

# BDDC Preconditioned P-Multigrid for High-Order Finite Elements

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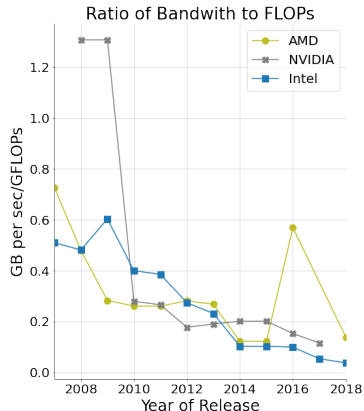
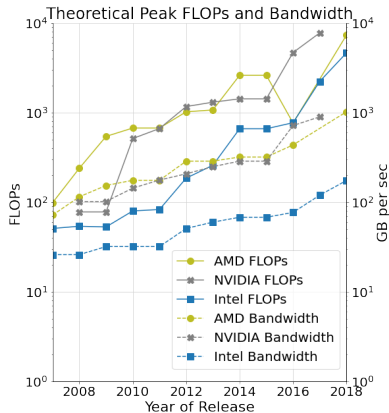
# Overview

- 1 Introduction
- 2 High-Order Matrix-Free FEM
- 3 LFA of High-Order FEM
- 4 LFA of P-Multigrid Methods
- 5 LFA of BDDC
- 6 Summary

# Big Picture

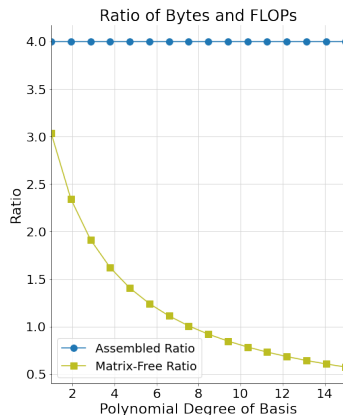
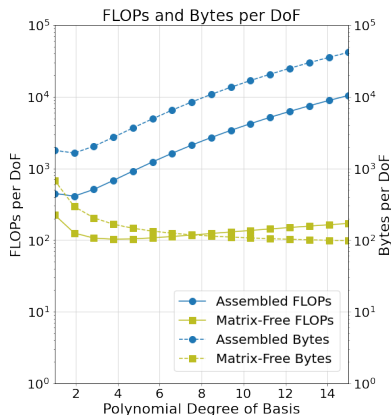
- High-order matrix-free representations of PDEs are better suited to modern hardware than sparse matrices
- High-order matrix-free representations require preconditioned iterative solvers
- Local Fourier Analysis (LFA) provides sharp convergence estimates for these preconditioners
- We investigate LFA of Balancing Domain Decomposition by Constraints (BDDC) for high-order element subdomains
- We investigate LFA of  $p$ -multigrid with a BDDC smoother

# Modern Hardware



Modern hardware has lower memory bandwidth than FLOPs  
 (<https://github.com/karlrupp/cpu-gpu-mic-comparison>)

# Benefits of Matrix-Free



Requirements for matrix-vector product with sparse matrix vs matrix-free  
for screened Poisson  $\nabla^2 u - \alpha^2 u = f$  in 3D

For more details - see Rezgar Shakeri's talk Thursday, 13:05 in Session 10B

# Matrix-Free Representation

Weak form for an arbitrary second order PDE:

$$\begin{aligned} & \text{find } u \in V \text{ such that for all } v \in V \\ \langle v, u \rangle = \int_{\Omega} v \cdot f_0(u, \nabla u) + \nabla v : f_1(u, \nabla u) = 0 \end{aligned} \quad (1)$$

where

- $\cdot$  - contraction over fields
- $:$  - contraction over fields and spatial dimensions

Note: pointwise functions  $f_0$  and  $f_1$  don't depend upon the discretization

# Matrix-Free Representation

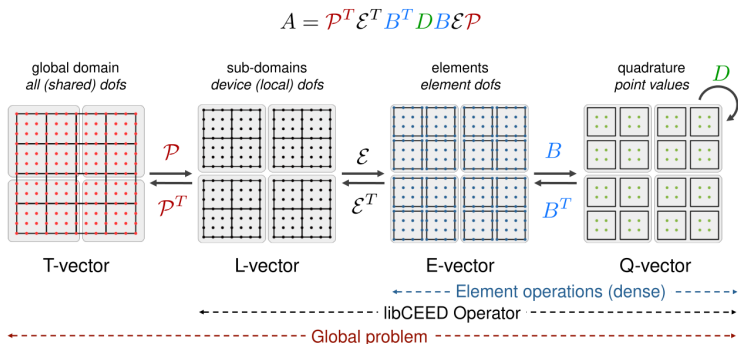
Galerkin form for an arbitrary second order PDE:

$$\sum_e \mathcal{E}^T \left[ (\mathbf{B}_I^e)^T \mathbf{W}^e \Lambda(f_0(u^e, \nabla u^e)) + \sum_{i=0}^{d-1} (\mathbf{B}_{\xi,i}^e)^T \mathbf{W}^e \Lambda(f_1(u^e, \nabla u^e)) \right] = 0 \quad (2)$$

- $\mathcal{E}$  - element assembly/restriction operator
- $\mathbf{B}_I^e$  - interpolation to quadrature points
- $\mathbf{B}_{\xi,i}^e$  - derivatives at quadrature points
- $\mathbf{W}^e$  - quadrature weights
- $\Lambda$  - pointwise multiplication at quadrature points
- $u^e = \mathbf{B}_I^e \mathcal{E}^e u$  and  $\nabla u^e = \{\mathbf{B}_{\xi,i}^e \mathcal{E}^e u\}_{i=0}^{d-1}$



## libCEED Representation



- $\mathcal{P}$  - parallel element assembly operator
- $\mathcal{E}$  - local element assembly operator
- $B$  - basis action operator
- $D$  - weak form and geometry at quadrature points

# Preconditioning Required

- Matrix-free representations require iterative solvers
- Iterative solvers are sensitive to conditioning of the operator (among other factors)
- High-order operators are ill-conditioned
- Preconditioners are required for good convergence
- LFA helps us tune these preconditioners

# LFA Background

Consider a scalar Toeplitz operator  $L_h$  on the infinite 1D grid  $G_h$

$$\begin{aligned}
 L_h &\triangleq [s_\kappa]_h \ (\kappa \in V) \\
 L_h w_h(x) &= \sum_{\kappa \in V} s_\kappa w_h(x + \kappa h)
 \end{aligned} \tag{3}$$

where

- $V \subset \mathbb{Z}$  is an index set
- $s_\kappa \in \mathbb{R}$  are constant coefficients
- $w_h(x)$  is an  $l^2$  function on  $G_h$

# LFA Background

Our function can be diagonalized by the standard Fourier modes:

If for all grid functions  $\varphi(\theta, x)$

$$L_h \varphi(\theta, x) = \tilde{L}_h(\theta) \varphi(\theta, x) \quad (4)$$

then  $\tilde{L}_h(\theta) = \sum_{\kappa \in V} s_\kappa e^{i\theta \kappa}$  is the **symbol** of  $L_h$

# LFA Background

For a  $q \times q$  system of equations, the matrix symbol is given by:

$$\mathbf{L}_h = \begin{bmatrix} L_h^{1,1} & \cdots & L_h^{1,q} \\ \vdots & \vdots & \vdots \\ L_h^{q,1} & \cdots & L_h^{q,q} \end{bmatrix} \Rightarrow \tilde{\mathbf{L}}_h = \begin{bmatrix} \tilde{L}_h^{1,1} & \cdots & \tilde{L}_h^{1,q} \\ \vdots & \vdots & \vdots \\ \tilde{L}_h^{q,1} & \cdots & \tilde{L}_h^{q,q} \end{bmatrix} \quad (5)$$

# LFA of High-Order FEM

For a scalar PDE operator on a single 1D finite element

$$\tilde{\mathbf{A}}(\theta) = \mathbf{Q}^T \left( \mathbf{A}^e \odot \left[ e^{i(x_j - x_i)\theta/h} \right] \right) \mathbf{Q} \quad (6)$$

where

$$\mathbf{A}^e = \mathbf{B}^T \mathbf{D} \mathbf{B}, \quad \mathbf{Q} = \begin{bmatrix} \mathbf{I} \\ \mathbf{e}_0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \\ 1 & 0 & \cdots & 0 \end{bmatrix} \quad (7)$$

# LFA of High-Order FEM

Natural extension to multiple components and higher dimensions:

$$\tilde{\mathbf{A}}(\boldsymbol{\theta}) = \mathbf{Q}^T \left( \mathbf{A}^e \odot \left[ e^{i(\mathbf{x}_j - \mathbf{x}_i) \cdot \boldsymbol{\theta} / h} \right] \right) \mathbf{Q} \quad (8)$$

Multiple Components:

$$\mathbf{Q}_n = \mathbf{I}_n \otimes \mathbf{Q} \quad (9)$$

Multiple Dimensions:

$$\mathbf{Q}_{nd} = \mathbf{Q} \otimes \mathbf{Q} \otimes \cdots \otimes \mathbf{Q} \quad (10)$$

# Example: Scalar Poisson

$$\int \nabla v \nabla u = \int f v \quad (11)$$

- **B** - given by tensor  $H^1$  Lagrange basis
- **D** - given by quadrature weights and product

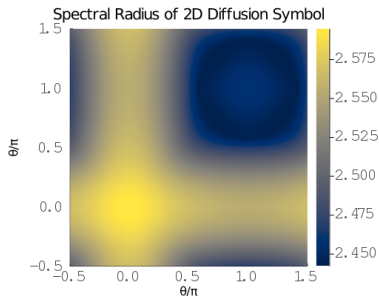
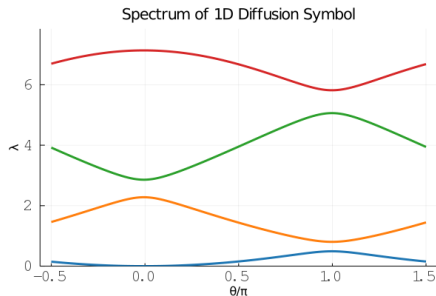
```
# mesh
dim = 1
mesh = Mesh1D(1.0)

# basis
p = 3
ncomp = 1
basis = TensorH1LagrangeBasis(p+1, p+1, ncomp, dim)

# weak form
function diffusionweakform(du::Array{Float64}, w::Array{Float64})
    return dv = du*w[1]
end
```



# Example: Scalar Poisson



Scalar Poisson problem on quartic elements

Goal: decrease spectral radius with preconditioners

# LFA of High-Order Smoothers

Error propagation operator for smoothers given by

$$S = I - M^{-1} \mathbf{A} \quad (12)$$

with a symbol given by

$$\tilde{S}(\boldsymbol{\theta}, \omega) = I - \tilde{M}^{-1}(\boldsymbol{\theta}, \omega) \tilde{\mathbf{A}}(\boldsymbol{\theta}) \quad (13)$$

# Two-Grid Multigrid Error

Multigrid methods target the low frequency error

$$E_{2MG} = S_f (I - P_{ctof} A_c^{-1} R_{ftoc} A_f) S_f \quad (14)$$

- $A_f$  - fine grid operator
- $A_c^{-1}$  - coarse grid solve (low frequency error)
- $S_f$  - fine grid smoother (high frequency error)
- $P_{ctof}$  - coarse to fine grid prolongation operator
- $R_{ftoc}$  - fine to coarse grid restriction operator

Grid transfer operators and coarse representation differentiate  
*h*-multigrid and *p*-multigrid

# Two-Grid Multigrid Error

The definition of the symbol follows naturally:

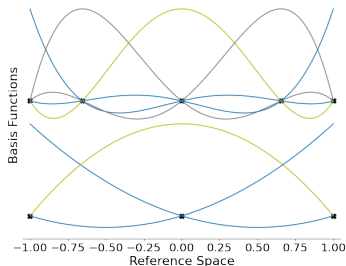
$$\tilde{E}_{2MG}(\theta) = \tilde{S}_f(\theta, \omega) \left( I - \tilde{P}_{ctof}(\theta) \tilde{A}_c^{-1}(\theta) \tilde{R}_{ftoc}(\theta) \tilde{A}_f(\theta) \right) \tilde{S}_f(\theta, \omega) \quad (15)$$

- $\tilde{A}_f$  - fine grid symbol
- $\tilde{A}_c^{-1}$  - coarse grid symbol inverse (low frequency error)
- $\tilde{S}_f$  - fine grid smoother symbol (high frequency error)
- $\tilde{P}_{ctof}$  - coarse to fine grid prolongation symbol
- $\tilde{R}_{ftoc}$  - fine to coarse grid restriction symbol

# P-Multigrid Transfer Operators

$p$ -multigrid prolongation can be represented as an interpolation from the coarse to fine grid

P-Prolongation from Coarse Basis to Fine Nodes

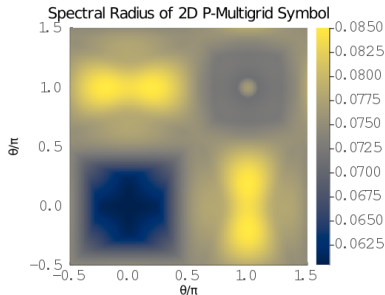
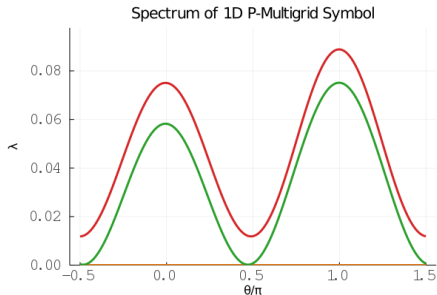


$$\mathbf{P}_{\text{ctof}} = \mathcal{P}_f^T \mathcal{E}_f^T \mathbf{P}^e \mathcal{E}_c \mathcal{P}_c \quad (16)$$

$$\mathbf{P}^e = \mathbf{I} \mathbf{D}_{\text{scale}} \mathbf{B}_{\text{ctof}}$$

$\mathbf{D}$  scales for node multiplicity

# Example: $P$ -Multigrid



$p$ -multigrid with third order Chebyshev on quartic to quadratic elements

Significant reduction in spectral radius

# Aggressive Coarsening

- High-order fine grid is most efficient representation
- Linear coarse grid is easier to solve with traditional methods
- Want to reduce number of intermediate grids
- Typical smoothers do not respond well to aggressive coarsening

Experiments:  $P$ -Multigrid with Chebyshev

$p_{\text{fine}}$ to $p_{\text{coarse}}$	$k = 3$			$k = 4$		
	LFA	libCEED	its	LFA	libCEED	its
$p = 2$ to $p = 1$	0.076	0.058	9	0.041	0.033	7
$p = 4$ to $p = 2$	0.111	0.097	10	0.062	0.050	8
$p = 4$ to $p = 1$	0.416	0.398	25	0.295	0.276	18
$p = 8$ to $p = 4$	0.197	0.195	15	0.121	0.110	11
$p = 8$ to $p = 2$	0.611	0.603	46	0.506	0.469	31
$p = 8$ to $p = 1$	0.871	0.861	154	0.827	0.814	112

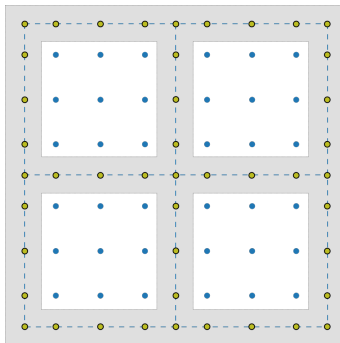
LFA and experimental two-grid convergence factors with  
Chebyshev smoothing for 3D Laplacian

3D manufactured solution on the domain  $[-3, 3]^3$  with Dirichlet boundaries:

$$f(x, y, z) = xyz \sin(\pi x) \sin(\pi(1.23 + 0.5y)) \sin(\pi(2.34 + 0.25z)) \quad (17)$$



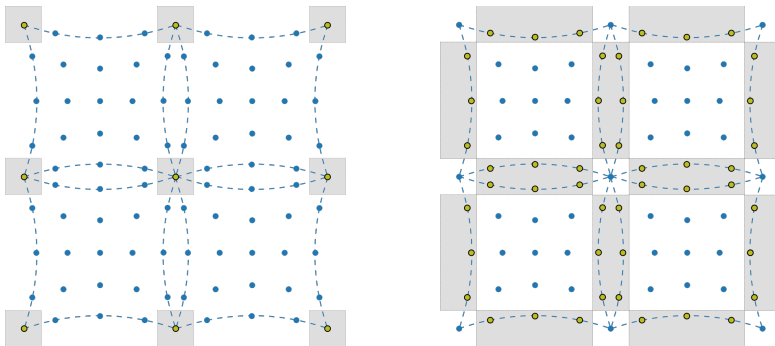
# BDDC Overview



High-order single element subdomains

BDDC - non-overlapping domain decomposition method by Dohrmann

# Broken Subdomains



Non-overlapping domain decomposition of high-order mesh

Global problem only "partially subassembled" on primal ( $\Pi$ ) vertices

Remaining interface nodes replicated across broken interface

# Subassembled Problem

$$\hat{\mathbf{A}}^{-1} = \sum_{e=1}^N \mathbf{R}_i^{e,T} \hat{\mathbf{A}}^{e,-1} \mathbf{R}_i^e, \quad \hat{\mathbf{A}}^e = \begin{bmatrix} \mathbf{A}_{r,r}^e & \hat{\mathbf{A}}_{\Pi,r}^{e,T} \\ \hat{\mathbf{A}}_{\Pi,r}^e & \hat{\mathbf{A}}_{\Pi,\Pi}^e \end{bmatrix} \quad (18)$$

Partially subassembled problem is easier to invert

Injection operator  $\mathbf{R}_i$  maps from global space to broken space  
and provides different BDDC variants

# Injection Operators

$$R_1 = \text{diag} \left( \left[ \frac{1}{|\mathcal{N}(x_i)|} \right] \right) \quad (19)$$

where  $|\mathcal{N}(x_i)|$  is node multiplicity  
across broken spaces

$$R_2 = R_1 - J^T \mathcal{H}^T \quad (20)$$

$$\mathcal{H}^e = -\mathbf{A}_{l,l}^{e,-1} \mathbf{A}_{\Gamma,l}^{e,T}$$

where  $\mathcal{H}$  is a harmonic extension,  
 $J$  a map over the interfaces

Lumped BDDC with  $R_1$  cheaper to setup but poorer conditioning

Dirichlet BDDC with  $R_2$  equivalent to Dirichlet FETI-DP

# Subassembled Inverse

$$\hat{\mathbf{A}}^e = \begin{bmatrix} \mathbf{A}_{r,r}^e & \hat{\mathbf{A}}_{\Pi,r}^{e,T} \\ \hat{\mathbf{A}}_{\Pi,r}^e & \hat{\mathbf{A}}_{\Pi,\Pi}^e \end{bmatrix} \quad \hat{\mathbf{S}}_{\Pi} = \mathbf{A}_{\Pi,\Pi} - \hat{\mathbf{A}}_{\Pi,r} \mathbf{A}_{r,r}^{-1} \hat{\mathbf{A}}_{\Pi,r}^T \quad (21)$$

- Subassembled problem inverted with Schur complement
- Coarse grid problem  $\hat{\mathbf{S}}_{\Pi}$  is easier to solve with traditional methods
- Dense high-order element interior inverse  $\mathbf{A}_{r,r}^{-1}$  can be expensive
- Fast diagonalization can provide efficient approximate solver

# Fast Diagonalization

For separable problems of the form

$$A = aM + bK \quad (22)$$

Fast Diagonalization provides fast approximate solver

$$A^{-1} = S^T (aI + b\Lambda)^{-1} S \quad (23)$$

where

$$SMS^T = I, \quad SKS^T = \Lambda \quad (24)$$

# Fast Diagonalization

- Tensor product bases have tensor product diagonalizations
- Convergence impact of approximate solver formulations is ongoing research
- Cheaper to compute Fast Diagonalization solver than invert assembled subdomain matrices
- Reusing diagonalization for injection subdomain operator inverse mitigates expensive setup cost of Dirichlet BDDC

## LFA of BDDC

$$\tilde{\tilde{\mathbf{A}}}^{-1} = \begin{bmatrix} \mathbf{I} & -\tilde{\mathbf{A}}_{r,r}^{-1} \tilde{\tilde{\mathbf{A}}}_{\Pi,r}^T \\ 0 & \mathbf{I} \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{A}}_{r,r}^{-1} & 0 \\ 0 & \tilde{\tilde{\mathbf{S}}}_{\Pi}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{I} & 0 \\ -\tilde{\tilde{\mathbf{A}}}_{\Pi,r} \tilde{\mathbf{A}}_{r,r}^{-1} & \mathbf{I} \end{bmatrix} \quad (25)$$

$$\begin{aligned} \tilde{\tilde{\mathbf{A}}}_{r,r}^{-1}(\boldsymbol{\theta}) &= \mathbf{A}_{r,r}^{-1} \odot \left[ e^{i(\mathbf{x}_j - \mathbf{x}_i) \cdot \boldsymbol{\theta}/h} \right], & \tilde{\tilde{\mathbf{A}}}_{r,\Pi}(\boldsymbol{\theta}) &= \left( \hat{\mathbf{A}}_{r,\Pi} \odot \left[ e^{i(\mathbf{x}_j - \mathbf{x}_i) \cdot \boldsymbol{\theta}/h} \right] \right) \mathbf{Q}_{\Pi}, \\ \tilde{\tilde{\mathbf{S}}}_{\Pi}^{-1}(\boldsymbol{\theta}) &= \left( \mathbf{Q}_{\Pi}^T \left( \hat{\mathbf{S}}_{\Pi} \odot \left[ e^{i(\mathbf{x}_j - \mathbf{x}_i) \cdot \boldsymbol{\theta}/h} \right] \right) \mathbf{Q}_{\Pi} \right)^{-1} \end{aligned} \quad (26)$$

Only primal modes are localized for subassembled operator symbol

Symbols of injection operators are relatively straightforward



# Low-Order Validation

$m$	Lumped BDDC			Dirichlet BDDC		
	$\lambda_{\min}$	$\lambda_{\max}$	$\kappa$	$\lambda_{\min}$	$\lambda_{\max}$	$\kappa$
$m = 4$	1.000	4.444	4.444	1.000	2.351	2.351
$m = 8$	1.000	12.269	12.269	1.000	3.196	3.196
$m = 16$	1.000	31.179	31.179	1.000	4.188	4.188
$m = 32$	1.000	75.761	75.761	1.000	5.335	5.335

Condition numbers and maximal eigenvalues  
for low-order macro-elements

Exactly reproduces original work on LFA of low-order subdomains (Brown and He)

# High-Order Experiments

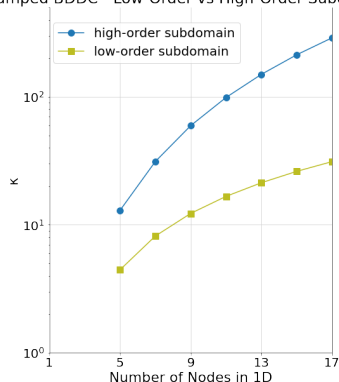
$p$	Lumped BDDC			Dirichlet BDDC		
	$\lambda_{\min}$	$\lambda_{\max}$	$\kappa$	$\lambda_{\min}$	$\lambda_{\max}$	$\kappa$
$p = 2$	1.000	2.800	2.800	1.000	2.042	2.042
$p = 4$	1.000	12.948	12.948	1.000	3.242	3.242
$p = 8$	1.000	59.563	59.563	1.000	5.197	5.197
$p = 16$	1.000	289.678	289.678	1.000	7.761	7.761

Condition numbers and maximal eigenvalues  
for single high-order element subdomains

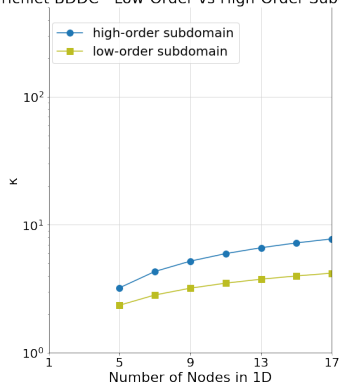
Single high-order element subdomains less well conditioned

# Low vs High-Order BDDC

Lumped BDDC - Low-Order vs High-Order Subdomain



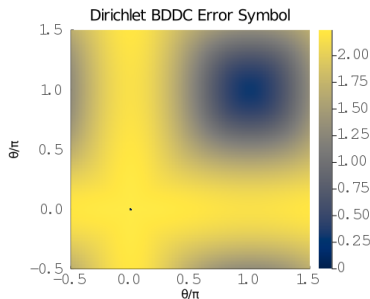
Dirichlet BDDC - Low-Order vs High-Order Subdomain



Low-order and high-order subdomain condition number

Dirichlet BDDC important for single high-order element subdomains

# BDDC Smoother for $P$ -Multigrid



Symbol of error operator for Dirichlet BDDC of 2D Laplacian for  $p = 4$

Dirichlet BDDC smoother still has large spectral radius, so we introduce relaxation parameter

$$\tilde{\mathbf{E}}(\boldsymbol{\theta}, \omega) = \mathbf{I} - \omega \tilde{\mathbf{M}}_2^{-1} \tilde{\mathbf{A}}(\boldsymbol{\theta}) \quad (27)$$

# BDDC Smoother for $P$ -Multigrid

$p_{\text{fine}}$ to $p_{\text{coarse}}$	Dirichlet BDDC			Chebyshev	
	$\rho$	$\omega_{\text{opt}}$	its	$\rho$	its
$p = 2$ to $p = 1$	0.121	0.66	11	0.075	9
$p = 4$ to $p = 2$	0.272	0.48	18	0.085	10
$p = 4$ to $p = 1$	0.281	0.47	19	0.219	16
$p = 8$ to $p = 4$	0.409	0.38	26	0.110	11
$p = 8$ to $p = 1$	0.462	0.32	30	0.795	101
$p = 16$ to $p = 8$	0.504	0.32	34	0.435	28
$p = 16$ to $p = 1$	0.597	0.23	45	0.959	551

Two-grid convergence factor for  $p$ -multigrid with BDDC  
vs cubic Chebyshev smoothing for 2D Laplacian

Weighted Dirichlet BDDC smoother better supports rapid coarsening

BDDC Smoother for  $P$ -Multigrid

$p_{\text{fine}}$ to $p_{\text{coarse}}$	Dirichlet BDDC			Chebyshev	
	$\rho$	$\omega_{\text{opt}}$	its	$\rho$	its
$p = 2$ to $p = 1$	0.121	0.66	11	0.252	17
$p = 4$ to $p = 2$	0.272	0.48	18	0.281	19
$p = 4$ to $p = 1$	0.281	0.47	19	0.424	27
$p = 8$ to $p = 4$	0.409	0.38	26	0.278	18
$p = 8$ to $p = 1$	0.462	0.32	30	0.873	170
$p = 16$ to $p = 8$	0.504	0.32	34	0.613	48
$p = 16$ to $p = 1$	0.597	0.23	45	0.975	910

Two-grid convergence factor for  $p$ -multigrid with BDDC  
vs quadratic Chebyshev smoothing for 2D Laplacian

Weighted Dirichlet BDDC smoother better supports rapid coarsening

# Summary

- High-order matrix-free representations of PDEs are better suited to modern hardware than sparse matrices
- High-order matrix-free representations require preconditioned iterative solvers
- Local Fourier Analysis (LFA) provides sharp convergence estimates for these preconditioners
- We investigated LFA of Balancing Domain Decomposition by Constraints (BDDC) for high-order element subdomains
- Finally, we investigated LFA of  $p$ -multigrid with a BDDC smoother

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