# Algorithms in Lattice QCD V

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What happened so far

Methods for Markov Chain Monte Carlo

Sequence of field configuratoins

 $\rightarrow$ MC time series of measurements

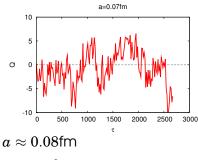
Field updates are expensive  $\rightarrow$  limited statistics

#### Outline for today

Methods to deal with autocorrelations

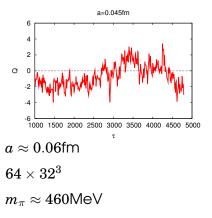
## Bad start

#### Topological charge

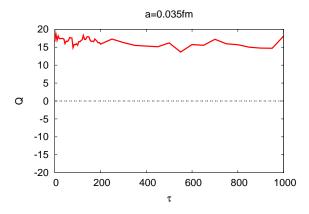


 $64 imes 32^3$ 

 $m_\pi pprox 360 {
m MeV}$ 



## A bad start



a pprox 0.04fm

 $128\times 64^3$ 

 $m_\pi pprox 480 {
m MeV}$ 

# Markov Chain Monte Carlo

Sequence of field configurations

$$U_1 o U_2 o U_3 o \cdots o U_N$$

Generated by a **transition probability** density

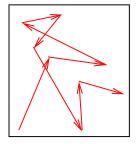
$$T(U' \leftarrow U) \geq 0$$
 for all  $U, U'$ 

Stability

$$\int [dU] \, T(U' \leftarrow U) \, P[U] = P[U']$$

Normalization

$$\int [dU'] \, T(U' \leftarrow U) = 1$$



#### **Autocorrelations**

Sequence of field configurations

$$U_1 o U_2 o U_3 o \cdots o U_N$$

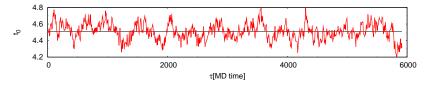
Measurements of observables are correlated

$$A_1 o A_2 o A_3 o \cdots o A_N$$

Estimates

$$ar{A}pprox ilde{A}_N = rac{1}{N}\sum_{i=1}^N A_i$$

How far is this off?



## Autocorrelations

Variance of estimator

$$\langle\!\langle ( ilde{A}_N-ar{A})^2
angle\!
angle=rac{1}{N^2}\sum_{i,j=1}^N\langle\!\langle (A_i-A)(A_j-A)
angle\!
angle$$

For N large, this depends only on the difference in simulation time

$$egin{aligned} &\langle ( ilde{A}_N-ar{A})^2 
angle = rac{1}{N}\sum_{t=-\infty}^{\infty}\Gamma_A(t) \ &\Gamma_A(t) = \langle (A_0-ar{A})(A_t-ar{A}) 
angle \end{aligned}$$

Note:

again substitution average over simulations

 $\rightarrow$  average in simulation time

## Error of the measurement

$$egin{aligned} &\langle ( ilde{A}-ar{A})^2 
angle &= rac{1}{N}\sum_{t=-\infty}^{\infty}\Gamma_A(t) \;; & \Gamma_A(t) \;\; = \langle (A_0-A)(A_t-A) 
angle \ &= rac{ ext{var}(A)}{N}\sum_{t=-\infty}^{\infty}
ho_A(t) \end{aligned}$$

Integrated autocorrelation time

$$\tau_{\rm int}(A) = \frac{1}{2} + \sum_{t=1}^{\infty} \frac{\Gamma_A(t)}{\Gamma_A(0)} \equiv \frac{1}{2} + \sum_{t=0}^{\infty} \rho_A(t)$$

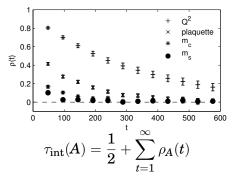
Error of the measurement

$$\sigma_A = \sqrt{rac{\mathrm{var}(A)}{N/(2 au_{\mathrm{int}}(A))}}$$

Measures efficiency of algorithm.

 $\rightarrow$  eff. statistics reduced by  $2\tau_{int}$ 

### Measuring autocorrelations



We only have a limited precision estimate of the integrand.

Summing to  $t = \infty$  leads to diverging variance.

 $\rightarrow$  need to cut the summation

 $\rightarrow$  biased estimate

Need to find a balance between stat. and syst. error.

# Error of $\tau_{int}$

$$au_{ ext{int}}(A) = rac{1}{2} + \sum_{t=1}^W rac{\Gamma_A(t)}{\Gamma_A(0)}$$

#### Systematic error

Summation truncated at W  $\rightarrow$  neglect potentially large tail. Particular problem in presence of slow modes.

Statistical error

Madras,Sokal

$$\langle [ ilde{ au}_{ ext{int}}(A,W) - au_{ ext{int}}(A,W)]^2 
angle pprox rac{4}{N} (W + rac{1}{2} - au_{ ext{int}}(A)) au_{ ext{int}}(A)^2$$

Infinite variance for  $W \to \infty$ .

# Criteria for W

All automatic methods are problematic.

• Cut where  $\delta\Gamma > \Gamma$ 

 $\rightarrow$  large systematic error

Madras-Sokal criterion

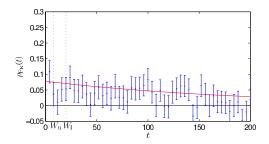
 $\rightarrow$  minimum of sum of systematic and statistical error

$$rac{\delta\sqrt{ au}}{\sqrt{ au}} \propto \min_W \left( e^{-W/ au} + 2\sqrt{W/N} 
ight)$$

ALPHA method (2010) Estimate  $\tau_{exp}$  from various (slow) observables Add tail to all other observables before losing signal  $\Gamma_A$ 

# Attaching the tail

ALPHA collaboration, 2012



Use slow observables and scaling laws to estimate tail.

#### Detailed balance

#### Detailed balance

$$T(U' \leftarrow U) P[U] = P[U'] T(U' \leftarrow U)$$

Implies stability

$$\int [dU]T(U' \leftarrow U)P[U] = \int [dU]P[U']T(U' \leftarrow U) = P[U']$$

Elementary steps frequently fulfill this condition.

As a consequence we have a symmetric matrix M

$$M(U' \leftarrow U) = P[U']^{-1/2}T(U' \leftarrow U)P[U]^{1/2}$$

#### Detailed balance

Detailed balance

$$T(U' \leftarrow U) \, P[U] = P[U'] \, T(U' \leftarrow U)$$

Associated symmetric matrix M

$$M(U' \leftarrow U) = P[U']^{-1/2}T(U' \leftarrow U)P[U]^{1/2}$$

If  $\eta$  eigenvector of T

$$\xi(U)=P^{-1/2}(U)\eta(U)$$

is eigenvector of M with the same eigenvalue  $\lambda$ .

Spectral decomposition

$$M = \sum_i \lambda_i \, \xi_i \xi_i^\dagger$$

### Autocorrelation

Spectral decomposition

$$egin{aligned} &\Gamma_A(t) \ &= \ \langle \left(A_t - ar{A}
ight) \left(A_0 - ar{A}
ight) 
ight
angle \ &= \ \int [dU] [dU'] \delta A(U') \, T^t(U' \leftarrow U) \, \delta A(U) \, P[U] \ &= \ \int [dU] [dU'] P^{1/2} [U'] \delta A(U') M^t(U' \leftarrow U) \delta A(U) P^{1/2} [U] \ &= \ \sum_{n > 0} (\lambda_n)^t \, [c_n(A)]^2 \end{aligned}$$

With "matrix elements"

$$c_n(A) = \int [dU] \xi_n(U) [P[U]]^{1/2} (A(U) - ar{A})$$

### Spectral representation

$$egin{aligned} \Gamma_A(t) &= \sum_n \left(\lambda_n
ight)^t \left[c_n(A)
ight]^2 \ &= \sum_n \mathrm{sign}\lambda_n \, e^{-t/ au_n} \, [c_n(A)]^2 \end{aligned}$$

- For the analysis of algorithms it is useful to think of Monte Carlo time t as a fifth dimension.
- Autocorrelation function is a 2pt function.
- time constants  $au_n 
  ightarrow$  inverse masses
- Slowest decay  $\tau_1 \rightarrow$  exponential AC time

## Comments

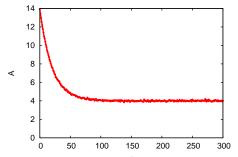
$$\Gamma_A(t) = \sum_n e^{-t/ au_n} c_n^2(A)$$

- $\tau_n$  depend only on algorithm
- Matrix elements  $c_n$  depend on observable.
- All observables affected by slow modes.

#### Length of a simulation

- Simulation must have length of at least  $O(100) \times \tau_1$ .
- $au_{\mathrm{int}}(A)$  can be much smaller than  $au_1$
- Danger of
  - Incomplete thermalization.
  - Bias.
  - Wrong estimate of autocorrelations.

## Thermalization



Same decay rates contribute  $\dot{a}$ s in  $\tau_{int}$  different initial distribution/matrix elements

$$\begin{split} &\int [dU] [dU'] P^{1/2} [U'] \delta A(U') M^t (U' \leftarrow U) \delta A(U) \frac{P_0[U]}{P^{1/2}[U]} \\ &= \sum_{n > 0} (\lambda_n)^t \, [c_n(A)] [c_n^{(0)}(A)] \end{split}$$

Opportunity to learn about largest  $\tau_1$ .

#### Observables

Need to look at observables with large  $c_i(A_k)$ , i small. Hunt for slow quantities.

Noise can cover up auto-correlations ( $\eta$  Gaussian noise)

$$egin{array}{rl} A o A + c\eta & \Rightarrow & \Gamma(t) o \Gamma(t) + c^2 \delta_{t,0} \ & \Rightarrow & au_{ ext{int}}(A) o au_{ ext{int}}(A) rac{ ext{var}(A)}{ ext{var}(A) + c^2} \end{array}$$

Look at low-noise observables

Take into consideration expected scaling properties

$$au_{
m int} \propto rac{1}{a^2}$$
 for  $a 
ightarrow 0$ 

### Wilson flow

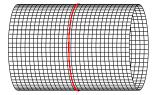
LÜSCHER'10, LÜSCHER&WEISZ'11

Smoothing with gradient flow at fixed flow time  $t = t_0$ .

 $\partial_t V_t(x,\mu) = -g_0^2 \left[\partial_{x,\mu} S(V_t)\right] V_t(x,\mu); \quad V_t(x,\mu)|_{t=} = U(x,\mu)$ 

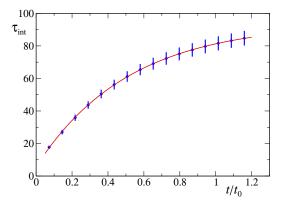
- Gaussian smoothing over  $r \sim \sqrt{8t}$ .
- Renormalized quantities with continuum limit.
- $\blacksquare$  Smooth observables  $\rightarrow$  long autocorrelations.

$$ar{E}=-rac{a^3}{2L^3}\sum_{ec{x}}\mathrm{tr}\,G_{\mu
u}G_{\mu
u}ig|_{x_0=T/2} 
onumber\ egin{array}{c} \overline{Q}=-rac{a^3}{32\pi^2}\sum_{ec{x}}\mathrm{tr}\, ilde{G}_{\mu
u}G_{\mu
u}ig|_{x_0=T/2} 
onumber\ Q=-rac{a^4}{32\pi^2}\sum_{ec{x}}\mathrm{tr}\, ilde{G}_{\mu
u}G_{\mu
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# Effect of the smoothing

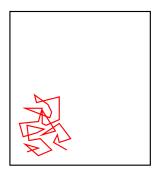
Autocorrelation time of  $\overline{E}$  vs smoothing range (a=0.05fm).



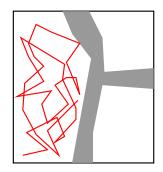
■  $\sqrt{8t}$  smoothing radius  $\rightarrow t = t_0$  smoothing over  $r \approx r_0$ ■  $\tau_{\text{int}}$  saturates with  $\tau_{\text{int}} = 93 + ae^{-ct}$ .

### Dangers

- Algorithm is slow.
- Detectable by measuring autocorrelations.



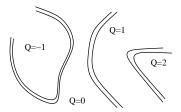
- There are barriers in field space.
- Hard to detect.



## Topological charge

$$Q=-rac{1}{32\pi^2}\int d\,x\,\epsilon_{\mu
u
ho\sigma}{
m tr}\,F_{\mu
u}F_{
ho\sigma}$$

- In continuum limit, disconnected topological sectors emerge.
- The probability of configurations "in between" sectors drops rapidly.
   M. LÜSCHER, '10
- Simulations get stuck in one sector.



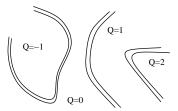
### Topological charge

#### Tunneling is a cut-off effect.

- Quasi continuous algorithms will not cure it.
- Problem for interpretation of data.
- Fixed topology introduces finite volume effects.

$$\langle A 
angle = \langle A 
angle_{oldsymbol{Q} = oldsymbol{Q}_0} \, \cdot \, \{1 + \mathcal{O}(V^{-1})\}$$

Prevents simulations on fine lattices.



#### Topological charge

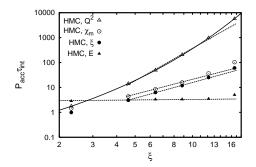
Best recipe: Avoid large autocorrelations  $\tau_n$ .

Special case: Topological charge

In the continuum, topological sectors form. Consequence of the periodic boundary conditions.

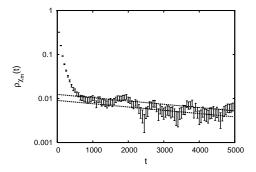
Happens very quickly as  $a \rightarrow 0$ .

Engel, S.'10



# Example from $CP^{N-1}$ model

Topological charge AC dominates other observables.



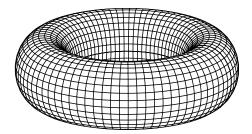
#### Solution:

Use setup without topological sectors

 $\rightarrow$  Open boundary conditions.

### Boundary conditions

- Periodic boundary conditions do not let charge flow out of the volume.
- Field space is disconnected in continuum.



## Open boundary conditions

#### Proposed solution

- open boundary condition in time direction → same transfer matrix, same particle spectrum
- periodic boundary condition in spatial directions → momentum projection possible



## Open boundary conditions

• Lattices of size  $T \times L^3$ .

Neumann boundary conditions in time.



Gauge fields

$$F_{0k}|_{x_0=0}=F_{0k}|_{x_0=T}=0, \ \ k=1,2,3$$

#### Fermion fields

$$\begin{split} P_{+}\psi(x)|_{x_{0}=0} &= P_{-}\psi(x)|_{x_{0}=T} = 0 \qquad P_{\pm} = \frac{1}{2}(1\pm\gamma_{0})\\ \bar{\psi}(x)P_{-}|_{x_{0}=0} &= \bar{\psi}(x)P_{+}|_{x_{0}=T} = 0 \end{split}$$

# Critical slowing down

Continuum limit = continuous phase transition

Universal dynamical critical behavior

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How do \Gamma(t) and \tau_{\text{int}} scale as a \rightarrow 0?
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Free field expectation

$$au_{
m int} \propto a^{-1}$$

broken by interactions

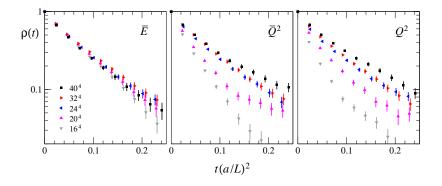
Lüscher, S '11

Expected Langevin (random walk) scaling

$$au_{
m int} \propto a^{-2}$$

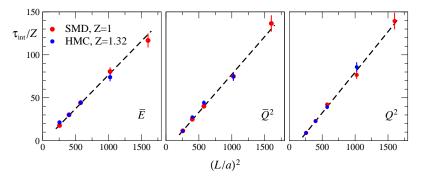
## Scaling towards continuum limit

Autocorrelation function vs scaled MC time



Energy (on time slice) shows very good scaling.Large cut-off effects in topological observables.

### Scaling towards continuum limit: $au_{ m int}$ vs $a^{-2}$



- $\blacksquare$  HMC and SMD\_{0.3} show same scaling up to a constant.  $\rightarrow$  universal behavior
- Topological observables well described by

$$au_{
m int}=c_1+c_2/a^2$$

Also  $Q^2$  and  $\bar{Q}^2$  show  $a^2$  scaling for a 
ightarrow 0.

#### Experience

- Improved Wilson fermions, Iwasaki gauge action.
- $64 imes 32^3$  lattice, a = 0.09 fm
- lacksquare physical light and strange quark mass,  $m_\pi L=2$
- $\tau_{\rm int}(E) \sim {\rm O}(20)$

#### Estimate

- Twice larger lattice for  $m_{\pi}L = 4$ ,  $L \approx 6$  fm.
- Run length  $100 \cdot \tau_{\text{int}}(E) = 2000 \cdot (a/0.09 \text{fm})^{-2}$ .

 $cost = 3 T flops \cdot years \cdot (a/0.09 fm)^{-7}$ 

■ a = 0.045 fm still cost 400 Tflops years.