
On Adaptive-Multilevel BDDC

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1 Introduction

The BDDC method [2] is one of the most advanced methods of iterative substructuring. In the case of many substructures, solving the coarse problem exactly becomes a bottleneck. Since the coarse problem has the same structure as the original problem, it is straightforward to apply the method recursively to solve it only approximately. The two-level BDDC analysis has been extended into three-levels in a pioneering work in [14, 15], and into a general multilevel method in [11]. The methods for the adaptive selection of constraints for the two-level BDDC method have been studied in [9, 12]. Here we combine the two approaches into a new method preserving parallel scalability with increasing number of subdomains and excellent convergence properties.

The theoretical aspects of the design of the BDDC and a closely related FETI-DP on irregular subdomains in the plane has been studied in [6]. The authors in particular demonstrated that a proper choice of a certain scaling can significantly improve convergence of the methods. Our goal here is different. We consider only the standard stiffness scaling and we look for a space where the action of the BDDC preconditioner is defined. A combination of these two approaches, also with the proper choice of initial constraints [1], would be of independent interest. The presented algorithm has been recently extended into 3D in [13].

[4, 5] have recently successfully developed and extensively used several inexact solvers for the FETI-DP method, and [16] has extended the three-level BDDC methods to the saddle point problems.

All abstract spaces in this paper are finite dimensional. The dual space of a linear space U is denoted by U' , and $\langle \cdot, \cdot \rangle$ is the duality pairing.

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2 Abstract BDDC for a Model Problem

Let $\Omega \subset \mathbb{R}^2$ be a bounded polygonal domain, decomposed into N nonoverlapping polygonal substructures Ω_i , $i = 1, \dots, N$, which form a conforming triangulation. That is, if two substructures have a nonempty intersection, then the intersection is a vertex, or a whole edge. Substructure vertices will also be called corners. Let W_i be the space of Lagrangean $P1$ or $Q1$ finite element functions with characteristic mesh size h on Ω_i , and which are zero on the boundary $\partial\Omega$. Suppose that the nodes of the finite elements coincide on edges common to two substructures. Let

$$W = W_1 \times \dots \times W_N,$$

and let $U \subset W$ be the subspace of functions that are continuous across the substructure interfaces. We wish to solve the abstract linear problem

$$u \in U : a(u, v) = \langle f, v \rangle, \quad \forall v \in U, \quad (1)$$

for a given $f \in U'$, where a is a symmetric positive semidefinite bilinear form on some space $W \supset U$ and positive definite on U . The form $a(\cdot, \cdot)$ is called the energy inner product, the value of the quadratic form $a(u, u)$ is called the energy of u , and the norm $\|u\|_a = a(u, u)^{1/2}$ is called the energy norm. The operator $A : U \mapsto U'$ associated with a is defined by

$$a(u, v) = \langle Au, v \rangle, \quad \forall u, v \in U.$$

The values of functions from W at the corners and certain averages over the edges will be called the *coarse degrees of freedom*. Let $\widetilde{W} \subset W$ be the space of all functions such that the values of any coarse degrees of freedom have a common value for all relevant substructures and vanish on $\partial\Omega$. Define $U_I \subset U \subset W$ as the subspace of all functions that are zero on all substructure boundaries $\partial\Omega_i$, $\widetilde{W}_\Delta \subset W$ as the subspace of all function such that their coarse degrees of freedom vanish, \widetilde{W}_Π as the subspace of all functions such that their coarse degrees of freedom between adjacent substructures coincide and such that their energy is minimal. Then

$$\widetilde{W} = \widetilde{W}_\Delta \oplus \widetilde{W}_\Pi, \quad \widetilde{W}_\Delta \perp_a \widetilde{W}_\Pi. \quad (2)$$

The component of the BDDC preconditioner computed in \widetilde{W}_Π is called the coarse problem, cf. [11, Algorithm 11]. Functions that are a -orthogonal to U_I are called discrete harmonic. In [7, 9], the analysis was done in spaces of discrete harmonic functions after eliminating U_I ; this is not the case here, so \widetilde{W} does not consist of discrete harmonic functions only. Denote by P the energy orthogonal projection from U to U_I . Then $I - P$ is known as the projection onto the discrete harmonic functions. Finally, let E be a projection from \widetilde{W} onto U defined by taking some weighted average over the substructure interfaces.

Let us briefly describe the construction of the space \widetilde{W} using the coarse degrees of freedom. Suppose we are given a space X and a linear operator $C : W \mapsto X$ and define

$$\widetilde{W} = \{w \in W : C(I - E)w = 0\}. \quad (3)$$

The values Cw will be called the local coarse degrees of freedom. To represent their common values, i.e. the global coarse degrees of freedom, suppose there is a space U_c and linear operators

$$Q_P^T : U \rightarrow U_c \quad R_c : U_c \rightarrow X \quad R : U \rightarrow W,$$

such that

$$CR = R_c Q_P^T.$$

The operator Q_P^T selects global coarse degrees of freedom in U_c as linear combinations of global degrees of freedom; a global coarse degree of freedom is given by a row of Q_P . The operator R (resp. R_c) restricts a vector of global (coarse) degrees of freedom into a vector of local (coarse) degrees of freedom. See [9] for more details.

2.1 Multilevel BDDC

The substructuring components (the domains, spaces and operators) from the previous section will be denoted by an additional subscript $_1$, as Ω_1^i , $i = 1, \dots, N_1$, etc., and called level 1. We will call the coarse problem in $\widetilde{W}_{\Pi 1}$ the level 2 problem. It has the same finite element structure as the original problem (1) on level 1, so we have $U_2 = \widetilde{W}_{\Pi 1}$. Level 1 substructures are level 2 elements, level 1 coarse degrees of freedom are level 2 degrees of freedom. The shape functions on level 2 are the coarse basis functions in $\widetilde{W}_{\Pi 1}$, which are given by the conditions that the value of exactly one coarse degree of freedom is one and the others are zero, and that they are energy minimal in \widetilde{W}_1 . Note that the resulting shape functions on level 2 are in general discontinuous between level 2 elements. Level 2 elements are then agglomerated into nonoverlapping level 2 substructures, etc. Level ℓ elements are level $\ell - 1$ substructures, and the level ℓ substructures are agglomerates of level ℓ elements. Level ℓ substructures are denoted by Ω_ℓ^i , and they are assumed to form a quasiuniform conforming triangulation with characteristic substructure size H_ℓ . The degrees of freedom of level ℓ elements are given by level $\ell - 1$ coarse degrees of freedom, and shape functions on level ℓ are determined by minimization of energy on each level $\ell - 1$ substructure separately, so $U_\ell = \widetilde{W}_{\Pi, \ell - 1}$. The averaging operators on level ℓ , $E_\ell : \widetilde{W}_\ell \rightarrow U_\ell$, are defined by averaging of the values of level ℓ degrees of freedom between level ℓ substructures Ω_ℓ^i . The space $U_{I\ell}$ consists of functions in U_ℓ that are zero on the boundaries of all level ℓ substructures, and $P_\ell : U_\ell \rightarrow U_{I\ell}$ is the a -orthogonal projection in U_ℓ onto $U_{I\ell}$. For convenience, let Ω_0^i be the original finite elements, $H_0 = h$.

Algorithm 1 (Multilevel BDDC, [11], Algorithm 17) Define the preconditioner $r_1 \in U_1^i \mapsto u_1 \in U_1$ as follows:
for $\ell = 1, \dots, L - 1$,

Compute interior pre-correction on level ℓ ,

$$u_{I\ell} \in U_{I\ell} : a(u_{I\ell}, z_{I\ell}) = \langle r_\ell, z_{I\ell} \rangle, \quad \forall z_{I\ell} \in U_{I\ell}. \quad (4)$$

Get updated residual on level ℓ ,

$$r_{B\ell} \in U_\ell, \quad \langle r_{B\ell}, v_\ell \rangle = \langle r_\ell, v_\ell \rangle - a(u_{I\ell}, v_\ell), \quad \forall v_\ell \in U_\ell. \quad (5)$$

Find the substructure correction on level ℓ ,

$$w_{\Delta\ell} \in W_{\Delta\ell} : a(w_{\Delta\ell}, z_{\Delta\ell}) = \langle r_{B\ell}, E_\ell z_{\Delta\ell} \rangle, \quad \forall z_{\Delta\ell} \in W_{\Delta\ell}. \quad (6)$$

Formulate the coarse problem on level ℓ ,

$$w_{\Pi\ell} \in W_{\Pi\ell} : a(w_{\Pi\ell}, z_{\Pi\ell}) = \langle r_{B\ell}, E_\ell z_{\Pi\ell} \rangle, \quad \forall z_{\Pi\ell} \in W_{\Pi\ell}, \quad (7)$$

If $\ell = L - 1$, solve the coarse problem directly and set $u_L = w_{\Pi L - 1}$,
otherwise set up the right-hand side for level $\ell + 1$,

$$r_{\ell+1} \in \widetilde{W}'_{\Pi\ell}, \quad \langle r_{\ell+1}, z_{\ell+1} \rangle = \langle r_{B\ell}, E_\ell z_{\ell+1} \rangle, \quad \forall z_{\ell+1} \in \widetilde{W}_{\Pi\ell} = U_{\ell+1}, \quad (8)$$

end.

for $\ell = L - 1, \dots, 1$,

Average the approximate corrections on substructure interfaces on level ℓ ,

$$u_{B\ell} = E_\ell(w_{\Delta\ell} + u_{\ell+1}). \quad (9)$$

Compute the interior post-correction on level ℓ ,

$$v_{I\ell} \in U_{I\ell} : a(v_{I\ell}, z_{I\ell}) = a(u_{B\ell}, z_{I\ell}), \quad \forall z_{I\ell} \in U_{I\ell}. \quad (10)$$

Apply the combined corrections,

$$u_\ell = u_{I\ell} + u_{B\ell} - v_{I\ell}. \quad (11)$$

end.

A condition number bound follows, cf. [[11], Lemma 20].

Lemma 1. *If for some $\omega_\ell \geq 1$, for all $\ell = 1, \dots, L - 1$,*

$$\omega_\ell = \sup_{w_\ell \in (I - P_\ell)\widetilde{W}_\ell} J_\ell(w_\ell), \quad J_\ell(w_\ell) = \frac{\|(I - E_\ell)w_\ell\|_a^2}{\|w_\ell\|_a^2}, \quad (12)$$

then the multilevel BDDC preconditioner satisfies $\kappa \leq \omega = \prod_{k=1}^{L-1} \omega_k$.

3 Indicator of the Condition Number Bound

As in [9], we propose as an indicator of the condition number the maximum of the bounds from Lemma 1 computed by considering on each level ℓ only one pair of adjacent substructures s and t at a time:

$$\omega \approx \tilde{\omega} = \Pi_{\ell=1}^{L-1} \max_{st} \omega_{\ell}^{st}, \quad \omega_{\ell}^{st} = \sup_{w_{\ell}^{st} \in (I - P_{\ell}^{st}) \widetilde{W}_{\ell}^{st}} J_{\ell}^{st}(w_{\ell}^{st}), \quad (13)$$

where a pair of substructures is called adjacent if they share an edge, and the quantities with the superscript st are defined using the domain consisting of the level ℓ substructures s and t only.

The quantity $\tilde{\omega}$ is called an *indicator* of the condition number bound.

Let S_{ℓ}^{st} be the Schur complement operator associated with the bilinear form $a(\cdot, \cdot)$ on the space $(I - P_{\ell}^{st}) \widetilde{W}_{\ell}^{st}$. The next theorem is [8, Theorem 2] written in a way suitable for our purposes.

Theorem 1 *Let for $a > 0$, Π_{ℓ}^{st} be the orthogonal projection onto $(I - P_{\ell}^{st}) \widetilde{W}_{\ell}^{st}$, and $I - \overline{\Pi}_{\ell}^{st}$ the orthogonal projection onto*

$$\text{null} \left(\Pi_{\ell}^{st} S_{\ell}^{st} \Pi_{\ell}^{st} + a (I - \Pi_{\ell}^{st}) \right).$$

Then the stationary values $\omega_{\ell,1}^{st} \geq \omega_{\ell,2}^{st} \geq \dots$ and the corresponding stationary vectors $w_{\ell,k}^{st}$ of the Rayleigh quotient J_{ℓ}^{st} in (13) satisfy

$$X_{\ell}^{st} w_{\ell,k}^{st} = \omega_{\ell,k}^{st} Y_{\ell}^{st} w_{\ell,k}^{st} \quad (14)$$

with Y_{ℓ}^{st} positive definite, where

$$\begin{aligned} X_{\ell}^{st} &= \Pi_{\ell}^{st} (I - E_{\ell}^{st})^T S_{\ell}^{st} (I - E_{\ell}^{st}) \Pi_{\ell}^{st}, \\ Y_{\ell}^{st} &= \left(\overline{\Pi}_{\ell}^{st} (\Pi_{\ell}^{st} S_{\ell}^{st} \Pi_{\ell}^{st} + a (I - \Pi_{\ell}^{st})) \overline{\Pi}_{\ell}^{st} + a (I - \overline{\Pi}_{\ell}^{st}) \right). \end{aligned}$$

The eigenvalue problem (14) is obtained by projecting the gradient of the Rayleigh quotient $J_{\ell}^{st}(w_{\ell}^{st})$ onto the complement in $(I - P_{\ell}^{st}) \widetilde{W}_{\ell}^{st}$ of the subspace, where its denominator is zero, in two steps. Both projections Π_{ℓ}^{st} and $\overline{\Pi}_{\ell}^{st}$ are computed by matrix algebra, which is straightforward to implement numerically. The projection Π_{ℓ}^{st} projects onto $\text{null } C_{\ell}^{st} (I - E_{\ell}^{st})$, and $I - \overline{\Pi}_{\ell}^{st}$ projects onto a subspace of $\text{null } S_{\ell}^{st}$, which can be easily constructed computationally if a matrix Z_{ℓ}^{st} is given such that $\text{null } S_{\ell}^{st} \subset \text{range } Z_{\ell}^{st}$. For this purpose, the rigid body modes are often available directly or they can be computed from the geometry of the finite element mesh. For levels $\ell > 1$, we can use the matrix Z_{ℓ}^{st} with columns consisting of coarse basis functions, because the span of the coarse basis functions contains the rigid body modes. In this way, we can reduce (14) to a symmetric eigenvalue problem, which is easier and more efficient to solve numerically.

4 Optimal Coarse Degrees of Freedom

Writing $\widetilde{W}_\ell^{st} = \text{null } C_\ell^{st} (I - E_\ell^{st})$ suggests how to add coarse degrees of freedom to decrease the value of indicator $\widetilde{\omega}$. The following theorem is an analogy of [8, Theorem 3]. It follows immediately from the standard characterization of eigenvalues as minima and maxima of the Rayleigh quotient on subspaces spanned by eigenvectors, applied to (14).

Theorem 2. *Suppose $n_\ell^{st} \geq 0$ and the coarse dof selection matrix $C_\ell^{st} (I - E_\ell^{st})$ is augmented by the rows $w_{\ell,k}^{stT} (I - E_\ell^{st})^T S_\ell^{st} (I - E_\ell^{st})$, where $w_{\ell,k}^{st}$ are the eigenvectors from (14). Then $\omega_\ell^{st} = \omega_{\ell, n_\ell^{st}+1}^{st}$, and $\omega_\ell^{st} \geq \omega_{\ell, n_\ell^{st}+1}^{st}$ for any other augmentation by at most n_ℓ^{st} columns.*

In particular, if $\omega_{\ell, n_\ell^{st}+1}^{st} \leq \tau$ for all pairs of adjacent substructures s, t and for all levels $\ell = 1, \dots, L-1$, then $\widetilde{\omega} \leq \tau^{L-1}$.

Theorem 2 allows us to guarantee that the condition number indicator $\widetilde{\omega} \leq \tau^{L-1}$ for a given target value τ , by adding the smallest possible number of coarse degrees of freedom.

The primal coarse space selection mechanism that corresponds to this augmentation can be explained as follows. Let us write the augmentation as

$$c_{\ell,k}^{st} = [c_{\ell,k}^s \ c_{\ell,k}^t] = w_{\ell,k}^{stT} (I - E_\ell^{st})^T S_\ell^{st} (I - E_\ell^{st}),$$

where $c_{\ell,k}^s$ and $c_{\ell,k}^t$ are blocks corresponding to substructures s and t . It should be noted that the matrix E_ℓ^{st} is constructed for a pair of substructures s, t in such a way that, cf., e.g. [10, Eq. (7)],

$$B_{D,\ell}^{stT} B_\ell^{st} = I - E_\ell^{st},$$

where $B_{D,\ell}^{st}$ and B_ℓ^{st} are matrices known from the FETI-DP method. In particular, the entries of B_ℓ^{st} are $+1$ for substructure s and -1 for substructure t . This also relates our algorithm to the one from [8, 9]. Next, let us observe that, due to the application of $I - E_\ell^{st}$, for the two blocks of $c_{\ell,k}^{st}$ it holds that $c_{\ell,k}^s = -c_{\ell,k}^t$, i.e., for the two substructures the constraint weights have the same absolute values and opposite sign. Hence it is sufficient to consider only one of the two blocks, e.g., $c_{\ell,k}^s$. The augmentation of the global coarse degrees of freedom selection matrix $[Q_{P,\ell}, q_{k,\ell}]$ is constructed by adding a block of k columns computed as

$$q_{k,\ell} = R_\ell^{sT} c_k^{sT}.$$

Each column of $q_{k,\ell}$ defines a coarse degree of freedom associated with the interface of level ℓ substructures s and t . Because R_ℓ^s is a $0-1$ matrix, it means that columns in $q_{k,\ell}$ are formed by a scattering of the entries in c_k^{sT} .

5 Adaptive-Multilevel BDDC in 2D

We describe in more detail the implementation of the algorithm. It consists of two main steps: (i) setup, and (ii) the loop of preconditioned conjugate gradients with the Adaptive-Multilevel BDDC as a preconditioner. The setup was outlined in the previous section, and it can be summarized as follows:

Algorithm 2 *Adding of coarse degrees of freedom in order to guarantee that the condition number indicator $\tilde{\omega} \leq \tau^{L-1}$, for a given a target value τ :*

for levels $\ell = 1 : L - 1$,

Create substructures with roughly the same numbers of degrees of freedom, minimizing the number of “cuts” (use a graph partitioner, e.g., METIS 4.0 ([3]) with weights on both, vertices and edges).

Find a suitable set of initial constraints (corners in 2D), and set up the BDDC structures for the adaptive algorithm.

for all edges \mathcal{E}_ℓ **on level** ℓ ,

Compute the largest local eigenvalues and corresponding eigenvectors, until the first m^{st} is found such that $\lambda_{m^{st}}^{st} \leq \tau$, put $k = 1, \dots, m^{st}$.

Compute the constraint weights $c_k^{st} = [c_k^s \ c_k^t]$ as

$$c_k^{st} = w_k^{stT} \Pi_\ell^{Ist} (I - E_\ell^{st})^T S_\ell^{st} (I - E_\ell^{st}) \Pi_\ell^{Ist}, \quad (15)$$

where Π_ℓ^{Ist} is a projection constructed using the set of initial constraints.

Take one block, e.g., c_k^s and keep nonzero weights for the edge \mathcal{E}_ℓ .

Add to the global coarse dofs selection matrix $Q_{P,\ell}$ the k columns $q_{k,\ell}$ as

$$q_{k,\ell} = R_\ell^{sT} c_k^{sT}. \quad (16)$$

end.

Setup the BDDC structures for level ℓ and check size of the coarse problem: if small enough, call this the level L problem, factor it directly, and exit from the loop.

end.

We remark that the adaptive algorithm is significantly simpler and easier to implement compared to our previous algorithm from [8, 9]. The constraints in (15) are generated from the eigenvectors by the same function that evaluates the left hand side in (14). Then they are “torn” into two blocks, and entries of one of them, that correspond to a particular edge shared by substructures s and t on the level ℓ , are scattered into additional columns of the matrix $Q_{P,\ell}$.

The adaptive algorithm uses matrices and operators that are readily available in our implementation of the standard BDDC method (unlike in [12] this time with an explicit coarse space solve) with one exception: in order to satisfy the local partition of unity property, cf. [10, Eq. (9)],

$$E_\ell^{st} R_\ell^{st} = I,$$

we need to generate locally the weight matrices E_ℓ^{st} .

The substructures on higher levels are then treated as (coarse) elements with energy minimal basis functions. However, the number of added constraints is a-priori unknown. For this reason, the coarse elements must allow for variable number of nodes per element, and also for variable number of degrees of freedom per node. It is also essential to generate a sufficient number of corners as initial constraints, in particular to prevent rigid body motions between any pair of adjacent substructures. This topic has been addressed several times cf., e.g., a recent contribution in [1].

Finally, we remark that instead of performing interior pre-correction and post-correction on level $\ell = 1$, cf. Eqs. (4)–(5) and (10)–(11), we can benefit from reducing the problem to interfaces in the pre-processing step.

6 Numerical Examples and Conclusion

The adaptive-multilevel BDDC preconditioner was implemented in Matlab for the 2D linear elasticity problem (with $\lambda = 1$, and $\mu = 2$) on a square domain discretized by finite elements with 1,182,722 degrees of freedom. The domain was decomposed into 2,304 subdomains on the second level and into 9 subdomains on the third-level. Such decomposition leads to the coarsening ratio $H_\ell/H_{\ell-1} = 16$, with $\ell = 1, 2$. In order to test the adaptive selection of constraints, one single edge has been jagged on both decomposition levels, see Fig. 1. We have computed the eigenvalues and eigenvectors of (14) by setting up the matrices and using standard methods for the symmetric eigenvalue problem in Matlab, version 7.8.0.347 (R2009a).

In the first set of experiments, we have compared performance of the non-adaptive BDDC method with 2 and 3 decomposition levels. The results are presented in Tables 1 and 2. As expected from the theory the convergence of the algorithm deteriorates when additional levels are introduced.

In the next set of experiments, we have tested the adaptive algorithm for the two-level BDDC. The results are summarized in Table 6. The algorithm performs consistently with our previous formulation in [9]. The eigenvalues associated with edges between substructures clearly distinguish between the single problematic edge and the others (Table 3). Adding the coarse dofs created from the associated eigenvectors according to Theorem 2 decreases the value of the condition number indicator $\tilde{\omega}$ and improves convergence at the cost of increasing the number of coarse dofs.

Finally, we have tested the performance of the Adaptive-Multilevel BDDC for the model problem with three-level decomposition (Fig. 1). Because the number of coarse degrees of freedom depends on an a-priori chosen value of τ and the coarse

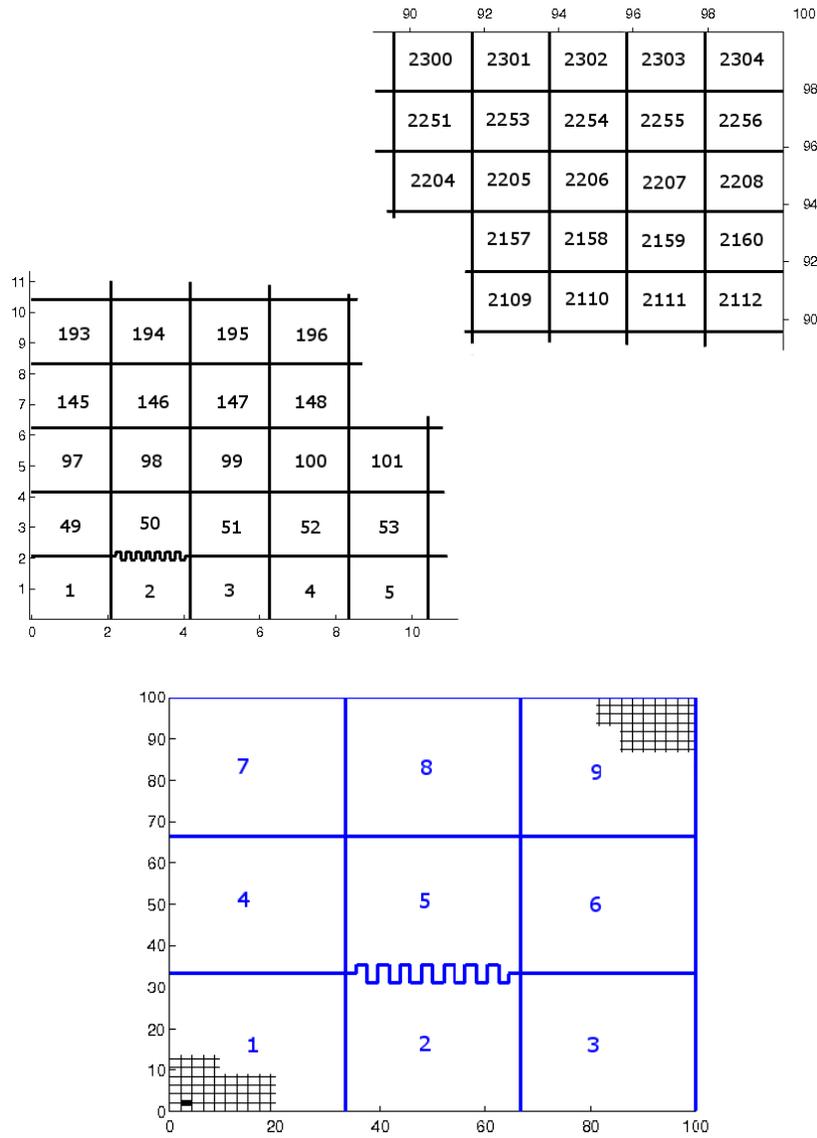


Fig. 1. The two-level decomposition into $48 \times 48 (= 2,304)$ subdomains (*top*), and the decomposition into 9 subdomains for the three-level method (*bottom*); the jagged edge from the lower decomposition level is indicated here by a *thick line*.

Table 1. Results for non-adaptive 2-level method. Constraints are corners, or corners and arithmetic averages over edges, denoted as c, c+e, resp. N_c is number of constraints, \mathcal{C} is size of the coarse problem related to size of a subdomain problem, κ is the condition number estimate, it is number of iterations (tol. 10^{-8}).

Constraint	N_c	\mathcal{C}	κ	it
c	4794	9.3	18.41	43
c+e	13818	26.9	18.43	32

Table 2. Results for non-adaptive 3-level method. Headings are as in Table 1.

Constraint	N_c	\mathcal{C}	κ	it
c	4794 + 24	1.0	67.5	74
c+e	13818 + 48	3.0	97.7	70

Table 3. Eigenvalues of the local problems for several pairs of subdomains s and t on the decomposition level $\ell = 1$ (the jagged edge is between subdomains 2 and 50).

s	t	$\lambda_{st,1}$	$\lambda_{st,2}$	$\lambda_{st,3}$	$\lambda_{st,4}$	$\lambda_{st,5}$	$\lambda_{st,6}$	$\lambda_{st,7}$	$\lambda_{st,8}$
1	2	3.8	2.4	1.4	1.3	1.2	1.1	1.1	1.1
1	49	6.0	3.5	2.7	1.4	1.3	1.1	1.1	1.1
2	3	5.4	2.6	1.6	1.3	1.2	1.1	1.1	1.1
2	50	24.3	18.4	18.3	16.7	16.7	14.7	13.5	13.1
3	4	3.4	2.4	1.4	1.3	1.1	1.1	1.1	1.1
3	51	7.4	4.6	3.7	1.7	1.4	1.3	1.2	1.1
49	50	12.6	5.1	4.3	1.9	1.6	1.3	1.2	1.2
50	51	8.7	4.8	3.9	1.8	1.5	1.3	1.2	1.2
50	98	7.5	4.6	3.7	1.7	1.4	1.3	1.2	1.1

basis functions on level ℓ become shape basis functions on level $\ell + 1$, the solutions of local eigenvalue problems will depend on τ as well. This fact is illustrated by Table 4 for $\tau = 2$, and by Table 5 for $\tau = 10$ (the local eigenvalues for $\tau = 3$ were essentially same as for $\tau = 2$). Comparing the values in these two tables, we see that lower values of τ result in worse conditioning of the local eigenvalue problems on higher decomposition level. This immediately gives rise to a conjecture that it might not be desirable to decrease the values of τ arbitrarily low in order to achieve a better convergence of the method. On the other hand, for the model problem, comparing the convergence results for the two-level method (Table 6) with the three-level method (Table 7), we see that with the adaptive constraints we were able to achieve nearly the same convergence properties for the two methods.

Table 4. Eigenvalues of the local problems for several pairs of subdomains s, t on level $\ell = 2$ with $\tau = 2$ (the jagged edge is between subdomains 2 and 5).

s	t	$\lambda_{st,1}$	$\lambda_{st,2}$	$\lambda_{st,3}$	$\lambda_{st,4}$	$\lambda_{st,5}$	$\lambda_{st,6}$	$\lambda_{st,7}$	$\lambda_{st,8}$
1	2	16.5	9.0	5.4	2.6	2.1	1.4	1.3	1.3
1	4	6.5	4.7	1.9	1.7	1.3	1.2	1.2	1.1
2	3	23.1	9.4	4.6	3.2	2.1	1.6	1.4	1.3
2	5	84.3	61.4	61.4	55.9	55.8	49.3	48.0	46.9
3	6	13.7	8.8	4.4	2.2	1.9	1.4	1.3	1.2
4	7	6.5	4.7	1.9	1.7	1.3	1.2	1.2	1.1
5	6	18.9	13.1	11.3	3.8	2.6	2.1	1.9	1.5
5	8	17.3	12.9	10.8	3.6	2.3	2.0	1.8	1.4
8	9	13.7	8.8	4.4	2.2	1.9	1.4	1.3	1.2

Table 5. Eigenvalues of the local problems for several pairs of subdomains s, t on level $\ell = 2$ with $\tau = 10$ (the jagged edge is between subdomains 2 and 5).

s	t	$\lambda_{st,1}$	$\lambda_{st,2}$	$\lambda_{st,3}$	$\lambda_{st,4}$	$\lambda_{st,5}$	$\lambda_{st,6}$	$\lambda_{st,7}$	$\lambda_{st,8}$
1	2	7.7	4.5	2.7	1.6	1.4	1.2	1.2	1.1
1	4	3.6	3.0	1.5	1.5	1.2	1.2	1.1	1.1
2	3	10.9	4.8	2.7	1.7	1.5	1.2	1.2	1.1
2	5	23.2	17.2	13.7	13.7	12.7	12.4	11.0	10.9
3	6	6.1	4.2	2.5	1.5	1.3	1.2	1.1	1.1
4	7	3.6	3.0	1.5	1.5	1.2	1.2	1.1	1.1
5	6	9.8	6.2	4.1	2.1	1.6	1.5	1.3	1.2
5	8	8.6	5.9	3.9	2.0	1.5	1.4	1.2	1.2
8	9	6.1	4.2	2.5	1.5	1.3	1.2	1.1	1.1

Table 6. Results for the adaptive 2-level method. Headings are same as in Table 1, and τ is the condition number target, $\tilde{\omega}$ is the condition number indicator.

τ	N_c	\mathcal{C}	$\tilde{\omega}$	κ	it
$\infty(=c)$	4,794	9.3	-	18.41	43
10	4,805	9.4	8.67	8.34	34
3	18,110	35.3	2.67	2.44	15
2	18,305	35.7	1.97	1.97	13

Table 7. Results for the adaptive 3-level method. Headings are as in Table 6, but the threshold τ is now used on each of the two decomposition levels and so $\tilde{\omega} \leq \tau^2$.

τ	N_c	\mathcal{C}	$\tilde{\omega}$	κ	it
$\infty(=c)$	4,794 + 24	1.0	-	67.5	74
10	4,805 + 34	1.0	84.97	37.42	60
3	18,110 + 93	3.9	7.88	3.11	19
2	18,305 + 117	4.0	3.84	2.28	15

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