, ϕ'_i depend on two parameters, ν and η . Defining as usual $e^{-\Gamma}$

one can define a coupling throu

$$\begin{split} \hat{D} &= \int D[u] e^{-S[U]} \\ \text{Jgh} \left. \frac{\partial \Gamma}{\partial \eta} \right|_{\eta = \nu = 0} = \frac{k}{\overline{g}^2} \end{split}$$

Lüscher, Sommer, Weisz, Wolff, 1993

O(a) improvement

To implement O(a) improvement, one defines

$$S[U] = \frac{1}{g^2} \sum_{p} w(p) \, \operatorname{tr}[1 - U(p)]$$

For plaquettes at the boundary, containing a spatial link, one sets

$$w(p) = c_t(g_0) = 1 + c_t^{(1)}g^2 + c_t^{(1)}g^2$$

... else w(p) = 1

Lüscher, Sommer, Weisz, Wolff, 1993

The Coupling Order by Order

We then write the coupling order by order,

$$\bar{g}^2(L) = g_0^2 + m_1(L/a)g_0^4$$

with

$$m_1 = m_1^a + c_t^{(1)} m_1^b,$$

$$m_2 - m_1^2 = m_2^a + c_t^{(1)} m_2^b + \left[c_t^{(1)}\right]^2 m_2^c + c_t^{(2)} m_2^d$$

We use this as a check for our quenched code!

 $+ m_2(L/a)g_0^6 + \dots$

Bode, Weisz, Wolff, 1997

Data Analysis

- We use a custom implementation of the method presented by Ulli Wolff (CPC 156, 2004) to control autocorrelation and determine the errors.
- The extrapolation $\tau \to 0$ was done using a bootstrap analysis.
- We performed a cross check using binned Jackknife samples.

Our Data

- Did **test runs** on L/a = 4,8,12
- Coupling is known to be hard to measure
- We experience
 - Lots of **noise**
 - Slow thermalization
 - Long autocorrelation times

Simulation Details

• L = 4 on TURING (Univ. of Milan Bicocca) • L = 8, 12 on FERMI (CINECA)

This is an ideal test case to **check core performances**!

	L = 12			L = 8			L = 4		
au	0.0015	0.003	0.005	0.0015	0.003	0.005	0.001	0.002	0.003
$N_{\rm eff}$	270(50)	540(80)	800(100)	I.6(2)k	3.3(2)k	6.3(3)k	6.6(3)k	l4.9(5)k	21.8(6)k
CORE-H	2k			I.6k			33		

www.hpc.cineca.it www.mib.infn.it



Known results (red) form: Bode, Weisz, Wolff: hep-lat/9809175

Results - Two Loops



Known results (red) form: Bode, Weisz, Wolff: hep-lat/9809175

Conclusions

- Quenched NSPT for the SF is implemented.
- Applications: Cut-off effects.
- Next Step: Fermions!

A word about logs.

An observable with at most a logarithmic divergence looks like this



One can extract the coefficients using successive fits.

But: One requires many data points with high precision.

M. Lüscher, P. Weisz, 1996,

- A. Bode, P. Weisz, U. Wolff, 2000.