# Multigrid Methods and Lattice QCD 

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## Outline

- Lattice QCD - solvers
- Review of Multigrid iterative solvers
- Bootstrap / adaptive Multigrid solvers for the Dirac system


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## The Dirac PDE

$$
\mathcal{D}[\mathcal{A}] \psi=\sum_{\mu} \gamma^{\mu} \partial_{\mu} \psi+m \psi=f
$$

- $\gamma_{\mu} \in \mathbb{C}^{4 \times 4}$ satisfy $\left\{\gamma_{\mu}, \gamma_{\nu}\right\}=\delta_{\mu, \nu} I$ and $\gamma_{5}=\prod_{\mu=1}^{4} \gamma_{\mu}=\gamma_{5}^{*}$
- $\psi_{s, c}^{j}: s=1,2,3,4, c=1,2,3, j=1, . ., n_{f}\left(n_{f}=1\right)$

Wilson's discretization

$$
D_{x, y}=\delta_{x, y}-\kappa \sum_{\mu=1}^{d}\left(1-\gamma_{\mu}\right) \otimes U_{x}^{\mu} \delta_{x+\mu, y}+\left(1+\gamma_{\mu}\right) \otimes U_{x-\mu}^{\mu *} \delta_{x-\mu, y}
$$

- Removes spurious zero modes, but breaks chiral symmetry
- Basic building block of chiral Overlap and Domain Wall operators


## The Dirac-Wilson matrix

$$
D_{x, y}=\delta_{x, y}-\kappa \sum_{\mu=1}^{d}\left(1-\gamma_{\mu}\right) \otimes U_{x}^{\mu} \delta_{x+\mu, y}+\left(1+\gamma_{\mu}\right) \otimes U_{x-\mu}^{\mu *} \delta_{x-\mu, y}
$$

- $D=I-\kappa D_{0}$ is positive real for $0 \leq \kappa<\kappa_{c}$
- nearest neighbor coupling on hypercubic lattice embedded in a 4 d torus
- 12 variables per grid point
- $n=12 \cdot n_{1} \cdot n_{2} \cdot n_{3} \cdot n_{4}$
- $n_{i}=16 \ldots 128$
- Interesting case: $\kappa \rightarrow \kappa_{c} \Rightarrow$

$$
m=\frac{1}{2}\left(\frac{1}{\kappa}-\frac{1}{\kappa_{c}}\right) \approx 0
$$

- Performance of Krylov methods degrades
 as $\kappa \rightarrow \kappa_{c r}$


## 2d Dirac-Wilson matrix: block-spin form

Wilson system consists of a sum of two parts, a stabilization term

$$
A_{x y}=-\frac{1}{2} \sum_{\mu=1}^{d}\left(U_{x}^{\mu} \delta_{x+\mu, y}+U_{x-\mu}^{\mu *} \delta_{x-\mu, y}\right)+(d+m) \delta_{x, y}
$$

referred to as the Gauge Laplacian, and a central covariant difference approximation of the Dirac system

$$
B_{x y}=\frac{1}{2} \sum_{\mu=1}^{d} \gamma_{\mu} \otimes U_{x}^{\mu} \delta_{x+\mu, y}-\gamma_{\mu} \otimes U_{x-\mu}^{\mu *} \delta_{x-\mu, y}
$$

where for a 2d lattice

$$
\gamma_{1}=\left(\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right) \quad \text { and } \quad \gamma_{2}=\left(\begin{array}{cc}
0 & i \\
-i & 0
\end{array}\right)
$$

and $D$ can be written as a block matrix

$$
D=\left(\begin{array}{cc}
A & B \\
-B^{*} & A
\end{array}\right)
$$

## Symmetries of the Wilson fermion matrix

$\gamma_{5}$-Symmetry:

$$
\Gamma_{5} D=D^{*} \Gamma_{5},
$$

where

$$
\Gamma_{5}=I \otimes\left(\gamma_{5}\right), \quad \gamma_{5}=\prod_{i=1}^{n_{s}} \gamma_{i}
$$

- $\lambda \in \operatorname{spec}(D) \Rightarrow \bar{\lambda} \in \operatorname{spec}(D)$
- $Q=\Gamma_{5} D$ is hermitian and maximally indefinite worst case for BiCGSTAB, GMRES, etc...

S. M. Pickles, Algorithms in Lattice QCD, Ph.D., University of Edinburgh, 1998, UKQCD


## Odd-even system

- grid points $x$ are odd or even ( $=$ red or green).
- odd-even-ordering yields

$$
D_{0}=\left(\begin{array}{cc}
0 & D_{e o} \\
D_{o e} & 0
\end{array}\right)
$$



## Spectrum of the Dirac-Wilson matrix and the odd-even Schur-complement



Specta of $D$ and odd-even Schur compl. for $4^{4}$ grid (realistic configuration)

## Multigrid for QCD circa 2000

- Gauge field $U$ is not geometrically smooth $\Rightarrow$ near kernel is locally oscillatory

$\Rightarrow$ Constant preserving (algebraic) multigrid methods completely fail


## Multigrid for QCD circa 2000

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Many others ...

## Multigrid for QCD circa 2000




R. C. Brower, R. G. Edwards, C. Rebbi, and E. Vicari, Projective multigrid for Wilson fermions, Nucl. Phys. B366 (1991), pp. 689-705.

## Basic Multigrid components



## Multigrid Methods

- Geometric Multigrid methods
- Specialized, e.g., for PDEs they are integrated with the finite element / volume / difference discretization
- Highly (maximally) efficient iterative solvers
- Limited applicability, e.g., with respect to the geometry and parameters of the problem
- Single-grid methods: geometric-algebraic methods
- More generally applicable
- Slight loss in efficiency vs. GMG resulting from use of auxiliary-grid (mesh) and some algebraic techniques, as needed
- Intended for grid-based problems
- No-grid methods: algebraic methods
- Black-box iterative solver for sparse $M$-matrix systems
- Convergence and complexity difficult to control in practice


## A two-grid method

For a given fine-level system of equations $D \psi=f$ ( $D \mathrm{HPD}$ ) defined on a space $V_{h}$, a two-level solver is described in terms of its two main components
(1) a smoother $M$
(2) a coarse space $V_{H}$ related to $P$ the prolongation operator and $R$ the restriction operator

Given an initial guess $u^{0}$, a single iteration of a two-grid method is as follows:
(9) Fine-level smoothing: $\tilde{u}=u^{0}+M^{-1}\left(f-D u^{0}\right)$
(2) Coarse-level correction: solve $D_{H} e_{H}=r_{H}$ with $r_{H}=R(f-D \tilde{u})$
(3) Update: $u^{1}:=\tilde{u}+P e_{H}$

Oftentimes, $D_{H}=R D P$ and $R=P^{*}$ if $D$ HPD

## Two-grid theory

It follows that $u-u^{1}=E_{T G}\left(u-u^{0}\right)$ where

$$
E_{T G}=\left(I-\pi_{D}\right)\left(I-M^{-1} D\right), \quad \pi_{D}:=P D_{H}^{-1} P^{*} D
$$

A sharp estimate of the convergence of a two-grid method is $\left\|E_{T G}\right\|_{D}^{2}=1-1 / K(P)$, where

$$
K(P)=\sup _{v} \frac{\left\|\left(I-\pi_{\widetilde{M}}\right) v\right\|_{\widetilde{M}}^{2}}{\|v\|_{D}^{2}} \quad \widetilde{M}:=M^{*}\left(M^{*}+M-D\right)^{-1} M, \quad\|v\|_{D}^{2}:=(D v, v)
$$

At least three different approaches are possible in choosing the components of a two-level method:
(1) For a fixed $P$, construct a suitable $M$ - geometric methods
(2) For a fixed $M$, optimize the choice of $P$ - (adaptive) algebraic methods
(3) Given certain measures on the suitability of $M$ and $P$, simultaneously construct both

## Weak approximation property

The suitability of $V_{H}$ is measured via an approximation property. Assuming $\widetilde{M} \approx\|D\| I$ and noting that $\pi_{\widetilde{M}}$ is the $\widetilde{M}$-orthogonal projector onto $\operatorname{Range}(P)$, we have

$$
\begin{aligned}
\left\|\left(I-\pi_{\widetilde{M}}\right) v\right\|_{\widetilde{M}}^{2} & \leq\|(I-\pi) v\|_{\widetilde{M}}^{2} \\
& \lesssim\|D\|\|(I-\pi) v\|^{2} \quad \text { all } \quad v \in V
\end{aligned}
$$

Thus,

$$
K(P)=\sup _{v \neq 0} \frac{\left\|\left(I-\pi_{\tilde{M}}\right) v\right\|_{\tilde{M}}^{2}}{\|v\|_{D}^{2}} \lesssim \sup _{v \neq 0} \frac{\|D\|\|(I-\pi) v\|^{2}}{\|v\|_{D}^{2}}
$$

where $\pi:=P\left(P^{*} P\right)^{-1} P^{*}$
Smooth error $w$ satisfies

$$
\frac{\|D w\|}{\|w\|} \approx \min _{v} \frac{\|D v\|}{\|v\|}
$$

## No-grid (algebraic) methods

- "Algebraic" stands for the fact that all the tools of the method are constructed solely on the basis of the original matrix $M$ in a setup phase
- Coarse space is constructed automatically within the algorithm, level by level, in a (hopefully) computationally optimal setup procedure which involves
(1) Picking a set of coarse variables, i.e., set of indices $\Omega_{H}=\left\{i_{1}, \ldots, i_{n_{H}}\right\} \rightarrow$ graph theoretic approaches
(2) Definiing $V_{H}=\operatorname{span}\left\{\psi_{k}\right\}_{k=1}^{n_{H}}$ such that each $\psi_{k}$ is supported in $\Omega_{k}$, for a vector: $\Omega_{k} \subset\{1, \ldots, n\} \rightarrow$ null space of the system matrix
- Each of the $V_{H}$ (or $V_{i}$ obtained recursively) must satisfy certain properties, related to the convergence of the overall algorithm. As subspaces are built "on the fly", multilevel theory for the convergence of such algorithms is very difficult


## Operator-dependent interpolation

Setup algorithm

- Given $D$ HPD and $\Omega=\{1, \ldots, n\}$, select $\Omega_{H}=\left\{1, \ldots, n_{H}\right\}, n_{H}<n$
- Compute entries of $P: V_{H} \mapsto V_{h}, R: V_{h} \mapsto V_{H}$, and $D_{H}=R A P$

1. Classical AMG: $\Omega_{H} \subset \Omega$ and

$$
\left.P=\left[\begin{array}{c}
W \\
I
\end{array}\right]\right\} \Omega_{H},
$$

where $W \in \mathbb{C}^{m \times n_{H}}$ with $m=n-n_{H}$
2. Aggregation AMG: $\Omega=\cup_{i=1}^{n_{H}} \Omega_{i}$,

$$
P_{j i}=\left\{\begin{array}{lll}
1 & \text { for } & j \in \Omega_{i} \\
0 & \text { for } & j \notin \Omega_{i}
\end{array} \quad i=1, \ldots, n_{H}\right.
$$

In Smoothed Aggregation, additional smoothing step applied to interpolation:

$$
P \leftarrow S P, \quad S=I-\tau D
$$

- Typically, for PDEs use constant preserving $P$, i.e., select $P$ to ensure that there exists $v_{H}$ such that $P v_{H}=1$ for some vector of coefficients $v_{H}$


## Multilevel iterative solvers in lattice computations

Solver challenges:

- Systems are nearly singular
- Non-hermitian and positive real or hermitian and maximally indefinite
- Near kernel is unknown: highly oscillatory with oscillations dependent upon fluctuations in background gauge fields $\leftarrow$ heterogeneity of covariant derivatives
- Large near kernel dependent upon on topology

What is needed? A method that can

- Approximate several "arbitrary" kernel components to within desired level of accuracy
- Extract the components from the algebraic problem (suitable smoother required)
- Automatically construct coarse-level basis


## Aggregation MG solver for Dirac-Wilson system

Given standard geometric blocking into $m^{d}$ aggregates and matrix $X:=\left[x^{(1)}, \ldots, x^{(r)}\right]$ such that $D x^{(i)} \approx 0$, interpolation defined as

$$
P=\left(\begin{array}{c|c|c}
X_{1} & & \\
& \ddots & \\
& & \\
\hline & & X_{n_{H}}
\end{array}\right) \quad \begin{gathered}
\\
\end{gathered}
$$

- $\left|\Omega_{i}\right|=m^{d}, i=1, \ldots, n_{H}$
- $P^{*} P=Q^{*} Q=I$
- $P X_{H}=P R=X$
- $P=S P, S=I-\tau D$
(smoothed aggregation)


## Basic Adaptive (A)MG Algorithm

For $\ell=0, \ldots, J$
While $\left\|I-B_{\ell}^{-1} D_{\ell}\right\|_{\text {est }}$ increasing: $x_{\ell} \leftarrow\left(I-B_{\ell}^{-1} D_{\ell}\right) x_{\ell}$
If $\left\|I-B_{\ell}^{-1} D_{\ell}\right\|_{\text {est }}$ is large
recalibrate interpolation based on (new) $x_{\ell}$
recompute coarse-grid operator
recurse


- Relax on $D x=0, x$ rand, if $\left\|I-B^{-1} D\right\|_{\text {est }}>$ tol, set $X=[x]$ update $P$
- Relax on $D x=0$
$\square$ Iterate on $D x=0$
- Iterate on $D x=0$, if $\left\|I-B^{-1} D\right\|_{\text {est }}>$ tol, set $X=\left[\begin{array}{ll}X & x\end{array}\right]$ update $P$

2d Dirac-Wilson system: adaptive Smoothed Aggregation MG solver

$$
D_{x, y}=-\frac{1}{2} \sum_{\mu=1}^{2}\left(1-\gamma_{\mu}\right) \otimes U_{x}^{\mu} \delta_{x+\mu, y}+\left(1+\gamma_{\mu}\right) \otimes U_{x-\mu}^{\mu *} \delta_{x-\mu, y}+(4+m) \delta_{x, y}
$$

with $U(x) \in U(1)$ on $2 d$ space-time lattice

- Solve $D^{*} D \psi=D^{*} f$
- Use adaptive $V(4,4)$-cycle setup with Gauss Seidel smoother based on current hierarchy to test solver
- Previously found error components quickly reduced and "new" error vector rich in unresolved components of the error
- Augment hierarchy to preserve additional vector space
- Use SA framework to cut vectors $x_{1}, x_{2}, \ldots, x_{r}$ into blocks and on each block use $Q R$ to define augmented - multiple vector preserving - $P$
- Add more vectors until satisfactory solver found, where each vector corresponds to an extra dof per coarse site

Results for 2d Dirac-Wilson system with $U(1)$ background

3 level, exact coarse solve $128^{2}, \mathrm{~m}=10^{-7}, \beta=1,4 \times 4$ blocking


Number of iterations versus residual for different number of TVs

## Adaptive SA results...



Number of applications of $D^{*} D$ needed to reduce relative residual to $O\left(10^{-8}\right)$

- $128 \times 128$ lattice, $\beta=1,6,4 \times 4$ blocking, 3 levels, 8 vectors
- Use GS for smoothing with exact solve on coarse grid
- Standard algorithm requires hpd $\Longrightarrow D^{*} D$
- Apply MG V-cycle as a preconditioner to CG
- Compare total number of $D^{*} D$ applications on fine level only with plain CG

[^0]
## Solving the non-hermitian system

- Solve $D$ directly, instead of $D^{*} D$
- Better sparsity of $D_{H}=R M P, R \neq P^{*}$ and less vectors required to define $P$
- Reduce the setup cost; for normal equations setup requires equivalent of 3-4 CG inversions
- $D_{H}=R D P, R \neq P^{*}$
- Use variant of MG solver for $D^{*} D$ for $D$
- $\gamma_{5}$ symmerty: $Q=\Gamma_{5} D=D^{*} \Gamma_{5}=Q^{*}$ such that

$$
Q=\sum_{i=1}^{N} \lambda_{i} v_{i} v_{i}^{*} \Rightarrow D=\sum_{i=1}^{n} \lambda_{i}\left(\Gamma_{5} v_{i}\right) v_{i}^{*}
$$

so that left and right eigenvectors (singular) vectors are related by $u_{i}=\Gamma_{5} v_{i}$

- Coarse-grid operator $R D P$, with $P$ based on $v_{i}$ and $R$ based on $u_{i}=\Gamma_{5} v_{i}$
- Leads to Galerkin coarse-level operator and preserves $\gamma_{5}$-symmetry on coarse levels
- Use adaptivity with MinRes smoother to compute the nearly singular vectors and incorporate them into unsmoothed aggregation solver
- Compare with previous results obtained for the normal equations


## 2d Dirac-Wilson system


(a) Number of iterations versus residual for different number of TVs

(b) Flops needed in setup and solve to reduce residual to tol $=10^{-8}$ for various values of the mass

Figure. 2d Dirac Wilson system with $\beta=1$ on $128 \times 128$ lattice

## Bootstrap MG solver for Dirac-Wilson system



Figure : Coarsening of the grid of even points and the odd-even reduction

Given matrix $X=\left[x^{(1)}, \ldots, x^{(r)}\right]=[V W]$, rows of interpolation, $p_{i}, i \in \Omega \backslash \Omega_{H}$, are defined such that

$$
\mathcal{L}\left(p_{i}\right)=\sum_{\kappa=1}^{r} \omega_{\kappa}\left(x_{\{i\}}^{(\kappa)}-\sum_{j \in C_{i} \cup\{i\}}\left(p_{i}\right)_{j} x_{\{j\}}^{(\kappa)}\right)^{2} \mapsto \min
$$

where weights $\omega_{\kappa} \sim \frac{\left\|x^{(\kappa)}\right\|}{\left\|D x^{(\kappa)}\right\|} \in \mathbb{R}^{+}$and $r=|\mathcal{V}|+|\mathcal{W}|$

## Bootstrap setup - multilevel eigensolver

Assuming no a priori information on low modes available

- Smoother and initial test vectors given

$$
v^{(s)}=G^{\eta} \widetilde{v}^{(s)}, \quad \widetilde{v}^{(s)} \text { random }
$$

- Observation $\left(P_{\ell}=P_{1}^{0} \cdots P_{\ell}^{\ell-1}, D_{\ell}=P_{\ell}^{H} D P_{\ell}, T_{\ell}=P_{\ell}^{H} P_{\ell}\right)$

$$
\frac{\left\langle w_{\ell}, w_{\ell}\right\rangle_{D_{\ell}}}{\left\langle w_{\ell}, w_{\ell}\right\rangle_{T_{\ell}}}=\frac{\left\langle P_{\ell} w_{\ell}, P_{\ell} w_{\ell}\right\rangle_{D}}{\left\langle P_{\ell} w_{\ell}, P_{\ell} w_{\ell}\right\rangle_{2}}
$$

Bootstrap Idea

$$
\begin{gathered}
\text { Eigenpairs } \\
\left(w_{\ell}, \lambda_{\ell}\right) \text { of }\left(D_{\ell}, T_{\ell}\right)
\end{gathered} \quad \begin{gathered}
\text { Eigenpairs } \\
\left(P_{\ell} w_{\ell}, \lambda_{\ell}\right) \text { of } D \\
+ \text { interpolation error }
\end{gathered}
$$

## Bootstrap multilevel eigensolver - cycling




## Solving the non-hermitian system: BAMG

Recall the abstract smoothing property: for smooth error, $e, \frac{\|D e\|}{\|e\|} \approx \min _{v} \frac{\|M v\|}{\|v\|}$. It is simple to show that

$$
\sigma_{1} \leq|\lambda| \leq \sigma_{n},
$$

where $\sigma_{1} \leq \sigma_{1} \leq \ldots \leq \sigma_{n}$ are the ordered singular values of $D$ and $\lambda$ is any of its eigenvalues. Thus, smooth error dominated by singular vectors (and eigenvectors?), suggesting use of a multilevel SVD solver in bootstrap cycle

Modifications to the setup:

- Kaczmarz smoother for $D^{*}$ and $D$ in the bootstrap and adaptive cycles to compute left and right singular vectors, respectively
- Weighted LS formulation to compute restriction (left singular vectors) and interpolation (right singular vectors)
- Multilevel singular value solver in the bootstrap approach where on the coarsest level we solve the symmetric eigenvalue problem directly:

$$
\left(\begin{array}{cc} 
& D_{L} \\
D_{L}^{H} &
\end{array}\right)\left(\begin{array}{cc}
U & U \\
V & -V
\end{array}\right)=\left(\begin{array}{ll}
T_{L} & \\
& Q_{L}
\end{array}\right)\left(\begin{array}{cc}
U & U \\
V & -V
\end{array}\right)\left(\begin{array}{cc}
\Sigma & 0 \\
0 & -\Sigma
\end{array}\right)
$$

where $D_{l}=R_{l} D P_{l}, Q_{l}=R_{l} R_{l}^{H}$, and $T_{l}=P_{l}^{H} P_{l}$ with

$$
\begin{aligned}
P_{l} & =P_{1}^{0} \cdot \ldots \cdot P_{l}^{l-1} \\
R_{l} & =R_{l-1}^{l} \cdot \ldots \cdot R_{0}^{1}, \quad l=2, \ldots, L
\end{aligned}
$$

## BAMG variant for 4d Dirac-Wilson system of QCD



Figure. Flops needed to reduce residual to $10^{-8}$ versus mass

- $32^{3} \times 96$ lattice, $\beta=6,4^{4}$ blocking, 3 levels
- Apply aggregation-based MG solver as preconditioner to GCR


## Concluding remarks

## Summary

- Bootstrap / Adaptive MG provide effective solvers for Dirac-Wilson systems effectively removing critical slowing down (Rottman)
- Extensions of methodology to chiral models (and other stochastic PDEs) underway with promising results (Kahl)


## Outlook

- Parallel version(s) of Bootstrap algorithm under development (e.g., in Hypre)
- Smoothing analysis and two-grid theory in progress, both complicated by fact that $D$ is non-normal matrix


[^0]:    J. Brannick, R. Brower, M. Clark, J. Osborn, and C. Rebbi, Adaptive multigrid algorithm for lattice QCD, Phys. Rev. Lett., published 28 January 2008, Issue 4, Volume 100, article 041601

