

# Parallel Computing and Iterative Solution of Linear Systems

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## Outline of the talk

Aim: describe parallel iterative solvers for PDE-FEM linear algebraic systems, discuss their mathematical/computer implementation aspects

- Paralelization of CG and other iterative methods
- Block diagonal and full block preconditioners
- Overlapping Schwarz DD methods
- Convergence analysis
- Two-level Schwarz methods
- Algebraic constructions of coarse grid problem
- Nonoverlapping DD methods
- What is not considered

## Literature

- Chan, T.F., Mathew, T.P., Domain Decomposition Algorithms. Acta Numerica 1994, 61-143
- Smith, B.F., Bjørstad, P.E., Gropp, W.D., Domain Decomposition Parallel Multilevel Methods for Elliptic Partial Differential Equations. Cambridge University Press, Cambridge 1996
- A. Toselli, O. Widlund, Domain Decomposition Methods - Algorithms and Theory, Springer 2004
- R. Blaheta, Space decomposition Preconditioners and Parallel Solvers. In: M. Feistauer et al. eds. Numerical Mathematics and Advanced Appl., Springer 2004, pp. 20-38 (ENUMATH 2003 invited lecture)
- R. Blaheta, O. Jakl, J. Starý, Linear system solvers based on space decompositions and parallel computations, Inženýrská Mechanika, 10(2003), pp. 439-454

## The Problem

For modelling of different phenomena as diffusion etc. , it is necessary to solve a symmetric elliptic boundary value problem in  $\Omega$  like:

$$\begin{aligned} - \sum_{ij} \frac{\partial}{\partial x_i} \left( k_{ij} \frac{\partial u}{\partial x_j} \right) &= f && \text{in } \Omega \\ u &= \hat{u} && \text{on } \Gamma_0 \subset \partial\Omega \\ \sum_{ij} k_{ij} \frac{\partial u}{\partial x_j} n_i &= \hat{f} && \text{on } \Gamma_1 \subset \partial\Omega \end{aligned}$$

Discretization by FEM leads to an algebraic problem

$$Au = b$$

with a matrix  $A$ , which is SPD, large scale (dimension  $n \sim 10^5 - 10^7$ ) and ill conditioned.

# Linear System Solvers

- direct methods (Gauss elimination type)
- iterative methods - history
  - relaxation (Gauss, Jacobi, Southwell)
  - SOR (Young 1954, Varga 1962)
  - CG (Hestenes, Stiefel, Lanczos, 1952)
  - practical CG (Reid 1971)
  - ILU preconditioning (Axelsson 1972, van der Vorst 1977,...)
  - multigrid (Fedorenko 1964, Hackbusch 1977, ...), AMG
  - GMRES (Saad, Schulz 1986)
  - DD, //iterative methods (Dryja, Widlund, ...1st DD conf. 1989)
- iterative methods - today: **Numerical scalability** (optimal linear complexity), **Parallel scalability**, Robustness, Nonsymmetric Systems..

## Sequential CG

given  $u^0$

compute  $r^0 = b - Au^0$ ,

$g^0 = B^{-1}r^0$ ,  $r^0 = g^0$ ,  $\sigma = \langle r^0, g^0 \rangle$

**for**  $i = 0, 1, \dots$  **until** convergence **do**

$w^i = Av^i$

$\alpha_i = \sigma / \langle v^i, w^i \rangle$

$u^{i+1} = u^i + \alpha_i v^i$

$r^{i+1} = r^i - \alpha_i w^i$

$g^{i+1} = B^{-1}r^{i+1}$

$\sigma_0 = \sigma$ ,  $\sigma = \langle r^{i+1}, g^{i+1} \rangle$

$\beta_i = \sigma / \sigma_0$

$v^{i+1} = g^{i+1} + \beta_i v^i$

**end**

**CG iteration needs**

- matrix-by-vector multiplication
- vector updates
- two inner products

**Preconditioning**

$g = G(r) = B^{-1}r$

standard B - SPD

#it  $\sim$

$\sqrt{\text{cond}(B^{-1}A)} \ln(1/\varepsilon)$

**Convergence test**

$\| r^i \| \leq \varepsilon \| b \|$

## Parallelization by Splitting to Blocks

$$A = \left[ \begin{array}{cccc} A_{11} & A_{12} & A_{13} & A_{14} \\ & A_{22} & A_{23} & A_{24} \\ & & A_{33} & A_{34} \\ sym & & & A_{44} \end{array} \right] \left. \begin{array}{l} \\ \\ \\ \end{array} \right\} \begin{array}{l} A_1 \\ \\ A_2 \end{array} \quad v = \left[ \begin{array}{c} v_1 \\ v_2 \\ v_3 \\ v_4 \end{array} \right] \left. \begin{array}{l} \\ \\ \\ \end{array} \right\} \begin{array}{l} \underline{v}_1 \\ \\ \underline{v}_2 \end{array} \quad w = \left[ \begin{array}{c} w_1 \\ w_2 \\ w_3 \\ w_4 \end{array} \right] \left. \begin{array}{l} \\ \\ \\ \end{array} \right\} \begin{array}{l} \underline{w}_1 \\ \\ \underline{w}_2 \end{array}$$

$v, w, A$  are full (**accumulated**) vectors and matrix ,

$\underline{v} = (\underline{v}_1, \underline{v}_2)^T, w = (\underline{w}_1, \underline{w}_2)^T, \underline{A} = (A_1, A_2)^T$  are **distributed** data.

Distribution on two processors  $\mathbb{P}_1$  and  $\mathbb{P}_2$ ,  $\mathbb{P}_k \Leftrightarrow \{A_k, \underline{v}_k, \underline{w}_k, v, w\}$ .

- Matrix-by-vector multiplication  $\underline{w}_k = A_k \cdot v$
- inner product  $\sigma = \langle v, w \rangle$  or  $\sigma_k = \langle v, \underline{w}_k \rangle \rightarrow \sigma = \sum \sigma_k$
- vector updates  $v = v + \alpha \cdot w$  or  $\underline{v}_k = \underline{v}_k + \alpha \cdot \underline{w}_k$

## Preconditioning by Splitting to Blocks

$$A = \begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ & A_{22} & A_{23} & A_{24} \\ & & A_{33} & A_{34} \\ \text{sym} & & & A_{44} \end{bmatrix} = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \Rightarrow B = \begin{bmatrix} B_{11} & \\ & B_{22} \end{bmatrix}$$

$B$  additive (block diagonal, block Jacobi) preconditioner

$$A = \begin{bmatrix} I & & \\ B_{21}B_{11}^{-1} & I & \\ & & \end{bmatrix} \begin{bmatrix} B_{11} & \\ & S \end{bmatrix} \begin{bmatrix} I & B_{11}^{-1}B_{12} \\ 0 & I \end{bmatrix}$$

$$\Rightarrow B_M = \begin{bmatrix} I & & \\ B_{21}B_{11}^{-1} & I & \\ & & \end{bmatrix} \begin{bmatrix} B_{11} & \\ & B_{22} \end{bmatrix} \begin{bmatrix} I & B_{11}^{-1}B_{12} \\ & I \end{bmatrix}$$

$B_M$  symmetric multiplicative (full block, Gauss-Seidel) preconditioner



## Parallel CG

given  $u^0$

compute  $\mathbf{r}^0 = \mathbf{b} - \mathbf{A}u^0$ ,

$\mathbf{g}^0 = B^{-1}\mathbf{r}^0$ ,  $v^0 = \mathbf{g}^0$ ,  $\sigma = \langle \mathbf{r}^0, \mathbf{g}^0 \rangle$

for  $i = 0, 1, \dots$  until convergence do

$\mathbf{w}^i = \mathbf{A}v^i$

$\alpha_i = \sigma / \langle v^i, \mathbf{w}^i \rangle$  (C)

$u^{i+1} = u^i + \alpha_i v^i$

$\mathbf{r}^{i+1} = \mathbf{r}^i - \alpha_i \mathbf{w}^i$

$\mathbf{g}^{i+1} = B^{-1}\mathbf{r}^{i+1}$

$\sigma_0 = \sigma$ ,  $\sigma = \langle \mathbf{r}^{i+1}, \mathbf{g}^{i+1} \rangle$  (C)

$\beta_i = \sigma / \sigma_0$

$v^{i+1} = \mathbf{g}^{i+1} + \beta_i v^i$  (C)

end

Parallel  
computation:

multiple arithmetic  
units & better memory  
access

$v$  accumulated

$\mathbf{v} = \underline{v}$  distributed

(C)

communication points

Block diagonal  
preconditioner

## Properties of the //CG & Preconditioning

- parallel implementation of the block diagonal preconditioning, **no communication**, parallel scalability depends on the cost of the necessary communications,
- numerical scalability (efficiency):  
number of iterations vs. complexity of one iteration.
- for a general splitting into  $m$  blocks and  $m$  processors:
  1. for  $m$  small, the solution of systems with  $B_k$  is expensive,
  2. for  $m$  big, the preconditioner is not efficient.

Note:  $m \rightarrow n: B = \text{diag}(A) \Rightarrow \text{cond}(B^{-1}A) = O(h^{-2})$ . **Remedies:**

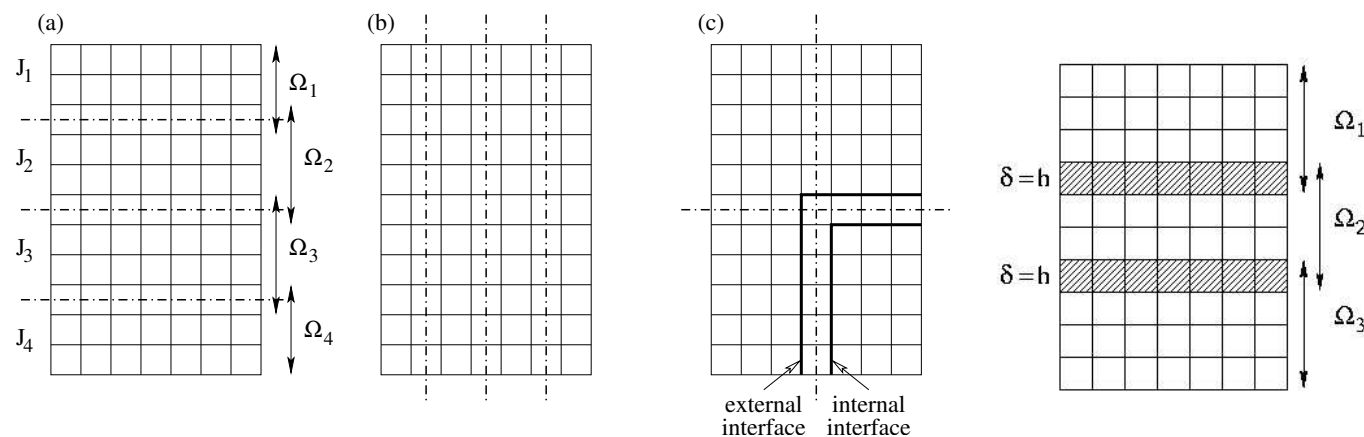
1. for  $m$  small, use an approximation to  $B_k$ , e.g. incomplete factorization,
2. for  $m$  big, use more efficient efficient methods, e.g. Schwarz overlapping DD methods.

# Block diagonal - overlapping DD preconditioners

$$u = \begin{bmatrix} \underline{u}_1 \\ \vdots \\ \underline{u}_m \end{bmatrix} \quad A = \begin{bmatrix} A_{11} & \dots & A_{1m} \\ \vdots & & \vdots \\ A_{m1} & \dots & A_{mm} \end{bmatrix}$$

- additive block diagonal preconditioner  $B$ ,
- $R_k$  Boolean  $\dots n_k \times n$ ,
- $\underline{u}_k = R_k u$ ,
- $B^{-1}r = \sum R_k^T A_{kk}^{-1} R_k r$

## Block splitting - DD:



increase of  $\delta$ :

overlapping DD

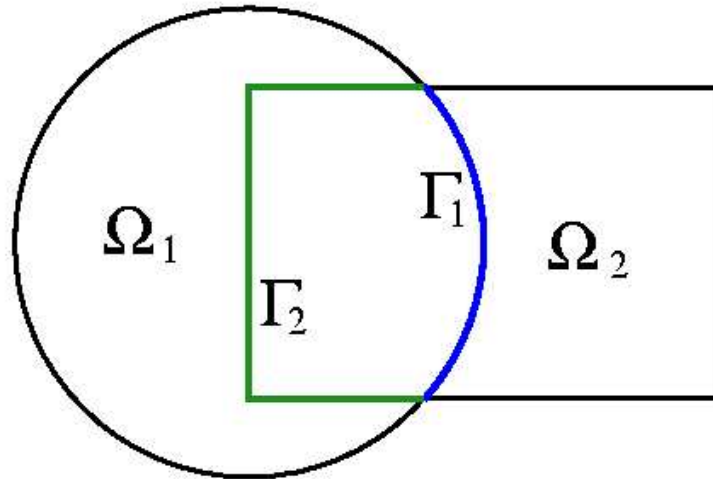
Schwarz methods:

$$B^{-1}r$$

$$= \sum R_k^T A_{kk}^{-1} R_k r$$

space decomposition - subspace correction methods

## Schwarz method/preconditioner



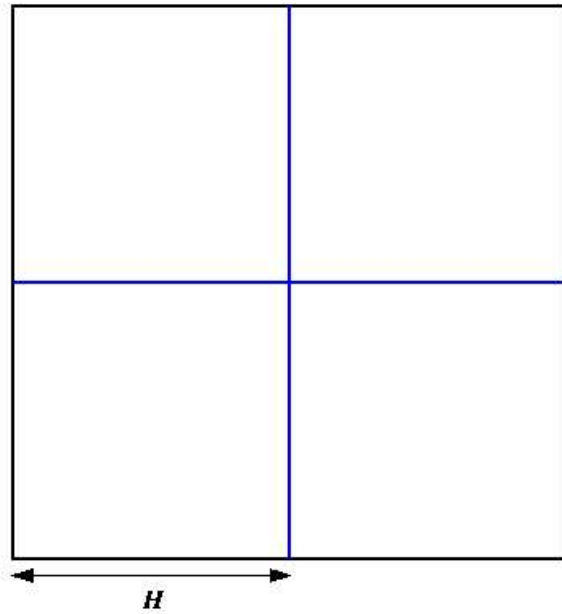
Alternating Schwarz method - existence of the solution of BVPs in more general domains, H.A. Schwarz 1870

Preconditioner

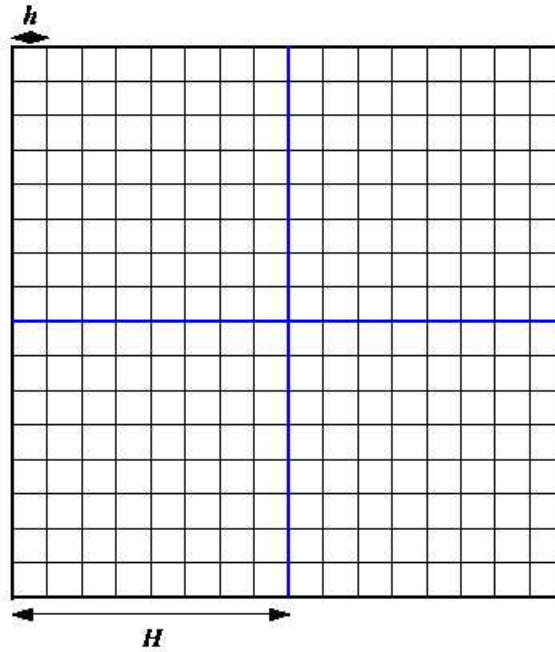
$$B^{-1}r = \sum R_k^T A_{kk}^{-1} R_k r$$

- A.M. Matsokin, S.V. Nepomnyaschikh (1985)
- Lions, P.L. (1987): On the Schwarz alternating methods *I*. In: *1st Internat. Symposium on Domain Decomposition Methods for PDE*, SIAM, pp. 1-42
- M. Dryja, O.B. Widlund (1989), Towards a unified theory of domain decomposition algorithms for elliptic problems, *3rd Internat. Symposium on Domain Decomposition Methods for PDE*, SIAM, pp. 3-21

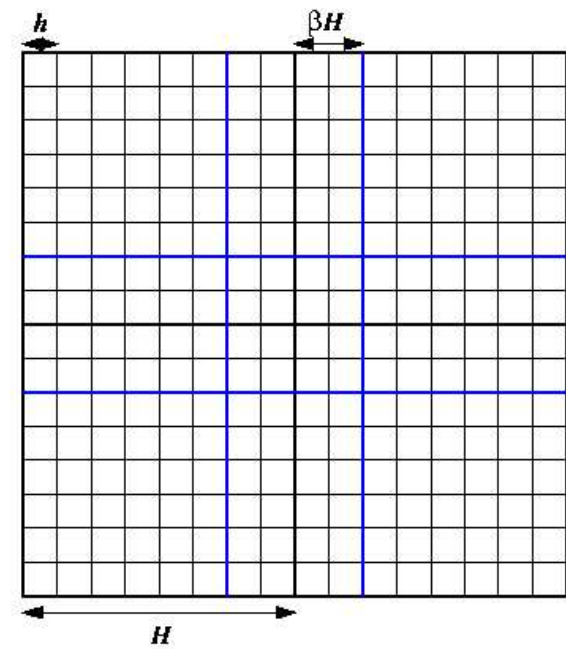
# Overlapping Subdomains



Coarse partition of  $\Omega$  into  $\Omega_k^H$  with auxiliary grid of size  $H$ .

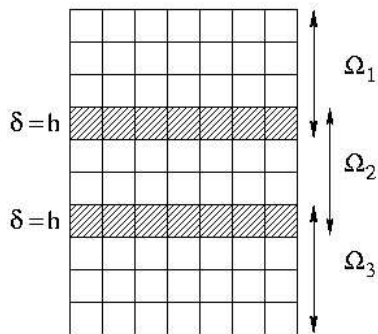


Refinement of the domain partition with grid size  $h$ .



Extension of  $\Omega_k^H$  with size  $\beta H$  into overlapping subdomains  $\Omega_k$ .

## Parallelization of the overlapping DD



**Subdomain matrix**  $\underline{A}_k$  e.g. FEM matrix corresponding to the problem in  $\Omega_k$  with the Dirichlet BC on inner boundary  $\Gamma_k$  realized by penalty scaling of diagonal elements.  $A_k = R_k^T A R_k$  is involved in  $\underline{A}_k$ .

**Subdomain vector**  $\underline{v}_k$  includes all DOFs in the subdomain  $\Omega_k \setminus \partial\Omega_{Dirichlet}$  including the DOFs on  $\Gamma_k$ .

$v, w, A$  are full (**global**) vectors and matrix ,

Distribution on processors  $\mathbb{P}_k \Leftrightarrow \{\underline{A}_k, \underline{v}_k, \underline{w}_k\}$  (only **local** data).

- Matrix-by-vector multiplication  $\underline{w}_k = \underline{A}_k \cdot \underline{v}_k$  and transfer of data to DOFs on  $\Gamma_k$ .
- inner product  $\sigma_k = \langle \underline{v}_k, \underline{w}_k \rangle - \sum_{j \text{ in overlap}} \frac{1-K_j}{K_j} \underline{v}_{k,j} \cdot \underline{w}_{k,j} \rightarrow \sigma = \sum \sigma_k$
- vector updates  $\underline{v}_k = \underline{v}_k + \alpha \cdot \underline{w}_k$

## Parallel CG with overlapping DD preconditioner

given  $u^0$

compute  $\mathbf{r}^0 = \mathbf{b} - \mathbf{A}u^0$ ,

$\mathbf{g}^0 = B^{-1}\mathbf{r}^0$ ,  $\mathbf{v}^0 = \mathbf{g}^0$ ,  $\sigma = \langle \mathbf{r}^0, \mathbf{g}^0 \rangle$

for  $i = 0, 1, \dots$  until convergence do

$\mathbf{w}^i = \mathbf{A}\mathbf{v}^i$  (C)

$\alpha_i = \sigma / \langle \mathbf{v}^i, \mathbf{w}^i \rangle$  (C)

$\mathbf{u}^{i+1} = \mathbf{u}^i + \alpha_i \mathbf{v}^i$

$\mathbf{r}^{i+1} = \mathbf{r}^i - \alpha_i \mathbf{w}^i$

$\mathbf{g}^{i+1} = B^{-1}\mathbf{r}^{i+1}$  (C)

$\sigma_0 = \sigma$ ,  $\sigma = \langle \mathbf{r}^{i+1}, \mathbf{g}^{i+1} \rangle$  (C)

$\beta_i = \sigma / \sigma_0$

$\mathbf{v}^{i+1} = \mathbf{g}^{i+1} + \beta_i \mathbf{v}^i$

end

CG with additive  
Schwarz preconditioner

Only local vectors  $\mathbf{v} = \underline{v}$   
and local matrices  $\mathbf{A}$

(C)

communication points

# Effect of the overlap

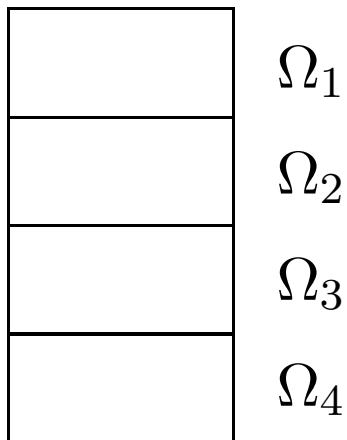
Model problem:

$$\Omega = \langle 0, 2 \rangle \times \langle 0, 3 \rangle$$

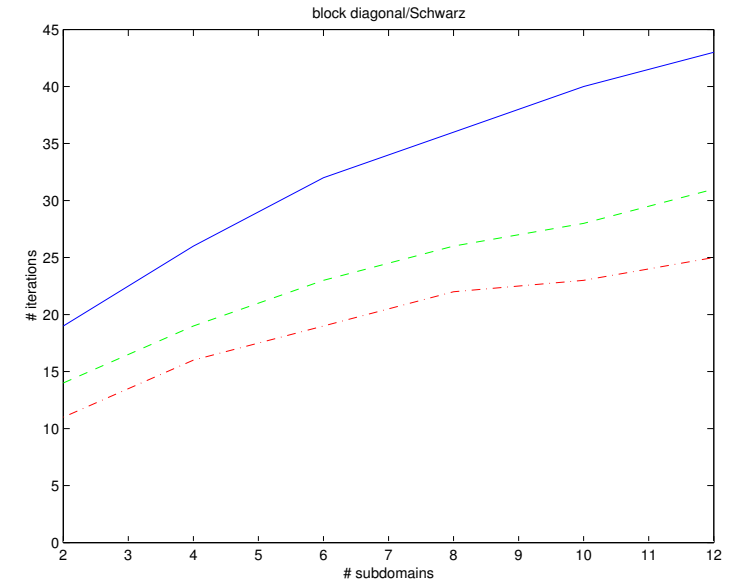
$$-\Delta u = f \text{ in } \Omega$$

$$u = 0 \text{ on } \partial\Omega$$

$$h=1/30, n=5100$$



	$\delta:$	$h$	$2h$	$3h$
$m$	$H:$	—	—	—
2		19	14	11
4		26	19	16
6		32	23	19
8		36	26	22
10		40	28	23
12		43	31	25



Numbers of iterations for  $\varepsilon = 10^{-4}$  with additive preconditioning, overlap  $\delta = h$  - Block Jacobi,  $\delta > h$  Schwarz



## Space decompositions

$$R^n, R_k : R^n \rightarrow R^{n_k} \quad (k = 1, \dots, m), \quad I_k : R^{n_k} \rightarrow R^n, \quad I_k = R_k^T$$

$$R^n = \sum_k \text{range}(I_k)$$

$$A_k = R_k^T A R_k$$

Abstract setting - finite dimensional Hilbert space  $V$

$$V = V_1 + \dots + V_m$$

$$V, R_k : V \rightarrow V_k \quad (k = 1, \dots, m), \quad I_k : V_k \rightarrow V, \quad I_k = R_k^T$$

$$Au = b, \quad u \in V, \quad b \in V', \quad A \in L(V, V')$$

$$A \text{ is SPD} \iff \langle Au, v \rangle = \langle Av, u \rangle, \quad \langle Av, v \rangle > 0 \Rightarrow \langle u, v \rangle_A$$

$$A_k = R_k^T A R_k$$

- analysis in the FE spaces,
- investigation of DD and other decomp. (DiD) in a unique framework.

# Space decomposition preconditioners

$$G : r \mapsto g, \quad g \sim A^{-1}r \text{ (pseudores.)}$$

$$g = 0$$

**for**  $k = 1, \dots, m$

$$g \leftarrow g + I_k \tilde{A}_k^{-1} R_k z_k$$

**end**

$$\tilde{A}_k \sim R_k A I_k = A_k$$

$$S_k(z) \sim A_k^{-1} z$$

exact

inexact – linear

– nonlinear

**additive** preconditioner  $G_A$ :

$$z_k = r, \quad k = 1, \dots, m$$

**multiplicative** preconditioner

$G_M$ :

$$z_k = r - Ag$$

**hybrid** preconditioner  $G_H$ , e.g.:

$$z_k = r \quad k = 1, \dots, m-1$$

$$z_m = r - Ag$$

symmetric SD preconditioner

nonsymmetric

symmetrization

$$k = 1, \dots, m-1, m, m-1, \dots, 1$$

## SD analysis, $V = V_1 + \dots + V_m$

- Assumption **A1** (stability)  $\forall v \in V \exists v_k \in V_k : v = v_1 + \dots + v_m$   

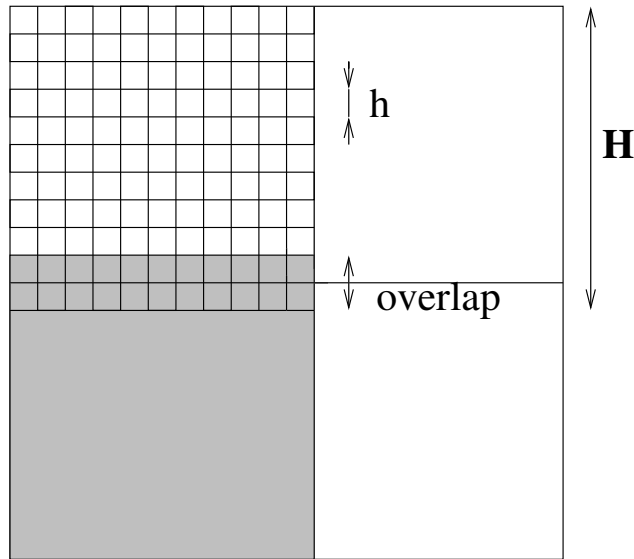
$$\sum_k \|v_k\|_A^2 \leq K_0 \|v\|_A^2.$$
- Assumption **A2**  $\forall v \in V \forall v_k \in V_k : v = v_1 + \dots + v_m$   

$$\|v\|_A^2 \leq K_1 \sum_k \|v_k\|_A^2.$$
- **Note:** Trivial bound  $K_1 = m$ . Potentially m-independent bounds:  
 If  $\mathcal{E} = (\varepsilon_{kl})$ ,  $\varepsilon_{kl} = \cos(V_k, V_l)_A$ , then  $K_1 \leq \rho(\mathcal{E}) \leq \max \sum_l \varepsilon_{kl}$ .  
 If  $\mathcal{E}_0 = (\varepsilon_{kl} : k, l \neq j)$ , then  $K_1 \leq 2(1 + \rho(\mathcal{E}_0))$ .
- **Theorem:** Let A1, A2 hold. Then  
 $\lambda_{\min}(G_A A) \geq 1/K_0$ ,  $\lambda_{\max}(G_A A) \leq K_1$ ,  $\text{cond}(G_A A) \leq K_0 K_1$   

$$\|I - G_S A\|_A = \|I - G_M A\|_A^2 \leq \left(1 - \frac{1}{K_0(1+K_1)^2}\right)^2.$$

Matsokin, Nepomnyaschikh 1985, Lions 1988, Dryja, Widlund 1987,1989, Bramble, Pasciak,  
 Wang, Xu 1991, Bjørstad, Mandel 1991

# Domain decomposition (DD)



$\Omega = \Omega_1 \cup \dots \cup \Omega_m$  aligned with  $\mathcal{T}_h$

overlap  $\delta = \text{dist}(\partial\Omega_k \cap \Omega_l, \partial\Omega_l \cap \Omega_k)$

subdomain size  $\mathbf{H} = \max \text{diam}(\Omega_k)$

$\delta = \beta\mathbf{H}$ ,  $\mathbf{H} \rightarrow 0 \Rightarrow \delta \rightarrow 0$

$\Omega_k \rightarrow V_k = \{v \in V_h : v = 0 \text{ on } \Omega \setminus \Omega_k\}$

- ex. partition of unity  $\theta_1, \dots, \theta_m$

$$\sum \theta_k = 1 \text{ on } \Omega$$

$$\theta_k \in C^\infty(R^d), \theta_k = 0 \text{ on } R^d \setminus \Omega_k \Rightarrow$$

$$\|\nabla \theta_k\|_{L^\infty(\Omega)} \leq c/\delta$$

- ex. interpolation  $\Pi_h : C(\Omega) \rightarrow V_h$

- $V_h = V_1 + \dots + V_m, v \in V_h \Rightarrow$

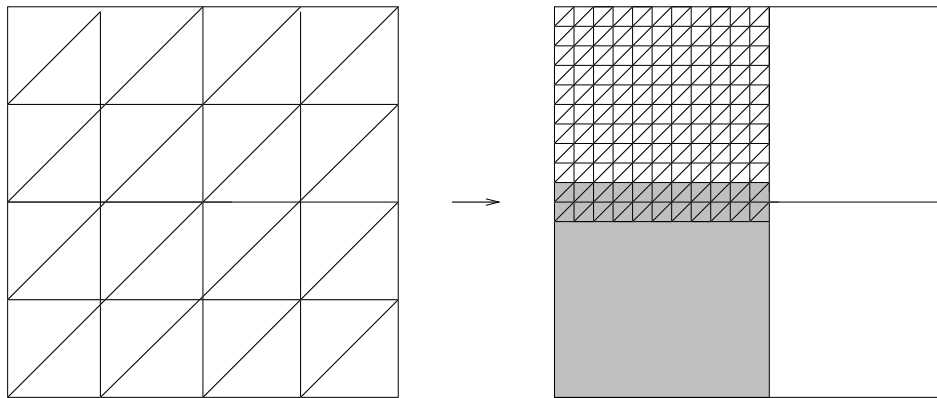
- $v = \sum_k v_k, v_k = \Pi_h(\theta_k v)$

- $K_0 = C(1 + \delta^{-2})$

- $K_1 \leq \rho(\mathcal{E})$  independent on m

$$\text{cond}(G_A A) \leq C(1 + \delta^{-2}), \text{ etc.}$$

## Two-level domain decomposition



$$\Omega = \Omega_1 \cup \dots \cup \Omega_m$$

$$V_h = V_0 + V_1 + \dots + V_m$$

$$V_0 = \{v \in V : v|_T \in P_1 \forall T \in \mathcal{T}_H\}$$

$$H \leq \mathbf{H}, \text{ qualitative analysis } H = \mathbf{H}$$

$$v \in V_h \Rightarrow \text{ex. } v_0 = Qv \in V_0$$

$$(1) |v_0|_{H^1(\Omega)} \leq c_1 |v|_{H^1(\Omega)}$$

$$(2) \|v - v_0\|_{L_2(\Omega)} \leq c_2 H |v|_{H^1(\Omega)}$$

Properties (1), (2) are valid if  $Q$  is  $L_2$ -orthogonal projection onto  $V_0$  (Bramble, Xu 1991)

$$\bullet v \in V \Rightarrow v = v_0 + v_1 + \dots + v_m$$

$$\bullet v_0 = Qv, \quad v_k = \Pi_h(\theta_k(v - v_0))$$

$$\bullet K_0 = C(1 + \delta^{-2}H^2)$$

$$\bullet K_1 \leq 2(1 + \rho(\mathcal{E}))$$

$$\text{cond}(G_A A) \leq C(1 + \delta^{-2}H^2), \text{ etc.}$$

# Overlapping DD preconditioners: A model problem

Model problem:

$$\Omega = \langle 0, 2 \rangle \times \langle 0, 3 \rangle$$

$$-\Delta u = f \text{ in } \Omega$$

$$u = 0 \text{ on } \partial\Omega$$

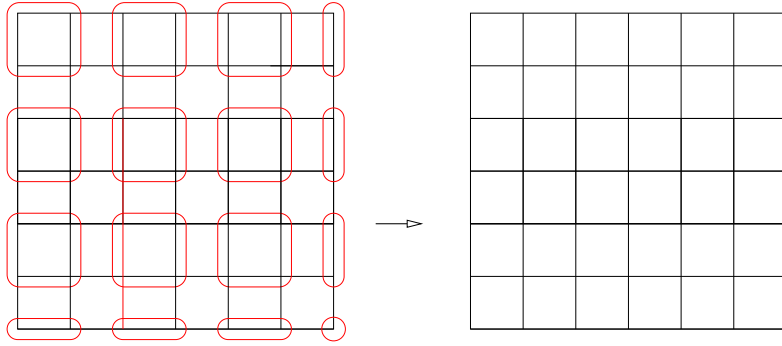
$$h=1/30, n=5100$$


 $\Omega_1$ 
 $\Omega_2$ 
 $\Omega_3$ 
 $\Omega_4$ 

	$\delta:$	$h$	$2h$	$3h$	$h$	$2h$	$3h$	$3h$	$3h$	$3h$
$m$	$H:$	—	—	—	$\frac{1}{15}$	$\frac{1}{15}$	$\frac{1}{15}$	$\frac{1}{15}$	$\frac{1}{5}$	$\frac{1}{4}$
2		19	14	11	9	8	8	8	9	9
4		26	19	16	9	8	8	8	10	10
6		32	23	19	9	9	9	9	10	10
8		36	26	22	9	9	9	9	10	10
10		40	28	23	10	9	9	9	10	10
12		43	31	25	9	9	9	9	10	10

Numbers of iterations for  $\varepsilon = 10^{-4}$  with additive  
precond., overlap  $\delta = h$  - Block Jacobi,  $\delta > h$  Schwarz

# Two-level domain decomposition with aggregations



$$\Omega = \Omega_1 \cup \dots \cup \Omega_m$$

$$V_h = \text{span} \{ \phi_i^h \}_{i=1}^n$$

$$\{1, \dots, n\} = J_1 \cup \dots \cup J_N, \text{ disjoint}$$

$$V_0 = \text{span} \{ \psi_i \}, \quad \psi_i = \sum_{j \in J_i} \phi_j^h$$

$$c_1 H^d \leq |\text{supp} \psi_i| \leq c_2 H^d, \quad H \leq \{ \mathbf{H}, kh \}$$

$$Q : V_h \rightarrow V_0, \quad Qv = \sum \alpha_k(v) \psi_k$$

$$\alpha_k = \frac{1}{\mu(\text{supp} \psi_k)} \int_{\text{supp} \psi_k} u(x) dx$$

Then:

- $\| Qv \|_{H^1(\Omega)}^2 \leq c \frac{H}{h} \| v \|_{H^1(\Omega)}^2$
- $\| v - Qv \|_{L_2(\Omega)} \leq CH \| v \|_{H^1(\Omega)}$

(weak approximation property)

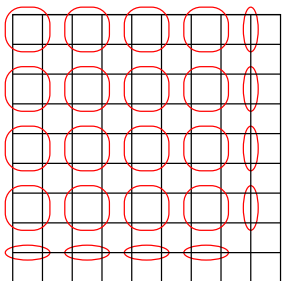
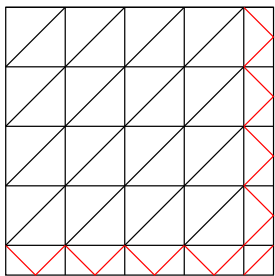
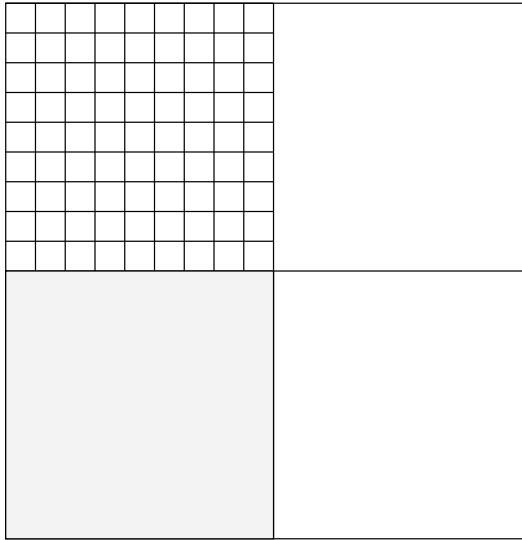
- $V_h = V_0 + V_1 + \dots + V_m$
- $v \in V \Rightarrow v = v_0 + v_1 + \dots + v_m$
- $v_0 = Qv, \quad v_k = \Pi_h(\theta_k(v - v_0))$

$$\bullet K_0 = C(1 + h^{-1}H + \delta^{-2}H^2)$$

$$\bullet K_1 \leq 2(1 + \rho(\mathcal{E}))$$

$$\text{cond}(G_A A) \leq K_0 K_1, \text{ etc.}$$

## Two-level DD with interfaces on coarse grid



$$\bar{\Omega} = \bar{\Omega}_1 \cup \dots \cup \bar{\Omega}_m, \quad \Omega_k \cap \Omega_l = \emptyset \text{ for } k \neq l$$

$$\Omega_k \rightarrow V_k : \varepsilon_{kl} = \cos(v_k, v_l)_A = 0 \text{ for } k \neq l$$

$$V_1 + \dots + V_m = W \neq V_h$$

$V_0 \dots$  by coarse triangulation or by aggregations with missing DOF  $\Rightarrow V_h = V_0 + W$ ,

$R_0 : V_h \rightarrow V_0$  interpolation to nodes of  $\mathcal{T}_H$  or selected nodes from  $J_k$

$$R_k : V_h \rightarrow V_k, \quad R_k v(x) = v(x) \text{ for nodes from } \Omega_k$$

$$v \in V_h \rightarrow v = v_0 + v_1 + \dots + v_m$$

$$v_0 = R_0 v, \quad v_k = R_k (v - v_0)$$

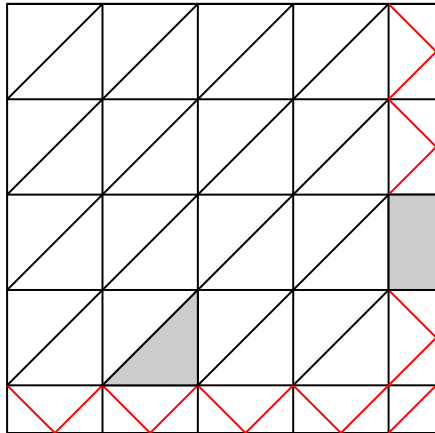
$$V_h = V_0 \oplus W_0, \quad W_0 \subset W, \quad W_0 = \sum R_k (I - R_0) V_h$$

$$\gamma = \cos(V_0, W_0)_A$$

$$K_0 = 1 / (1 - \gamma), \quad K_1 = 1 + \gamma$$



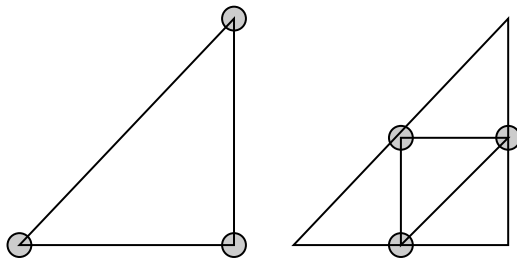
# CBS constant - boundary macroelements



$$\gamma = \sup \left\{ \frac{a(v,w)}{\sqrt{a(v,v)}\sqrt{a(w,w)}} : v \in V_0, w \in W_0, v, w \neq 0 \right\}$$

$$a(v, w) = \sum_E a_E(v, w) = \sum_E \int_E \langle D \nabla v, \nabla w \rangle dx$$

$\gamma = \max \gamma_E$ , E - inner/interface macroelement

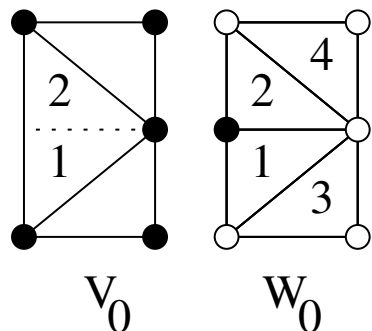


**Interface:**  $v \in V_0(E) \rightarrow \nabla v = (\delta_x, \delta_y)$  in  $T_1, T_2$ .

$w \in W_0(E) \rightarrow \nabla w = (d_x, d_y)$  in  $T_1$ ,

$\nabla w = (d_x, -d_y)$  in  $T_2$ ,  $\nabla w = 0$  in

$T_3, T_4$ . **If**  $D = \text{diag}[k_x, k_y]$ , **then**

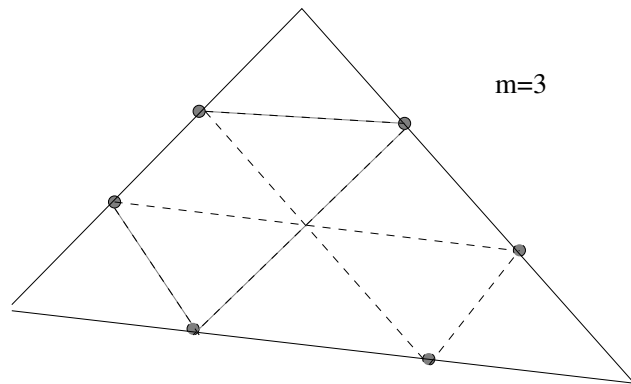


$$\left. \begin{aligned} a_E(v, w) &= 2k_x \delta_x d_x \Delta \\ a_E(v) &\leq 2k_x \delta_x^2 \Delta \\ a_E(w) &\leq 2(k_x d_x^2 + k_y d_x^2) \Delta \end{aligned} \right\} \Rightarrow \gamma_E \leq \sqrt{\frac{k_x}{k_x + k_y}}$$

# CBS constant - inner macroelements

Maitre, Musy 1981:

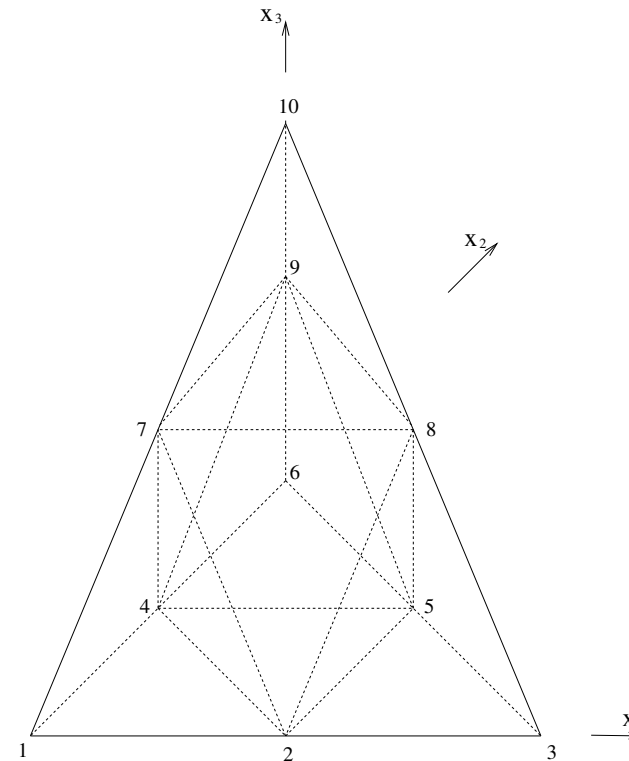
$\gamma = \sqrt{1/2}$  for isotropic Laplacian  
and rectangular elements,  $m_p = 2$ .



- $m_p$  fold refinement!
- anisotropic Laplacian
- general elasticity  $c_{i,j,k,l}$ !
- arbitrary element shape

$$\gamma \leq \sqrt{\frac{m_p^2 - 1}{m_p^2}} \quad \text{Axelsson, RB 2001}$$

$m_p^3$  tetrahedra



$$\gamma \leq \sqrt{1 - \frac{2}{m_p^4 + m_p^2}}$$

RB NLAA 2003

## Nonlinear and Nonsymmetric Preconditioners

- For  $A_k$  corresponding to subdomain, we successfully replace  $A_k$  by incomplete factorization.
- For coarse grid subproblem  $A_0$ , the same strategy is inefficient. More accurate approximation to  $A_0$  is necessary, e.g. by inner PCG iterations. Then the pseudoresidual is  $g = G(r)$  but  $G$  is *not linear*.
- It can be difficult to construct  $A_0$  and solve this subproblem in a time comparable with the solution of subdomain problems  $A_k$ . Then it may be advantageous to use hybrid additive-multiplicative algorithm:
  - coarse grid correction is computed individually,
  - residual is updated and subdomain problems are solved in parallel.

The hybrid algorithm gives  $g = G(r)$  with  $G$  linear (for linear subproblem solvers) but not symmetric. The symmetrization is relative expensive and not necessary, an alternative is *nonsymmetric preconditioner* with GPCG.

## Inner iterations GPCG[s] method

given  $u^0 \rightarrow r^0 = b - Au^0$ ,  $g^0 = G(r^0)$ ,  $v^0 = g^0$

**for**  $i = 0, 1, \dots$  **until**  $\| r^i \| \leq \varepsilon \| b \|$  **do**

$$w^i = Av^i$$

$$\alpha_i = \sigma_i / \langle w^i, v^i \rangle$$

$$u^{i+1} = u^i + \alpha_i v^i$$

$$r^{i+1} = r^i - \alpha_i w^i$$

$$v^{i+1} = g^{i+1} = G(r^{i+1})$$

**for**  $k = 1, \dots, \min\{i + 1, s\}$  **do**

$$\beta_{i+1}^{(k)} = (\langle g^{i+1}, r^{i+2-k} \rangle - \langle g^{i+1}, r^{i+1-k} \rangle) / \sigma_{i+1-k}$$

$$v^{i+1} = v^{i+1} + \beta_{i+1}^{(k)} v^{i+1-k}$$

**end**  $\Rightarrow \sigma_{i+1}$

**Extra:**

$s - 1 \times$  vector storage

$r^i, r^{i-1}, \dots, r^{i+2-s}(r^0)$

$s \times$  inner products

$\langle g^{i+1}, r^{i+1-k} \rangle,$

$k = 1, \dots, \min\{i + 1, s\}$

For  $s=1$ : one extra  
inner product

**end**

# DD preconditioners: A model problem

Model problem:

$$\Omega = \langle 0, 2 \rangle \times \langle 0, 3 \rangle$$

$$-\Delta u = f \text{ in } \Omega$$

$$u = 0 \text{ on } \partial\Omega$$

$$h=1/30, n=5100$$


 $\Omega_1$ 
 $\Omega_2$ 
 $\Omega_3$ 
 $\Omega_4$ 

Overlap 2h, #subdomains:	2	4	8	12	16	24
no coarse grid	12	16	22	23	31	37
c-grid H=3h, AP	7	7	8	8	7	8
c-grid H=3h, HP	6	6	6	6	6	6
aggreg. 2h, AP	12	13	15	16	17	17
aggreg. 2h, HP	9	10	11	11	11	11
interface & aggreg. 2h, AP	13	14	14	14	14	14
interface & aggreg. 2h, HP	7	8	7	8	8	8

Numbers of iterations for  $\varepsilon = 10^{-3}$ . AP=additive preconditioner, HP=hybrid preconditioner + GPCG[1]

## DD preconditioners: Another model problem

–  $\Delta u = 2\pi^2 \sin(\pi x) \sin(\pi y)$  in  $\Omega = \langle 0, 1 \rangle^2$  and  $u = 0$  on  $\partial\Omega$

Fine grid	$p=4$		$p=16$		$p=64$	
	$\mathbf{B}^{\text{as}}$	$\mathbf{B}^{\text{ms}}$	$\mathbf{B}^{\text{as}}$	$\mathbf{B}^{\text{ms}}$	$\mathbf{B}^{\text{as}}$	$\mathbf{B}^{\text{ms}}$
$h = 1/48$	37	20	110	58	377	193
$h = 1/96$	41	22	124	65	423	221
$h = 1/192$	45	24	138	72	478	248

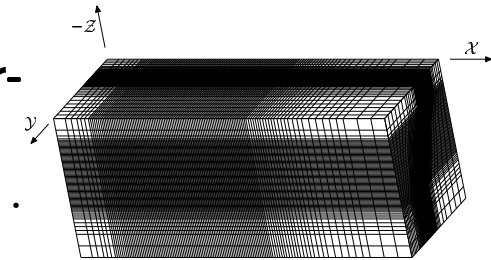
Table 1: overlap  $\beta = 1/6$ , accuracy  $\varepsilon = 10^{-4}$ , (see Knut-Andreas Lie, Uni. Oslo, 2001)

Fine grid	$p=4$		$p=16$		$p=64$	
	$\mathbf{B}_H^{\text{as}}$	$\mathbf{B}_H^{\text{ms}}$	$\mathbf{B}_H^{\text{as}}$	$\mathbf{B}_H^{\text{ms}}$	$\mathbf{B}_H^{\text{as}}$	$\mathbf{B}_H^{\text{ms}}$
$h = 1/48$	9	6	12	5	7	4
$h = 1/96$	10	6	13	5	9	5
$h = 1/192$	11	7	15	6	10	5

Table 2:  
two-level Schwarz preconditioners

## AS preconditioners - a geotechnical problem.

Overlap 2h. The sub-solvers: subdomain = incomplete factorization, the aggregated problem = inner PCG with  $\varepsilon_0 = 10^{-1}$ .



**Left:** #iterations. **Right:** times [s] on THEA.

# subd.	one-level method	two-level method	
		$3x3x3$	$6x6x6$
2	92	45	56
3	102	47	60
4	110	51	64
6	121	55	70
7	125	57	72
8	128	–	–

one-level method	two-level method	
	$3x3x3$	$6x6x6$
386	267	242
289	241	175
242	240	145
190	244	115
170	265	111
161	–	–

# Robustness of DD preconditioners: anisotropy

Model problem:

$$\Omega = \langle 0, 2 \rangle \times \langle 0, 3 \rangle$$

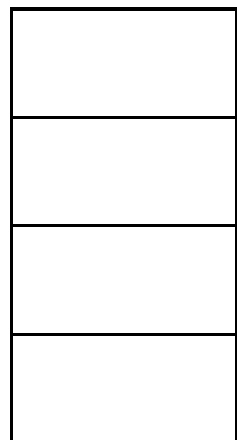
$$-k_x \frac{\partial^2 u}{\partial x^2} - k_y \frac{\partial^2 u}{\partial y^2} =$$

$f$  in  $\Omega$

$u = 0$  on  $\partial\Omega$

$h=1/30, n=5100$

$\delta$ 2h, #subd's:	4			8			16		
no coarse grid	16	$\frac{8}{27}$	$\frac{4}{26}$	22	$\frac{10}{37}$	$\frac{4}{38}$	31	$\frac{14}{52}$	$\frac{5}{61}$
c-grid H=3h, A	7	$\frac{7}{10}$	$\frac{7}{16}$	8	$\frac{7}{10}$	$\frac{7}{20}$	7	$\frac{7}{11}$	$\frac{7}{26}$
c-grid H=3h, H	6	$\frac{6}{8}$	$\frac{6}{13}$	6	$\frac{6}{8}$	$\frac{7}{16}$	6	$\frac{6}{10}$	$\frac{7}{22}$
aggreg. 2h, A	13	$\frac{10}{16}$	$\frac{8}{19}$	15	$\frac{12}{16}$	$\frac{8}{23}$	17	$\frac{13}{19}$	$\frac{9}{30}$
aggreg. 2h, H	10	$\frac{8}{11}$	$\frac{7}{15}$	11	$\frac{9}{12}$	$\frac{7}{19}$	11	$\frac{9}{13}$	$\frac{8}{25}$
if. & agg. 2h, A	14	$\frac{11}{14}$	$\frac{8}{22}$	14	$\frac{12}{16}$	$\frac{8}{30}$	14	$\frac{12}{19}$	$\frac{9}{40}$
if. & agg. 2h, H	8	$\frac{6}{8}$	$\frac{4}{14}$	7	$\frac{6}{9}$	$\frac{5}{16}$	8	$\frac{7}{10}$	$\frac{5}{22}$



$\Omega_1$

$\Omega_2$

$\Omega_3$

$\Omega_4$

Numbers of iterations for  $\epsilon = 10^{-3}$ . A=additive preconditioner, H=hybrid preconditioner + GPCG[1].

Columns: (1)= isotropy, (2)=  $\frac{k_x/k_y=10}{k_x/k_y=0.1}$ , (3)=  $\frac{k_x/k_y=100}{k_x/k_y=0.01}$ .



# DD preconditioners: A model parabolic problem

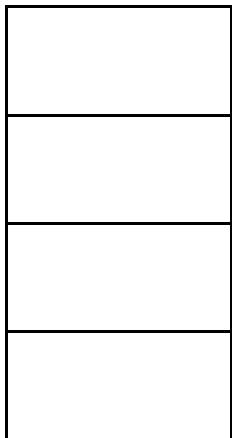
Model problem:

$$\Omega = \langle 0, 2 \rangle \times \langle 0, 3 \rangle$$

$$\frac{\partial u}{\partial t} - \Delta u = f$$

$$u = 0 \text{ on } \partial\Omega$$

$$h=1/30, n=5100$$



$\Omega_1$

$\Omega_2$

$\Omega_3$

$\Omega_4$

Overlap 2h, #subd's:	4			8			16		
no coarse grid	16	8	3	22	9	3	31	13	3
c-grid H=3h, AP	7	8	4	8	8	4	7	10	5
c-grid H=3h, HP	6	8	3	6	8		6	10	4
aggred. 2h, AP	13	8	5	15	8	5	17	10	5
aggred. 2h, HP	10	6		11	8		11	8	
interf. & aggr. 2h, AP	14	9	5	14	10	5	14	10	5
interf. & aggr. 2h, HP	8	5	3	7	5	3	8	5	3

Numbers of iterations for  $\varepsilon = 10^{-3}$ . AP=additive preconditioner, HP=hybrid preconditioner + GPCG[1].

Column 1: matrix  $K$ , columns 2,3: matrix  $M + \xi K$  with  $\xi = h, h^2$ , respectively.

## Overlapping DD methods

- :-) overlapping DD is an efficient tool for data decomposition, building preconditioners and construction of parallel algorithms with a small amount of communications
- :-) the efficiency can be substantially increased by adding a rough global problem, which can be defined by coarse grid or aggregations
- :-) there is a variety of DD methods: overlap, nested/non-nested coarse grid, simple/smoothed aggregations, RAS and RASHO, etc.
- :-) new CGR with interface: robustness w.r.t. coefficient jumps between macroelements, no communication between subdomain problems in preconditioning, efficient hybrid version, clear quantitative analysis
- :-) inexact sub-solvers, automatic partition/aggreg. (not PDE systems)
- :-( increase of subproblems due to overlap, decrease of efficiency due to anisotropy

## Final remarks

- overlapping DD - can be applied to other classes of problems (nonsymmetric, parabolic, saddle point etc), are easy to implement, potentially fully algebraical (black box),
- there are also other classes of nonoverlapping DD methods handling the interface through Schur complement or Lagrangian multipliers (Neumann-Neumann, FETI),
- there are many possible decompositions, which can be used for efficient parallel solvers: beside DD, also composite grid FEM, HB decomposition, AMLI, DiD etc.
- it is possible to combine different decompositions and different (additive/multiplicative) algorithms,
- the two-level concept can be developed to multi-level one for better balance, better efficiency and development of optimal solvers.

**Thank you for your attention.**