

# Computational Aspects Related to the Matrix Sign Function in Lattice QCD



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# Outline

- 1. the setting
  - the Wilson fermion matrix
  - overlap fermions and the sign function
  - partial fraction expansions and multishift CG
- 2. inner-outer schemes
  - relaxation
  - recursive preconditioning
  - deflation
- 3. error estimates and bounds
  - Gaussian quadrature
  - estimates from CG





- 4. non-zero chemical potential
  - the sign function revisited
  - the Arnoldi process
  - deflation
  - outlook





# 1. The Setting

Wilson fermion matrix: intro

# Lattice Gauge Theory

- QCD = standard theory of strong interaction between quarks
- lattice gauge theory = discretization of QCD

 approximation of gauge fields by configurations U of gauge links

$$\mathcal{U} = \{U_{\mu}(x) \mid x \in G, \mu = 1, \dots, 4\}$$





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## Wilson fermion matrix: details 1

- $M = I \kappa D$
- $M \in \mathbb{C}^{n \times n}$
- nearest neighbor coupling on 4-dimensional torus
- 12 variables per grid point
- $n = 12 \cdot n_1 \cdot n_2 \cdot n_3 \cdot n_4$
- $n_i = 8 \dots 128$





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#### Wilson fermion matrix: detail 2

$$(M\psi)_x = \psi_x - \kappa \left( \sum_{\mu=1}^4 \left( (I - \gamma_\mu) \otimes U_\mu(x) \right) \psi_{x+e_\mu} + \sum_{\mu=1}^4 \left( (I + \gamma_\mu) \otimes U^H_\mu(x - e_\mu) \right) \psi_{x-e_\mu} \right)$$





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• 
$$U_{\mu}(x) \in SU(3)$$

• 
$$\gamma_{\mu} \in \mathbb{C}^{4 \times 4}$$

# • $I \pm \gamma_{\mu}$ is projector on 2-dimensional subspace

• =  $\gamma_1 \gamma_2 \gamma_3 \gamma_4$  satisfies  $\gamma_5 \gamma_\mu = \gamma_\mu \gamma_5 = 0$ .

$$\gamma_5 = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \quad \text{or} \quad \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{pmatrix}$$





# **Overlap fermions & sign function: intro**

Chiral symmetry is an important physical property which should be reflected in the discretized operator.

- Wilson fermion matrix: No chiral symmetry
- **Ginsparg-Wilson relation (GW):** establishes a version of chiral symmetry on the lattice
- Overlap fermions (Neuberger, 1998): satisfy GW.





## **Overlap fermions & sign function: overlap operator**

#### Neuberger's overlap operator:

$$N = \rho \cdot I + M \cdot (M^H M)^{-1/2}$$
  
=  $\rho \cdot I + \gamma_5 \cdot \operatorname{sign}(Q)$ 

where

- $Q = \gamma_5 \cdot M \Rightarrow Q^H = Q$  hermitian Wilson matrix
- sign(Q) = Vsign $(\Lambda)V^H$  where  $Q = V\Lambda V^H$
- $\rho \ge 1$  ( $\rho = 1$ : massless operator)
- $\kappa = \frac{4}{3}\kappa_c$





# **Overlap fermions & sign function: inner/outer**

Computational work in simulation: solve

 $N\psi = \phi$  $\Leftrightarrow \quad (\rho \cdot I + \gamma_5 \operatorname{sign}(Q))\psi = \phi$ 

- N is represented by a dense matrix  $\Rightarrow$  cannot be determined explicitly
- nested iteration for

$$\underbrace{(\rho I + \gamma_5 \operatorname{sign}(Q))}_{=N} \psi = \phi$$

- outer iteration: MVM with N
- inner iteration: approximate sign(Q)b in  $N \cdot b$





#### Lanczos approach

Krylov subspace  $K_m(Q, b) = \langle b, Qb, Q^2b, \dots, Q^{m-1}b \rangle$ Lanczos method generates basis  $v_1, \dots, v_m$ : Put  $V_m = [v_1|v_2|\dots v_m]$ . Then

 $QV_m = V_m T_m + \beta_{m+1} v_{m+1} e_m^T$ ,  $T_m$  tridiagonal.

Note:  $T_m = V_m^H Q V_m$ Approximate via the Galerkin approximation  $\operatorname{sign}(Q)b \approx V_m \operatorname{sign}(T_m)e_1 \cdot ||b||.$ 





#### Improvement:

- Lanczos for  $Q^2$ , start with Qb
- use sign $(t) = t \cdot (t^2)^{-1/2}$
- approximate sign(Q) $b = V_m(T_m)^{-1/2}e^1 \cdot \beta_0$

# Advantages:

- smooth convergence
- less vectors to store
- easily computable error bound





# Partial fraction expansions & multishift cg: Zolotarev

**Zolotarev:**  $l_{\infty}$  best approx. of sign on  $[-b, -a] \cup [a, b]$ Assume spec $(Q) \subset [-b, -a] \cup [a, b]$ . Then

$$Z_p = \delta \cdot Q \prod_{i=1}^{p-1} (Q^2 + c_{2i}I) \cdot \prod_{i=1}^p (Q^2 + c_{2i-1}I)^{-1}$$
  
=  $\delta \cdot Q \sum_{i=1}^p \omega_i (Q^2 + \tau_i I)^{-1},$ 

where

$$c_i = \frac{\operatorname{sn}^2\left(iK/(2m); \sqrt{1-(b/a)^2}\right)}{1-\operatorname{sn}^2\left(iK/(2m); \sqrt{1-(b/a)^2}\right)},$$

K is the complete elliptic integral.

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#### Partial fraction expansions & multishift cg: cg

$$\operatorname{sign}(Q)v \approx \sum_{i=1}^{p} \omega_{i}Q \left(Q^{2} - \sigma_{i}I\right)^{-1} v.$$

 $(\sigma_i < 0).$ 

Solve all p systems  $(Q^2 - \sigma_i I) x_i = v$  in one stroke ('multishift CG'), since

$$K_m(Q^2, b) = K_m(Q^2 - \sigma_i I, b), \ i = 1, 2, \dots, m.$$





## **Summary of methods**

- 1. both (Lanczos and multishift CG) compute  $sign(Q)b \approx p_m(Q)b$ ,  $p_m$  polynomial
- 2. Zolotarev needs storage prop. to number of poles
- 3. Lanczos needs storage prop. to m
- 4. Lanczos adapts itself to b (finite termination)
- 5. Zolotarev: converged systems can be removed for efficiency
- 6. both benefit from deflation







# 2. Inner-outer scheme

Shifted unitary form of  $N = \rho I + \gamma_5 \cdot \text{sign}(Q)$ 

**Method:** 'SUMR' = GMRES for shifted unitary matrices (Reichel and Jagels, 1995)

- isometric Arnoldi
- minimal residual property
- short (coupled) recurrence





#### relaxation 1

Each iterative step needs an evaluation of sign(Q)x.

Relaxation: Relax accuracy condition for sign(Q)x as iteration proceeds.





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**Relaxation:** Relax accuracy condition for sign(Q)x as iteration proceeds.

Theory [Simoncini & Szyld, v. d. Eshof & Sleijpen, 2003]:

System Ax = b.

Investigate

$$\|\underbrace{b-Ax^{k}}_{k}\| \leq \|\underbrace{r^{k}-(b-Ax^{k})}_{residual gap}\| + \|\underbrace{r^{k}}_{computed residual}\|$$

Develop strategy to bound residual gap below required accuracy  $\epsilon$ .



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matrix properties	method	rel. tolerance $\eta_j$
herm. pos. def. $(N^H N)$	CG	$\eta_j = \epsilon \sqrt{\sum_{i=0}^j \ r^i\ ^{-2}}$
herm. indefinite $\gamma_{\rm F} N$	MINRES	$\eta_j = \epsilon / \ r^j\ $
shifted unitary $(N)$	SUMR	$\eta_j = \epsilon / \ r^j\ $





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## SUMR: recursive preconditioning

Idea: Relaxation pays more if convergence is fast.

- Use low accuracy SUMR as preconditioner
- outer: take adequate iterative method like GMRESR





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 $\mathsf{relGMRESR}(A, b, \epsilon)$ 

{computes x with  $||Ax-b|| < \epsilon \cdot ||b||$  via relaxed GMRESR} x = 0, r = b {initial values} C = [], U = []; {empty matrix} while  $||r|| > \epsilon \cdot ||b||$  do solve Au = r to relative accuracy  $\xi \quad \{\text{preconditioner}\}$ (for example  $u = \text{relSUMR}(A, r, \xi)$ ) compute c with  $||Au - c|| < \epsilon \cdot ||b|| \cdot ||u|| / ||r||$ for i=1:size(C,2) do  $\beta = C[:,i]^H \cdot c$  $c = c - \beta \cdot C[:, i]$  $u = u - \beta \cdot U[:, i]$ end for  $c = c/||c||, \ u = u/||c||$ C = [C, c], U = [U, u] $\alpha = c^H \cdot r$  $x = x + \alpha \cdot u$  $r = r - \alpha \cdot c$ end while



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#### **SUMR: numerical results**



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Method	$\mu = 0.03$	$\mu = 0.1$	$\mu = 0.3$
SUMR	31550	8312	3200
relSUMR	18840(1.87)	6038(1.38)	2656(1.20)
relGMRESR(SUMR)	5974(5.82)	2252(3.69)	1382(2.32)

Times (in seconds) on (quenched) 16<sup>4</sup> configuration, run on 16 processors of ALiCE.



## **Deflation: intro**

# Features:

- precompute some ( $\approx 30)$  smallest eigenvalues and -vectors of  $Q^2$
- 'project those out' (effect on sign function is known)  $b = b^+ + b^- + b^{\perp} \Rightarrow \operatorname{sign}(Q)b = b^+ - b^- + \operatorname{sign}(Q)b^{\perp}$   $\operatorname{sign}(Q)b^{\perp} = \operatorname{sign}(\Pi^H Q \Pi)b^{\perp}$ 
  - improves cond. no. of Q
  - significant decrease in no. of poles in Zolotarev PFE (for example 28 for  $10^{-10}$ )

- decreases no. of iterations in multishift CG

relaxed GMRES(SUMR)



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#### **Deflation: results**



$n_p$	Inversion	Calls to Wilson op.	Eigenval. calc.	Total time
1	9144	1032172	0	9144
10	1269	189514	111	1380
20	796	112862	118	914
30	568	78548	172	740
40	459	63566	274	733
50	387	52758	361	748
60	340	45732	410	750

total time for one relGMRESR(CG) + projection of  $n_p$  eigenmodes, 8<sup>4</sup> lattice,  $\mu = 0.1$ 





# 3. Error Estimates and Bounds

(Aggressive) relaxation requires good estimates or upper bounds for approximation error

 $\|\operatorname{sign}(Q)b - p_m(Q)b\|$ 

Lanczos

Lanczos for  $Q^2$ :

 $\|\operatorname{sign}(Q)b - p_m(Q)b\| \le \rho_m,$ 

where  $\rho_m$  is norm of *m*-th CG residual for  $Q^2x = b$  (initial guess 0) [van den Eshof et al 2002] Back
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#### **Zolotarev I: basics**

**Notation:** Zolotarev =  $t \cdot g(t^2)$  with

$$g(t) = \sum_{i=1}^{s} \omega_i \frac{1}{t - \sigma_i}.$$

**Remember:** For all poles  $\sigma_i$ , the *m*-th CG residuals for  $(Q^2 - \sigma_i I)x = b$  are collinear to the Lanczos vector  $v_m$ ,

$$r_i^m = b - (Q^2 - \sigma_i I) x_i^m = \rho_i^m v_m.$$

Approximation and error:

$$x^m = \beta_0 V_m g(T_m) e_1 = \sum_{i=1}^s \omega_i x_i^m, \quad e^m = x^m - g(A) b.$$

**Classical estimate**: If convergence is monotone or even superlinear

 $||e^m|| \approx ||x^m - x^{m+d}||, \ d \ge 1$  moderately large





#### Zolotarev: Gaussian quadrature

Expand error in terms of residuals:

$$e^{m} = \sum_{i=1}^{s} \rho_{i}^{m} \omega_{i} (Q^{2} - \sigma_{i}I)^{-1} v^{m}, \quad \|e^{m}\|^{2} = (v^{m})^{H} h(Q^{2}) v^{m},$$

where

$$h(Q^{2}) = \sum_{i,j=1}^{s} \rho_{i}^{m} \rho_{j}^{m} \omega_{i} \omega_{j} (Q^{2} - \sigma_{i}I)^{-1} (Q^{2} - \sigma_{j}I)^{-1}$$

Golub/Meurant (1994, 1997): Use Gaussian quadrature w.r.t. discrete measure to get upper and lower bounds for the moment  $(v^m)^H h(Q^2) v^m$ .

- Elegant theory, lower and upper bounds
- $\bullet$  One more node in quadrature rule amounts to one further step of Lanczos for  $Q^2$  and  $v^m$
- MVMs cannot be recycled to improve the solutions to the systems



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#### Zolotarev: example

Zolotarev for  $(Q^2)^{-1/2}Qb$  with spec $(Q) \subset [-32, -1] \cup [1, 32]$ .





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#### New estimates based on CG coefficients

#### Recall CG algorithm:

Choose 
$$x^{0} = 0$$
, set  $r^{(0)} = b$ ,  $p^{0} = r^{0}$   
for  $k = 1, 2, ...$  do  
 $\gamma^{k-1} = \langle r^{k-1}, r^{k-1} \rangle / \langle p^{k-1}, Ap^{k-1} \rangle$   
 $x^{k} = x^{k-1} + \gamma^{k-1}p^{k-1}$   
 $r^{k} = r^{k-1} - \gamma^{k-1}Ap^{k-1}$   
 $\delta^{k} = \langle r^{k}, r^{k} \rangle / \langle r^{k-1}, r^{k-1} \rangle$   
 $p^{k} = r^{k} + \delta^{k}p^{k-1}$ 

end for





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 $r^{k} = r^{k-1} - \gamma^{k-1}Ap^{k-1}$   
 $\delta^{k} = \langle r^{k}, r^{k} \rangle / \langle r^{k-1}, r^{k-1} \rangle$   
 $p^{k} = r^{k} + \delta^{k}p^{k-1}$   
end for

For  $d \in \mathbb{N}$ , denote

$$\begin{split} \eta^{k,d} &:= \sum_{i=0}^{d-1} \gamma^{k+i} \langle r^{k+i}, r^{k+i} \rangle \\ \varphi^{k,d} &:= \sum_{i=0}^{d} \frac{\langle p^{k+i}, p^{k+i} \rangle}{\langle p^{k+i}, Ap^{k+i} \rangle} \cdot \left( \langle r^{k+i}, e^{k+i} \rangle + \langle r^{k+i+1}, e^{k+i+1} \rangle \right). \end{split}$$



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#### Lemma:

$$\begin{array}{lll} \langle r^k, e^k \rangle &=& \langle r^{k+d}, e^{k+d} \rangle + \eta^{k,d} \geq \eta^{k,d}, \\ \langle e^k, e^k \rangle &=& \langle e^{k+d}, e^{k+d} \rangle + \varphi^{k,d} \geq \varphi^{k,d}. \end{array}$$

**Note:**  $\langle r^{k+i}, e^{k+i} \rangle$  in  $\varphi^{k,d}$  is not available. Replacing by  $\eta^{k+i,d}$  gives the estimate

$$\tau^{k,d} = \sum_{i=0}^{d} \frac{\langle p^{k+i}, p^{k+i} \rangle}{\langle p^{k+i}, Ap^{k+i} \rangle} (\eta^{k+i,d} + \eta^{k+i+1,d})$$
  
$$\leq \langle e^k, e^k \rangle.$$

[Hestenes-Stiefel 1952, Strakos-Tichy 2002, Meurant 2005]



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For Galerkin approximation to g(A)b we have

$$||e^k||^2 = \sum_{i,j=1}^s \omega_i \omega_j \langle e_i^k, e_j^k \rangle.$$

For  $\sigma_i \neq \sigma_j$  one has

$$\frac{1}{(t-\sigma_i)(t-\sigma_j)} = \frac{1}{\sigma_i-\sigma_j} \cdot \left(\frac{1}{t-\sigma_i} - \frac{1}{t-\sigma_j}\right),$$

thus

$$\langle e_i^k, e_j^k \rangle = \frac{1}{\sigma_i - \sigma_j} \cdot \left( \langle r_i^k, e_j^k \rangle - \langle r_j^k, e_i^k \rangle \right)$$



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Theorem: We have

$$\|g(A)b - \sum_{i=1}^{s} \omega_i x_i^k\|_2^2 \ge \boldsymbol{\eta}^{k,d} + \boldsymbol{\tau}^{k,d},$$

where

$$\boldsymbol{\eta}^{k,d} = \sum_{i,j=1,\sigma_i\neq\sigma_j}^{s} \frac{\omega_i \omega_j}{\sigma_i - \sigma_j} \left( \frac{\rho_j^k}{\rho_i^k} \eta_i^{(k,d)} - \frac{\rho_i^k}{\rho_j^k} \eta_j^{k,d} \right),$$
$$\boldsymbol{\tau}^{(k,d)} = \sum_{i,j=1,\sigma_i=\sigma_j}^{s} \omega_i \omega_j \tau_j^{k,d}.$$

**Corollary:** If estimates are positive, error norm decreases.

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#### Error estimates: numerical results

**Example 1:** sign(A)b, including deflation of small eigenvalues. Configuration conf5.4-0018x8-2000.mtxfrom MatrixMarket, d = 5









# 4. Nonzero chemical potential

Wilson matrix is modified:

$$(M(\mu)\psi)_{x} = \psi_{x} - \kappa \left(\sum_{\nu=1}^{3} \left((I - \gamma_{\nu}) \otimes U_{\nu}(x)\right)\psi_{x+e_{\nu}} + \sum_{\nu=1}^{3} \left((I + \gamma_{\nu}) \otimes U_{\nu}^{H}(x - e_{\nu})\right)\psi_{x-e_{\nu}}\right)$$
$$-\kappa \left(e^{-\mu} \left(I - \gamma_{4}\right) \otimes U_{4}(x)\right)\psi_{x+e_{4}} - \kappa \left(e^{\mu} \left(I + \gamma_{4}\right) \otimes U_{4}^{H}(x - e_{4})\right)\psi_{x-e_{4}}\right)$$

**Consequence:**  $Q = \gamma_5 M$  is not hermitian any more.



#### sign function revisited

We need **alternatives** to the spectral definition.

**Function theory**: f analytic in neighborhood of spec(A),  $\Gamma$  contour:

$$f(A) = \frac{1}{2\pi i} \oint_{\Gamma} f(z)(zI - A)^{-1} dz.$$

A is **diagonalizable**,  $A = U \wedge U^{-1}$ , then

 $f(A) = Uf(\Lambda)U^{-1}$  with  $f(\Lambda) = \text{diag}(f(\lambda_i))$ .



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## sign function revisited II

A not diagonalizable, Jordan decomposition

$$A = U(\bigoplus_{i} J_{i})U^{-1}, \quad J_{i} = \begin{pmatrix} \lambda_{i} & 1 & \cdots & 0 \\ 0 & \lambda_{i} & \ddots & \vdots \\ \vdots & \ddots & \ddots & 1 \\ 0 & \cdots & 0 & \lambda_{i} \end{pmatrix}$$

Then

$$f(A) = U\left(\bigoplus_{i} f(J_i)\right) U^{-1},$$

where

$$f(J_i) = \begin{pmatrix} f(\lambda_i) & f^{(1)}(\lambda_i) & \cdots & \frac{f^{(m_i-1)}(\lambda_i)}{(m_i-1)!} \\ 0 & f(\lambda_i) & \ddots & \vdots \\ \vdots & \ddots & \ddots & f^{(1)}(\lambda_i) \\ 0 & \cdots & 0 & f(\lambda_i) \end{pmatrix}$$



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sign function revisited: Galerkin

**Consequence:** 

 $\operatorname{sign}(Q) = p(Q),$ 

p Hermite interpolating polynomial on spec(Q).

**Problem:** Fix  $m \in \mathbb{N}$ . Find "best" approximating polynomial  $p_{m-1}$  s.t.

 $p_{m-1}(Q)b - \operatorname{sign}(Q)b \to \min!$  for all  $p_{m-1} \in \mathcal{P}_{m-1}$ .

Solution: Use Galerkin condition

 $p_{m-1}(Q)b - \operatorname{sign}(Q)b \perp K_m(Q,b)$ 





# sign function revisited: computation

Use Arnoldi process to construct orthogonal basis  $v_1, \ldots, v_m$  of  $K_m(Q, b)$ :

 $QV_m = V_m H_m + \beta_m v_{m+1} e_1^T$ ,  $H_m$  upper Hessenberg.

Results in long recurrence!.

The Galerkin approximation can be computed as

 $p_{m-1}(Q)b = V_m \operatorname{sign}(H_m)e_1 \cdot \|b\| = V_m \operatorname{sign}(V_m^H Q V_m) V_m^H b.$ 

**Note:**  $H_m$  is "small", sign $(H_m)$  can be computed using Roberts' iteration

 $S_{k+1} = S_k + S_k^{-1}, \quad S_0 = H_m.$ 

**Option**: Compute QR-factorization first.





#### Deflation

**Experience:** Deflation of small eigenvalues is mandatory.

Our test cases:  $\mu = 0.3$ , lattice 4<sup>4</sup> (left) and 6<sup>4</sup> (right)



spectrum of  $Q = \gamma_5 M$ 





#### Deflation: augmented subspaces

**Problem:**  $w \perp S$  and  $QS = S \Rightarrow Qw \perp S!!$ 

**Consequence**: Let S be spanned by "small" eigenvectors. Decompose

$$b = b_{\parallel} + b_{\perp}.$$

Then  $K_m(Q, b_{\perp}) \cap S \neq \{0\}$ .

Solution: Augmented Krylov subspace approach.

Let AS = ST,  $T \in \mathbb{R}^{k \times k}$  (T is usually triangular).

Compute orthogonal basis for  $K_m(Q, b_{\perp}) + S$  similarly to Arnoldi:

$$Q\left(S \ V_m\right) = \begin{pmatrix} S \ V_m \end{pmatrix} \begin{pmatrix} T \ S^H A V_m \\ 0 \ H_m \end{pmatrix} + \beta_m v_{m+1} e_{k+m+1}^T.$$





#### **Deflation: Galerkin**

Define

$$W_m = \begin{pmatrix} S & V_m \end{pmatrix}, \quad G_m = \begin{pmatrix} T & S^H A V_m \\ 0 & H_m \end{pmatrix}.$$

# Imposing the Galerkin condition gives

 $\operatorname{sign}(Q)b_{\perp} \approx W_m \operatorname{sign}(G_m)e_{m+1} \cdot \|b_{\perp}\|.$ 

#### Note:

$$\operatorname{sign}(G_m) = \begin{pmatrix} \operatorname{sign}(T) & Y \\ 0 & \operatorname{sign}(H_m) \end{pmatrix}$$

where Y solves the Sylvester equation

 $TY - YH_m = \operatorname{sign}(T)X - X\operatorname{sign}(H_m), \quad X = S^H QV_m.$ 





#### Numerical results

# Convergence histories (left: 4<sup>4</sup> lattice, right: 6<sup>4</sup> lattice):







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#### **Alternatives**

# Simplify your life: Two-sided deflation! Specifically: colspan(V) right evs, colspan(W) left evs $Qv_i = \lambda_i v_i, \quad w_i^H Q = \lambda_i w_i^H, \quad w_i^H v_i = \delta_{ij}, i = 1, ..., m.$

 $VW^H$  projects on colspan(V) along colspan(W) Decompose:  $b = VW^Hb + (I - VW^H)b$ .

Then:  $\operatorname{sign}(Q) \cdot VW^H b = \operatorname{diag}(\operatorname{sign}(\lambda_i)) \cdot V \cdot (W^H b),$  $VW^H \cdot Q \cdot (I - VW^H)b = 0.$ 

#### **Consequence:**

 $K_k(Q, (I - VW^H)b) = K_k((I - VW^H)Q(I - VW^H), b).$ 





#### Performance one-sided vs two-sided deflation





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## Computational cost

4 <sup>4</sup> lattice, $m = 32$			
Schur deflation			
init	ializatior	n time: 14	.1 s
k	Arnoldi	$sign(H_k)$	total
100	0.18	0.03	0.23
200	0.59	0.21	0.81
300	1.22	0.52	1.75
400	2.05	1.08	3.16
500	3.12	1.79	4.93
600	4.37	2.90	7.31
700	5.88	4.57	10.49
800	7.56	6.69	14.28
900	9.50	9.38	18.92
1000	11.63	12.68	24.36





$4^4$ lattice, $m = 32$			
	LR-de	eflation	
init	ialization	time: 27	′.5 s
k	Arnoldi	sign(H)	total
100	0.12	0.03	0.15
200	0.45	0.20	0.66
300	1.01	0.49	1.51
400	1.77	1.02	2.82
500	2.76	1.69	4.47
600	3.94	2.77	6.74
700	5.36	4.40	9.79
800	6.96	6.44	13.44
900	8.84	9.10	17.98
1000	10.84	12.33	23.21





$6^4$ lattice, $m = 128$				
	Schur deflation			
ini	tializatio	n time: 88	84 s	
k	Arnoldi	$sign(H_k)$	total	
100	2.03	0.05	2.13	
200	5.16	0.22	5.45	
300	9.27	0.56	9.91	
400	14.59	1.15	15.85	
500	20.95	2.09	23.17	
600	28.12	3.35	31.61	
700	36.81	5.17	42.15	
800	46.32	7.39	53.88	
900	56.83	10.37	67.39	
1000	68.29	13.88	82.39	





6 <sup>4</sup> lattice, $m = 128$			
	LR-de	eflation	
initi	alization	time: 17	13 s
k	Arnoldi	sign(H)	total
100	0.66	0.03	0.75
200	2.39	0.15	2.62
300	5.16	0.42	5.69
400	9.01	0.94	10.06
500	13.96	1.73	15.84
600	20.03	2.80	22.98
700	27.09	4.44	31.70
800	35.09	6.49	41.78
900	44.38	9.10	53.70
1000	54.74	12.36	67.34



