

## Plan of presentation

## Part 1:

## Motivation of DDM

Partitioning algorithms

Classical Schwarz algorithm

Matrix/discrete level
Convergence
Two level approach

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## DDM Motivation

The global problem cannot fit into main memory, out of core computations: very slow swapping to disk
(AND $\mid$ OR) Concurrency can be exploited to solve the global problem: solving problem faster on parallel computers
(AND | OR) The solution of the subproblems is "easier" than the global problem: direct methods on smaller subproblems cache friendly

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## Domain Decomposition



Divide and conquer applied to PDEs:


Decompose domain into many sub-domains
Solve independently each smaller problem
Glue the solutions together: convergence?

## Mesh partifioning: decompose the domain

- Geometric Based Algorithms


## Practical DDM

Each part of problem solved on a compute node:


- Coordinate bisection
- Inertia bisection
- Graph Theory Based Algorithms
- Graph bisection
- Greedy algorithm
- Spectral bisection
- K-L algorithm
- Other Partitioning Algorithms
- Global optimization algorithms
- Reducing the bandwidth of the matrix
- Index based algorithms
- The State of the Art
- Hybrid approach
- Multilevel approach
- Parallel partitioning algorithms



## Example: spectral bisection



Laplacian

$$
\begin{aligned}
& L x=\left[\begin{array}{ccccccc}
\ddots & \ddots & \ddots & \ddots & \ddots & & \\
& -1 & -1 & 4 & -1 & -1 & \\
& & \ddots & \ddots & \ddots & \ddots & \ddots
\end{array}\right] x=\lambda x \\
& \sum_{i=1}^{n}\left(x_{i}\right)=0 \quad \sum_{i=1}^{n}\left(x_{i}\right)^{2}=n
\end{aligned}
$$

- Needs to be an eigenvector of Laplacian
- If composed of half +1 and half- 1 it satisfies the two constraints
- Finding the Fielder vector: Lanczos algorithm
- Proceed recursively...


## Partifioned meshes

Examples from Computational Fluid Dynamics: ParMetis


## Mesh partitioning

Represents only the technical part of DDM
Has deep ties with parallel computing: MIMD
DDM denotes also the development of special algorithms to solve decomposed problems

Algorithms: Schwarz, FETI, sub-structuring ...

## Basic DD methods


(Overlapping) Schwarz (1870): existence of elliptic problems on non trivial domains
(Non-overlapping) Schur / sub-structuring methods
Kron (53)


Przemieniecki (63)

2 classes of methods: overlapping and non-overlapping

## Domain Decomposition

## Divide and conquer applied to PDEs

Decompose domains into many sub-domains
Solve independently each smaller problem
Glue the solutions together: convergence?

## Classical Schwarz

## Suppose we need to solve:

$$
\mathcal{L} u=f \quad \text { in } \Omega, \quad \mathcal{B} u=g \quad \text { on } \partial \Omega
$$

Partition the original domain into 2 domains:



## Schwarz with large overlap


$\Delta u=0$, on $[-1,1]$ with $u(-1)=u(1)=0$

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## Schwarz no overlap


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## Schwarz no overlap



## Schwarz no overlap


$\Delta u=0$, on $[-1,1]$ with $u(-1)=u(1)=0$

## Matrix formulation

Continuous problem:

$$
\begin{aligned}
& \begin{array}{rllllll}
\mathcal{L} u_{1}^{n+1} & =f & \text { in } \Omega_{1}, & \mathcal{L} u_{2}^{n+1} & =f & \text { in } \Omega_{2}, \\
\mathcal{B}\left(u_{1}^{n+1}\right) & =g & \text { on } \partial \Omega_{1}, & \mathcal{B}\left(u_{2}^{n+1}\right) & =g & \text { on } \partial \Omega_{2}, \\
u_{1}^{n+1} & =u_{2}^{n} & \text { on } \Gamma_{12}, & u_{2}^{n+1} & =u_{1}^{n} & \text { on } \Gamma_{21},
\end{array} \\
& A_{1} \underline{u}_{1}^{n+1}=\underline{\mathrm{f}}_{1}+B_{21} \underline{\mathrm{u}}_{2}^{n} \quad A_{2} \underline{\mathrm{u}}_{2}^{n+1}=\underline{\mathrm{f}}_{2}+B_{12} \underline{\mathrm{u}}_{1}^{n}
\end{aligned}
$$

Partition of unity: $\quad e=\tilde{R}_{1}^{T} e_{1}+\tilde{R}_{2}^{T} e_{2} \rightarrow \underline{u}^{n+1}=\tilde{R}_{1}^{T} \underline{u}_{1}^{n+1}+\tilde{R}_{2}^{T} \underline{u}_{2}^{n+1}$ Restriction operator:

$$
\underline{\mathrm{u}}_{k}^{n}=R_{k} \underline{\mathbf{u}}
$$

## Matrix formulation

If consistent then: $A_{1} R_{1}-B_{21} R_{2}=R_{1} A, \quad A_{2} R_{2}-B_{12} R_{1}=R_{2} A$. Leading to:

$$
\underline{\mathrm{u}}^{n+1}=\underline{\mathrm{u}}^{n}+\sum_{i=1}^{2} \tilde{R}_{i}^{T} \underbrace{A_{i}^{-1} R_{i}\left(f-A \underline{u}^{n}\right)}
$$

RAS: Restricted Additive Schwarz

- Nonsymmetric
- Default option in PETSC
- Cai and Sarkis (1997)
- Equivalent to continuous

AS: Additive Schwarz $\tilde{R}_{i}^{T} \rightarrow R_{i}^{T}$

- Symmetric
- No continuous equivalent (EGO2)
- Use with Krylov accelerator
- Nepomnyaschikh (86)


## Convergence theory

For symmetric positive definite matrices
No results for Restricted Additive Schwarz
Solved by using
Convergence rate not optimal a coarse solver

Convergence rate not scalable
Developed by Lions, Dryia, Widlund, BRamble, Pasciak, Wang, Xu, Zhang etc ...

## Matrix formulation

On multiple domains:

$$
\underline{\mathrm{u}}^{n+1}=\underline{\mathrm{u}}^{n}+\sum_{i=1}^{K} \tilde{R}_{i}^{T} A_{i}^{-1} R_{i}\left(f-A \underline{\mathrm{u}}^{n}\right)
$$

Preconditioning in Krylov methods:

$$
M_{R A S}^{-1}=\sum_{i=1}^{K} \tilde{R}_{i}^{T} A_{i}^{-1} R_{i} \quad M_{A S}^{-1}=\sum_{i=1}^{K} R_{i}^{T} A_{i}^{-1} R_{i}
$$

- In practice the restriction and extension are not created
- Matrix Problem can be reformulated: lower operation counts


## Convergence addifive Schwarz

$$
\begin{array}{r}
\text { Convergence of PCG: }\left\|\underline{u}^{(k)}-\underline{u}^{*}\right\| \leq 2 \gamma^{k}\left\|\underline{u}^{(0)}-u^{*}\right\| \\
\text { where } \gamma=\frac{\sqrt{\kappa\left(M^{-1} A\right)}-1}{\sqrt{\kappa\left(M^{-1} A\right)}+1}
\end{array}
$$

| Subdomain diameter: | $H=\max _{1 \leq i \leq K} \operatorname{diam}\left(\Omega_{i}\right)$ |
| :--- | :---: |
| Mesh size: | $h$ |
| Overlap size: | $\beta H, \beta \in(0,1]$ |

## Convergence addifive Schwarz

$$
\kappa\left(M_{A S}^{-1} A\right) \leq C H^{-2}\left(1 \pm \beta^{-1}\right)
$$

Diameter tends to zero as the number of subdomain increases
The overlap size does not remove the diameter problem
Estimate worsen when A has varying coefficients: $\nabla \cdot(a(\mathbf{x}) \nabla u)$
Diameter dependence prevents algorithmic scalability
Schwarz didn't care about the scalability!

## Practical scalability



If scalable the solution is reached 4 times faster!

If additive Schwarz is used it takes the half time to solve!!

## Scalable/Opimal DDM algorihm

A DDM is scalable if its rate of convergence does not deteriorate when the number of subdomains grows.

A DDM for the solution of a linear system is optimal if its rate of convergence to the exact solution is independent of the size of the system.

The first definition involves $H$
The second involves $\quad h$

## Two level methods

Add a very coarse problem solved on the entire domain
Removes completely the subdomain diameter problem
Not easy to parallelize! (Duplication of coarse solves)
$M_{A S, 2}^{-1}=R_{H}^{T} A_{H}^{-1} R_{H}+\sum_{i=1}^{K} R_{i}^{T} A_{i}^{-1} R_{i}=\sum_{i=0}^{K} R_{i}^{T} A_{i}^{-1} R_{i}$
Condition number:

$$
\kappa\left(M_{A S, 2}^{-1} A\right) \leq C\left(1+\beta^{-1}\right)
$$

Varying coefficients:
$C(\beta)(1+\log (H / h), C(\beta)(H / h)$

## Two level methods

Still not perfect since overlap must be kept constant!
The perfect method would have zero overlap and a condition number independent of $H$ and $h$

Is it possible to construct such a Schwarz method ??
If not how close can we get?

## DDM sources

## Part 2.

## The Robin method

# Fourier analysis of Classical Schwarz 

Fourier analysis for optimized Schwarz
Optimization over all Fourier modes
Examples FDM
High-Order methods (HOMs)
Optimized Schwarz in a massively parallel GCM
Conclusion

## DDM sources

## Domein

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DDM sources


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