

A Highly Scalable Implementation of the BDDC Preconditioner

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Outline

- ① Introduction and motivation
- ② BDDC preconditioner
- ③ BDDC MPI-parallel implementation
- ④ MultiLevel BDDC MPI-parallel implementation
- ⑤ Conclusions and future work

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Problem statement

- **Variational problem:** Find $u \in H^1(\Omega)$ s.t.:

$$a(u, v) = (f, v), \quad \forall v \in H_0^1(\Omega),$$

assuming $a(\cdot, \cdot)$ **symmetric, coercive** (e.g., Laplacian, Linear Elasticity)

- **Discrete problem:** Find $u_h \in V_h \subset H^1(\Omega)$ (conforming FE space) s.t.:

$$a(u_h, v_h) = (f, v_h), \quad \forall v_h \in V_h^0.$$

- **Algebraic problem:** Find $x \in \mathbb{R}^n$ s.t.:

$$Ax = b,$$

where A is a large, sparse, SPD matrix

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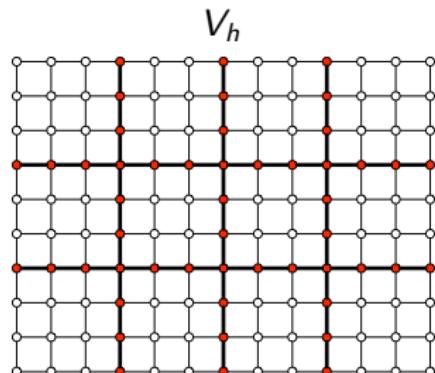
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Motivation:

Efficient exploitation of current petascale supercomputers (and beyond)

Domain decomposition framework

○: interior DoFs (I); ●: interface DOFs (Γ)



Numerical solution of $Ax = b$

- Solve iteratively $M^{-1}Ax = M^{-1}b \rightarrow \tilde{A} = \tilde{b}$
- Preconditioner M aims at **accelerating convergence**
- M should be a parallel, cheap-to-invert, “good” approximation of A

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Solve $Ax = b$ by M -PCG

Set-up M

call PCG(A, M, b, x^0)

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PCG (In: (A, M, f, x^0) , Out: x)

$$r^0 := f - Ax^0$$

$$z^0 := M^{-1}r^0$$

$$p^0 := z^0$$

for $j = 0, \dots$, till convergence do

$$s^{j+1} = Ap^j$$

$$\alpha^j := (r^j, z^j) / (s^{j+1}, p^j)$$

$$x^{j+1} := x^j + \alpha^j p^j$$

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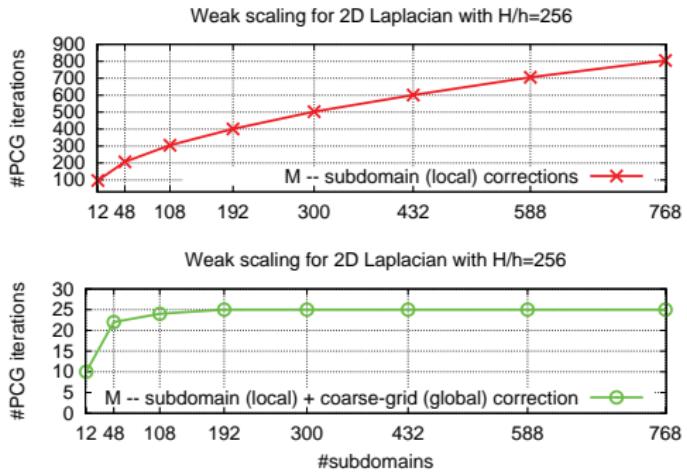
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Weak scaling: $\uparrow P$, fixed $\frac{N}{P}$

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BDDC Balancing DD by constraints

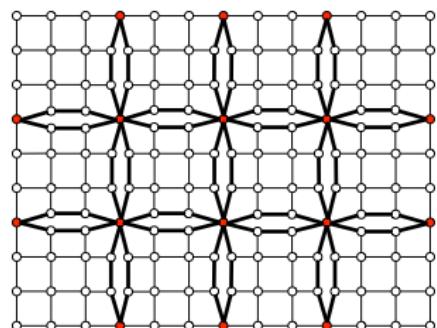
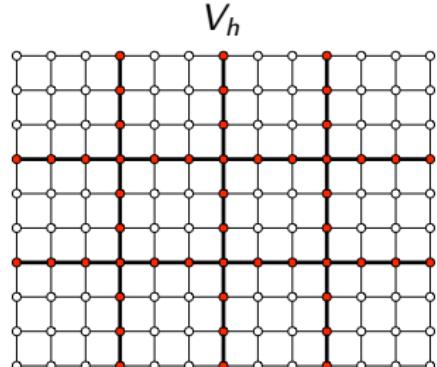
BDDC preconditioner [Dohrmann, 2003]

- Replace V_h by \bar{V}_h (reduced continuity)
- Define the injection $E : \bar{V}_h \longrightarrow V_h$
(weighted average; involves nearest neighbours communication)
- Find $\bar{z}_h \in \bar{V}_h$ such that:

$$a(\bar{z}_h, \bar{v}_h) = (E^T r_h, \bar{v}_h), \quad \forall \bar{v}_h \in \bar{V}_h$$

and obtain $z_h = M_{BDDC}^{-1} r_h = \mathcal{E} E \bar{z}_h$

- Last correction \mathcal{E} is the harmonic extension of the interface values
(involves local Dirichlet solvers)



\bar{V}_h

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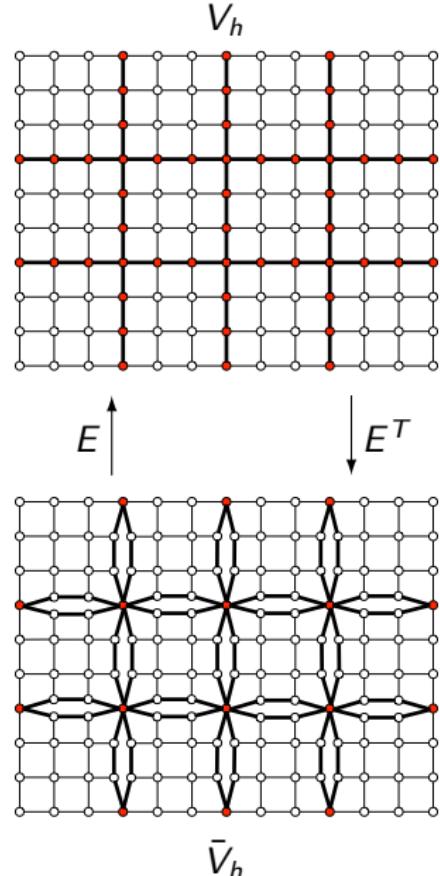
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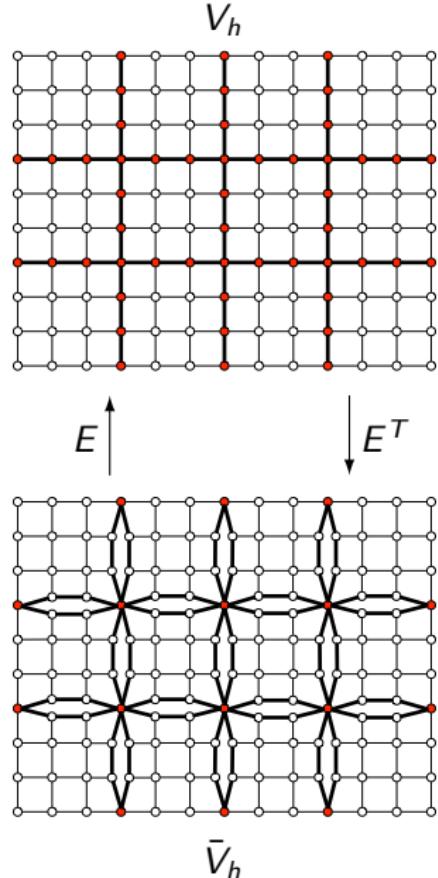
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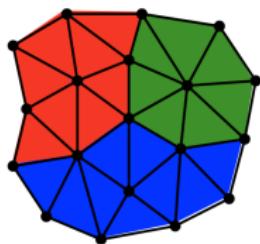
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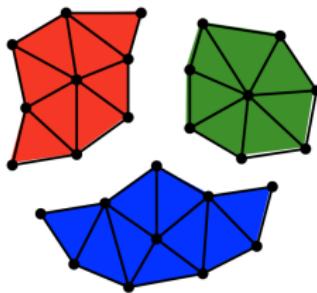
BDDC preconditioning

Given a conforming FE space V_h , we have:

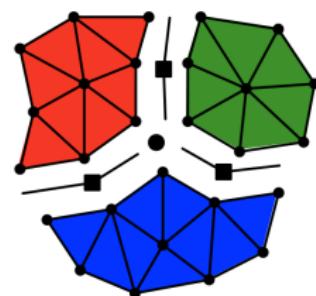
- The unassembled space $\tilde{V}_h = V_h^1 \times \dots \times V_h^n$ ($V_h \subset \tilde{V}_h$)
- The BDDC (under-assembled) space $\bar{V}_h = \{\tilde{v}_h \in \tilde{V}_h \mid \text{continuous } \mathcal{F}(\tilde{v}_h)\}$



V_h



\tilde{V}_h

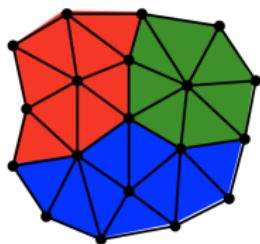


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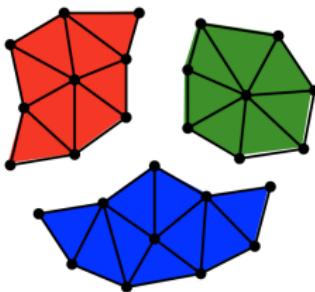
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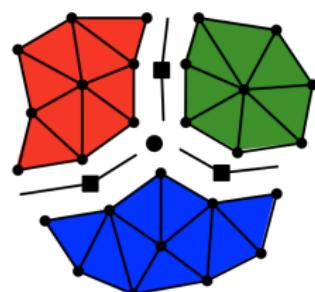
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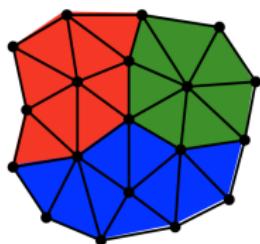


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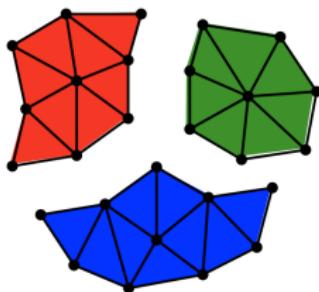
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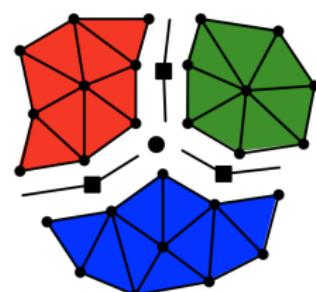
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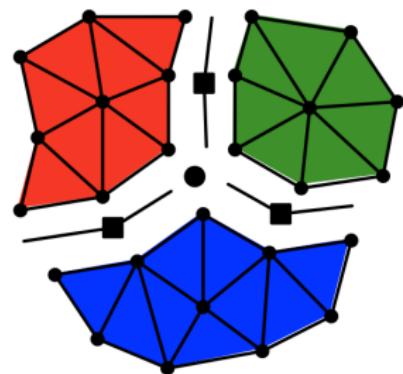
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- The under-assembled space \bar{V}_h can be decomposed as:

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- Sought-after correction split into fine-grid (\bar{z}_F) + coarse-grid (\bar{z}_C) correction



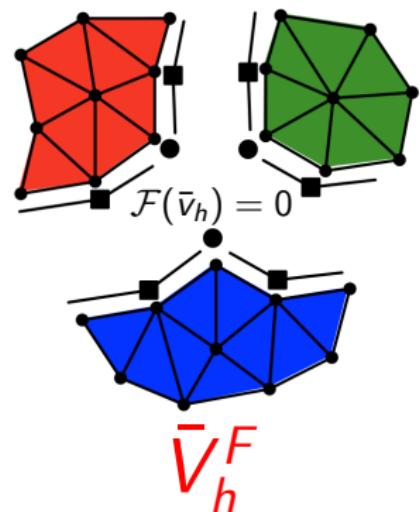
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Fine-grid correction (\bar{z}_F)

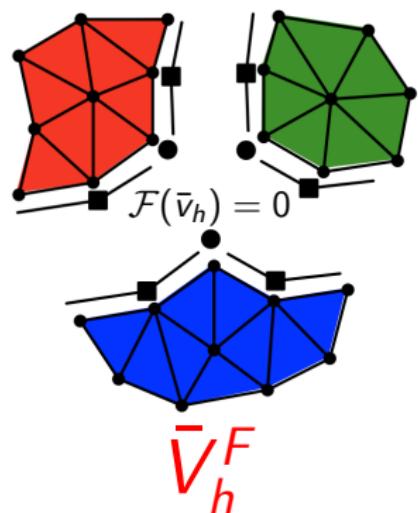
- Find $\bar{z}_F \in \mathbb{R}^{\tilde{n}}$ such that

$$\begin{bmatrix} \tilde{A} & C^T \\ C & 0 \end{bmatrix} \begin{bmatrix} \bar{z}_F \\ \lambda \end{bmatrix} = \begin{bmatrix} E^T r \\ 0 \end{bmatrix}$$

- Equivalent to P independent problems

Find $\bar{z}_F^{(i)} \in \mathbb{R}^{\tilde{n}^{(i)}}$ such that

$$\begin{bmatrix} \tilde{A}^{(i)} & C^{(i)T} \\ C^{(i)} & 0 \end{bmatrix} \begin{bmatrix} \bar{z}_F^{(i)} \\ \lambda^{(i)} \end{bmatrix} = \begin{bmatrix} E^{(i)T} r^{(i)} \\ 0 \end{bmatrix}$$

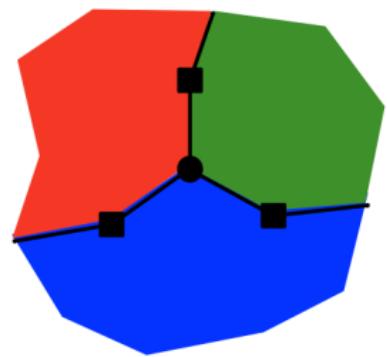


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Coarse-grid correction (\bar{z}_C)

Computation of $\bar{V}_h^C = \text{span}\{\Phi_1, \Phi_2, \dots, \Phi_{n_C}\}$

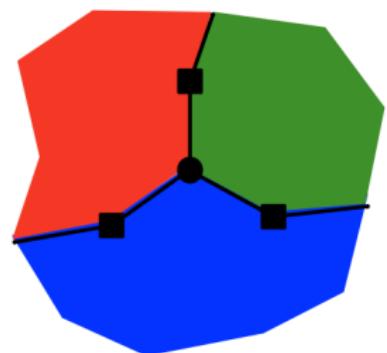
- Find $\Phi \in \mathbb{R}^{\tilde{n} \times n_C}$ such that

$$\begin{bmatrix} \tilde{A} & C^T \\ C & 0 \end{bmatrix} \begin{bmatrix} \Phi \\ \Lambda \end{bmatrix} = \begin{bmatrix} 0 \\ I_{n_C} \end{bmatrix}$$

- Equivalent to P independent problems

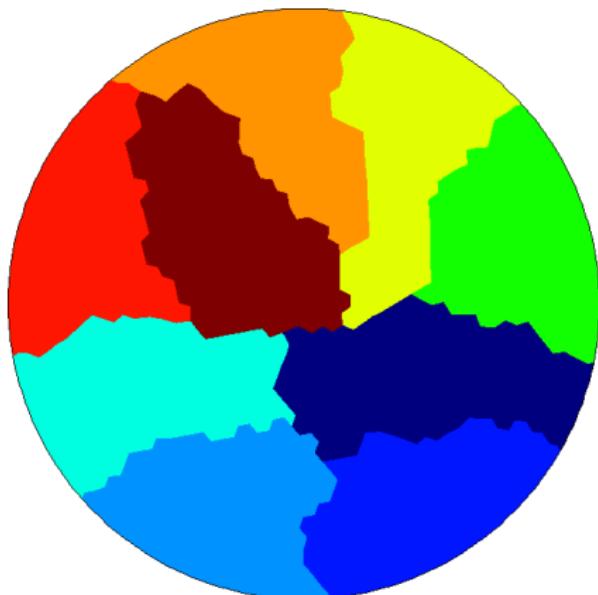
Find $\Phi^{(i)} \in \mathbb{R}^{\tilde{n} \times n_C^{(i)}}$ such that

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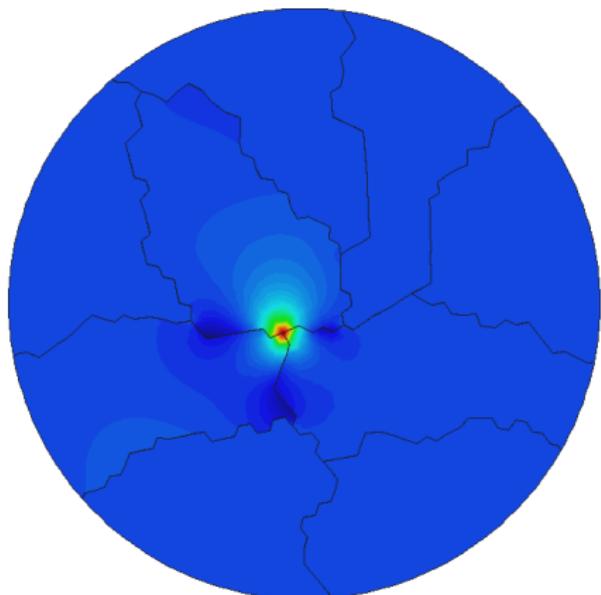


\bar{V}_h^C

BDDC coarse space basis functions

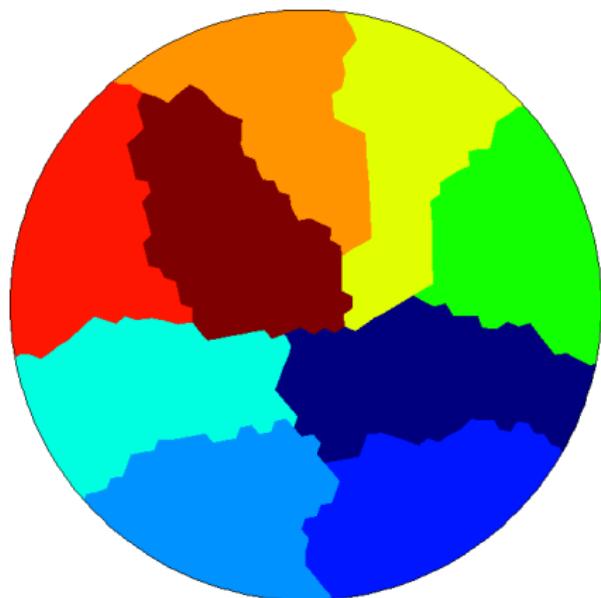


Circle domain partitioned into 9 subdomains

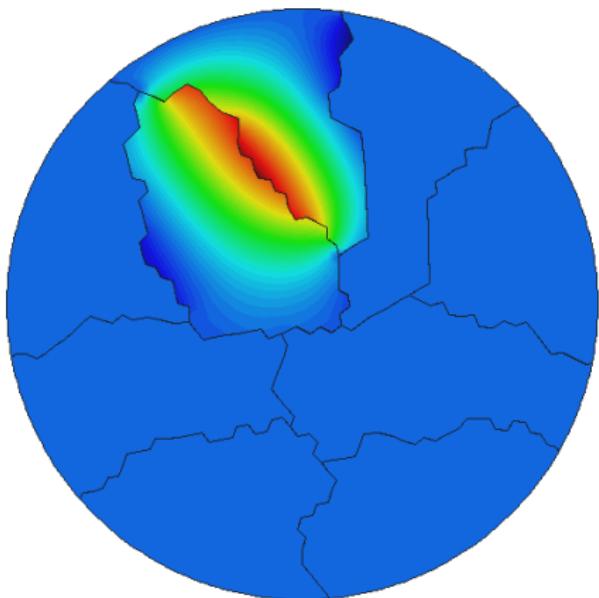


Φ_j (\bar{V}_h^C 's corner basis vector)

BDDC coarse space basis functions



Circle domain partitioned into 9 subdomains



Φ_j (\bar{V}_h^C 's edge basis vector)

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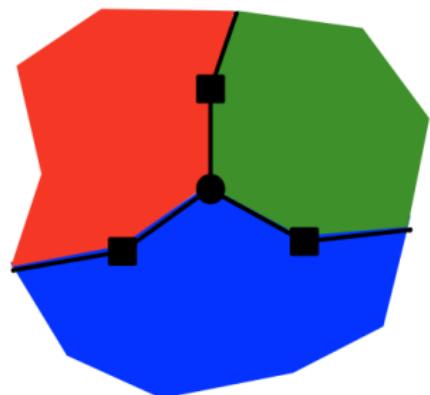
Coarse-grid correction (\bar{z}_C)

Assembly and solution of coarse-grid problem:

- $A_C = \text{assembly}(\Phi^{(i)T} \tilde{A}^{(i)} \Phi^{(i)})$
- $r_C = \text{assembly}(\Phi^{(i)T} E^{(i)T} r^{(i)})$
- Solve $A_C \alpha_C = r_C$
- $\bar{z}_C = \Phi \alpha_C$

Coarse-grid problem is:

- Global, i.e. couples all subdomains
- But much smaller than original problem
- Potential loss of parallel efficiency with P



\bar{V}_h^C

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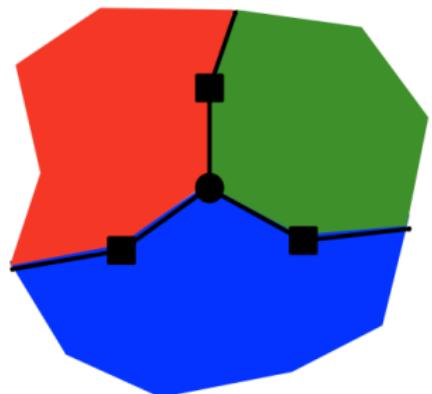
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\bar{V}_h^C

Coarse DoFs definition

Key aspect: Selection of coarse DoFs, i.e. continuity among subdomains

- Weak scalability ($\kappa(M_{BDDC}^{-1}A)$ constant for fixed N/P and $\uparrow P$)
- N/P large in practice $\sim \mathcal{O}(10^{4-5})$
- BDDC(ce) and BDDC(cef) require much less iterations in 3D
- But at the expense of a more costly coarse-grid problem

Coarse DoFs vs. $\kappa(M_{BDDC}^{-1}A)$:

$$d = 2$$

$$d = 3$$

Continuity on corners

$$\left[1 + d^{-2} \log^2 \left(\frac{N}{P}\right)\right] \quad \frac{N^{\frac{1}{d}}}{P} \left[1 + d^{-2} \log^2 \left(\frac{N}{P}\right)\right]$$

Continuity of mean value on edges too

$$\left[1 + d^{-2} \log^2 \left(\frac{N}{P}\right)\right] \quad \left[1 + d^{-2} \log^2 \left(\frac{N}{P}\right)\right]$$

Continuity of mean value on faces too

-

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Why BDDC for extreme scales?

- ① (Mathematically supported) **extremely aggressive coarsening** ($10^5 - 10^6$ size reduction between fine/coarse level)
- ② The coarse matrix has a similar **sparsity** as the original matrix
- ③ Coarse/local components can be computed **in parallel** (like additive)
- ④ ALL local + coarse problems can be solved **inexactly** (AMG-cycle)
- ⑤ A **multilevel** extension is possible (for extreme core counts)

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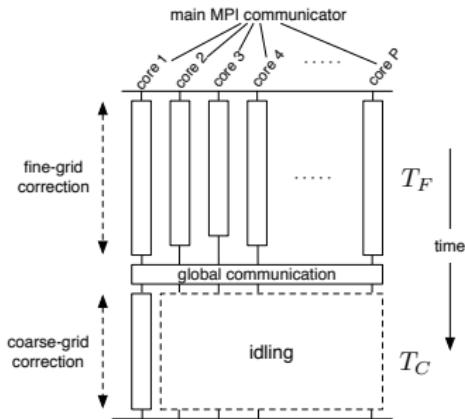
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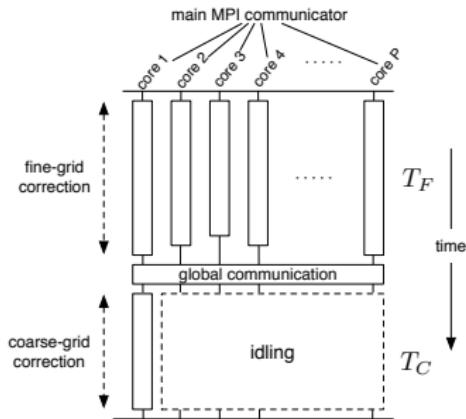
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 - (1)-(2) always exploited in BDDC implementations
 - Let us see **how to exploit (3)**, in order to introduce **asynchronicity** and boost scalability (**overlapped** implementation)

Naive parallel implementation

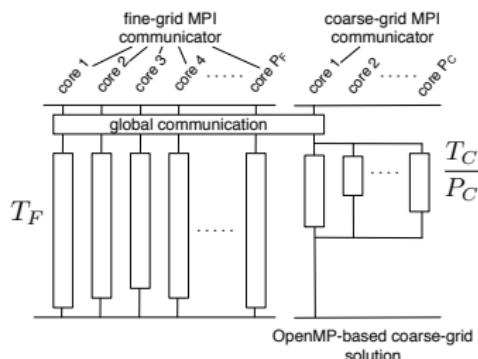


- All MPI tasks have f-g duties and one/several have also c-g duties
- Computation of f-g/c-g duties serialized (but they are independent!)
- $T_C \propto O(P^2) \rightarrow$ idling $\simeq PT_C$
- $\text{mem} \propto \mathcal{O}(P^{\frac{4}{3}}) \rightarrow$ mem per core rapidly exceeded

Naive parallel implementation



Overlapped implementation Our approach



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- $\text{mem} \propto \mathcal{O}(P^{\frac{4}{3}}) \rightarrow \text{mem per core rapidly exceeded}$

- MPI tasks have either f-g OR c-g duties
- f-g/c-g corrections OVERLAPPED in time
- c-g duties can be MASKED with f-g ones
- OpenMP/MPI-based (**MPI later on this talk**) solutions possible for c-g correction

Overlapping regions

Solve $Ax = b$ by BDDC-PCG

Set-up M_{BDDC}

$$r^0 := b - Ax^0$$

$$x^0 := x^0 + R_I A_{II}^{-1} R_I^t r^0$$

call PCG($A, M_{\text{BDDC}}, b, x^0$)

PCG

$$r^0 := b - Ax^0$$

$$z^0 := M_{\text{BDDC}}^{-1} r^0$$

$$p^0 := z^0$$

for $j = 0, \dots$, till CONV **do**

$$s^{j+1} = Ap^j$$

...

$$z^{j+1} := M_{\text{BDDC}}^{-1} r^{j+1}$$

...

end for

Overlapping regions

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...

end for

Fine-grid tasks	Coarse-grid task
Identify local coarse DoFs	
Gather coarse-grid DoFs	
Symb fact($G_{A_F^{(i)}}$)	$\mathcal{O}(n_i^{\frac{4}{3}})$
Symb fact($G_{A_H^{(i)}}$)	$\mathcal{O}(n_i^{\frac{4}{3}})$
Num fact($A_F^{(i)}$)	$\mathcal{O}(n_i^2)$
Compute Φ_i	$\mathcal{O}(n_i^{\frac{4}{3}})$
$A_C^{(i)} := \Phi_i^t A^{(i)} \Phi_i$	
Gather $A_C^{(i)}$	
Num fact($A_{II}^{(i)}$)	$\mathcal{O}(n_i^2)$
$x_0 := x_0 + R_I A_{II}^{-1} R_I^t r_0$	$\mathcal{O}(n_i^{\frac{4}{3}})$
$r_0 := b - Ax_0$	
$r^{(i)} := E_i^t r$	
$r_C^{(i)} := \Phi_i^t r^{(i)}$	
Gather $r_C^{(i)}$	
Compute $s_F^{(i)}$	$\mathcal{O}(n_i^{\frac{4}{3}})$
Solve $A_C z_C = r_C$	
Scatter z_C into $z_C^{(i)}$	
$s_C^{(i)} := \Phi_i z_C^{(i)}$	
$z^{(i)} := E_i(s_F^{(i)} + s_C^{(i)})$	
$z_I^{(i)} := -(A_{II}^{(i)})^{-1} A_{I\Gamma}^{(i)} z_\Gamma^{(i)}$	$\mathcal{O}(n_i^{\frac{4}{3}})$

Weak scalability for 3D Laplacian

Target machine: [HELIOS@IFERC-CSC](#)

4,410 bullx B510 compute blades (2 Intel Xeon E5-2680 8-core CPUs; 64GB)

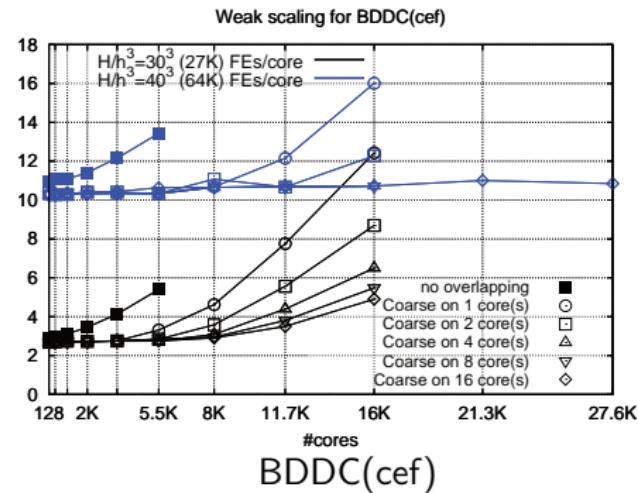
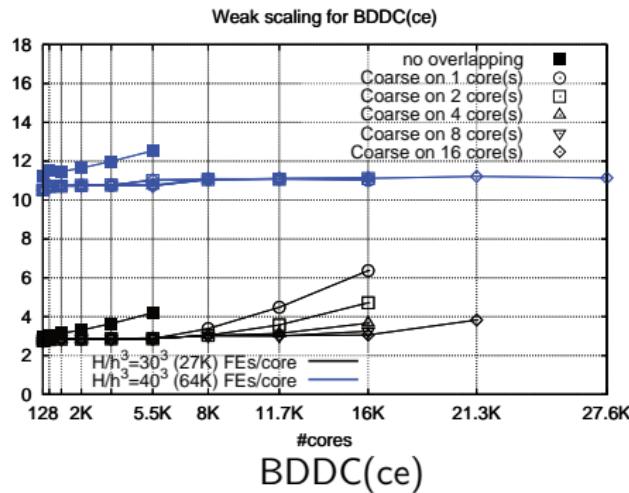
- Target problem: $-\Delta u = f$ on $\bar{\Omega} = [0, 2] \times [0, 1] \times [0, 1]$
- Uniform global mesh (Q1 FEs) + Uniform partition (cubic local meshes)
- 8,432, ..., 27648 cores for fine duties
- Direct solution of Dirichlet/Neumann/coarse problems (PARDISO)
- Entire 16-core blade for coarse-grid duties (multi-threaded PARDISO)
- Gradually larger local problem sizes: $\frac{H}{h} = 30^3, 40^3$ FEs/core

Weak scaling 2-level BDDC

3D Laplacian problem on HELIOS@ifERC

4,410 bullx B510 compute blades (2 Intel Xeon E5-2680 8-core CPUs; 64GB)

Largest problem size is 1.8 billion DoFs



Total time (secs.)

Outline

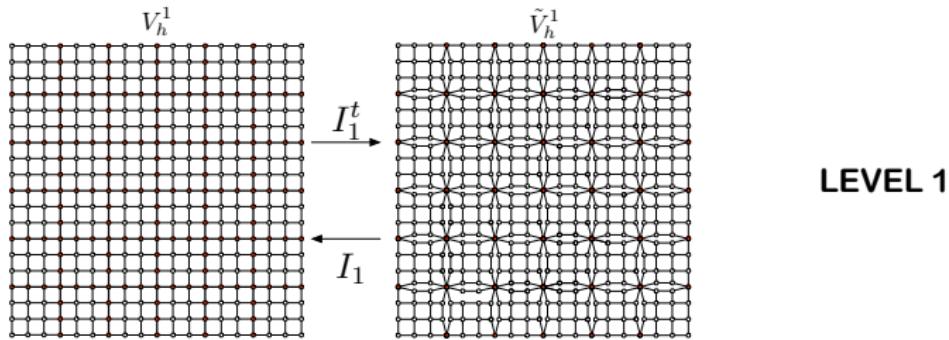
- 1 Introduction and motivation
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- ⑤ A **multilevel** extension is possible (for extreme core counts)
 - (1)-(2)-(3)-(4) already exploited in our codes
 - Let us see **how to exploit (5)**, in order to boost scalability even further (for exact version for the moment)

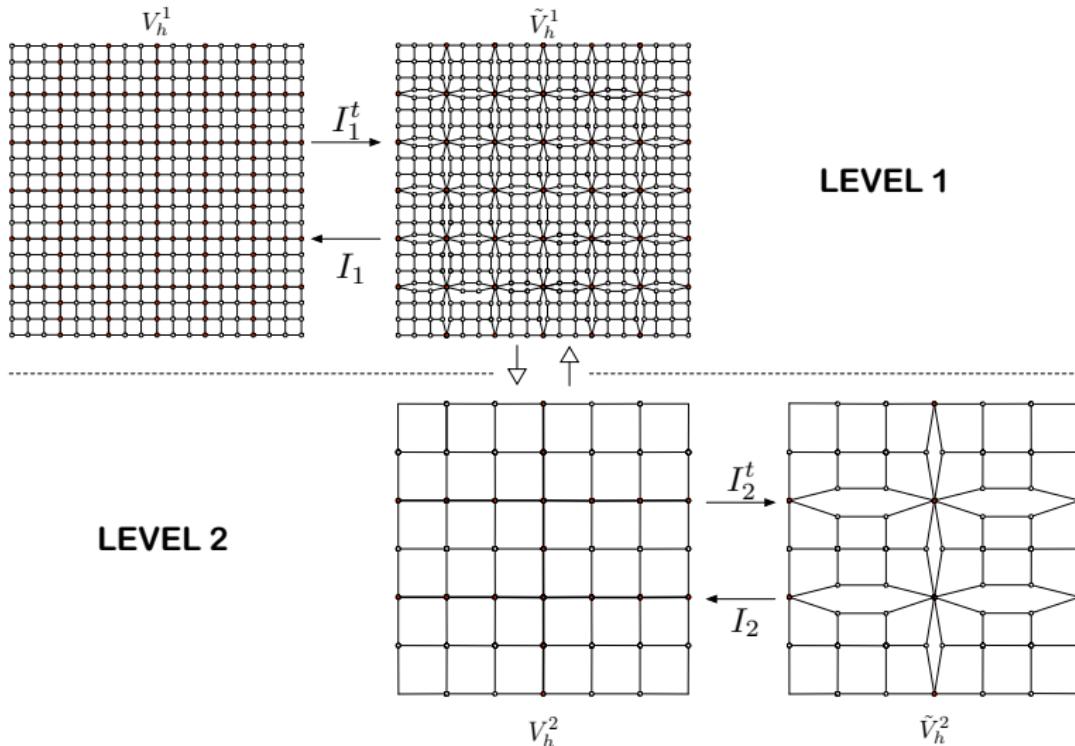
MLBDDC basic idea

MLBDDC [Mandel et. al., 2008]: Replace coarse problem by BDDC precond



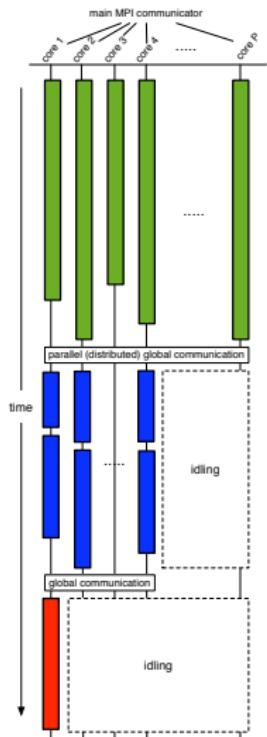
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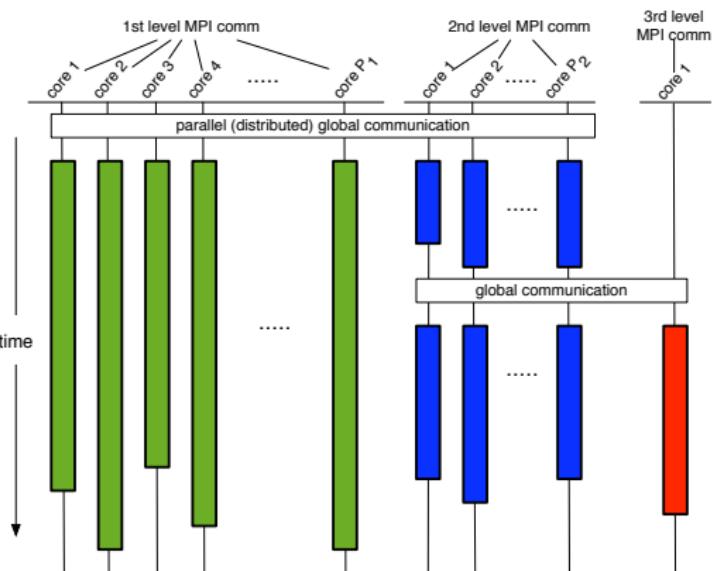


Highly scalable implementation of MLBDDC

Naive parallel implementation



Overlapped implementation Our approach

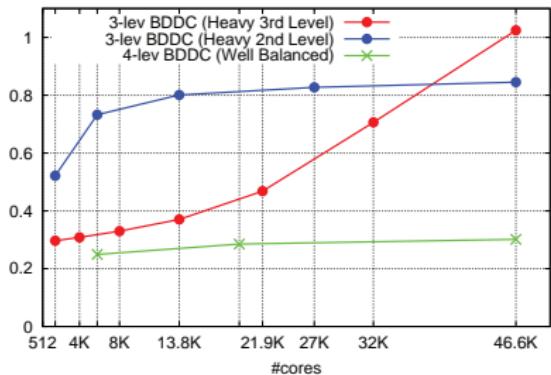


Goal: strike a balance such that blue/red areas are kept below green ones!

L_1 MPI tasks	L_2 MPI tasks	L_3 MPI task
Identify local coarse DoFs		
Gather coarse-grid DoFs		
Algorithm 1 ($k \equiv i_{L_1}$)	Build $G_{AC}^{(i_{L_2})}$	
Algorithm 2 ($k \equiv i_{L_1}$)	Identify local coarse DoFs	
Compute $\Phi_{i_{L_1}}$	Gather coarse-grid DoFs	
$A_C^{(i_{L_1})} \leftarrow \Phi_{i_{L_1}}^t (-C_{i_{L_1}}^T \Lambda_{i_{L_1}})$	Algorithm 1 ($k \equiv i_{L_2}$)	Build G_{AC}
Gather $A_C^{(i_{L_1})}$		Re+Sy fact(G_{AC})
Algorithm 3 ($k \equiv i_{L_1}$)	$A_C^{(i_{L_2})} := \text{assemb}(A_C^{(i_{L_1})})$	
Algorithm 4 ($k \equiv i_{L_1}$)	Algorithm 2 ($k \equiv i_{L_2}$)	
	Compute $\Phi_{i_{L_2}}$	
	$A_C^{(i_{L_2})} \leftarrow \Phi_{i_{L_2}}^t (-C_{i_{L_2}}^T \Lambda_{i_{L_2}})$	
	Gather $A_C^{(i_{L_2})}$	
	Algorithm 3 ($k \equiv i_{L_2}$)	$A_C := \text{assemb}(A_C^{(i_{L_2})})$
		Num fact(A_C)
Gather $r_C^{(i_{L_1})}$		
Algorithm 5 ($k \equiv i_{L_1}$)	$r_C^{(i_{L_2})} := \text{assemb}(r_C^{(i_{L_1})})$	
	Algorithm 4 ($k \equiv i_{L_2}$)	
	Gather $r_C^{(i_{L_2})}$	
	Algorithm 5 ($k \equiv i_{L_2}$)	$r_C := \text{assemb}(r_C^{(i_{L_2})})$
		Solve $A_C z_C = r_C$
		Scatter z_C into $z_C^{(i_{L_2})}$
	Algorithm 6 ($k \equiv i_{L_2}$)	
Scatter $z_C^{(i_{L_2})}$ into $z_C^{(i_{L_1})}$		
Algorithm 6 ($k \equiv i_{L_1}$)		

Weak scaling (1K FEs/core)

Weak scaling for MLBDDC(cef) solver with 1K FEs/core



Total Time (secs.)

Algorithm 1
 $\text{Re+Sy fact}(G_{AF}^{(k)})$
 $\text{Re+Sy fact}(G_{AH}^{(k)})$

Algorithm 2
 $\text{Num fact}(A_F^{(k)})$

Algorithm 3
 $\text{Num fact}(A_{II}^{(k)})$

Algorithm 4
 $\delta_I^{(k)} \leftarrow (A_{II}^{(k)})^{-1} r_I^{(k)}$
 $r_I^{(k)} \leftarrow r_I^{(k)} - A_{II}^{(k)} \delta_I^{(k)}$
 $r_I^{(k)} \leftarrow E_k^{(k)} r$

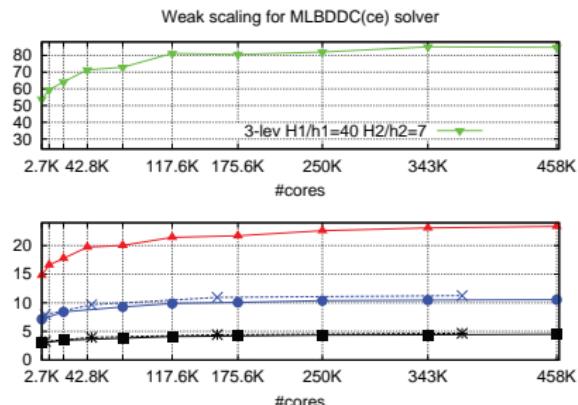
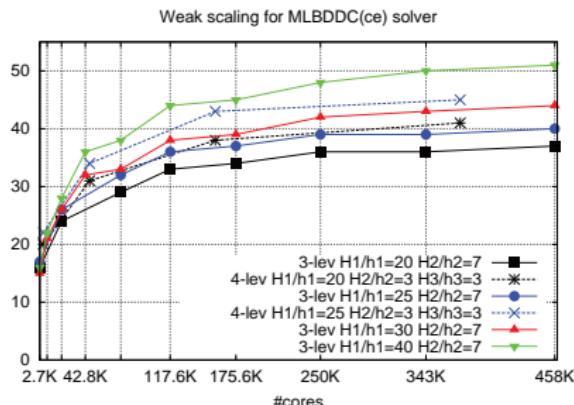
Algorithm 5
 $\text{Solve } A_F^{(k)} \begin{bmatrix} t \\ s_F^{(k)} \\ \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ r^{(k)} \\ 0 \end{bmatrix}$

Algorithm 6
 $s_C^{(k)} \leftarrow \Phi_k z_C^{(k)}$
 $z^{(k)} \leftarrow E_k(s_C^{(k)} + s_C^{(k)})$
 $z_I^{(k)} \leftarrow -(A_{II}^{(k)})^{-1} A_{I\Gamma}^{(k)} z_\Gamma^{(k)}$
 $z_I^{(k)} \leftarrow z_I^{(k)} + \delta_I^{(k)}$

Weak scaling 3-lev BDDC(ce) solver

3D Laplacian problem on IBM BG/Q (JUQUEEN@JSC)
 16 MPI tasks/compute node, 1 OpenMP thread/MPI task

Largest problem size is 29.2 billion DoFs



#PCG iterations

Total time (secs.)

Experiment set-up

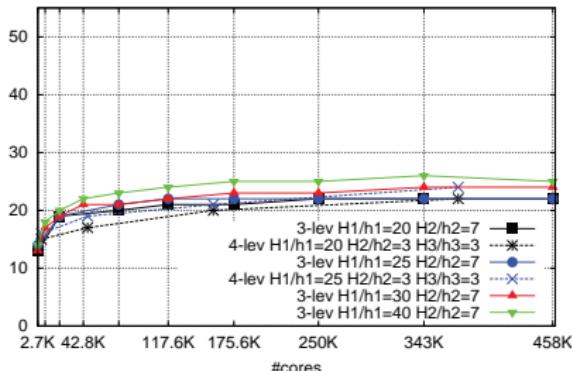
Lev.	# MPI tasks						$F_Es/core$
	1st	42.8K	74.1K	117.6K	175.6K	250K	
1st	42.8K	74.1K	117.6K	175.6K	250K	343K	$20^3/25^3/30^3/40^3$
2nd	125	216	343	512	729	1000	7^3
3rd	1	1	1	1	1	1	n/a

Weak scaling 3-lev BDDC(cef) solver

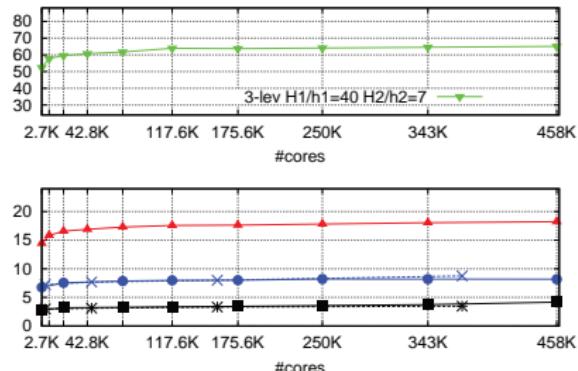
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Weak scaling for MLBDDC(cef) solver



Weak scaling for MLBDDC(cef) solver



#PCG iterations

Total time (secs.)

Experiment set-up

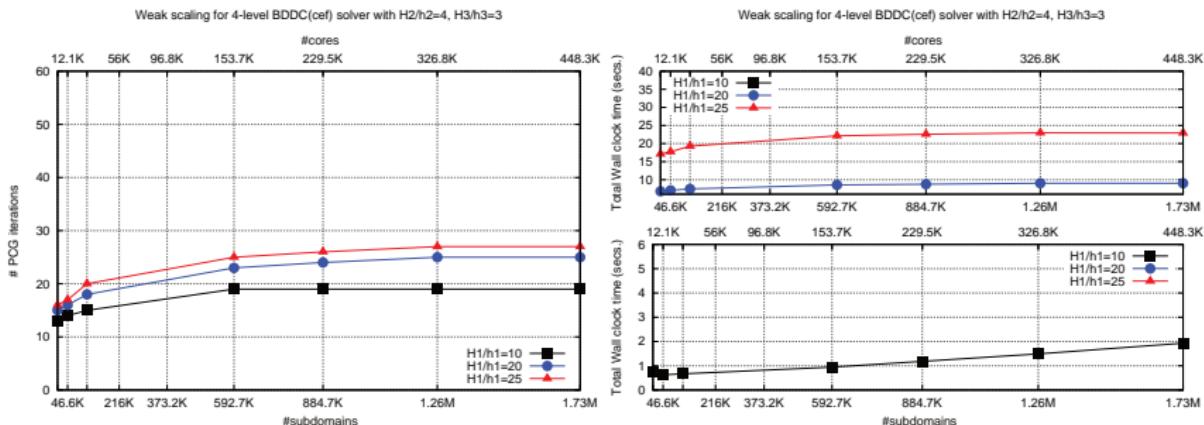
Lev.	# MPI tasks						FEs/core		
	1st	42.8K	74.1K	117.6K	175.6K	250K	343K	456.5K	$20^3/25^3/30^3/40^3$
2nd	125	216	343	512	729	1000	1331	1	7^3
3rd	1	1	1	1	1	1	1	1	n/a

Weak scal. 4-lev BDDC(cef)+4 MPI tasks/core

3D Laplacian problem on IBM BG/Q (JUQUEEN@JSC)

64 MPI tasks/compute node, 1 OpenMP thread/MPI task

Largest problem size is 27 billion DoFs



PCG iters.

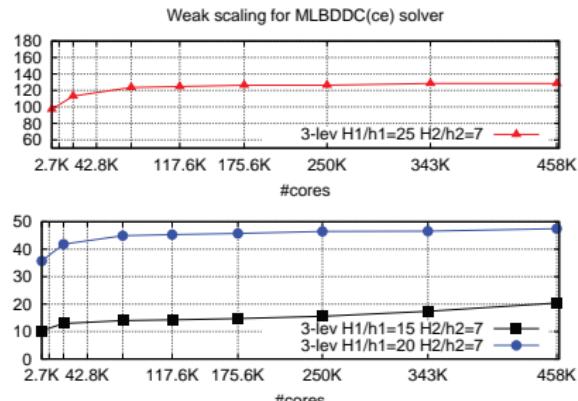
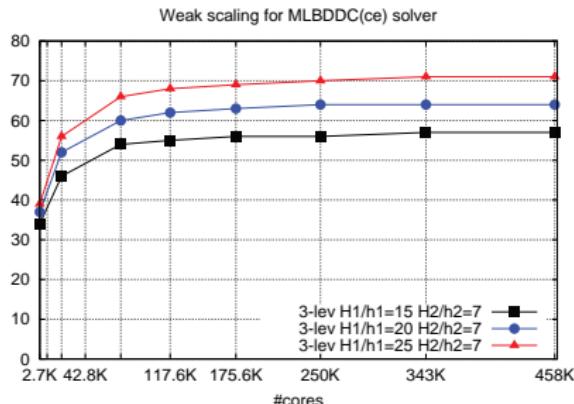
Total time (secs.)

Lev.	# MPI tasks							FEs/core
1st	110.6K	216K	373.2K	592.7K	884.7K	1.26M	1.73M	$10^3 / 20^3 / 25^3$
2nd	1.73K	3.38K	5.83K	9.26K	13.8K	19.7K	64K	4^3
3rd	64	125	216	343	512	729	1K	3^3
4th	1	1	1	1	1	1	1	n/a

Weak scaling 3-lev BDDC(ce) solver

3D Linear Elasticity problem on IBM BG/Q (JUQUEEN@JSC)
16 MPI tasks/compute node, 1 OpenMP thread/MPI task

Largest problem size is 21.4 billion DoFs



#PCG iterations

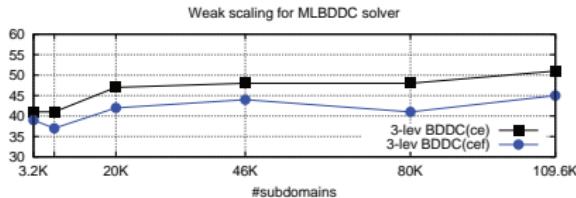
Total time (secs.)

Experiment set-up

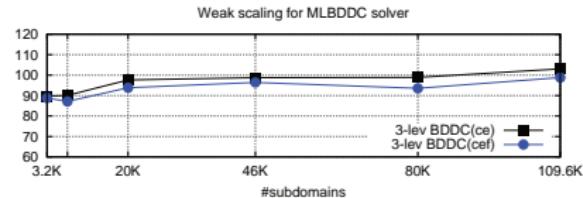
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1st	42.8K	74.1K	117.6K	175.6K	250K	343K	$15^3 / 20^3 / 25^3$
2nd	125	216	343	512	729	1000	7^3
3rd	1	1	1	1	1	1	n/a

Weak scaling 3-lev BDDC

3D Laplacian problem + **unstructured mesh discretizations**
16 MPI tasks/compute node, 1 OpenMP thread/MPI task



#PCG iterations



Total time (secs.)

- Unstructured meshes of tetrahedra
- Each L_2 subdomain aggregates ≈ 384 L_1 subdomains
- Largest unstructured problem has ~ 3.6 billion FEs
- Remarkable scalability despite underlying irregularity of the problem

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- Highly scalable implementation of (linear/exact) MLBDDC
 - Fully-distributed
 - Communicator-aware
 - Interlevel-overlapped (coarse-grain comput./comput./comm. overlap)
 - Recursive (extensible to arbitrary # levels)
- Remarkable scalability
 - 3D Laplacian and Linear Elasticity PDEs
 - 3/4 levels are sufficient to (efficiently) scale till full JUQUEEN
 - Largest scaling/problem sizes reported so far with (linear/exact) DDM

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- The algorithms presented herein have been implemented in FEMPAR:
 - OO framework for the development of parallel multiphysics FE solvers
 - From desktops/laptops to extreme scale supercomputers
 - More than >200K Fortran2008 code long
 - JSC High-Q club member since 2014
 - Open source project (GNU GPLv3)
- At the solver's level, FEMPAR provides an advanced OO user-customizable framework for the development of highly scalable MLBDDC precond's:
 - Tight integration among discretization/preconditioner (not black-box)
 - Up-to-know: "standard" MLBDDC [this talk], (r)PB-BDDC [Badia, Martín, Nguyen, Submitted'16], BDD by constraints+perturbation [Badia, Nguyen, SISC'16], Inexact BDDC [Badia, Martín, Principe, ParCo'15]
 - On-going: (ML)BDDC for unfitted FEM [Badia's talk SIAM CSE17], Inexact MLBDDC (hybrid DD/AMG)
 - Future plans: MLBDDC suitable for $\mathcal{H}(\mathbf{curl})$ and $\mathcal{H}(\mathbf{div})$ FE spaces
 - Other extensions that we envision possible are: deluxe scaling, spectral methods ... (volunteers?)

<https://gitlab.com/fempar/fempar>

Thank you!

-  Santiago Badia, Alberto F. Martín and Hieu Nguyen Physics-based balancing domain decomposition by constraints for heterogeneous problems. *Submitted*, 2016.
-  Santiago Badia, A. F. Martín and J. Principe. Multilevel balancing domain decomposition at extreme scales. *SIAM Journal on Scientific Computing*. Vol. 38(1), pp. C22–C52, 2016.
-  Santiago Badia, A. F. Martín and J. Principe. A highly scalable parallel implementation of balancing domain decomposition by constraints. *SIAM Journal on Scientific Computing*. Vol. 36(2), pp. C190-C218, 2014.