ADAPTIVE FETI-DP AND BDDC METHODS WITH A GENERALIZED TRANSFORMATION OF BASIS FOR HETEROGENEOUS PROBLEMS*

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Abstract. In FETI-DP (Finite Element Tearing and Interconnecting) and BDDC (Balancing Domain Decomposition by Constraints) domain decomposition methods, the transformation-of-basis approach is used to improve the convergence by combining the local assembly with a change of basis. Suitable basis vectors can be constructed by the recently introduced adaptive coarse space approaches. The resulting FETI-DP and BDDC methods fulfill a condition number bound independent of heterogeneities in the problem. The adaptive method with a transformation of basis presented here builds on a recently introduced adaptive FETI-DP approach for elliptic problems in three dimensions and uses a coarse space constructed from solving small, local eigenvalue problems on closed faces and on a small number of edges. In contrast to our earlier work on adaptive FETI-DP, the coarse space correction is not implemented by using balancing (or deflation), which requires the use of an exact coarse space solver, but by using local transformations. This will make it simpler to extend the method to a large number of subdomains and large supercomputers. The recently established theory of a generalized transformation-of-basis approach yields a condition number estimate for the preconditioned operator that is independent of jumps of the coefficients across and inside subdomains when using the local adaptive constraints. It is shown that all results are also valid for BDDC. Numerical results are presented in three dimensions for FETI-DP and BDDC. We also provide a comparison of different scalings, i.e., deluxe, rho, stiffness, and multiplicity for our adaptive coarse space in 3D.

Key words. domain decomposition, FETI-DP, BDDC, coarse space, adaptive, eigenvalue problem, elliptic partial differential equations

AMS subject classifications. 65N30,65N55,65F08,65F10

1. Introduction. Domain decomposition methods [49, 59, 62] are widely-used iterative methods for the parallel solution of implicit finite element problems. In these methods, the finite element problem is decomposed into parallel local problems. Additionally, to obtain scalability in the number of iterations, a coarse space ensures the global transport of information.

Originally, the coarse space of FETI-DP and BDDC methods [11, 13, 17, 18, 19, 62] is formed by coupling the subdomains in a few primal variables, i.e., a partial finite element assembly is used to enforce continuity across subdomain boundaries in a few degrees of freedom such as corner nodes.

In three dimensions, this is not sufficient to obtain a good condition number bound, and, e.g., continuous edge averages have to be enforced throughout the iteration additionally. A transformation of basis to explicitly introduce the averages as new variables and a subsequent partial finite element assembly is a technique to implement such average constraints in FETI-DP and BDDC methods; see, e.g., [35, 40, 41, 45]. An alternative is the use of the edge characteristic functions as deflation vectors in a deflation (also known as projector preconditioning) or balancing approach; see, e.g., [38]. Indeed, for different problems, many different FETI-DP and BDDC coarse spaces have been developed over time.

It has been shown in [24, 38] for homogeneous problems that for any FETI-DP method with a transformation of basis there exists a FETI-DP method using deflation or balancing with essentially the same spectrum [38, Theorem 6.7]. In [30], we have shown that for heterogeneous problems, a generalized transformation-of-basis approach is necessary when

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a second coarse space is introduced to improve a domain decomposition method with an initial (*a priori*) coarse space and a corresponding scaling. Such *a posteriori* constraints, i.e., constraints that are computed after the choice of an initial coarse space and a scaling, are typical for domain decomposition methods with automatic or so-called adaptive coarse spaces. These approaches have been introduced for highly heterogeneous problems, notably in compressible or almost incompressible elasticity with coefficient jumps. For the resulting coarse spaces the condition number then is independent of the heterogeneities. Of course, this comes at the price of a larger coarse problem. It is therefore important that an automatic coarse space as space.

The adaptive (i.e., problem-specific) coarse spaces are constructed automatically, i.e., during the computation typically by (approximately) solving local eigenvalue problems. Since these computations are performed after the introduction of the scaling, the constraints are scaling dependent, and the enforcement of these constraints has to be carried out by a *generalization* of the transformation-of-basis approach as introduced in [30]. A subtle detail of this new method, which is relevant for the theory, is that we observe nontrivial values at coarse nodes of the a posteriori or second coarse space when the FETI-DP or BDDC operators P_D or $I - E_D$ are applied to functions from the new constraint space; cf. [30].

Recently, such domain decomposition methods with adaptive coarse spaces have gained considerable interest. In [6, 7], adaptive Neumann-Neumann methods were considered. Adaptive Schwarz methods were considered later by [15, 16, 20, 21, 60], where one of these methods was generalized to adaptive BDD and FETI; see [61].

For FETI-DP and BDDC, adaptive methods were introduced or considered in [5, 9, 12, 26, 27, 28, 31, 33, 34, 47, 48, 52, 53]. In adaptive FETI-DP and BDDC methods, typically (approximate) eigenvectors from local eigenvalue problems on, e.g., faces or edges, are integrated into the coarse space. A deflation or balancing approach, as in [38], can then be used to implement the coarse space correction using the eigenvectors as deflation vectors. This approach can be used with different automatic coarse space strategies and was used in [33] for a comparison of adaptive approaches in two dimensions. As shown by the authors in [28], deflation can also be used in three dimensions to create an adaptive FETI-DP method with a condition number bound for heterogeneous problems that is independent of the jumps of the coefficients. The approach in [28] thus sets the successful heuristics introduced in [47], using only face eigenvalue problems, on a firm theoretical ground by adding a small number of edge eigenvalue problems.

However, in the deflation and balancing approach as used in [28], the coarse correction has to be computed quite exactly to ensure convergence to the correct solution; see [38]. When the coarse problem becomes large, an exact solution of the coarse problem can be inefficient.

Using a transformation-of-basis (or change-of-variables) approach combined with partial assembly can be an efficient alternative; see also, e.g., [35, 40]. Building on this, highly scalable inexact FETI-DP methods have been constructed as in [32, 36] that scale to hundreds of thousands of processor cores using a single cycle of an AMG method for the coarse problem.

In this current paper, we show that for adaptive coarse spaces, the generalized transformation-of-basis approach [30] can be used instead of deflation as in [28]. As in [28], we obtain a condition number bound independent of heterogeneities. This paper thus combines the generalized transformation-of-basis approach for FETI-DP with the adaptive coarse space [28] for problems in three dimensions and gives a condition number bound for this new approach. We will argue that the explicit condition number of the adaptive FETI-DP or BDDC method with the generalized transformation-of-basis approach is always smaller or equal to the condition number of FETI-DP with balancing or deflation.

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Adaptive coarse spaces for BDDC methods, e.g., [5, 9, 52, 53], are related to our approach, and we will also give a condition number bound and present numerical results for the coarse space of [28] with BDDC using the generalized transformation-of-basis approach. This extends [28], where only FETI-DP was considered because deflation for FETI-DP and BDDC are not equivalent [38].

The remainder of the paper is organized as follows. In Section 2 we introduce the model problem considered and the geometry used. In Section 3, we shortly rephrase the standard FETI-DP and BDDC methods. In Section 4, we introduce the generalized eigenvalue problems already considered in [28] that can result in adaptive constraints for FETI-DP and BDDC enforceable by the generalized transformation of basis as presented in [30]. This section will be concluded by a condition number bound for both the adaptive FETI-DP and the adaptive BDDC method. A vast overview on the numerical behavior of the new algorithm and of heuristic modifications thereof (cf. [28]) will be presented in Section 5. In Section 6, we provide some preliminary results using a first parallel implementation of our adaptive FETI-DP methods. Finally, we draw a conclusion in Section 7.

2. Model problem and geometry. Given a bounded polyhedral domain $\Omega \subset \mathbb{R}^3$ with $\partial \Omega_D \subset \partial \Omega$ a closed subset of positive surface measure and $\partial \Omega_N := \partial \Omega \setminus \partial \Omega_D$, we consider the weak formulation of (compressible) linear elasticity. We introduce the space $H_0^1(\Omega, \partial \Omega_D)^3 := \{v \in H^1(\Omega)^3 : v = 0 \text{ on } \partial \Omega_D\}$ as well as the Young Modulus E(x) > 0 and the Poisson ratio $0 < \nu(x) < 1/2$ for all $x \in \Omega$, and we are interested in finding $u \in H_0^1(\Omega, \partial \Omega_D)^3$ such that

$$a(u,v) = F(v) \qquad \forall v \in H_0^1(\Omega, \partial \Omega_D)^3,$$

where

$$\begin{split} a(u,v) &:= \int_{\Omega} 2\mu \varepsilon(u) : \varepsilon(v) \mathrm{d}x + \int_{\Omega} \lambda \mathrm{div}(u) \mathrm{div}(v) \mathrm{d}x, \\ F(v) &:= \int_{\Omega} f \cdot v \mathrm{d}x + \int_{\partial \Omega_N} g \cdot v \mathrm{d}s. \end{split}$$

Here, λ and μ are the Lamé constants, and

$$\varepsilon(u):\varepsilon(v)=\mathrm{tr}(\varepsilon(u)^T\varepsilon(v))\qquad \text{with}\quad \varepsilon(u)=\frac{1}{2}(\nabla u+\nabla u^T).$$

The functions $f : \Omega \to \mathbb{R}^3$ and $g : \partial \Omega_N \to \mathbb{R}^3$ denote volume and surface force, respectively. Let us note that our theory also holds for the case of almost incompressible linear elasticity; see [28] and the numerical results therein.

The domain Ω will be decomposed into N nonoverlapping, open subdomains Ω_i , i = 1, ..., N. The interface Γ is defined as the union of the subdomain boundaries without taking $\partial\Omega$ into account, i.e., $\Gamma := \{x \in \overline{\Omega}_i \cap \overline{\Omega}_j; i \neq j\}$. Then, the subdomains are each triangulated and then discretized by the finite element method with matching nodes on the interface. For simplicity, we use piecewise linear conforming finite elements.

In three dimensions, the interface can be decomposed into vertices, edges, and faces; for a detailed definition of these sets, cf. [40]. Edges and faces are defined as open sets. In a regular decomposition, vertices are the endpoints of edges. In general, this also applies to irregular decompositions. However, in these cases, edges with less than two vertices or vertices not being the endpoint of an edge can appear. A face between two arbitrary subdomains Ω_i and Ω_j will be denoted by \mathcal{F}^{ij} , while we denote edges between Ω_i , Ω_j , Ω_l and possibly more subdomains by \mathcal{E}^{il} . Vertices of Ω_i that belong to multiple subdomains are denoted by \mathcal{V}^{ik} .

Finally, we introduce the finite element spaces used for the FETI-DP and BDDC methods. For i = 1, ..., N, by $W^h(\Omega_i)$ we denote the local finite element space on Ω_i . The local trace space $W_i := W^h(\Gamma_i)$ is defined on $\Gamma_i := \overline{\Omega_i} \cap \Gamma$. We also introduce the global product space $W := \prod_{i=1}^N W_i$ and denote the space of functions that are continuous on the interface by $\widehat{W} \subset W$.

3. FETI-DP and BDDC methods with a transformation of basis.

3.1. Standard FETI-DP and BDDC. The FETI-DP method was originally introduced in [17, 18]. To obtain the BDDC method, different authors proposed similar ways; see [11, 13, 19].

We partition the degrees of freedom on the subdomains $\Omega_i \subset \Omega$, i = 1, ..., N, into interior, dual, and primal degrees of freedom, denoted by I, Δ' , and Π' , respectively. For subdomains Ω_i that share a part of the global boundary, i.e., $\partial \Omega_i \cap \partial \Omega \neq \emptyset$, their degrees of freedom on $\partial \Omega$ are assigned to the interior index set I. The assembled local stiffness matrices and the local load vectors of the subdomains Ω_i , i = 1, ..., N, then can be partitioned correspondingly. We need two different kinds of groupings of the previously introduced index sets I, Δ' , and Π' , i.e., on the one hand, we group interior and dual variables denoted by the index $(\cdot)_{B'}$, and, on the other hand, we group dual and primal indices denoted by the index $(\cdot)_{\Gamma}$. Then, there are multiple partitionings of the local stiffness matrices,

$$K^{(i)} =: \begin{bmatrix} K_{II}^{(i)} & K_{\Delta'I}^{(i)T} & K_{\Pi'I}^{(i)T} \\ K_{\Delta'I}^{(i)} & K_{\Delta'\Delta'}^{(i)} & K_{\Pi'\Delta'}^{(i)T} \\ K_{\Pi'I}^{(i)} & K_{\Pi'\Delta'}^{(i)} & K_{\Pi'\Pi'}^{(i)} \end{bmatrix} \begin{cases} =: \begin{bmatrix} K_{B'B'}^{(i)} & K_{\Pi'B'}^{(i)T} \\ K_{\Pi'B'}^{(i)} & K_{\Pi'\Pi'}^{(i)} \end{bmatrix} \\ =: \begin{bmatrix} K_{II}^{(i)} & K_{I\Gamma}^{(i)T} \\ K_{II}^{(i)} & K_{\Gamma\Gamma}^{(i)} \end{bmatrix}. \end{cases}$$

In the same manner, the load vectors $f^{(i)}$ and the displacements $u^{(i)}$, i = 1, ..., N, can be partitioned. We then introduce the block diagonal matrices

$$K_{II} := \operatorname{diag}_{i=1}^{N} K_{II}^{(i)}, \qquad K_{B'B'} := \operatorname{diag}_{i=1}^{N} K_{B'B'}^{(i)}, \qquad K_{\Gamma\Gamma} := \operatorname{diag}_{i=1}^{N} K_{\Gamma\Gamma}^{(i)}$$

as well as the corresponding offdiagonal block $K_{\Gamma I}$. Again, the global right-hand side f and the global displacement vector u can be partitioned accordingly.

The inter-subdomain assembly operator $R_{\Pi'}^T := \left[R_{\Pi'}^{(1)T}, \ldots, R_{\Pi'}^{(N)T} \right]$, which performs the partial finite element assembly in the primal variables $u_{\Pi'}^{(i)}$, is central to FETI-DP and BDDC. For BDDC, we also need $R_{\Delta'}^T := \left[R_{\Delta'}^{(1)T}, \ldots, R_{\Delta'}^{(N)T} \right]$, which performs the finite element assembly in the dual variables $u_{\Delta'}^{(i)}$. For FETI-DP, instead of $R_{\Delta'}^T$, we need a signed Boolean jump operator B and its restriction to the interface $B_{\Gamma} = [B_{\Gamma}^{(1)}, \ldots, B_{\Gamma}^{(N)}]$, which has one +1 and one -1 per row such that $B_{\Gamma}u_{\Gamma} = 0$ if and only if u_{Γ} is continuous on the interface Γ .

Then, the FETI-DP master system is given by

$$\begin{bmatrix} K_{B'B'} & K_{\Pi'B'}^T & B_{B'}^T \\ \widetilde{K}_{\Pi'B'} & \widetilde{K}_{\Pi'\Pi'} & 0 \\ B_{B'} & 0 & 0 \end{bmatrix} \begin{bmatrix} u_{B'} \\ \widetilde{u}_{\Pi'} \\ \lambda \end{bmatrix} = \begin{bmatrix} \widetilde{f} \\ 0 \\ 0 \end{bmatrix},$$

where we performed the partial assembly in the primal variables to couple the subdomain problems, i.e.,

$$\widetilde{K}_{\Pi'\Pi'} = \sum_{i=1}^{N} R_{\Pi'}^{(i)T} K_{\Pi'\Pi'}^{(i)} R_{\Pi'}^{(i)},$$

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$$\widetilde{K}_{\Pi'B'} = \left[R_{\Pi'}^{(1)T} K_{\Pi'B'}^{(1)}, \dots, R_{\Pi'}^{(N)T} K_{\Pi'B'}^{(N)} \right],$$
$$\widetilde{f} = \left[f_{B'}^T, (\sum_{i=1}^N R_{\Pi'}^{(i)T} f_{\Pi'}^{(i)})^T \right]^T,$$

and

$$B_{B'} = \left[B_{B'}^{(1)}, \dots, B_{B'}^{(N)} \right],$$

the part of B where the columns corresponding to primal variables were removed.

By proper block Gaussian elimination, we obtain the (unpreconditioned) standard FETI-DP system $F\lambda = d$, where

$$F\lambda := \left(B_{B'}K_{B'B'}^{-1}B_{B'}^{T} + B_{B'}K_{B'B'}^{-1}\widetilde{K}_{\Pi'B}^{T}\widetilde{S}_{\Pi'\Pi'}^{-1}\widetilde{K}_{\Pi'B'}K_{B'B'}^{-1}B_{B'}^{T}\right)\lambda$$

$$(3.1) \qquad = \left(B_{\Gamma}\widetilde{S}^{-1}B_{\Gamma}^{T}\right)\lambda,$$

$$d := B_{B'}K_{B'B'}^{-1}f_{B'} + B_{B'}K_{B'B'}^{-1}\widetilde{K}_{\Pi'B'}^{T}\widetilde{S}_{\Pi'\Pi'}^{-1}\left(\left(\sum_{i=1}^{N}R_{\Pi'}^{(i)T}f_{\Pi'}^{(i)}\right) - \widetilde{K}_{\Pi'B'}K_{B'B'}^{-1}f_{B'}\right).$$

The Schur complement $\widetilde{S}_{\Pi'\Pi'} := \widetilde{K}_{\Pi'\Pi'} - \widetilde{K}_{\Pi'B'} (K_{B'B'})^{-1} \widetilde{K}_{\Pi'B'}^{T}$ represents the initial (a priori) coarse space. The Schur complement \widetilde{S} is obtained from the local contributions $S^{(i)} := \operatorname{diag}_{i=1}^{N} (K_{\Gamma\Gamma}^{(i)} - K_{\Gamma I}^{(i)} (K_{II}^{(i)})^{-1} K_{\Gamma I}^{(i)T})$ assembled in the primal variables Π' .

$$\begin{split} S^{(i)} &:= \operatorname{diag}_{i=1}^{N} \left(K_{\Gamma\Gamma}^{(i)} - K_{\Gamma I}^{(i)} (K_{II}^{(i)})^{-1} K_{\Gamma I}^{(i)T} \right) \text{ assembled in the primal variables } \Pi'. \\ & \text{To define the BDDC method, it is convenient to introduce a modified ordering, i.e.,} \\ & \left[u_{\Delta'}^{(1)}, \ldots, u_{\Delta'}^{(N)}, u_{\Pi'}^{(1)}, \ldots, u_{\Pi'}^{(N)} \right] \text{ instead of } \left[u_{\Delta'}^{(1)}, u_{\Pi'}^{(1)}, \ldots, u_{\Delta'}^{(N)}, u_{\Pi'}^{(N)} \right]. \\ & \text{We use the block diagonal matrix } \mathcal{K}_{\Gamma\Gamma} \text{ and } \mathcal{K}_{\Gamma I} \text{ to define the Schur complement on the interface } \Gamma, \end{split}$$

$$\mathcal{S}_{\Gamma\Gamma} := \mathcal{K}_{\Gamma\Gamma} - \mathcal{K}_{\Gamma I} K_{II}^{-1} \mathcal{K}_{\Gamma I}^{T}.$$

The right-hand side is obtained by the corresponding elimination of the interior degrees of freedom from f and is denoted by $[g_{\Delta'}^T, g_{\Pi'}^T]^T$. The (unpreconditioned) BDDC system for $u_{\Gamma}^T = [u_{\Delta'}^T, u_{\Pi'}^T]^T$ with right-hand side $g^T = [g_{\Delta'}^T, R_{\Delta'}, g_{\Pi'}^T, R_{\Pi'}]^T$ writes

$$\mathcal{S}u_{\Gamma} := \begin{bmatrix} R_{\Delta'}^T & 0\\ 0 & I_{\Pi'} \end{bmatrix} \begin{bmatrix} I_{\Delta'} & 0\\ 0 & R_{\Pi'}^T \end{bmatrix} \mathcal{S}_{\Gamma\Gamma} \begin{bmatrix} I_{\Delta'} & 0\\ 0 & R_{\Pi'} \end{bmatrix} \begin{bmatrix} R_{\Delta'} & 0\\ 0 & I_{\Pi'} \end{bmatrix} u_{\Gamma} = g.$$

For heterogeneous problems, the use of an appropriate scaling is important. To be effective, this scaling has to depend on the coefficient. The scaling is of equal importance in adaptive FETI-DP and BDDC methods since, for a bad scaling, the resulting automatic coarse space can be very large [33].

In this paper, we will consider four different scalings. First, we introduce the standard ρ -scaling; see, e.g., [37, 39, 54, 57, 62]. For $x \in \Gamma_i$, we define $\mathcal{N}_x := \{i \mid x \in \partial\Omega_i\}$. The coefficient evaluation is given by $\hat{\rho}_i(x) := \sup_{x \in \operatorname{supp}(\varphi_x) \cap \Omega_i} \rho(x)$, where φ_x is the nodal finite element function at x and $\operatorname{supp}(\varphi_x)$ its support. Let Ω_j and Ω_i share either a face or an edge, and let $x \in \partial\Omega_i \cap \partial\Omega_j$. The corresponding nontrivial row of $B^{(j)}$ is then multiplied by the scaling $\delta_i^{\dagger}(x) := \hat{\rho}_i(x) / \left(\sum_{k \in \mathcal{N}_x} \hat{\rho}_k(x)\right)$ and vice versa. By this, we obtain the local scaling $D^{(j)}$. In BDDC the degrees of freedom on $\partial\Omega_i$ are scaled by $\delta_i^{\dagger}(x)$. This defines the BDDC scaling $D_u^{(i)}$.

Second, we introduce the stiffness- (or K-, or super-lumped-) scaling [55], which is a heuristic approximation of the ρ -scaling. There, we replace the coefficient $\hat{\rho}_i(x)$ by the corresponding diagonal element of the local stiffness matrix $K^{(i)}$.

Third, we introduce multiplicity-scaling [17, 55], where the inverse of the multiplicity is used as a scaling. We obtain multiplicity-scaling from ρ -scaling by setting the coefficient to one.

Last, we consider deluxe-scaling, which was introduced recently; see, e.g, [4, 8, 10, 12, 14]. Deluxe-scaling is computationally expensive. Let us define the Schur complement of the stiffness matrix on the interface Γ , i.e.,

$$S_{\Gamma\Gamma} := \operatorname{diag}_{i=1}^N S^{(i)}.$$

Then, consider either a face \mathcal{F}^{ij} shared by the two subdomains Ω_i and Ω_j or an edge \mathcal{E}^{ik} shared by the subdomains Ω_i , Ω_j , Ω_k . Multiplicities greater than three can be handled analogously. Define for $l \in \{i, j\}$ the matrix $S_{\mathcal{F}^{ij},0}^{(l)}$ as the restriction of $S^{(l)}$ to the (open) face \mathcal{F}^{ij} . For a face \mathcal{F}^{ij} , in deluxe-scaling the nontrivial rows of $B^{(j)}$ corresponding to the Lagrange multipliers on this face are multiplied by $D_{u,\mathcal{F}^{ij}}^{(i)T} = \left((S_{\mathcal{F}^{ij},0}^{(i)} + S_{\mathcal{F}^{ij},0}^{(j)})^{-1} S_{\mathcal{F}^{ij},0}^{(i)} \right)^T$ if the orientation of the constraints in B is chosen consistently. Otherwise, certain entries of $D_{u,\mathcal{F}^{ij}}^{(i)}$ have to be scaled by -1. For $l \in \{i,j,k\}$, define $S_{\mathcal{E}^{ik},0}^{(l)}$ as the restriction of $S^{(l)}$ to the (open) edge \mathcal{E}^{ik} . For an edge \mathcal{E}^{ik} , in deluxe-scaling the nontrivial rows of $B^{(j)}$ corresponding to the Lagrange multipliers coupling Ω_i and Ω_j on this edge are multiplied by $D_{u,\mathcal{F}^{ij}}^{(i)T} = \left((S_{\mathcal{E}^{ik},0}^{(i)} + S_{\mathcal{E}^{ik},0}^{(j)} + S_{\mathcal{E}^{ik},0}^{(k)})^{-1} S_{\mathcal{E}^{ik},0}^{(i)} \right)^T$. Again, a consistent orientation of the Lagrange multipliers is assumed.

Choosing either of the scalings we can define the standard FETI-DP Dirichlet preconditioner (see [62])

(3.2)
$$M_D^{-1} := B_{D,\Gamma} R_{\Gamma}^T S_{\Gamma\Gamma} R_{\Gamma} B_{D,\Gamma}^T = B_{D,\Gamma} \widetilde{S} B_{D,\Gamma}^T.$$

The standard BDDC preconditioner can be written

(3.3)
$$M_{\text{BDDC}}^{-1} := \begin{bmatrix} R_{\Delta',D_u}^T R_{\Gamma} & 0\\ 0 & I_{\Pi'} \end{bmatrix} \begin{bmatrix} K_{B'B'} & \tilde{K}_{\Pi'B'}^T\\ \tilde{K}_{\Pi'B'} & \tilde{K}_{\Pi'\Pi'} \end{bmatrix}^{-1} \begin{bmatrix} R_{\Gamma}^T R_{\Delta',D_u} & 0\\ 0 & I_{\Pi'} \end{bmatrix},$$

where $B_{D,\Gamma}$ and $R_{\Delta',D_{\mu}}$ are the scaled variants B_{Γ} and $R_{\Delta'}$.

3.2. A standard transformation of basis for FETI-DP and BDDC. We briefly recall the standard transformation of basis (see, e.g., [35, 38, 40, 41, 45]), which, in the adaptive context, will be replaced by the generalized transformation of basis as described in [30].

Consider an edge \mathcal{E} shared by Ω_i , Ω_j , and some other subdomains. It is intended to enforce the Krylov iterates u to fulfill a constraint of the form

(3.4)
$$c^{T}\left(u_{\mathcal{E}}^{(i)} - u_{\mathcal{E}}^{(j)}\right) = 0 \quad \Leftrightarrow \quad c^{T}u_{\mathcal{E}}^{(i)} = c^{T}u_{\mathcal{E}}^{(j)}$$

for a normalized constraint vector c. For c different from any Euclidean basis vector, this corresponds to the use of a non-nodal basis function in the transformation of basis.

We then define a (square) transformation matrix $T_{\mathcal{E}}^{(l)} = [c, C^{(l)\perp}], l \in \{i, j\}$, where $C^{(l)\perp}$ is constructed such that the transformation is orthogonal. We then define the transformation matrix $T^{(l)}$ by extending $T_{\mathcal{E}}^{(l)}$ by the identity matrix outside of the edge. The transformed

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variables $\overline{u}^{(l)}$, stiffness matrices $\overline{K}^{(l)}$, and load vectors $\overline{f}^{(l)}$ on Ω_l then are

$$\overline{K}^{(l)} = T^{(l)T} K^{(l)} T^{(l)}, \qquad \overline{u}^{(l)} = T^{(l)T} u^{(l)}, \qquad \overline{f}^{(l)} = T^{(l)T} f^{(l)}, \qquad l \in \{i, j\}.$$

After the basis transformation, the constraint is explicitly enforced by a partial subassembly in the new primal variables.

Applying the constraint vector to the displacements on the edge yields

$$c^{T}u_{\mathcal{E}}^{(l)} = c^{T}T_{\mathcal{E}}^{(l)}\overline{u}_{\mathcal{E}}^{(l)} = c^{T}\left[c, C^{(l)\perp}\right]\overline{u}_{\mathcal{E}}^{(l)} = \overline{u}_{\mathcal{E},1}^{(l)}, \qquad l \in \{i,j\},$$

where $\overline{u}_{\mathcal{E},1}^{(l)}$ is the displacement at the first degree of freedom on the edge \mathcal{E} . A finite element assembly in this degree of freedom gives continuous values

$$\widehat{u}_{\mathcal{E},1} := \widehat{u}_{\mathcal{E},1}^{(i)} := \widehat{u}_{\mathcal{E},1}^{(j)} := \frac{1}{2}\overline{u}_{\mathcal{E},1}^{(i)} + \frac{1}{2}\overline{u}_{\mathcal{E},1}^{(j)}$$

and (possibly) noncontinuous values $\widehat{u}_{\mathcal{E},k}^{(l)} = \overline{u}_{\mathcal{E},k}^{(l)}$ for $k > 1, l \in \{i, j\}$. For the vectors that are transformed back to the original basis, we see that the constraint is enforced:

$$c^{T}\left(T_{\mathcal{E}}^{(i)}\widehat{u}_{\mathcal{E}}^{(i)} - T_{\mathcal{E}}^{(j)}\widehat{u}_{\mathcal{E}}^{(j)}\right) = \frac{1}{2}\overline{u}_{\mathcal{E},1}^{(i)} + \frac{1}{2}\overline{u}_{\mathcal{E},1}^{(j)} - \left(\frac{1}{2}\overline{u}_{\mathcal{E},1}^{(i)} + \frac{1}{2}\overline{u}_{\mathcal{E},1}^{(j)}\right) = 0.$$

This degree of freedom then belongs to the new set of primal variables Π .

3.3. An alternative formulation of the transformation of basis for FETI-DP and BDDC. In [40], a technique was introduced to avoid affecting the sparsity of the matrix $\overline{K}_{B'B'}^{(i)}$ by the explicit transformation of basis. This is important for face constraints. To resolve this issue, as in [40, Section 4.2.2], we can alternatively introduce additional, local Lagrange multipliers $\mu^{(i)}$ and consider local saddle point problems.

Let us briefly consider this in detail. By applying the operator F, the expression $B_{B'}\overline{K}_{B'B'}^{-1}\overline{v}_{B'}\left(\text{ or } B_{B'}^{(i)}\left(\overline{K}_{B'B'}^{(i)}\right)^{-1}\overline{v}_{B'}^{(i)} \right)$ has to be evaluated locally. Obviously, the minimization of $\overline{u}_{B'}^{(i)T}\overline{K}_{B'B'}^{(i)}\overline{u}_{B'}^{(i)}$ leads to the same result as the minimization of

(3.5)
$$\begin{bmatrix} \overline{u}_{B'}^{(i)T} & 0 \end{bmatrix} \begin{bmatrix} \overline{K}_{B'B'}^{(i)} & \overline{K}_{\Pi'B'}^{(i)T} \\ \overline{K}_{\Pi'B'}^{(i)} & \overline{K}_{\Pi'\Pi'}^{(i)} \end{bmatrix} \begin{bmatrix} \overline{u}_{B'}^{(i)} \\ 0 \end{bmatrix},$$

where the values at the additional primal variables Π (cf. the previous Section 3.2) are set to zero. This is admissible since the jump operator is applied afterwards, and thus the values at the primal variables are set to zero.

Then, instead of minimizing the expression (3.5) in the transformed variables $\overline{u}_{B'}^{(i)}$, we introduce a corresponding constraint $Q^{(i)T}u^{(i)} = 0$ for the nontransformed variables. This consequently leads to the following saddle point problem

(3.6)
$$\begin{bmatrix} K_{II}^{(i)} & K_{I\Gamma}^{(i)} & 0\\ K_{I\Gamma}^{(i)T} & K_{\Gamma\Gamma}^{(i)} & Q^{(i)}\\ 0 & Q^{(i)T} & 0 \end{bmatrix} \begin{bmatrix} u_I^{(i)}\\ u_\Gamma^{(i)}\\ \mu^{(i)} \end{bmatrix} = \begin{bmatrix} v_I^{(i)}\\ v_\Gamma^{(i)}\\ 0 \end{bmatrix}$$

with additional local Lagrange multipliers $\mu^{(i)}$ (cf. [40, Section 4.2.2] for more details), and where the right-hand side $(v_I^{(i)T}, v_{\Gamma}^{(i)T})$ corresponds to $(\overline{v}_{B'}^T, 0)$ transformed back to the initial basis and restricted to the local subdomain.

Similar techniques of using saddle point problems to enforce the primal constraints have been used by other authors: in [18] the Lagrange multipliers are global; in [46, 47] the saddle point problems are local, which is also the case in our approach. In our case, however, the coarse Schur complement operator is assembled from the local subdomain contributions rather than built from a triple matrix product.

4. Adaptive FETI-DP and BDDC methods with a generalized transformation-ofbasis approach. In this section, we revisit the adaptive FETI-DP method [28, Section 5] and construct a corresponding method without using deflation or balancing. For highly heterogeneous problems, the a priori coarse space might not be sufficient to ensure convergence, and therefore adaptive strategies such as [28, Section 5] are needed. We will explain how to define the new basis vectors (in the space of displacements) for the generalized transformationof-basis approach [30] and how they are related to the deflation vectors used in [28] (which, for FETI-DP, are defined in the Lagrange multiplier space).

It is well known that deflation vectors are typically chosen as (approximate) eigenvectors corresponding to extreme eigenvalues of the operator in order to reduce the (effective) condition number of the deflated operator [38, 50, 51]. In adaptive domain decomposition, the deflation vectors are obtained from (approximately) solving local eigenvalue problems; see, e.g., [28] and the many other references cited in the introduction.

4.1. Generalized eigenvalue problems. Let us consider either a face \mathcal{F}^{ij} between the subdomains Ω_i and Ω_j or an edge \mathcal{E}^{il} between the subdomains Ω_i and Ω_l , where Ω_i and Ω_l have no common face; cf. Figure 4.1 (left). As in [47] for faces and [34] for faces and edges, we define the operator on the closure of the face $B_{\overline{\mathcal{F}}^{ij}} = [B_{\overline{\mathcal{F}}^{ij}}^{(i)} B_{\overline{\mathcal{F}}^{ij}}^{(j)}]$ and $B_{\mathcal{E}^{il}} = [B_{\mathcal{E}^{il}}^{(i)} B_{\mathcal{E}^{il}}^{(l)}]$ as the submatrix of $[B^{(i)} B^{(j)}]$ and $[B^{(i)} B^{(l)}]$, respectively, consisting of all the rows that contain exactly one +1 and one -1. The operators $B_{\overline{\mathcal{F}}^{ij}}$ and $B_{\mathcal{E}^{il}}$ provide the local jump on the closure of the face or edge, respectively.

Analogously, $B_{D,\overline{\mathcal{F}}^{ij}} = [B_{D,\overline{\mathcal{F}}^{ij}}^{(i)} \ B_{D,\overline{\mathcal{F}}^{ij}}^{(j)}]$ and $B_{D,\mathcal{E}^{il}} = [B_{D,\mathcal{E}^{il}}^{(i)} \ B_{D,\mathcal{E}^{il}}^{(l)}]$ are the submatrices of $[B_D^{(i)} \ B_D^{(j)}]$ and $[B_D^{(i)} \ B_D^{(l)}]$, respectively. We also need

$$P_{D,\overline{\mathcal{F}}^{ij}} := B_{D,\overline{\mathcal{F}}^{ij}}^T B_{\overline{\mathcal{F}}^{ij}}, \ P_{D,\mathcal{E}^{il}} := B_{D,\mathcal{E}^{il}}^T B_{\mathcal{E}^{il}}, \quad \text{and} \quad S_{is} := \begin{bmatrix} S^{(i)} & 0\\ 0 & S^{(s)} \end{bmatrix}, \ s \in \{j,l\}.$$

We now establish and solve the following generalized eigenvalue problems that can be motivated by the localization of the global P_D operator from the standard FETI-DP theory. We define the bilinear form $s_{is}(u_{is}, v_{is}) := (S_{is}u_{is}, v_{is})$ with the displacement variables $u_{is}, v_{is} \in W_i \times W_s, s \in \{j, l\}$. The local generalized eigenvalue problem on either a face or an edge can then be formulated in variational form:

Find
$$w_{is}^k \in (\ker S_{is})^{\perp}$$
 with $\mu_{is}^k > \text{TOL}$, such that
(4.1) $s_{is}(P_{D,\overline{Z}^{is}}v_{is}, P_{D,\overline{Z}^{is}}w_{is}^k) = \mu_{is}^k s_{is}(v_{is}, w_{is}^k) \quad \forall v_{is} \in (\ker S_{is})^{\perp}, \ s \in \{j, l\};$

cf. [47, Sections 3 and 4] and [28, Section 5] for a more detailed description.

We assume that we have now computed certain eigenvectors w_{is}^k for $\mu_{is}^k > \text{TOL}$, $s \in \{j, l\}$, on closed faces or edges. Later, during the FETI-DP iteration, we then enforce constraints of the form (see (4.1) and [28, Section 5])

(4.2)
$$(w_{is}, P_{D,\overline{z}^{is}}^T S_{is} P_{D,\overline{z}^{is}} w_{is}^k) = 0 \qquad \forall w_{is} \in W_i \times W_s, \ s \in \{j,l\},$$

by deflation or balancing (see [28] and [38]) or, as presented in the following, by a partial assembly after a transformation of basis on edges and faces.

Eigenvectors on closed faces. We define $u_{ij}^k = B_{D,\overline{\mathcal{F}}^{ij}}S_{ij}P_{D,\overline{\mathcal{F}}^{ij}}w_{ij}^k$ and decompose it into edge components u_{ij,\mathcal{E}_m}^k and a component $u_{ij,\mathcal{F}}^k$ on the open face, all extended by zero to the closure (excluding the vertices) of the face; cf. [28, 29].

Eigenvectors on edges. The edge eigenvalue problems are crucial to obtain a convergence bound for heterogeneous problems in three dimensions [28]. For the theory in [28], edge eigenvalue problems are, however, only needed for edges which belong to more than three subdomains, i.e., with a multiplicity larger than three. If automatic mesh partitioners are used, then this applies to only a small percentage of edges. The methods proposed in [28], covered by a condition number bound, are therefore only slightly more expensive than the successful, classical face-based approaches from [47, 48], which only provided a condition number indicator.

4.2. Constraints for the generalized transformation of basis and solution spaces. For face eigenvalue problems, we obtain edge components and a component on the open face

$$q_{ij,\mathcal{E}_m}^k := B_{\overline{\mathcal{F}}^{ij}}^T u_{ij,\mathcal{E}_m}^k \qquad \text{and} \qquad q_{ij,\mathcal{F}}^k := B_{\overline{\mathcal{F}}^{ij}}^T u_{ij,\mathcal{F}}^k$$

and enforce the corresponding constraints on the displacement variable w_{ij} , i.e.,

(4.3)
$$(w_{ij}, q_{ij,\mathcal{E}_m}^k) = (w_{ij}, B_{\overline{\tau}^{ij}}^T u_{ij,\mathcal{E}_m}^k) = 0, \qquad m = 1, 2, \dots,$$

(4.4)
$$(w_{ij}, q_{ij,\mathcal{F}}^k) = (w_{ij}, B_{\overline{\mathcal{F}}^{ij}}^T u_{ij,\mathcal{F}}^k) = 0$$

Hence, we obtain the constraint vectors in the displacement space by a multiplication of the constraint vector in the Lagrange multiplier space by the localized version of the jump operator B, i.e., by $B_{\overline{\tau}^{ij}}^T$.

For the edge eigenvalue problems, we choose constraint vectors

$$q_{il}^k := B_{\mathcal{E}^{il}}^T u_{il}^k := P_{D,\mathcal{E}^{il}}^T S_{il} P_{D,\mathcal{E}^{il}} w_{il}^k$$

such that the constraint

(4.5)
$$(w_{il}, q_{il}^k) = (w_{il}, B_{\mathcal{E}^{il}}^T u_{il}^k) = 0$$

is satisfied. Thus, we obtain the constraint vectors in the displacement space by a multiplication of the constraint vector in the Lagrange multiplier space by the localized version of the jump operator B, i.e., by B_{sil}^T .

We extend all constraint vectors q_{ij,\mathcal{E}_m}^k , $q_{ij,\mathcal{F}}^k$, and q_{il}^k by zero to \widetilde{W} . These vectors define the columns of the matrix Q. Analogously, we extend all constraint vectors u_{ij,\mathcal{E}_m}^k , $u_{ij,\mathcal{F}}^k$, and u_{il}^k by zero to the space of the Lagrange multipliers, defining the columns of the matrix U.

Obviously, the spaces

(4.6)
$$\widetilde{W}_Q = \{ w \in \widetilde{W} : Q^T w = 0 \}$$
 and $\widetilde{W}_U = \{ w \in \widetilde{W} : U^T B w = 0 \}$

are the same but they correspond to two different approaches to implement the constraints. The space \widetilde{W}_U corresponds to the solution space for deflation or balancing; cf. [28]. In order to define the solution space for the generalized transformation of basis as introduced in [30], we have to introduce some notation and give some remarks. The solution space $\widetilde{W}_{\widehat{Q}}$ for the transformation-of-basis approach will result from a restriction of \widetilde{W}_Q .

We now explain the necessary details that have to be considered when our adaptive constraints are implemented such that the restrictions of the generalized transformation-of-basis approach [30] are satisfied.

Due to the form of the given constraints (see (4.3)–(4.5)), we have for q_{ij,\mathcal{E}_m}^k , $q_{ij,\mathcal{F}}^k$, and q_{il}^k

$$(4.7) \qquad (w_{ij}, q_{ij,\mathcal{E}_m}^k) = 0 \quad \Leftrightarrow \quad (B_{\overline{\mathcal{F}}^{ij}}^{(i)} w^{(i)}, u_{ij,\mathcal{E}_m}^k) = -(B_{\overline{\mathcal{F}}^{ij}}^{(j)} w^{(j)}, u_{ij,\mathcal{E}_m}^k), (u_{ij}, q_{ij,\mathcal{F}}^k) = 0 \quad \Leftrightarrow \quad (B_{\overline{\mathcal{F}}^{ij}}^{(i)} w^{(i)}, u_{ij,\mathcal{F}}^k) = -(B_{\overline{\mathcal{F}}^{ij}}^{(j)} w^{(j)}, u_{ij,\mathcal{F}}^k), (w_{il}, q_{il}^k) = 0 \quad \Leftrightarrow \quad (B_{\mathcal{E}^{il}}^{(i)} w^{(i)}, u_{il}^k) = -(B_{\mathcal{E}^{il}}^{(l)} w^{(l)}, u_{il}^k).$$

Since $B_{\overline{\mathcal{F}}^{ij}}^{(i)}$ and $B_{\overline{\mathcal{F}}^{ij}}^{(j)}$ are closely related, i.e., both operators only differ by their sign when restricted to the face (i.e., $B_{\overline{\mathcal{F}}^{ij}}^{(i)}$ and $B_{\overline{\mathcal{F}}^{ij}}^{(j)}$ differ by their sign when all zero columns are removed), the constraint vector on the face will be identical for both sides of the face. The same arguments apply to the edge eigenvalue problems and the edge constraints from face eigenvalue problems since, again, $B_{\mathcal{E}^{il}}^{(i)}$ and $-B_{\mathcal{E}^{il}}^{(l)}$ are identical if restricted to the corresponding edge. Given computed eigenvectors and resulting sets of orthonormalized constraints on a

Given computed eigenvectors and resulting sets of orthonormalized constraints on a certain open face \mathcal{F}^{ij} and on its related edges \mathcal{E}_m or just on an edge \mathcal{E}^{il} ; cf. (4.3)–(4.5). The orthonormalized result will be denoted by $T_{\mathcal{F}^{ij},\Pi}$ and $T_{\mathcal{E}_m,\Pi}$ or $T_{\mathcal{E}^{il},\Pi}$. The matrices $T_{\mathcal{F}^{ij}} := [T_{\mathcal{F}^{ij},\Pi}, T_{\mathcal{F}^{ij},\Delta}], T_{\mathcal{E}_m} = [T_{\mathcal{E}_m,\Pi}, T_{\mathcal{E}_m,\Delta}], \text{ and } T_{\mathcal{E}^{il}} = [T_{\mathcal{E}^{il},\Pi}, T_{\mathcal{E}^{il},\Delta}]$ then are computed such that they are orthogonal matrices.

Given a face \mathcal{F}^{ij} , the sets of orthogonalized constraints $T_{\mathcal{F}^{ij},\Pi}$ and $T_{\mathcal{E}_m,\Pi}$ can then be used as constraint vectors for both subdomains Ω_i and Ω_j . This results from the form of B and the fact that the constraint vector restricted to one subdomain equals (-1) times the constraint vector restricted to the other subdomain; cf. (4.7). Given an edge \mathcal{E}^{il} (\mathcal{E}_m can be handled likewise), the same applies for the two subdomains considered in the edge eigenvalue problem.

In order to satisfy the assumptions of the generalized transformation-of-basis approach, we also enforce the same constraints for all other jumps between two arbitrary subdomains at the considered face or edge. Then, for a face \mathcal{F}^{ij} with jumps $w^{(i)} - w^{(j)}$ and edges \mathcal{E}_m or an edge \mathcal{E}^{il} and all jumps $w^{(i)} - w^{(s)}$ across the edge, we have that the local transformations are equal for all subdomains sharing either the face or the edge, i.e.,

(4.8)
$$T_{|\mathcal{F}^{ij}}^{(i)} = T_{|\mathcal{F}^{ij}}^{(j)} = T_{\mathcal{F}^{ij}}, \quad T_{|\mathcal{E}_m}^{(i)} = T_{|\mathcal{E}_m}^{(s)} = T_{\mathcal{E}_m}, \quad T_{|\mathcal{E}^{il}}^{(i)} = T_{|\mathcal{E}^{il}}^{(s)} = T_{\mathcal{E}^{il}}^{(s)}$$

and all pairs of subdomains $\{\Omega_i, \Omega_s\}$ sharing the edges \mathcal{E}_m or \mathcal{E}^{il} .

Thus, the constraint set obtained from our local eigenvalue problems can be extended such that the local transformations satisfy condition (4.8) for the generalized transformation of basis as given in [30]. We then have $\widehat{Q} := [Q, *]$ with range $Q \subset \operatorname{range} \widehat{Q}$ and the solution space

(4.9)
$$\widetilde{W}_{\widehat{Q}} = \{ w \in \widetilde{W} : \widehat{Q}^T w = 0 \} \subset \widetilde{W}_Q = \widetilde{W}_U;$$

cf. the definitions in (4.6). The constraints are then enforced by a partial subassembly; cf. Figure 4.1. Therefore, the solution space $\widetilde{W}_{\widehat{Q}}$ is in general a strict subset of the solution space \widetilde{W}_U , and we obtain our P_D -estimate from [28, Lemma 6.1] for $\widetilde{W}_{\widehat{Q}}$, i.e., for $w \in \widetilde{W}_{\widehat{Q}}$, we have

(4.10)
$$|P_D w|_{\widetilde{S}}^2 \le 4 \max\{N_{\mathcal{F}}, N_{\mathcal{E}} M_{\mathcal{E}}\}^2 \operatorname{TOL} |w|_{\widetilde{S}}^2.$$

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FIGURE 4.1. Cross sectional view of four subdomains sharing an edge. Arrows symbolize redundant Lagrange multipliers in FETI-DP (left). Assume that, using deflation, one primal constraint is introduced, involving the Lagrange multiplier depicted in bold red color (middle). Using partial assembly, after a generalized transformation of basis, the primal constraint is now enforced between all four subdomains, effectively removing all six Lagrange multipliers (right).

In our implementation, the constrained elements of the space $\widetilde{W}_{\widehat{Q}}$ are represented by $\widetilde{W}_{T,a} := \{\widehat{w} = R_{\mu}^T T^T w : w \in \widetilde{W}\}$, where $R_{\mu}^T = (R^T R)^{-1} R^T$, $T = \text{blockdiag}_{i=1,...,N} T^{(i)}$. This requires the definition of R^T and $T^{(i)}$, i = 1, ..., n. The operator R^T assembles in all a posteriori degrees of freedom and leaves all other degrees of freedom (interior, a priori primal and remaining dual variables) unchanged. The operator $T^{(i)}$ reduces to the identity on all interior degrees of freedom and is blockdiagonal on the interface with blocks $T_{\mathcal{F}^{ij}}$ and $T_{\mathcal{E}^{il}}$ for all faces \mathcal{F}^{ij} and edges \mathcal{E}^{il} shared by Ω_i .

4.3. Adaptive FETI-DP and BDDC operators for the generalized transformation-ofbasis approach. In [30], FETI-DP and BDDC methods using a generalized transformationof-basis approach were introduced.

Using this approach and the notation from [30], the preconditioned adaptive FETI-DP method is then given by

(4.11)
$$\widehat{M}_T^{-1}\widehat{F}\lambda := (\widehat{B}_D\widehat{\widetilde{S}}\widehat{B}_D^T) \ (\widehat{B}\widehat{\widetilde{S}}^{-1}\widehat{B}^T)\lambda$$
$$:= (B_DTR_\mu (R^TT^T\widetilde{S}TR)R_\mu^TT^TB_D^T) \ (BTR(R^TT^T\widetilde{S}TR)^{-1}R^TT^TB^T)\lambda = \widehat{d}.$$

Introducing another operator R'^T which assembles in all remaining dual and a posteriori primal variables and leaves the a priori primal variables unchanged, the preconditioned adaptive BDDC method is given by

(4.12)
$$\widehat{M}_{\text{BDDC}}^{-1} \mathcal{S}u := (R'^T T \widehat{D}_u R \widehat{\widetilde{S}}^{-1} R^T \widehat{D}_u T^T R') (R'^T \widetilde{S} R')u$$
$$:= (R'^T D_u T R (R^T T^T \widetilde{S} T R)^{-1} R^T T^T D_u R') (R'^T \widetilde{S} R')u = \widehat{g}$$

with D_u the degree of freedom scaling in BDDC corresponding to the Lagrange multiplier scaling D in FETI-DP.

4.4. Condition number estimate for adaptive FETI-DP and BDDC. We can now formulate the following theorem for our adaptive algorithm using a generalized transformation of basis based on [30].

In [30], a correspondence of FETI-DP or BDDC methods using a generalized transformation of basis and of FETI-DP methods using deflation or balancing was shown. These methods have essentially the same eigenvalues, and as opposed to the previous theory, this result also applies to general scalings and heterogeneous problems including coefficient jumps inside subdomains.

THEOREM 4.1. Let N_F denote the maximum number of faces of a subdomain, $N_{\mathcal{E}}$ the maximum number of edges of a subdomain, $M_{\mathcal{E}}$ the maximum multiplicity of an edge, and TOL



FIGURE 5.1. Irregular decomposition of the unit cube with composite material (left), representative volume element (center), and randomized coefficients (right). High coefficients $E_2 = 1e+06$ (for composite and random) and $E_2 = 2.1e+05$ (for the RVE) are shown in a dark purple color in the picture, subdomains shown in different colors in the background and by half-transparent slices. Visualization for N = 8 and $1/h = 3N^{1/3}$ (left), N = 512 and $1/h = 4N^{1/3}$ (center), and N = 8 and $1/h = 5N^{1/3}$ (right).

a given tolerance for solving the local generalized eigenvalue problems (4.1). Furthermore, let all vertices chosen to be primal. Then, the condition number $\kappa(\widehat{M}_T^{-1}\widehat{F})$ of the FETI-DP algorithm with constraints enforced by the generalized transformation of basis as described satisfies

$$\kappa(\widehat{M}_T^{-1}\widehat{F}) \leq 4 \max\{N_{\mathcal{F}}, N_{\mathcal{E}}M_{\mathcal{E}}\}^2 \text{TOL}.$$

The condition number $\kappa(\widehat{M}_{BDDC}^{-1}S)$ of the BDDC algorithm with adaptive constraints enforced by the generalized transformation of basis as described above satisfies

$$\kappa(\widehat{M}_{BDDC}^{-1}\mathcal{S}) \le 4 \max\{N_{\mathcal{F}}, N_{\mathcal{E}}M_{\mathcal{E}}\}^2 \text{TOL}.$$

Proof. The proof is complete by acknowledging that this is a special case of [30, Theorem 6.7] and [30, Theorem 7.3] if the transformations are built according to [30] (i.e., they are identical for all sides of any considered edge) and by using (4.10).

Note that the bound in (4.1) is algebraic in the sense that the condition number bound holds under very weak assumptions. However, under unfavorable conditions, the coarse space can be so large that the method reduces to a direct solver.

REMARK 4.2. Note that the explicit condition number of the FETI-DP or BDDC method with our generalized transformation-of-basis approach is always smaller or equal to that of the corresponding balancing or deflation approach. This results from the fact that in our generalized transformation-of-basis approach in the partial assembly, we often enforce additional constraints compared to the deflation approach—without creating a larger coarse space; cf. (4.9), (4.8), and Figure 4.1.

Nevertheless, note that we can always find an equivalent balancing or deflation method by expanding the constraint columns U such that all the constraints from $\widetilde{W}_{\widehat{Q}}$ are implemented and such that $\kappa(M_{PP}^{-1}F) = \kappa(\widehat{M}_T^{-1}\widehat{F})$; cf. [30].

5. Numerical results. In this section, we provide numerical results for the adaptive coarse spaces presented in [28] (including the classical variants of [47, 48] relying on face eigenvalue problems only) but implemented by the generalized transformation-of-basis approach for FETI-DP and BDDC.

As in [28], we consider compressible linear elasticity on the unit cube $\Omega = [0, 1]^3$ as a model problem. The Poisson ratio is $\nu = 0.3$ and the Young Modulus is $E \in \{1, 1e+06\}$. We assume constant values of E and ν on each finite element.

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FIGURE 5.2. Local eigenvalues greater than 0.1 from all generalized eigenvalue problems (4.1) for some simulations for the three materials depicted in Figure 5.1: Composite material (left), representative volume element (center), and randomized coefficients for N = 216 and the first five runs (right).

The initial coarse space for all methods consists of vertex constraints. We then add several other vertices such that each nonstraight edge has at least three primal vertices. This choice is identical to that of [28].

We consider irregular decompositions of the unit cube. These decompositions are obtained from the METIS graph partitioner (see [25]) using the options -ncommon=3 and -contig to avoid noncontiguous subdomains and unwanted hinge modes between sets of tetrahedra inside single subdomains.

We test five different algorithms for FETI-DP and BDDC; cf. [28]. The first three, Algorithms Ia, Ib, and Ic originate from our recent paper [28] but are using a transformation of basis instead of balancing. Algorithm Ia uses all the constraints obtained from the local eigenvalue problems. Algorithm Ib and Algorithm Ic are heuristically reduced variants of Algorithm Ia. For Algorithm Ib, edge eigenvalue problems are discarded if there are no jumps of the Young Modulus present within a distance of one finite element around the edges. In Algorithm Ic not only edge eigenvalue problems are discarded on these edges but also edge constraints from face eigenvalue problems. The Algorithms II and III are the classical face-based algorithms from [47, 48]. There, Algorithm II uses all constraints (edge and face constraints) obtained from face eigenvalue problems but still lacks any edge eigenvalue problem. Algorithm III was already tested in [48] and just uses face constraints from face eigenvalue problems.

In all tables, we denote by κ the condition number of the preconditioned FETI-DP and BDDC operator, and *its* denotes the number of preconditioned conjugate gradient (pcg) iterations that are needed until convergence. The iteration is also stopped if no convergence is observed within 500 iterations. As a stopping criterion, we require the pcg algorithm to reduce the preconditioned starting residual by 10^{-10} . The condition numbers κ given in the table are the standard estimates obtained from pcg. The condition number estimates for FETI-DP and BDDC differ slightly since, in most cases, the estimates of the smallest eigenvalue differ slightly, typically starting in the second or third digit.

By $|\Pi'|$ we denote the size of the standard coarse space while by $|\Pi|$ we indicate the size of the corresponding adaptive coarse space implemented by the transformation of basis; see the previous sections and [30]. By N we denote the number of subdomains. For our coarse spaces introduced in [28], we also present $\#\mathcal{E}_{evp}$, the number of edge eigenvalue problems, and in parentheses the percentage of these in the total number of eigenvalue problems.

For all runs, we highlight small condition numbers, i.e., $\kappa \leq 50$, in bold face. If not stated otherwise, all eigenvalue problems are solved by the MATLAB builtin "eig" function; cf. Section 5.1.4 and Section 5.3 for a solution of the eigenvalue problems by the LOBPCG solver

[42, 43]. Except in Table 5.9, we always use TOL = 10. The resulting condition number is typically at the order of TOL; see also [28, p. A2893].

When ρ -scaling is used we apply the standard ρ -scaling, where the nodal coefficient values are given by the maximum value of the coefficient function on all neighboring tetrahedra inside the same subdomain.

We enforce zero Dirichlet boundary conditions for the face with x = 0 and zero Neumann boundary conditions elsewhere and apply the body force $f = (0.1, 0.1, 0.1)^T$.

The following section has three subsections. In Section 5.1, we test the transformation-ofbasis approach for materials already tested with ρ -scaling and the balancing approach in [28]. In Section 5.2, we test our composite material and randomly distributed coefficients with different scalings such as ρ -, deluxe-, stiffness-/K-, and multiplicity-scaling. In Section 5.3, we present more results in order to demonstrate the robustness of our algorithm with the LOBPCG eigensolver [42, 43].

5.1. Comparison to the balancing method. In this section, we test three different distributions of the Young Modulus E; see Figure 5.1, the more detailed descriptions in the following Sections 5.1.1-5.1.3, or our recent paper [28].

5.1.1. Composite material. In this section, we consider a composite material. The material consists of a soft matrix material with E = 1 that surrounds $N^{2/3}$ many beams that run in a straight line from the face with x = 0 to the face with x = 1 and that occupy 1/9 of that cross-section of the material; cf. Figure 5.1 and [28] where we referred to this material as composite material 1.

We see that FETI-DP and BDDC behave almost identically. Clearly, the coarse spaces are of the same size and the iteration count is almost the same for both methods. We see that the size of the second-level coarse space is quite modest compared to the standard coarse space. Compared to the coarse space implemented by the balancing approach (cf. [28, Table 8.2]), the coarse space by the transformation of basis in Table 5.1 is up to two times the size of the previous one. This is due to the fact that we always make all three degrees of freedom primal that belong to a node, even if only one or two constraints are required. On the other hand, the coarse spaces of [28, Table 8.3] and Table 5.2 do have a comparable size regardless of the implementation.

For a more detailed consideration of the computational overhead of additional edge eigenvalue problems, see the very detailed discussion in [28]. For a distribution of the local eigenvalues for some of the runs; see Figure 5.2 (left).

5.1.2. Representative volume element. In this section, we consider a representative volume element of a modern steel microstructure; see Figure 5.1. The RVE has been obtained from the one in [44, Fig. 5.5] by resampling; see [28, Section 8.2]. As in [44], we use $\nu = 0.3$, $E_1 = 210$, and $E_2 = 210\,000$ as artificial material parameters. The FETI-DP and BDDC algorithms tested here and the FETI-DP algorithm with a balancing approach (see [28]) show almost identical convergence behavior. The transformation-of-basis approach gives a larger coarse space. As mentioned in the previous section this could be improved so that the balancing approach does not have significant advantages over the transformation approach; we did not do so because of the ease of implementation. For a distribution of the local eigenvalues, see Figure 5.2 (center).

5.1.3. Randomly distributed coefficients. In this section, we test randomly distributed coefficients. We let 20% of the tetrahedra in the unit cube take the value E = 1e+06 while the other tetrahedra take E = 1. Both, FETI-DP and BDDC behave almost identically. The coarse spaces of transformation and balancing are of comparable size; cf. [28] for $N \in \{4^3, 5^3\}$. Compared to [28], we now show results for $N = 6^3$ where it can also be seen that not only

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TABLE 5.1

Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for TOL = 10 for all generalized eigenvalue problems.

				iposi	te mute	FET		ina 1/7	011	•		
			Algo	rithm	Ia, Ib, d			rithm L	I	Algor	ithm III	
\overline{N}	$ \Pi' $		ĸ	its	Π	$\# \mathcal{E}_{evp}$	ĸ	its	$ \Pi $	κ	its	$ \Pi $
		a)	8.44	29	144	7 (11.9%)						
3^{3}	615	b)	8.44	29	144	4 (7.1%)	8.44	29	144	8.31e+05	51	90
		c)	8.44	29	129	4 (7.1%)						
		a)	14.30	36	459	14 (5.2%)						
5^{3}	3084	b)	14.30	36	459	8 (3.0%)	14.30	36	453	3.29e+05	187	303
		c)	14.30	37	375	8 (3.0%)						
		a)	13.92	40	1098	48 (6.0%)						
7^{3}	8781	b)	13.92	40	1089	21 (2.7%)	2.93e+05	84	1074	2.96e+05	373	780
		c)	13.93	41	942	21 (2.7%)						
03	10020	a)	16.27	41	2070	90 (5.2%)	0.66 05		20.42	4.60 05	100	1570
9^{3}	19029	b)	16.28	42	2067	45 (2.7%)	2.66e+05	71	2043	4.69e+05	482	1572
		c)	16.28	42	1812	45 (2.7%)						
11^{3}	25214	a)	15.05	43	3582	167 (5.2%)	2 ((05	140	2504	2 (0- 105	500	2724
110	35214	b) c)	15.05 15.05	43 43	3570 3192	95 (3.0%) 95 (3.0%)	2.66e+05	142	3504	3.60e+05	500	2724
			15.05	43	5895	303 (5.6%)						
13^{3}	58179	a) b)	17.12	44	5889	303 (3.0%) 171 (3.3%)	2.74e+05	225	5739	3.01e+05	500	4557
15	38179	c)	17.12	44	5346	171 (3.3%)	2.740+03	223	5759	5.010+05	300	4337
		0)	17.15		5540	. ,						
			Alao	rithm	Ia, Ib, a	BD	-	rithm L	I	Algor	ithm III	r
N	Π'		πig0 κ	its	$ \Pi $	$\frac{\mathcal{U}_{unu} \mathcal{U}_{evp}}{\# \mathcal{E}_{evp}}$	κ κ	its	ι Π	κ κ	its	П
	11	a)	8.52	29	144	$\frac{\#c_{evp}}{7(11.9\%)}$	n	11.5	11	n	11.5	11
3^3	615	b)	8.52 8.52	29	144	4 (7.1%)	8.52	29	144	8.42e+05	66	90
0	015	c)	8.52	29	129	4 (7.1%)	0.02	2)	144	0.420105	00	70
		a)	14.44	35	459	14 (5.2%)						
5^{3}	3084	b)	14.44	35	459	8 (3.0%)	14.44	37	453	3.33e+05	241	303
-		c)	14.45	36	375	8 (3.0%)	-					
		a)	14.08	40	1098	48 (6.0%)						
7^{3}	8781	b)	14.08	40	1089	21 (2.7%)	2.97e+05	98	1074	3.00e+05	459	780
		c)	14.08	41	942	21 (2.7%)						
		a)	16.44	41	2070	90 (5.2%)						
9^{3}	19029	b)	16.44	41	2067	45 (2.7%)	2.69e+05	76	2043	4.75e+05	500	1572
		c)	16.44	42	1812	45 (2.7%)						
		a)	15.22	40	3582	167 (5.2%)						
11^{3}	35214	b)	15.22	41	3570	95 (3.0%)	2.69e+05	162	3504	3.72e+05	500	2724
		c)	15.22	42	3192	95 (3.0%)						
		a)	17.32	41	5895	303 (5.6%)						
						. ,						
13^{3}	58179	a) b) c)	17.32 17.32	41 41	5889 5346	171 (3.3%) 171 (3.3%)	2.77e+05	250	5739	3.40e+05	500	4557

Composite material, irregular partitioning, and $1/h = 3N^{1/3}$.

the condition number but also the iteration count of Algorithm II deteriorates with larger N. Therefore, we also include the maximum and minimum value in the table. For a distribution of the local eigenvalues for some of the runs, see Figure 5.2 (right).

5.1.4. Composite material using LOBPCG as an eigenvalue solver. In this section, we present results for the FETI-DP and BDDC algorithm in combination with the iterative LOBPCG (cf. [42, 43]) eigenvalue solver. We choose a block size of 10 and use a Cholesky decomposition of the right-hand side of the eigenvalue problem as local preconditioner. We limit the number of maximum iterations of the iterative solver as indicated in the table. We test a larger example than in the previous paper [28], i.e., $N = 5^3$, to show that a larger admissible number of iterations does not consequently lead to faster convergence; see the results for

TABLE 5.2 Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for TOL = 10 for all generalized eigenvalue problems.

			Comp	osite	materi	al , irregular pa	rtitioning	, and 1	/h = 61	$N^{1/3}$.		
						FETI-	DP					
			Algo	rithm	Ia, Ib, a	and Ic	Alg	orithn	ı II	Algor	ithm III	r
N	$ \Pi' $		κ	its	$ \Pi $	$\# \mathcal{E}_{evp}$	κ	its	$ \Pi $	κ	its	$ \Pi $
		a)	8.65	34	699	2 (2.0%)						
3^{3}	960	b)	8.65	34	699	1 (1.0%)	8.65	34	699	1.36e+06	68	243
		c)	8.67	34	450	1 (1.0%)						
		a)	9.16	35	3675	25 (4.2%)						
5^3	5433	b)	9.16	35	3675	12 (2.1%)	9.16	35	3669	5.50e+05	190	1242
		c)	10.49	36	2325	12 (2.1%)						
		a)	10.76	37	10101	65 (3.6%)						
7^{3}	16248	b)	10.76	37	10101	27 (1.5%)	10.76	37	10089	1.21e+06	424	3606
		c)	13.36	39	6693	27 (1.5%)						
						BDD	С					
			Algo	rithm	Ia, Ib, a	and Ic	Alg	orithm	ı II	Algor	ithm III	r
N	$ \Pi' $		κ	its	$ \Pi $	$\# \mathcal{E}_{evp}$	κ	its	$ \Pi $	κ	its	$ \Pi $
		a)	8.68	32	699	2 (2.0%)						
3^{3}	960	b)	8.68	32	699	1 (1.0%)	8.68	32	699	1.37e+06	77	243
		c)	8.69	32	450	1 (1.0%)						
		a)	9.22	33	3675	25 (4.2%)						
5^3	5433	b)	9.22	33	3675	12 (2.1%)	9.22	33	3669	5.53e+05	216	1242
		c)	10.53	34	2325	12 (2.1%)						
		a)	10.84	35	10101	65 (3.6%)						
7^{3}	16248	b)	10.84	35	10101	27 (1.5%)	10.84	35	10089	1.22e+06	492	3606
		c)	13.44	38	6693	27 (1.5%)						

LOBPCG with up to 200 iterations. We use a stopping criterion of 1e-05 for LOBPCG which, in combination with badly conditioned local matrices, already lead to instability of the solver. The implementation of LOBPCG already states that "excessively small requested tolerance may result in often restarts and instability"; see [42]. However, it should be noted that convergence does not seem to be necessary since 2–5 iterations already seem to give a stable algorithm with fast convergence.

5.2. Scaling comparisons. In this section, we compare the performance of the algorithms introduced above with four different kinds of scalings. Besides to ρ -scaling for which we already gave performance results before, we also test deluxe-, stiffness/K-, and multiplicity-scaling. As already reported in [28], neither Algorithm II nor III are sufficient to ensure convergence with ρ -scaling. But also for the other scalings as deluxe, Algorithm II is not robust. Concerning Algorithm Ia–c, as expected, the coarse space for deluxe-scaling is the smallest and the fewest iterations are needed for convergence. But deluxe-scaling is also far more costly than ρ - or stiffness-scaling, and these two only need an about 10–15% larger coarse space to achieve similar convergence results. Multiplicity-scaling cannot be recommended since it uses a coarse space that is almost two times as large as that of deluxe-scaling, i.e., an overhead of 70–90% is needed.

5.2.1. Composite material. First, we test the four different scalings for the composite material and $1/h = 6N^{1/3}$ as in Section 5.1.1; cf. Table 5.6 for the different scaling results.

5.2.2. Randomly distributed coefficients. Second, we test the four different scalings for randomly distributed coefficients with 20% high and 80% low coefficients as in Section 5.1.3 but with $1/h = 8N^{1/3}$ instead of $1/h = 5N^{1/3}$; cf. Tables 5.7 and 5.8 for the different scaling results. Again, Algorithm II and III are not robust for any scaling. Considering Algorithm Ia–c,

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TABLE 5.3

Compressible linear elasticity. Coarse spaces for TOL = 10 for all generalized eigenvalue problems. **Representative Volume Element** with $E_1 = 210$, $E_2 = 210000$,

regular and irregular partitioning, $N = 8^3$ and 1/h = 32.

			-	-		·		,			
					FETI-D	Р					
		Alge	orithm	Ia, Ib, an	nd Ic	Alg	orithn	ı II	Algo	rithm I	111
$ \Pi' $		κ	its	$ \Pi $	\mathcal{E}_{evp}	κ	its	$ \Pi $	κ	its	П
	a)	13.75	37	1275	114 (5.6%)						
18888	b)	13.75	37	1275	27 (1.4%)	13.75	37	1263	354.30	98	699
	c)	13.75	38	990	27 (1.4%)						
		-			BDDC		-				
		Alge	orithm	Ia, Ib, an	nd Ic	Alg	orithn	ı II	Algo	rithm I	III
$ \Pi' $		κ	its	$ \Pi $	\mathcal{E}_{evp}	κ	its	$ \Pi $	κ	its	$ \Pi $
	a)	13.94	31	1275	114 (5.6%)						

13.94

31

1263

359.20

699

27 (1.4%)

27 (1.4%)

18888

b)

c)

13.94

13.94

31

33

1275

990



FIGURE 5.3. Irregular decomposition of the unit cube into 64 subdomains (left), randomly distributed coefficients for seven materials with $E_1 = 1e+00$ in lightgray to $E_7 = 1e+06$ in black (center), and approximated local eigenvalues greater 0.1 from all generalized eigenvalue problems (4.1) (right: estimates from two iterations with LOBPCG on the local eigenvalue problems).

the coarse space for deluxe-scaling is again the smallest and the fewest iterations are needed for convergence. But deluxe-scaling is also far more costly than ρ - or stiffness-scaling, and these two only need an about 10–15% larger coarse space to achieve similar convergence results. Once again, multiplicity-scaling cannot be recommended since it uses a coarse space that is sometimes even two times as large as that of deluxe-scaling.

5.3. Verification of the robustness with LOBPCG eigensolver and randomly distributed coefficients. In this section, we test the LOBPCG eigensolver with blocksize 10, a Cholesky decomposition of the right-hand side of the eigenvalue problem as preconditioner and maximal two iterations on each eigenvalue problem.

We test three different random coefficient distributions for a heterogeneous material composed of seven different homogeneous materials. In these examples, 30% of the tetrahedra have a Young's Modulus of $E_1 = 1$, then 20% have a Young's Modulus of $E_2 = 10$, and another 10% each have a Young's Modulus of $E_3 = 100$, $E_4 = 1000$, $E_5 = 1e+04$, $E_6 = 1e+05$, and $E_7 = 1e+06$; see Table 5.9. As can be seen from Figure 5.3 (right) the approximated spectrum of the local eigenvalue problems is continuous such that we can expect different results for different tolerances.

We first observe that for all tolerances, convergence is achieved using just two iterations of the LOBPCG eigensolver. Considering the different tolerances, for all runs the approximated condition number is of the size of the chosen tolerance TOL. We state that only TOL = 10

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	6^3				თ. შ	I				4^{3}			N					6^3	I				53					4^{3}			N			
max σ	$\min_{x \in i} x \in I$	\overline{x}	σ	max	min	\widetilde{x}	\overline{x}	θ	max	min	\widetilde{x}	\overline{x}				٩	max	min	\widetilde{x}	\overline{x}	σ	max	min	\widetilde{x}	\overline{x}	θ	max	min	\widetilde{x}	\overline{x}				
I	9078		ı		0000	5058		ı		2 44 2	3443							2010	0078		1		0000	5050		ı		1447	2112 1					
10.93 0.48	9.12 8.40	9.18	0.56	11.13	7.90	8.88	8.94	0.48	9.91	7.48	8.78	8.82	к			0.48	10.93	8.40	9.12	9.18	0.56	11.13	7.90	8.88	8.94	0.48	9.92	7.48	8.78	8.83	к			
34 0.66	32 31	32.55	0.65	33	30	32	31.59	0.71	33	30	31	31.06	its	Alg		0.57	35	32	33	33.44	0.73	35	31	33	32.76	0.63	34	31	32	32.48	its	Al_{ξ}		
9015 114.73	8812.5 8454	8797.38	88.17	5364	4815	5143.5	5134.77	59.68	2553	2268	2421	2421.99	11	Algorithm Ib		114.73	9015	8454	8812.5	8797.38	88.17	5364	4815	5143.5	5134.77	59.68	2553	2268	2421	2421.99	$ \Pi' $	Algorithm Ib		irregula
ı	57 (5.8%)		ı		20(4.170)	73 (4 1%)		ı		9 (3.170)	0 (3 702)		$\# \mathcal{E}_{evp}$		BDDC			0, (0,0,0)	57 (5 80%)		-		23 (4.1%)	72 (1 102)		ı		(0.1.C) ¢	0 (2 702)		$\# \mathcal{E}_{evp}$		FET	irregular partitioning, and $1/h = 5N^{1/3}$.
7.41e+05 1.17e+05	2.03e+05 1.72e+04	2.06e+05	9.66e+04	4.32e+05	7.90	5.40e+04	8.40c+04	1.17e+05	5.93e+05	7.48	9.37	6.41c+04	к	A	DC	1.17e+05	7.41e+05	1.72e+04	2.03e+05	2.06e+05	9.66e+04	4.32e+05	7.90	5.40e+04	8.40c+04	1.17e+05	5.93e+05	7.48	9.37	6.42e+04	к	A	FETI-DP	, and $1/h = 1$
156 24.33	81.5 43	83.36	14.48	94	30	44.5	49.42	7.85	89	30	32	36.60	its	Algorithm II		20.60	139	41	74.5	76.19	11.99	87	31	44	46.92	5.71	61	31	33	36.28	its	Algorithm II		5N ^{1/3} .
8988 115.27	8794.5 8442	8778.66	88.32	5361	4806	5139	5128.53	59.68	2553	2265	2416.5	2420.10	11/	T		115.27	8868	8442	8794.5	8778.66	88.32	5361	4806	5139	5128.53	59.68	2553	2265	2416.5	2420.10	T/	Γ		
2.03e+06 2.49e+05	6.00e+05 3.51e+05	6.48e+05	1.37e+05	1.07e+06	2.87e+05	4.59e+05	4.94e+05	1.84e+05	1.34e+06	2.96e+05	5.20e+05	5.57e+05	к	Alg		2.54e+05	2.07e+06	3.53e+05	6.05e+05	6.55e+05	1.37e+05	1.07e+06	2.88e+05	4.61e+05	4.98e+05	1.85e+05	1.34e+06	2.97e+05	5.22e+05	5.57e+05	κ	Alg		
000	500 500	500	0	500	500	500	500	0	500	500	500	500	its	Algorithm III		0	500	500	500	500	0	500	500	500	500	0	500	500	500	500	its	Algorithm III		
5142 93.87	4957.5 4686	4949.40	70.15	3030	2583	2830.5	2831.28	46.06	1425	1200	1323	1317.60	11/	II		93.87	5142	4686	4957.5	4949.40	70.15	3030	2583	2830.5	2831.28	46.06	1425	1200	1323	1317.60	11/	III		

TABLE 5.4 Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for TOL = 10 for all generalized eigenvalue problems.

Randomly distributed coefficients with 20% high and 80% low coefficients,

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200	5	2	-	LOBPCG max. its				200			S			2			1		LOBPCG max. its		
6159	6159	6159	6159	11/				6159			6159			6159			6159		$ \Pi' $		
c) b) a)	c) b) a)	c b a	c b a				c)	<u>þ</u>	a)	c	<u>þ</u>	a)	c)	<u>b</u>	a)	c)	<u>b</u>	a)			
4.82e+04 4.82e+04 4.81e+04	18.43 18.43 18.42	15.05 15.05 15.05	43.72 43.72 43.79	к	Algori		4.78e+04	4.77e+04	4.77e+04	18.39	18.37	18.37	15.01	15.00	15.00	43.70	43.59	43.59	κ	Algori	
500 500	38 39	40 41	55 55	its	thm Ia,		500	500	500	41	40	40	42	42	42	59	59	59	its	thm Ia,	
13734 13734 7194	13437 13437 6954	12768 12768 6675	10677 10677 5673	Π	Algorithm Ia, Ib, and Ic	BDD	7194	13734	13734	6954	13437	13437	6675	12768	12768	5673	10677	10677	Π	Algorithm Ia, Ib, and Ic	FETI-
$8(1.2\%) \\0(0\%) \\0(0\%)$	$egin{array}{c} 8 \ (1.2\%) \ 0 \ (0\%) \ 0 \ (0\%) \ 0 \ (0\%) \end{array}$	$egin{array}{c} 8 \ (1.2\%) \ 0 \ (0\%) \ 0 \ (0\%) \ 0 \ (0\%) \end{array}$	$egin{array}{c} 8 \ (1.2\%) \ 0 \ (0\%) \ 0 \ (0\%) \ \end{array}$	$\# \mathcal{E}_{evp}$		BDDC with LOBPCG	0(0%)	0(0%)	8 (1.2%)	0(0%)	0(0%)	8 (1.2%)	0(0%)	0(0%)	8 (1.2%)	0(0%)	0(0%)	8 (1.2%)	$\# \mathcal{E}_{evp}$		FETI-DP with LOBPCG
4.82e+04	18.43	15.05	43.72	к	Algo	BPCG		4.77e+04			18.37			15.00			43.59		κ	Algo	
500	38	40	55	its	Algorithm II			500			40			42			59		its	Algorithm II	
13734	13437	12768	10677	Π	1			13734			13437			12768			10677		Π		
6.55e+05	6.83e+05	6.84e+05	6.84e+05	κ	Algo			6.16e+05			6.81e+05			6.82e+05			6.82e+05		к	Algo	
500	208	210	282	its	Algorithm III			500			184			189			269		its	Algorithm III	
3105	2928	2781	2259	11				3105			2928			2781			2259				

1	
ll generalized	
alized eigenvalue probl	TABLE 5.5
problems. Soi	

Compressible linear elasticity. Coarse spaces for TOL = 10 for all indicated number of maximum iterations. lution of the local eigenvalue problems by LOBPCG with blocksize 10 and

Composite material, irregular partitioning, $N = 5^3$ and 1/h = 75.

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TABLE 5.6

Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for TOL = 10 for all generalized eigenvalue problems.

			Con	posi	te materi	al, irregular pa	rtitioning, a	nd 1/	$h = 6N^2$	1/3.		
						TI-DP Scaling						
			Alge	orithn	n Ia, Ib, a		Algo	rithm	II	Algor	ithm II.	1
N	$ \Pi' $		κ	its	$ \Pi $	$\# \mathcal{E}_{evp}$	κ	its	$ \Pi $	κ	its	$ \Pi $
						ho-scali	ing					
		a)	8.65	34	699	2 (2.0%)						
3^3	960	b)	8.65	34	699	1 (1.0%)	8.65	34	699	1.36e+06	68	243
		c)	8.67	34	450	1 (1.0%)						
- 2		a)	9.16	35	3675	25 (4.2%)						
5^{3}	5433	b)	9.16	35	3675	12 (2.1%)	9.16	35	3669	5.50e+05	190	1242
		c)	10.49 10.76	36 37	2325	$\frac{12(2.1\%)}{65(2.6\%)}$						
7^{3}	16248	a) b)	10.76	37	10101	65 (3.6%) 27 (1.5%)	10.76	37	10089	1.21e+06	424	3606
1	10240	c)	13.36	39	6693	27 (1.5%)	10.70	57	10009	1.210+00	424	5000
		a)	10.13	36	19632	144 (3.6%)						
9^{3}	35838	b)	10.13	36	19632	52 (1.3%)	10.13	36	19626	7.77e+05	500	7053
U	55050	c)	12.85	39	12921	52 (1.3%)	10.10	50	19020	1.110105	500	1055
		-/				deluxe-so	caling					1
		a)	7.51	20	603	2 (2.0%)						
3^{3}	960	b)	7.51	20	603	1 (1.0%)	7.51	20	603	7.52	27	207
		c)	7.51	20	393	1 (1.0%)						
		a)	9.61	29	3129	25 (4.2%)						
5^3	5433	b)	9.61	29	3129	12 (2.1%)	9.61	29	3126	9.98e+03	77	1002
		c)	9.63	29	2004	12 (2.1%)						
_		a)	7.69	30	8721	65 (3.6%)						
7^{3}	16248	b)	7.69	30	8721	27 (1.5%)	7.69	30	8709	3.04e+04	178	2976
		c)	7.70	30	5859	27 (1.5%)						
. 9		a)	10.76	34	16671	144 (3.6%)						
9^{3}	35838	b)	10.76	34	16671	52 (1.3%)	10.76	34	16656	8.57e+04	221	5718
		c)	10.77	34	11022	52 (1.3%)						
			7.04	26	(54	stiffness-s	scaling				1	
3^3	960	a) b)	7.84 7.84	26 26	654 654	2(2.0%)	7.84	26	654	7.220104	55	228
3~	960	b)	7.84 7.85	20 27	654 423	1(1.0%)	/.84	20	034	7.23e+04	55	228
		c) a)	11.16	32	3393	$\frac{1(1.0\%)}{25(4.2\%)}$						
5^3	5433	a) b)	11.10	32	3393	12(2.1%)	11.16	32	3390	3.45e+04	148	1107
0	5455	c)	11.10	32	2151	12(2.1%) 12(2.1%)	11.10	52	5570	5.450104	140	1107
		a)	9.12	33	9255	65 (3.6%)						
7^{3}	16248	b)	9.12	33	9255	27 (1.5%)	9.12	33	9240	9.74e+04	342	3174
		c)	9.15	34	6132	27 (1.5%)						
		a)	9.92	34	17718	144 (3.6%)						
9^3	35838	b)	9.92	34	17718	52 (1.3%)	9.92	34	17712	1.04e+05	395	6138
		c)	9.93	34	11583	52 (1.3%)						
						multiplicity	-scaling					
		a)	8.63	33	1029	2 (2.0%)						
3^{3}	960	b)	8.63	33	1029	1 (1.01%)	5.51e+05	54	1026	1.36e+06	345	426
		c)	8.66	35	696	1 (1.01%)						
		a)	9.10	35	5172	25 (4.2%)						
5^{3}	5433	b)	9.10	35	5172	12 (2.07%)	9.10	35	5169	1.62e+06	500	2115
		c)	10.46	36	3420	12 (2.07%)						<u> </u>
7^3	16240	a) b)	10.73	37	14625	65(3.6%)	10.72	27	14610	1.72 06	500	6192
10	16248	b)	10.73	37	14625 10023	27 (1.54%) 27 (1.54%)	10.73	37	14619	1.72e+06	500	6183
		c) a)	13.34 10.09	39 36	29598	27 (1.54%) 144 (3.6%)						
9^{3}	35838	a) b)	10.09	36	29598	52 (1.3%)	10.09	36	29592	1.58e+06	500	12477
3	55656	c)	12.77	30 39	29398	52 (1.3%) 52 (1.3%)	10.07	50	27592	1.500-00	500	12+//
			14.11	57	20010	52 (1.570)	I				I	I

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TABLE 5.7

Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for TOL = 10 for all generalized eigenvalue problems.

Randomly distributed coefficients with 20% high and 80% low coefficients, irregular partitioning, $N = 4^3$, and 1/h = 32.

						ling comparis						
					n Ia, Ib, a			<i>p</i> and <i>rithm</i>		Algo	rithm III	r
run	'		κ κ	its	$\frac{ \Pi }{ \Pi }$	$\# \mathcal{E}_{evp}$	κ κ	its	Π Π	κ	its	Π
Tun	11		70	113	111	ρ -scali		ns	11	70	113	111
		a)	9.24	33	7350	8 (2.8%)					1	
1	2646	b)	9.24	33	7350	8 (2.8%)	12.21	34	7347	5.00e+05	500	3543
1	2040	c)	9.24	33	7350	8 (2.8%)	12.21	54	1341	5.000105	500	5545
		a)	9.40	33	7278	8 (2.8%)						
2	2646	b)	9.40	33	7278	8 (2.8%)	1.57e+05	52	7272	4.74e+05	500	3480
_		c)	9.40	33	7278	8 (2.8%)						
		a)	8.32	33	7320	8 (2.8%)						
3	2646	b)	8.32	33	7320	8 (2.8%)	6.52e+04	67	7308	4.72e+05	500	3525
		c)	8.32	33	7320	8 (2.8%)						
		a)	9.44	34	7230	8 (2.8%)						
4	2646	b)	9.44	34	7230	8 (2.8%)	9.44	34	7227	4.67e+05	500	3408
		c)	9.44	34	7230	8 (2.8%)						
		a)	9.26	33	7416	8 (2.8%)						
5	2646	b)	9.26	33	7416	8 (2.8%)	2.54e+05	73	7407	4.69e+05	500	3588
		c)	9.26	33	7416	8 (2.8%)						
		a)	9.73	34	7311	8 (2.8%)						
6	2646	b)	9.73	34	7311	8 (2.8%)	9.73	34	7308	5.80e+05	500	3477
		c)	9.73	34	7311	8 (2.8%)						
						deluxe-so	aling					
		a)	6.08	21	6174	8 (2.8%)						
1	2646	b)	6.08	21	6174	8 (2.8%)	7.09	22	6171	1.86e+05	358	2769
		c)	6.08	21	6174	8 (2.8%)						
		a)	5.65	22	5997	8 (2.8%)						
2	2646	b)	5.65	22	5997	8 (2.8%)	5.65	22	5997	6.90e+04	285	2643
		c)	5.65	22	5997	8 (2.8%)						
2	2646	a)	4.89	23	6069	8 (2.8%) 8 (2.8%)	2.12 .04	10	(057	0.05	422	0702
3	2646	b)	4.89 4.89	23 23	6069 6069	8 (2.8%) 8 (2.8%)	3.12e+04	42	6057	2.35e+05	433	2703
		c)				8 (2.8%)						
4	2646	a) b)	5.94 5.94	24 24	5979 5979	8 (2.8%) 8 (2.8%)	5.94	24	5976	2.07e+05	336	2601
4	2040	c)	5.94 5.94	24	5979 5979	8 (2.8%) 8 (2.8%)	3.74	24	3970	2.070+05	330	2001
		a)	4.70	24	6207	8 (2.8%)						
5	2646	a) b)	4.70	$\frac{22}{22}$	6207 6207	8 (2.8%) 8 (2.8%)	3.14e+04	36	6201	1.37e+05	320	2799
5	2040	c)	4.70	$\frac{22}{22}$	6207 6207	8 (2.8%) 8 (2.8%)	5.140+04	50	0201	1.376703	520	2177
		a)	4.43	22	6249	8 (2.8%)						
6	2646	b)	4.43	22	6249	8 (2.8%) 8 (2.8%)	4.43	22	6246	1.95e+05	346	2841
Ŭ	2010	c)	4.43	22	6249	8 (2.8%)			0210	1.750105		2011
					5217	5 (2.070)	I I		I I	I	I.	

ensures convergence within less than 50 iterations, but it also uses a coarse space that is three times as large as that of TOL = 100 and nine to ten times as large as that of TOL = 1000. A trade-off between fast convergence and a manageable size of the coarse space is a problem-and facility-dependent task.

6. Parallel results. We present some preliminary CPU timings using a first parallel implementation of our adaptive FETI-DP Algorithms Ia and Ic compared to the standard nonadaptive FETI-DP with a full vertex and edge average coarse space. In all FETI-DP methods, we use stiffness scaling, that is, we use the diagonal entries of the stiffness matrices. We use PETSc [2, 3] with the PARDISO solver from the Intel MKL [23, 58] as a direct solver, and the SLEPc software library [22, 56] to solve the local generalized eigenvalue problems.

TABLE 5.8

Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for TOL = 10 for all generalized eigenvalue problems.

Randomly distributed coefficients with 20% high and 80% low coefficients, irregular partitioning, $N = 4^3$, and 1/h = 32.

			FETI			omparison; Pa				icity)		
			Alge	orithn	ı Ia, Ib, a	nd Ic	Algo	rithm I	11	Algor	ithm III	
run	$ \Pi' $		κ	its	$ \Pi $	$\# \mathcal{E}_{evp}$	κ	its	$ \Pi $	κ	its	Π
						stiffness-s	scaling				•	
		a)	7.82	29	6921	8 (2.8%)						
1	2646	b)	7.82	29	6921	8 (2.8%)	7.82	29	6918	2.15e+05	500	3177
		c)	7.82	29	6921	8 (2.8%)						
		a)	7.40	27	6876	8 (2.8%)						
2	2646	b)	7.40	27	6876	8 (2.8%)	2.86e+04	43	6870	3.05e+05	500	3171
		c)	7.40	27	6876	8 (2.8%)						
		a)	9.97	30	6903	8 (2.8%)						
3	2646	b)	9.97	30	6903	8 (2.8%)	6.47e+04	61	6891	2.72e+05	500	3195
		c)	9.97	30	6903	8 (2.8%)						
		a)	7.80	30	6738	8 (2.8%)						
4	2646	b)	7.80	30	6738	8 (2.8%)	7.80	30	6735	2.19e+05	500	3033
		c)	7.80	30	6738	8 (2.8%)						
		a)	8.49	29	6987	8 (2.8%)						
5	2646	b)	8.49	29	6987	8 (2.8%)	3.63e+04	60	6978	2.15e+05	500	3228
		c)	8.49	29	6987	8 (2.8%)						
		a)	7.32	29	6888	8 (2.8%)						
6	2646	b)	7.32	29	6888	8 (2.8%)	7.32	29	6885	3.04e+05	500	3144
		c)	7.32	29	6888	8 (2.8%)						
						multiplicity	-scaling					
		a)	8.48	32	12744	8 (2.8%)						
1	2646	b)	8.48	32	12744	8 (2.8%)	6.12e+05	116	12732	9.70e+05	500	8574
		c)	8.48	32	12744	8 (2.8%)						
		a)	9.02	33	12900	8 (2.8%)						
2	2646	b)	9.02	33	12900	8 (2.8%)	9.69e+05	130	12885	1.31e+06	500	8721
		c)	9.02	33	12900	8 (2.8%)						
		a)	9.42	33	13092	8 (2.8%)						
3	2646	b)	9.42	33	13092	8 (2.8%)	6.32e+04	64	13080	1.42e+06		8922
		c)	9.42	33	13092	8 (2.8%)						
		a)	9.13	33	13074	8 (2.8%)						
4	2646	b)	9.13	33	13074	8 (2.8%)	2.69e+05	140	13059	1.35e+06	500	8895
		c)	9.13	33	13074	8 (2.8%)						
_		a)	7.78	31	12972	8 (2.8%)						
5	2646	b)	7.78	31	12972	8 (2.8%)	2.48e+05	67	12963	1.85e+06	500	8805
		c)	7.78	31	12972	8 (2.8%)						
		a)	8.82	32	12894	8 (2.8%)						
6	2646	b)	8.82	32	12894	8 (2.8%)	3.18e+05	104	12882	1.11e+06	500	8718
		(c)	8.82	32	12894	8 (2.8%)						

As before, we consider the unit cube and the linear elastic composite material as depicted in Figure 5.1 (left) with a METIS decomposition. We enforce zero Dirichlet boundary conditions on the face with x = 0 and apply the body force $f = (0.1, 0.1, 0.1)^T$. We use conforming \mathcal{P}_2 finite elements on a mesh with H/h = 6.

For Algorithm Ia, the eigenvalue problems are solved by the SLEPc Krylov-Schur method with block size 10. In Algorithm Ic, we use only two iterations of the Krylov-Schur method with block size 10.

In our parallel implementation, we make the more realistic assumption that the coefficient evaluation function is not known. Thus, the strategy of Algorithm Ic, to discard eigenvalue problems and constraints, is based on the stiffness scaling from which we try to deduce coefficient jumps. We also use $TOL = 50 \log(H/h)$ to reduce the adaptive coarse space

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TABLE 5.9

Compressible linear elasticity for a heterogeneous material composed of seven different homogeneous materials with $E_1 = 1$ to $E_7 = 1e+06$. Coarse spaces for TOL = 10, TOL = 100, and TOL = 1000 for all generalized eigenvalue problems. Solution of the local eigenvalue problems by LOBPCG with blocksize 10 and a maximum of two iterations.

Randomly distributed coefficients	with seven	different value	es for the	Young modulus E ,
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 $E_1 = 1$ (30%), $E_2 = 10$ (20%), $E_i = 10^{i-1}$, i = 3, ..., 7 (10% each),

$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$						irreg. pa	artition., $N =$	4^3 , and $1/l$	i = 48				
$\begin{array}{c c c c c c c c c c c c c c c c c c c $				Alg	orithm	Ia, Ib, a		Algo	rithm I	11	Algor	ithm III	
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	run	$ \Pi' $		κ	its	$ \Pi $	$\# \mathcal{E}_{evp}$	κ	its	$ \Pi $	κ	its	$ \Pi $
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$							TOL=	10					
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$			a)	14.11	41	12084	2 (0.7%)						
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	1	2820	b)	14.11	41	12084	2 (0.7%)	14.11	41	12084	1.12e+05	500	4842
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$			c)	14.11	41	12084	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			a)	13.18	40	12186	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	2	2820	b)	13.18	40	12186	2 (0.7%)	13.18	40	12186	2.12e+05	500	4920
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			c)	13.18	40	12186	2 (0.7%)						
$\begin{array}{c c c c c c c c c c c c c c c c c c c $			a)	18.89	42	12147	2 (0.7%)						
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	3	2820	b)	18.89	42	12147	2 (0.7%)	18.89	42	12147	1.04e+05	500	4863
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			c)	18.89	42	12147	2 (0.7%)						
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$							TOL=1	.00					
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			a)	107.26	113	4278	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	1	2820	b)	107.26	113	4278	2 (0.7%)	107.26	113	4278	1.11e+05	500	987
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$			c)	107.26	113	4278	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			a)	100.32	110	4299	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	2	2820	b)	100.32	110	4299	2 (0.7%)	100.32	110	4299	2.11e+05	500	1041
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			c)	100.32	110	4299	2 (0.7%)						
$\begin{array}{c c c c c c c c c c c c c c c c c c c $			a)	115.62	114	4329	2 (0.7%)						
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	3	2820	b)	115.62	114	4329	2 (0.7%)	115.62	114	4329	1.04e+05	500	1041
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$			c)	115.62	114	4329	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$							TOL=1	000					
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			a)	970.55	321	1311	2 (0.7%)						
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	1	2820	b)	970.55	321	1311	2 (0.7%)	970.55	321	1311	1.11e+05	500	321
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			c)		321	1311	2 (0.7%)						
c) 993.08 320 1260 2 (0.7%) a) 1609.10 343 1158 2 (0.7%) 3 2820 b) 1609.10 343 1158 2 (0.7%)			a)	993.08	320	1260	2 (0.7%)						
a) 1609.10 343 1158 2 (0.7%) 3 2820 b) 1609.10 343 1158 2 (0.7%) 1609.10 343 1158 2 (0.7%) 1609.10 343 1158 500 273	2	2820	b)	993.08	320	1260	2 (0.7%)	993.08	320	1260	2.08e+05	500	318
3 2820 b) 1609.10 343 1158 2 (0.7%) 1609.10 343 1158 1.03e+05 500 273			c)	993.08	320	1260	2 (0.7%)						
			a)	1609.10	343	1158	2 (0.7%)						
c) 1609.10 343 1158 2 (0.7%)	3	2820	b)	1609.10	343	1158	2 (0.7%)	1609.10	343	1158	1.03e+05	500	273
			c)	1609.10	343	1158	2 (0.7%)					[

dimension; cf. Table 5.9 for a comparison of different tolerances. If the local iterative solver breaks down due to the high condition numbers of the local matrices, then we use the SLEPc interface to the LAPACK solver [1].

We also report the size of the coarse space (nonadaptive and adaptive) denoted by $|\Pi|$ and the number of nonzeros in the coarse space matrix in parentheses. The computation is conducted with one subdomain per core on the supercomputer magnitUDE at the Center for Computational Sciences and Simulation (CCSS) of the University of Duisburg-Essen. We use a preconditioned conjugate gradients method with up to 10,000 iterations and a maximum wall time of one hour. We report the number of Krylov iterations, the total runtime, and the KSP residual norm ||res|| at convergence or cancellation when the wall time is exceeded.

Table 6.1 shows that the standard nonadaptive method does not converge as required and is thus not suited for this problem. The adaptive methods reduce the run time significantly and also reduce the residual to the given tolerance. For this problem and the strategy of Algorithm Ic, we obtain comparable results for both adaptive methods.

TABLE 6.1 Compressible linear elasticity with $E_1 = 1$, $E_2 = 1e+06$. Coarse spaces for $TOL = 50 \log(H/h)$ for all generalized eigenvalue problems.

			irregular	partitioning, an	d H/h =	6.		
			N = 64			Ν	I = 216	
	its	res	runtime	$ \widehat{\Pi} $ (nnz)	its	res	runtime	$ \widehat{\Pi} $ (nnz)
Standard FETI-DP	>6373	5.22e-5	>60 min	2346 (859,212)	>6033	7.56e-4	>60 min	9828 (4,466,412)
Adaptive FETI-DP (Alg. Ia)	63	4.68e-11	23.4 min	3258 (1,379,072)	75	8.83e-11	37.9 min	13627 (7,393,571)
Adaptive FETI-DP (Alg. Ic)	63	4.30e-11	21.7 min	3248 (1,374,056)	76	5.85e-11	35.7 min	13615 (7,384,271)

Composite material					
			TT /	1.	1

7. Conclusion. In [28], an adaptive coarse space for FETI-DP domain decomposition methods applied to heterogeneous elliptic problems in three dimension was introduced. The method is based on numerically solving local eigenvalue problems on faces and edges of subdomains and on using these eigenvectors as deflation vectors. The condition number of the resulting preconditioner operator using deflation is bounded independently of the heterogeneity. Then, in [30], for heterogeneous problems and general scalings, a correspondence was shown between FETI-DP methods using deflation and FETI-DP and BDDC methods using a generalized transformation of basis combined with a partial finite element assembly.

In this current paper, we now combine the approaches in [28] and [30] and obtain FETI-DP and BDDC methods with a condition number bound independent of heterogeneities but using a generalized transformation of basis instead of deflation. For the new approach, it will be easier to extend the parallel scalability to a large number of subdomains on large supercomputers, also for heterogeneous problems, by solving the coarse problem inexactly. This is not possible in projection approaches like deflation or balancing.

We also present comparisons of the adaptive method with different scalings such as ρ -, deluxe-, stiffness-, and multiplicity-scaling. For our test cases, we state that ρ -scaling just needs about 10% of additional constraints compared to deluxe-scaling. The findings of stiffness-scaling are comparable to those of ρ -scaling. Multiplicity-scaling on the other hand gives significantly larger coarse spaces.

We additionally show that, also for hard problems including those with random coefficients with seven different materials, only two iterations of an iterative solver on the local eigenvalue problems can be sufficient to obtain fast convergence of the overall method.

Eventually, we present some preliminary results for parallel adaptive FETI-DP to show that Algorithms Ia and Ic both excel the standard nonadaptive approach with respect to run time and residual norm reduction.

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