

# Exploiting the hopping parameter expansion in the HMC simulation of lattice QCD

Martin Hasenbusch

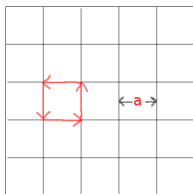
Humboldt-Universität zu Berlin

Frontiers in Lattice Quantum Field Theory, 1 June 2018  
Instituto de Física Teórica UAM-CSIC, Madrid

# Overview

M. Hasenbusch, [arXiv:1805.03560]

- ▶ The model; Improved pseudo fermions
- ▶ UV-filtering, rooted polynomials
- ▶ Numerical Results
- ▶ Conclusions



Lattice QCD:

- 4 dimensional hyper-cubic lattice
- $x$  sites of the lattice,  $\mu$  direction
- Gauge field  $U_{x,\mu} \in SU(3)$  lives on the link  $(x, \mu)$
- quark fields live on the sites

Wilson gauge action

$$S_G[U] = -\frac{\beta}{3} \sum_x \sum_{\mu > \nu} \text{Re Tr} \left( U_{x,\mu} U_{x+\hat{\mu},\nu} U_{x+\hat{\nu},\mu}^\dagger U_{x,\nu}^\dagger \right)$$

Wilson fermions

$$H[U] = \sum_{\mu} \left\{ (1 - \gamma_{\mu}) U_{x,\mu} \delta_{x+\hat{\mu},y} + (1 + \gamma_{\mu}) U_{x-\hat{\mu},\mu}^\dagger \delta_{x-\hat{\mu},y} \right\}$$

$$M[U] = 1 - \kappa H[U]$$

## QCD according to the particle data book 2017:

$$m_u = 2.2_{-0.4}^{+0.6} \text{ MeV} \quad m_d = 4.7_{-0.4}^{+0.5} \text{ MeV} \quad m_s = 96_{-4}^{+8} \text{ MeV}$$
$$m_c = 1.28 \pm 0.03 \text{ GeV} \quad m_b = 4.18_{-0.03}^{+0.04} \text{ GeV} \quad m_t = 173.1 \pm 0.6 \text{ GeV}$$

We simulate: Two degenerate flavours of dynamical quarks

$$m_u = m_d \text{ finite} \quad m_s = m_c = m_b = m_t = \infty$$

Fermions (Grassmann variables) can be **integrated out**:

$$Z = \int D[U] \exp(-S_G[U]) \det M[U]^2$$

# Hybrid Monte Carlo (HMC) Algorithm

determinant is too expensive ( $\propto \text{Volume}^3$ )  $\rightarrow$  pseudo-fermions

$$\det M^2 = \det M M^\dagger \propto \int D[\phi^\dagger] \int D[\phi] \exp(-|M^{-1}\phi|^2)$$

Problem:  $S_F = |M^{-1}\phi|^2$  is non-local

$\implies$  molecular dynamics evolution of all gauge fields.

Introduce conjugate momenta  $P$  for the gauge field

$\implies$  Hamiltonian:

$$H(U, \phi, \phi^\dagger, \Pi) = S_G(U) + S_F(U, \phi, \phi^\dagger) + \frac{1}{2} (\Pi, \Pi)$$

where  $(\Pi, \Pi) = -2 \sum_{x,\mu} \text{Tr} \Pi_{x,\mu}^2$

## What is the problem?

The simulation becomes **more expensive** as the **quark mass** becomes **smaller**, (Lattice 2001, Berlin “Berlin wall”):

$$\text{cost} = m_{PS}^{-2.8(2)} \quad (\text{for } \beta = 5.6, \text{ Lippert 2001})$$

- ▶ **Condition number** of  $M$  **increases**  
⇒ solver needs more iterations
- ▶ **step size must be decreased** to get constant acceptance

## Improved pseudo-fermions

Introduce  $N$  matrices  $W_i$  such that

$$M = \prod_{i=1}^N W_i$$

The  $W_i$  should have a smaller condition number than  $M$   
 Introduce pseudo-fermions for each  $W_i$

$$\det MM^\dagger \propto$$

$$\int D[\phi_1^\dagger] \int D[\phi_1] \dots \int D[\phi_N^\dagger] \int D[\phi_N] \exp\left(-\sum_{i=1}^N |W_i^{-1} \phi_i|^2\right)$$

- Mass-preconditioning (Hasenbusch 2001):

$$W_1 = M + \rho_1$$

$$W_i = (M + \rho_{i-1})^{-1}(M + \rho_i) \quad \text{for } 1 < i < N$$

$$W_N = (M + \rho_{N-1})^{-1}M$$

Alternative (R. Sommer; Hasenbusch and Jansen 2003): add  $\rho_i \gamma_5$

- Polynomial splitting (Peardon 2001)

formally the same splitting as in PHMC, however **low order polynomial**

$$W_1^{-1} = P(M) \approx M^{-1}$$

Easy to combine with “UV-filtering” or PHMC.



- **RHMC** (Clark, Kennedy 2004): take the  $n^{\text{th}}$  root of  $M$ , introduce a pseudo-fermion field for each of the roots. Technically done with a rational approximation. Requires a multi-mass solve for each of the roots.
  
- **Schwarz preconditioned HMC** (Lüscher 2004)

# The hopping parameter expansion

$$\det M^\dagger M = \exp(\text{Tr} \ln M^\dagger + \text{Tr} \ln M)$$

one expands

$$\ln M = \ln(1 - \kappa H) = - \sum_{n=1}^{\infty} \frac{1}{n} \kappa^n H^n .$$

For small  $n$ ,  $\text{Tr} H^n$  can be **evaluated analytically**.

$n = 4$ : shift in the gauge-coupling  $\beta$ .

**Clover-improvement**: already **non-trivial** contribution from  $n = 2$

K.-I. Ishikawa *et al.*,

arXiv:hep-lat/0610037, PoS LAT **2006**, 027 (2006).

- M.H., hep-lat/9807031, Phys.Rev. D 59 (1999) 054505;
- Ph. de Forcrand, *UV-filtered fermionic Monte Carlo*, hep-lat/9809145, Nucl.Phys.Proc.Suppl. 73 (1999) 822.

$$\tilde{M} = M \exp \left( \sum_{n=1}^k \frac{1}{n} \kappa^n H^n \right)$$

The inverse  $\tilde{M}^{-1}$  can be represented by a **polynomial** in  $M$ .

→ **Multiboson algorithm**

C. Alexandrou, Ph. de Forcrand, M. D'Elia, and H. Panagopoulos, *Efficiency of the UV-filtered Multiboson algorithm*, [arXiv:hep-lat/9906029], Phys. Rev. D **61**, 074503 (2000).

*... reduces the number of bosonic fields by a factor 3 or more ...*

→ Polynomial Hybrid Monte Carlo (PHMC)

K.-I. Ishikawa *et al.* [PACS-CS Collaboration], *An Application of the UV-filtering preconditioner to the polynomial hybrid Monte Carlo algorithm*, [arXiv:hep-lat/0610037] PoS LAT **2006**, 027 (2006).

*... UV-filtering reduces the magnitude of the molecular dynamics force from the pseudo fermion by a factor 3 by tuning the UV-filter parameter. Combining with the multi-time scale molecular dynamics integrator we achieve a factor 2 improvement.*

Here, simplest example:

$$\tilde{M}^{-1} = \exp(-\kappa H)(1 - \kappa H)^{-1} = \sum_{n=0}^{\infty} a_n \kappa^n H^n$$

$$a_n = \sum_{i=0}^n (-1)^i \frac{1}{i!} \quad , \quad \lim_{n \rightarrow \infty} a_n = \exp(-1)$$

Hence we can write

$$\tilde{M}^{-1} = \sum_{n=0}^{\infty} b_n \kappa^n H^n + \alpha M^{-1}$$

where  $\alpha = \exp(-1)$  and  $b_n = -\sum_{i=n+1}^{\infty} (-1)^i \frac{1}{i!}$

$\implies \sum_{n=0}^{\infty} b_n \kappa^n H^n$  can be truncated at low order

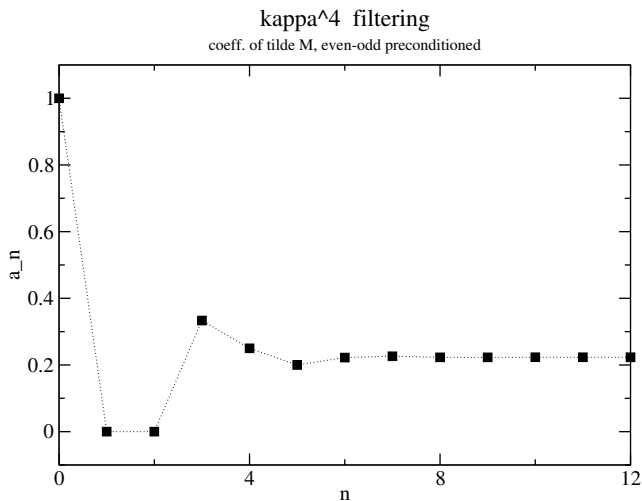
## Even-odd preconditioning

$$M_{oo} = 1 - \kappa^2 H_{oe} H_{eo}$$

Reduces the condition number of the fermion matrix  
Larger step size (de Forcrand, Takaishi 1996)

Order of polynomials in the following: powers of  $\kappa^2$

$M_{oo} = 1 - \kappa^2 H_{oe} H_{eo}$ . Coefficient  $n$  refers to powers of  $\kappa^2$



## Comparison with mass preconditioning

$$W_1 = M + \rho$$

$$W_2 = (M + \rho)^{-1}M$$

Taking the inverse

$$W_2^{-1} = 1 + \rho M^{-1}$$

Hence  $\alpha$  plays a similar role as  $\rho$



Generalization by using factorized rooted polynomials

Idea: **Noise reduction by rooting**, similar to **RHMC**

$M_1 = \tilde{M}$ , define recursively

$$M_{j+1} = W_j^{-N_j} M_j$$

where

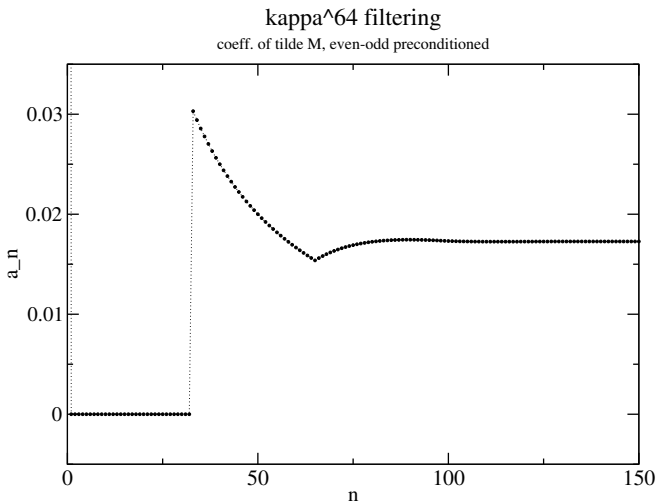
$$W_j^{-1} = \sum_{i=0}^{n_j} a_{j,i} \kappa^i H^i = M_j^{-1/N_j} + O(\kappa^{n_j+1})$$

The remainder

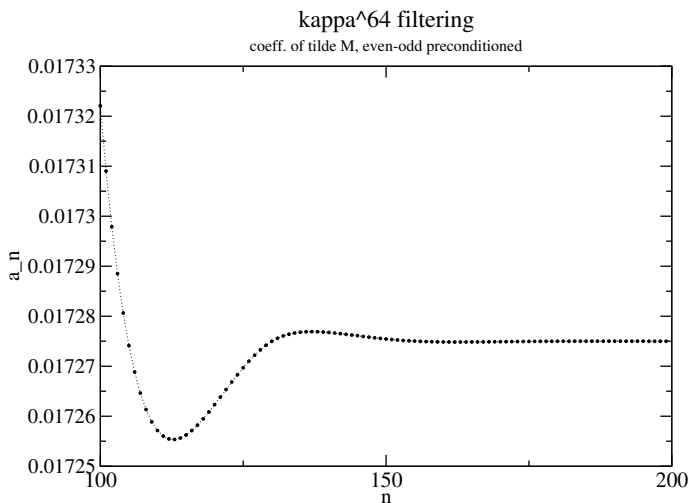
$$M_{j_{max}+1}^{-1} = \sum_{n=0}^{\infty} b_n \kappa^n H^n + \alpha M^{-1}$$

Computed by using an algebra program

$M_{oo} = 1 - \kappa^2 H_{oe} H_{eo}$ . Coefficient  $n$  refers to powers of  $\kappa^2$

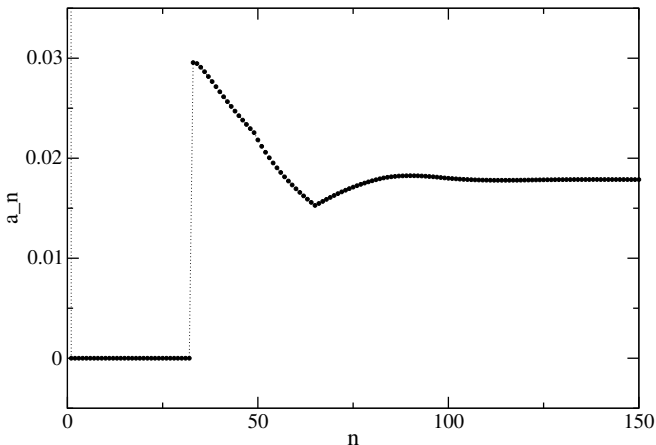


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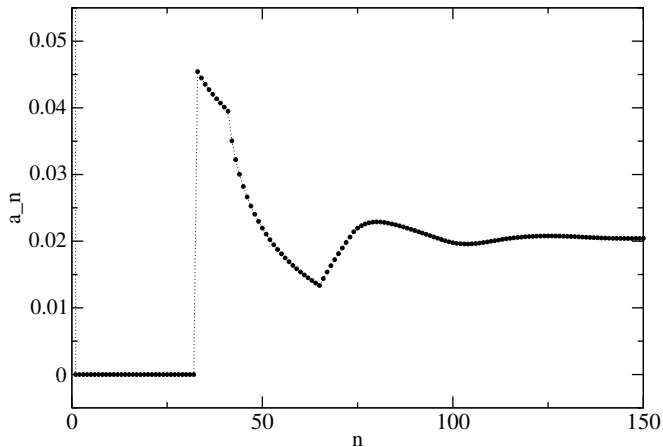
Polynomial filtering,  $n_1=8$ ,  $n_2=32$ ,  $N=8$   
coeff. of tilde M, even-odd preconditioning



$M_{oo} = 1 - \kappa^2 H_{oe} H_{eo}$ . Coefficient  $n$  refers to powers of  $\kappa^2$

Polynomial filtering  $n_1=8, n_2=32, N=3$

coeff. of tilde M, even-odd preconditioning



## Computing forces

Rooted Polynomials:

PHMC: Horner scheme; we need to store  $n$  vectors, where  $n$  is the order of the polynomial

Remainder  $M_{j_{max}+1}^{-1}$ :

extra effort due to the additive polynomial

$$\sum_{n=0}^{\infty} b_n \kappa^n H^n$$

That can be (hopefully) truncated at low order.

## Integration schemes

$$\begin{aligned}
 P(\delta\tau) : \Pi_{x,\mu} &\rightarrow \Pi'_{x,\mu} &= \Pi_{x,\mu} + \delta\tau \mathcal{F}_{x,\mu} , \\
 T(\delta\tau) : U_{x,\mu} &\rightarrow U'_{x,\mu} &= \exp(i\delta\tau \Pi_{x,\mu}) U_{x,\mu} .
 \end{aligned}$$

Second order Omelyan integrator

$$T_O = P(\lambda\delta\tau) T(\delta\tau/2) P([1 - 2\lambda]\delta\tau) T(\delta\tau/2) P(\lambda\delta\tau)$$

$\lambda = 1/6$  the scheme proposed by Sexton and Weingarten

$\lambda = 1/2$  leapfrog scheme

$$T_L = P(\delta\tau/2) T(\delta\tau) P(\delta\tau/2)$$

Multi-time scale following Sexton and Weingarten

At the end of the trajectory the new gauge field is accepted with the probability

$$P_{acc} = \min[1, \exp(-\Delta H)]$$

where

$$\Delta H = H(U', \phi, \phi^\dagger, \Pi') - H(U, \phi, \phi^\dagger, \Pi)$$

Various authors  $\approx$  1990

$$P_{acc} = \operatorname{erfc} \left( \sqrt{\operatorname{Var}(\Delta H)/8} \right) = 1 - \frac{1}{\sqrt{2\pi}} \sqrt{\operatorname{Var}(\Delta H)} + \dots$$

Bussone et al. 2018: For the second order Omelyan with  $\lambda = 1/6$ :

$$\operatorname{Var}(\Delta H) = \frac{2\delta\tau^4}{72^2} \left[ \operatorname{Var}(|\mathcal{F}_{i_{max}}|^2) + \frac{\operatorname{Var}(|\mathcal{F}_{i_{max}-1}|^2)}{(4m_{i_{max}-1}^2)^2} + \dots \right]$$



## Numerical test

Numerical studies at  $\beta = 5.6$ ,  $\kappa = 0.156$  and  $0.1575$ . Extensively studied by SESAM and Gal collaboration (See B. Orth, T. Lippert, and K. Schilling, [arXiv:hep-lat/0503016], Phys. Rev. D **72**, 014503 (2005).)

Used in algorithmic studies e.g. Lüscher 2004, Urbach, Jansen, Shindler and Wenger 2005

$\kappa = 0.156$ :  $a = 0.09796(56)$  fm,  $m_{PS} = 0.9002(69)$  GeV

$\kappa = 0.1575$ :  $a = 0.0839(11)$  fm,  $m_{PS} = 0.6524(86)$  GeV

Real world:  $m_{\pi^0} \approx 135$  MeV.

Trajectory length:  $\tau = \sqrt{2}$  throughout

$\kappa^k$ -filtering,  $12^3 \times 24$  lattice,  $\beta = 5.6$ ,  $\kappa = 0.156$ . Leapfrog scheme.

k	m	stat	$\langle P \rangle$	$P_{acc}$	$\text{Var}(\Delta H)$
0	42	2770	0.56982(7)	0.8006(43)	0.2673(54)
2	21	7050	0.56991(6)	0.7981(26)	0.2643(43)
4	16	7610	0.56995(4)	0.8106(24)	0.2264(40)

k	$\text{Var}( \mathcal{F}_{PF} ^2)$
0	11400000(200000)
2	344000(4000)
4	57500(1000)

$\kappa^4$ -filtering,  $16^3 \times 32$ ,  $\kappa = 0.1575$ : Speed-up by roughly a factor of 3 compared with SESAM, Gral

$\kappa^4$ -filtering; Truncation of  $\sum_{n=0}^{\infty} b_n \kappa^n H^n$ , force calculation

$n_t$	stat	$P_{acc}$	$\text{Var}(\Delta H)$
3	200	0.22(3)	4.98(59)
4	1030	0.177(8)	8.12(24)
5	6400	0.8631(22)	0.1180(31)
6	2200	0.8506(45)	0.1512(63)
7	2200	0.8868(32)	0.0920(36)
8	2000	0.8845(31)	0.0848(29)
9	2200	0.8851(36)	0.0904(33)
15	24500	0.8830(15)	0.0886(15)

## Rooted polynomials

$$16^3 \times 32, \kappa = 0.1575$$

Order of the polynomials  $n_1 = 8, n_2 = 32$   
(ad hoc choices to get a first idea)

Roots:  $N = 2, 3, 4, 6, 8,$  and  $16$

Leap-frog integration scheme with multiple time scales

$N$	stat	$m_0$	$m_1$	$m_2$	$m$	$\alpha$	$\langle P \rangle$	$P_{acc}$
2	290	6	6	4	8	0.022110...	0.57279(6)	0.870(11)
3	500	10	6	3	6	0.020504...	0.57255(6)	0.788(9)
4	910	6	5	4	5	0.019687...	0.57258(3)	0.793(9)
6	600	10	5	2	5	0.018872...	0.57256(5)	0.773(11)
8	500	6	5	2	5	0.018467...	0.57254(5)	0.792(12)
16	200	6	5	2	5	0.017866...	0.57259(6)	0.806(17)

Variances of the force:

$N$	$G$	$PF, 1$	$PF, 2$	$PF, 3$
2	85000000(4500000)	1110000(60000)	11000(900)	1300(120)
3	82000000(4700000)	290000(15000)	2020(180)	540(60)
4	84000000(3500000)	114000(4000)	710(100)	360(60)
6	77000000(4000000)	42400(2000)	197(17)	156(15)
8	79000000(4000000)	20200(1000)	81(8)	123(14)
16	83000000(6000000)	4860(400)	16.5(3.0)	156(40)

Cost index related to the terms  $S_{PF,1}$  and  $S_{PF,2}$

$N$	$8N \text{Var}( \mathcal{F}_{PF,1} ^2)^{1/4}$	$32N \text{Var}( \mathcal{F}_{PF,2} ^2)^{1/4}$
2	519(7)	655(13)
3	557(7)	644(14)
4	588(5)	661(22)
6	689(8)	719(15)
8	763(9)	768(18)
16	1069(21)	1032(44)

## Using $\kappa^4$ -filtering

$$n_1 = 12, n_2 = 42, N = 8$$

$\alpha = 0.01390254\dots$ ; For comparison  $N = \infty$ :  $\alpha = 0.01321050\dots$

$m = 4, m_2 = 2, m_1 = 3, m_0 = 40$ , and  $n_t = 160$ .

The acceptance rate is  $P_{acc} = 0.790(10)$  and  $\text{Var}(\Delta H) = 0.249(22)$ .

$$\text{Var}(|\mathcal{F}_{PF,1}|^2) = 2370(160)$$

$$\text{Var}(|\mathcal{F}_{PF,2}|^2) = 23.8(2.5)$$

$$\text{Var}(|\mathcal{F}_{PF,3}|^2) = 42(5)$$

## Conclusions and Outlook

- ▶ **Speed-up** of factor of **2** or **3** by using  $\kappa^2$  and  $\kappa^4$  filtering
- ▶ Solver can be used to compute the remainder
- ▶ **Extension to higher orders** by using **rooted polynomials**
- ▶ Promising results; strong reduction of forces and their variances
- ▶ Many parameters in the game; need more insight to fix them
- ▶ New chance for local finite step updates?