COMPATIBLE GAUGE APPROACHES FOR $H(\text{div})$ EQUATIONS

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Abstract. We are concerned with the compatible gauge reformulation for $H(\text{div})$ equations and the design of fast solvers of the resulting linear algebraic systems as in [5]. We propose an algebraic reformulation of the discrete $H(\text{div})$ equations along with an algebraic multigrid (AMG) technique for the reformulated problem. The reformulation uses discrete Hodge decompositions on co-chains to replace the discrete $H(\text{div})$ equations by an equivalent $2 \times 2$ block linear system whose diagonal blocks are discrete Hodge Laplace operators acting on 2-cochains and 1-cochains respectively. We illustrate the new technique, using the lowest order Raviart-Thomas elements on structured tetrahedral mesh in three dimension, and present computational results.

1. Introduction. In this paper, we consider general second order elliptic operators over the Lipschitz polyhedral domain $\Omega$ in 3D. Specifically, let $\Omega$ be a bounded, simply connected, and contractible domain in $\mathbb{R}^3$ with Lipschitz boundary $\partial \Omega$. We are looking at the compatible discretization of the following model equation:

$$\begin{cases}
-\nabla(\lambda \nabla \cdot u) + \frac{1}{\mu} u &= f & \text{in } \Omega, \\
\frac{1}{\mu} u \cdot n &= 0 & \text{on } \Gamma, \\
\lambda \nabla \cdot u &= 0 & \text{on } \Gamma^*,
\end{cases} \quad (1.1)$$

where $\partial \Omega = \Gamma \cup \Gamma^*$ and $\Gamma \cap \Gamma^* = \emptyset$. Here, we assume that $\lambda$ and $\mu$ are positive throughout the domain, but may possibly vary widely.

The variational formulation of problem (1.1) leads naturally to the Hilbert space $H(\text{div})$ given by

$$H(\text{div}) := \left\{ u \in (L^2(\Omega))^3 \mid \nabla \cdot u \in L^2(\Omega) \right\}.$$

It is ubiquitous in problems arising in fluid and solid mechanics [6, 10]. It occurs, in particular, in the solution of second order elliptic partial differential equations (PDE) by first order least-squares methods or by mixed methods with augmented Lagrangians, see [1, 11, 18, 19] and the references cited therein. The importance of $H(\text{div})$-related problems has prompted vigorous research into efficient multilevel schemes, see [1, 11, 12, 18, 19].

The method to be developed in the current paper follows closely the idea of the recent work of Bochev, Hu, Siefert and Tuminaro 2007 [5] for Maxwell’s equations. Specifically, we propose an algebraic reformulation of the discrete $H(\text{div})$ equations along with a new AMG technique for this reformulated problem. The reformulation process take advantage of a discrete Hodge decomposition on co-chains to replace the discrete $H(\text{div})$ equations by an equivalent $2 \times 2$ block linear system whose diagonal blocks are discrete Hodge Laplace operators acting on 2-cochains and 1-cochains, respectively. The new AMG algorithm in this paper makes use of the Hiptmair smoother ([11]) on the fine mesh, uses the canonical interpolations $\Pi^\text{div}_h$ and $\Pi^\text{curl}_h$ on $H(\text{div})$ and $H(\text{curl})$ to construct the grid-transfer operators, and then uses the standard AMG methods for Laplace-type problems on the coarse meshes.

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The rest of the paper is organized as follows. Section 2 reviews basic facts about the discretization framework used in the paper. In Section 3, we apply this framework to obtain a compatible discretization for the $H(\text{div})$ equations and its equivalent reformulation. AMG solvers for the reformulated system are developed in Section 4. In Section 5 we present computational result in three dimension that illustrate the new technique in the context of smoothed aggregation of AMG. In all experiments we use finite element discretizations based on the lowest order Raviart-Thomas element and lowest order Nédélec element on structured tetrahedral elements.

2. Compatible discretization framework. In this section, we give a short introduction of a general framework for compatible discretizations developed in [3]. This framework is based on algebraic topology and includes certain finite element [4, 17], finite volume [15], and finite difference [16] schemes as particular cases. As a result, the AMG algorithm developed in this paper is readily applicable to discrete problems generated by any of these schemes. The presentation here is almost the same as [5, Section 3]. We include this section just for the sake of completeness.

2.1. Computational grid. We consider computational grids $\Omega^h$ consisting of 0-cells (nodes), 1-cells (edges), 2-cells (faces), and 3-cells (volumes). Formal linear combinations of $k$-cells are called $k$-chains [8]. The sets of $k$-chains forming $\Omega^h$ are denoted by $C_k$. We will assume that $\Omega^h$ is such that the collection $\{C_0, C_1, C_2, C_3\}$ is a complex, i.e., for any $c \in C_k$, $\partial_k c \in C_{k-1}$, where $\partial_k : C_k \mapsto C_{k-1}$ is the boundary operator on $k$-chains [7]. Together with the identity $\partial_k \partial_{k+1} = 0$ this gives rise to the exact sequence

$$0 \leftarrow C_0 \xleftarrow{\partial_1} C_1 \xleftarrow{\partial_2} C_2 \xleftarrow{\partial_3} C_3 \leftarrow 0. \tag{2.1}$$

The dual of $C_k$ is denoted by $C^k$ and its members are called $k$-cochains [8]. While $C_k$ and $C^k$ are isomorphic, they have different meanings in our discretization framework. The sets $C_k$ represent the physical objects that form the grid, while $C^k$ are collections of real numbers associated with the grid objects. For example, $c_1 \in C_1$ is a formal sum of (oriented) grid edges, while its isomorphic image $c^1 \in C^1$ is a set of real numbers assigned to the edges of $c_1$.

Therefore, the elements of $C^0$ provide values associated with the 0-cells (grid nodes); the elements of $C^1$ are values associated with the 1-cells (grid edges); $C^2$ contains values assigned to the 2-cells (grid faces) of the grid, and $C^3$ are the values assigned to the 3-cells (grid volumes). We will use $C^0$ and $C^3$ to approximate scalar functions and $C^1$ and $C^2$ to approximate vector functions.

The symbols $C^k_{\Gamma}$ will denote the subspaces of $C^k$ constrained by zero on the Dirichlet boundary $\Gamma$ for $k = 0, 1, 2$. Such spaces are needed to approximate scalar and vector functions subject to appropriate boundary conditions.

2.2. Natural operators. Let $\langle \cdot, \cdot \rangle$ denote the duality pairing of $C_k$ and $C^k$. The adjoint of $\partial_k$, defined by $\langle a, \partial_k c \rangle = \langle \partial_k a, c \rangle$, induces an operator $\delta_k : C^k_{\Gamma} \mapsto C^{k+1}_{\Gamma}$ called coboundary. This operator satisfies $\delta_k \delta_{k+1} = 0$ and gives rise to the exact sequence

$$0 \leftarrow C^0_{\Gamma} \xleftarrow{\delta_1} C^1_{\Gamma} \xleftarrow{\delta_2} C^2_{\Gamma} \xleftarrow{\delta_3} C^3_{\Gamma} \leftarrow 0. \tag{2.2}$$

1Clearly, $C^k$ are isomorphic to $\mathbb{R}^\hat{k}$, where $\hat{k} = \dim C^k$. For simplicity, the isomorphic image of the cochain $c^k \in C^k$ in $\mathbb{R}^\hat{k}$ will be denoted by the same symbol.

2For example, $C^0_{\Gamma}$ approximates scalar functions such that $\phi = 0$ on $\Gamma$; $C^1_{\Gamma}$ can be used to approximate vector fields $E$ such that $n \times E = 0$ on $\Gamma$. The space $C^2_{\Gamma}$ is appropriate for vector fields $B$ that have a vanishing normal component on $\Gamma$. 


sequence
\[ \mathbb{R} \rightarrow C^0 \xrightarrow{\delta_0} C^1 \xrightarrow{\delta_1} C^2 \xrightarrow{\delta_2} C^3 \rightarrow 0. \]  
(2.2)

It is not hard to see that the matrix representation \(D_k\) of \(\delta_k\) is the signed incidence matrix between \(C^k\) and \(C^{k+1}\). Following [14] we call \(D_0, D_1,\) and \(D_2\) natural approximations of the gradient, curl and divergence operators. Note that from \(\delta_{k+1}\delta_k = 0\) it follows that
\[ D_{k+1}D_k = 0; \quad k = 0, 1, 2, \]  
(2.3)
and so our natural operators mimic the well-known vector calculus identities \(\nabla \times \nabla = 0,\) and \(\nabla \cdot \nabla \times = 0\). In [13], it is pointed out that natural operations are not enough to provide compatible discretizations of the basic second order operators because their ranges and domains do not match. For example, we cannot approximate \(\nabla \times \nabla \times\) by \(D_1D_1\) because \(D_1\) is in general a rectangular matrix. The number of its columns and rows equals the number of 1-cells and 2-cells in the grid, which are not the same.

2.3. Metric structures and derived operators. Let \(M_k : C_k^f \rightarrow C_{k+1}^f; \quad k = 0, 1, 2, 3\) denote symmetric positive definite matrices. The matrix \(M_k\) endows \(C_k^f\) with an inner product structure,
\[(a^k, b^k)_{C^k} = (a^k)^T M_k b^k. \]  
(2.4)
The matrices \(M_0\) and \(M_3\) approximate weighted \(L^2\) inner products of scalar functions:
\[ M_0 \rightarrow \int_\Omega \gamma p \hat{p} d\Omega; \quad M_3 \rightarrow \int_\Omega \lambda \phi \hat{\phi} d\Omega, \]
while \(M_1\) and \(M_2\) approximate the weighted \(L^2\) inner products of vector functions
\[ M_1 \rightarrow \int_\Omega \sigma \hat{E} E d\Omega; \quad M_2 \rightarrow \int_\Omega \mu^{-1} \hat{B} B d\Omega. \]
We will also use the notation \(M_0(\gamma), M_1(\sigma), M_2(\mu^{-1})\) and \(M_3(\lambda)\) to show the dependency of the coefficients of these mass matrices explicitly.

We define the derived operator \(D_k^* : C_{k+1}^f \rightarrow C_k^f\) as the adjoint of \(D_k\) with respect to the inner product (2.4):
\[(D^*_k a^{k+1}, b^k)_{C^k} = (a^{k+1}, D_k b^k)_{C_{k+1}}. \]  
(2.5)
From (2.5) it is easy to see that for \(k = 0, 1, 2\)
\[ D_k^* = M_k^{-1} D_k^T M_{k+1}. \]  
(2.6)
The matrices \(D_2^*\), \(D_1^*\) and \(D_0^*\) provide a second set of discrete differential operators. Specifically, they are approximations of scaled gradient, curl and divergence operators
\[ D_2^* \rightarrow -\mu \nabla \lambda; \quad D_1^* \rightarrow \sigma^{-1} \nabla \times \mu^{-1}; \quad D_0^* \rightarrow -\gamma^{-1} \nabla \cdot \sigma, \]
augmented with the boundary conditions
\[ \lambda \phi = 0; \quad n \times \mu^{-1} B = 0; \quad \text{and} \quad n \cdot \sigma E = 0 \quad \text{on } \Gamma^*, \]
Compatible Gauge Approaches for $H(\text{div})$

$$\mathbb{R} \xrightarrow{I} H^1(\Omega) \xrightarrow{\nabla} H(\text{curl}) \xrightarrow{\nabla \times} H(\text{div}) \xrightarrow{\nabla \cdot} L^2(\Omega) \longrightarrow 0$$

$$\mathbb{R} \xrightarrow{I} V_h(\text{grad}) \xrightarrow{D_0} V_h(\text{curl}) \xrightarrow{D_1} V_h(\text{div}) \xrightarrow{D_2} V_h(0) \longrightarrow 0$$

$$\mathbb{R} \xleftarrow{I} V_h(\text{grad}) \xleftarrow{D_0^*} V_h(\text{curl}) \xleftarrow{D_1^*} V_h(\text{div}) \xleftarrow{D_2^*} V_h(0) \longleftarrow 0$$

**Fig. 2.1. De Rahm Complex and the lowest order finite element spaces**

respectively. Using (2.6) and (2.3)

$$D_k D_k^* = M_k^{-1} D_k^T M_k + M_k^{-1} D_k^T M_{k+1} = M_k^{-1} D_k^T D_{k+1} M_{k+2} = 0,$$

and so, the basic vector calculus identities hold for the derived operators as well. The commuting diagram, and the relationships among the operators defined above can be summarized in Figure 2.1. Here, the operators $\Pi^\text{grad}_h$, $\Pi^\text{curl}_h$, $\Pi^\text{div}_h$, and $\Pi^0_h$ are the canonical interpolations on $H^1(\Omega)$, $H(\text{curl})$, $H(\text{div})$, and $L^2(\Omega)$ to the corresponding finite element spaces $V_h(\text{grad})$, $V_h(\text{curl})$, $V_h(\text{div})$, and $V_h(0)$ respectively. The lower half of the commuting diagram above presents the relationships among the operators. For example, from this diagram we can easily find out that

$$D_2^* = M_2^{-1} D_2^T M_3.$$

Because the range of $D_k$ is contained in the domain of $D_k^*$ and vice versa we can use the natural and the derived operators to define discrete versions of the basic second order differential operators, including a discrete Hodge Laplace operator. Specifically, for $k = 0, 1, 2$ we have the second order operators

$$D_k^* D_k = M_k^{-1} D_k^T M_{k+1} M_{k+2} = M_k^{-1} D_k^T D_{k+1} M_{k+2} = 0,$$

and the discrete Hodge Laplacian

$$L_k : C^k_\Gamma \hookrightarrow C^k_{\Gamma + 1} \quad L_k = D_k^* D_k + D_{k-1}^* D_{k-1}$$

with the understanding that $D_0 = 0$ and $D_{-1}^* = 0$.

The discrete operators in (2.7)-(2.9) approximate basic second order elliptic differential operators. In §3.1 we will use these operators to motivate and explain our reformulation strategy.
Similar to [5], we also introduce a second set of inner products defined by the matrices \( \mathbb{M}_k \), \((k = 0, 1, 2, 3)\) that uses a unit weight, i.e.,

\[
\mathbb{M}_k \rightarrow \int_{\Omega} u^k v^k d\Omega, \quad u^k, v^k \in C^k_\Gamma.
\]

These inner products can be used to define a second set of derived operators \( \bar{D}^*_k : C^{k+1}_\Gamma \rightarrow \bar{D}^*_k \) given

\[
\bar{D}^*_k = \mathbb{M}_k^{-1} D_k^T \mathbb{M}_{k+1}, k = 0, 1, 2
\]

respectively, and such that \( \bar{D}^*_k D^*_k = 0 \). These operators give rise to the discrete Hodge Laplace operators

\[
\bar{L}_k : C^k_\Gamma \rightarrow C^k_\Gamma; \quad \bar{L}_k = \bar{D}^*_k \bar{D}^*_k;
\]

that are different versions of \( L_k \) respectively.

The following general result from [3] provides the results needed for the reformulation of the discrete \( H(\text{div}) \) equation.

**Theorem 2.1.** The size of the kernel of the analytic and discrete Hodge Laplacians is the same.

Theorem 2.1 reveals that the null-space of the discrete Hodge Laplacian and, by extension the structure of the discrete Hodge decomposition of discrete functions in \( C^k_\Gamma \), are topological invariants that are independent of the particular choice of metric, i.e., the matrices \( \mathbb{M}_k \). As a result, the assertion of this theorem is valid for both \( \mathbb{L}_0, \mathbb{L}_1, \mathbb{L}_2 \), and \( \bar{L}_0, \bar{L}_1, \bar{L}_2 \). The properties of these operators, relevant to the reformulation process, are summarized in the following corollary, which is a generalization of [5, Corollary 3.2].

**Corollary 2.2.** Assume that \( \Omega \) is contractible. Then, every \( u^k \in C^k_\Gamma (k = 1, 2) \) has the discrete Hodge decomposition

\[
u^k = D_{k-1} p^{k-1} + D^*_k b^{k+1}
\]

where \( p^{k-1} \in C^{k-1}_\Gamma \) and \( b^{k+1} \in C^{k+1}_\Gamma \) solve the equations

\[
D_{k-1} D_{k-1} p^{k-1} = D^*_k u^k \quad \text{and} \quad D_k D_k^* b^{k+1} = D_k u^k,
\]

respectively.

3. **Compatible discretization of \( H(\text{div}) \) equation.** Using the discrete operators defined in the last section, a compatible discretization of the \( H(\text{div}) \) equation (1.1) is straightforward. Specifically, we approximate \( u \) by a 2-cochain \( u^2 \in C^2_\Gamma \) that is associated with the 2-cells (the faces) of the mesh that are not in \( \Gamma \). Then the compatible discrete version of the \( \nabla \nabla \cdot \) operator is provided by the second order discrete operator \( D^2 \). As a result, the compatible, fully discrete equation of (1.1) is given by

\[
(D^2_\Gamma M_3 D_2 + M_2) u^2 = f^2,
\]

with the matrix \( M_3 \) containing the material parameter \( \lambda \) and the matrix \( M_2 \) containing \( \mu^{-1} \) and \( f^2 \in C^2_\Gamma \) is a discrete version of \( f \) in (1.1). An equivalent “weak” form of (3.1) is given by the variational equation: seek \( u^2 \in C^2_\Gamma \) such that

\[
(u^2, \hat{u}^2)_{C^2} + (D_2 u^2, D_2 \hat{u}^2)_{C^2} = (f^2, \hat{u}^2) \quad \forall \hat{u}^2 \in C^2_\Gamma.
\]

(3.2)
3.1. Reformulation. Following [5] for Maxwell’s equations, we intend on forming the Hodge Laplacian, which here corresponds to adding a $\nabla \times \nabla \times$ term, namely

$$L_2 = D_2^2 D_2 + D_1 D_1^T. \quad (3.3)$$

The following main theorem states an analogue of Theorem 4.2 in [5].

**Theorem 3.1.** Assume that $u^2$ is a solution of (3.1) and let

$$u^2 = D_1 e^1 + \tilde{D}_2 b^3 \quad (3.4)$$

denote its discrete Hodge decomposition with respect to the inner product induced by $M_2$. The pair $(a^2, e^1)$, where $a^2 = \tilde{D}_2 b^3$, solves the linear system

$$
\begin{bmatrix}
M_2 + D_2^T M_3 D_2 + \tilde{M}_2 D_1 M_1^{-1} D_1^T \tilde{M}_2 & M_2 D_1 \\
D_1^T M_2 & D_1^T M_2 D_1
\end{bmatrix}
\begin{bmatrix}
a^2 \\
e^1
\end{bmatrix}
= \begin{bmatrix}
M_2 f^2 \\
D_1^T M_2 f^2
\end{bmatrix}. \quad (3.5)
$$

**Proof.** Denoting $a^2 = \tilde{D}_2 b^3$, and applying the decomposition (3.4) to the weak form (3.2) gives

$$(D_1 e^1 + a^2, \hat{u}^2)_{C_1^2} + (D_2 a^2, D_2 \hat{u}^2)_{C_3} = (f^2, \hat{u}^2)_{C_1^2}, \forall \hat{u}^2 \in C_1^2.$$  

In the above equality, we used the fact that $D_2 D_1 \equiv 0$. We note that the assumed Hodge decomposition implies that $D_1^T a^2 = 0$ (since $D_1^T D_2 = 0$), thus

$$(\tilde{D}_1^T a^2, \hat{D}_1^T \hat{u}^2)_{C_1^2} = 0, \forall \hat{u}^2 \in C_1^2.$$  

As a result, this term can be added to the last equation to obtain:

$$(D_1 e^1 + a^2, \hat{u}^2)_{C_1^2} + (D_2 a^2, D_2 \hat{u}^2)_{C_3} + (\tilde{D}_1^T a^2, \tilde{D}_1^T \hat{u}^2)_{C_1^2} = (f^2, \hat{u}^2)_{C_1^2}, \forall \hat{u}^2 \in C_1^2.$$  

It is easy to see that the above weak form is equivalent to the following linear system:

$$M_2 a^2 + \left(D_2^T M_3 D_2 + \tilde{M}_2 D_1 M_1^{-1} D_1^T \tilde{M}_2\right) + M_2 D_1 e^1 = M_2 f^2$$

which is the first equation in (3.5).

Applying the decomposition (3.4) to (3.1), and then multiplying by $D_1^T$ on both sides gives

$$D_1^T a^2 + D_1^T e^1 = D_1^T f^2.$$  

Noticing that by definition $D_1^T = M_1^{-1} D_1^T M_2$, the second set of equations in the block system follows by multiplying $M_1$ on both sides. This completes the proof. □

Here, we should notice that the (2,2) block $D_1^T D_1$ is singular. A further decomposition [5, Corollary 3.2] of

$$e^1 = D_0 e^0 + \tilde{D}_1 b^2 := D_0 e^0 + a^1$$

yields the following block system

$$
\begin{bmatrix}
A_{11} & M_2 D_1 \\
D_1^T M_2 & A_{22}
\end{bmatrix}
\begin{bmatrix}
a^2 \\
a^1
\end{bmatrix}
= \begin{bmatrix}
M_2 f^2 \\
D_1^T M_2 f^2
\end{bmatrix}. \quad (3.6)
$$
where $A_{11} = M_2 + D_2^T M_3 D_2 + \tilde{M}_2 D_1 M_1^{-1} D_1^T \tilde{M}_2$ and $A_{22} = D_1^T M_2 D_1 + \tilde{M}_1 D_0 M_0^{-1} D_0 \tilde{M}_1$.

In the above formulation, we used the fact that $D_1 D_0 = 0$ and $D_0^* D_0^* = 0$.

**Remark 3.2.** The reformulation (3.6) seems more complicated than the original equation (3.1) that we are actually solving. The idea here is to try to use the diagonal blocks $A_{11}$ and $A_{22}$ as preconditioner of (3.1), which is the main focus of the next section.

It is interesting to notice that during this reformulation procedure, the gauge term in the $A_{11}$ and $A_{22}$ blocks seems to be indispensable. As was pointed out in [5], these terms play an important role in avoiding the large null-space caused by $\nabla \nabla \cdot$ operator and $\nabla \times \nabla \times$ operator respectively. While to form $\tilde{M}_2 D_1 M_1^{-1} D_1^T \tilde{M}_2$ and $\tilde{M}_1 D_0 M_0^{-1} D_0 \tilde{M}_1$ requires the inversion of $M_1$ and $M_0$. Even if we can use mass lumping to simplify the computation, it makes the system more complicated and ruins the sparse pattern of the original system. The interesting fact is that according to the numerical tests (see Section 5 for more details), it is not so clear now if these gauge terms are necessary or not. We need a more rigorous investigation of the roles of these gauge terms for more complex problems.

4. Multigrid solvers. Now we are in position to combine the reformulation and preconditioning to develop a linear solver for the compatible discretization (3.1) of the $H(\text{div})$ equation (1.1). Similar to the algorithm in [5], we focus on developing the AMG block preconditioners.

The approach considered in this paper focuses on developing AMG methods for the (1,1) and (2,2) blocks in (3.6) separately. Note that these diagonal blocks are Laplace-like. Once constructed, these AMG solvers are combined as a Jacobi-like preconditioner to precondition (1.1).

We propose an AMG technique for the whole $2 \times 2$ system which employs a Hiptmair smoother (see for example [11]) at the finest level, but allows subsequent levels and transfers of the (1,1) and (2,2) blocks to be handled with the standard AMG method. To do this, the face element of the (1,1) block and the edge element version of the (2,2) block must be converted to a more standard nodal form on the coarse mesh. This is accomplished by two special prolongators that not only transfers solutions from a coarse to a fine solution but also transfers solutions from a nodal to a face or edge representation, respectively. The net effect of these special prolongators is that the corresponding Galerkin projection of the (1,1) and (2,2) block will, in fact, yield a coarse operator resembling a vector nodal Laplacian which is amenable to any standard AMG method for further coarsening.

4.1. The specialized prolongators. As discussed earlier, in order to use the standard AMG solvers for the (1,1) and (2,2) block, we must convert the face element (for the (1,1)-block) and the edge element (for the (2,2)-block) into the standard nodal form. To do this, we define specialized prolongators $P_{11}$ and $P_{22}$ to transfer solutions from a nodal to a face and edge representation respectively. Instead of introducing the near null-space to define the prolongators as was done in [5], here we make use of the interpolation $\Pi_h^{\text{div}}$ and $\Pi_h^{\text{curl}}$ (see Figure 2.1) as in [2] and [12].

There are many ways to obtain aggregates corresponding to nodes, see [5] for more details. In this paper, for simplicity we use perfect aggregation. By “perfect”, we mean that the aggregates are formed manually. Note that we only need to form these aggregates on the finest level. Once the aggregates are formed, $\Pi_h^{\text{div}}$ and $\Pi_h^{\text{curl}}$ must also be computed. The detailed construction of the special prolongators for the (1,1) and (2,2) block is given in Algorithm 1. Notice that the net effect of $P_{11}$ is to
interpolate coarse nodal quantities to fine face-oriented quantities, and the effect of \( P_{22} \) is to interpolate coarse nodal quantities to fine edge-oriented quantities.

**Algorithm 1**: \([P_{11}, P_{22}] = \text{Coarse\_Node\_Prolongators()}\)

1. \( \{A_i\} \leftarrow \text{Aggregate manually} \)
2. For each fine node \( n_i \) and each aggregate \( A_j \) define
   \[
   (P_{nf})_{i,j} = \begin{cases} 
   1, & \text{if } n_i \in A_j \\
   0, & \text{otherwise} 
   \end{cases}
   \]
3. \( P_{11} = \Pi_h^{\text{div}} P_{nf} \)
4. \( P_{22} = \Pi_h^{\text{curl}} P_{nf} \)

The Galerkin coarse discretizations are given by
\[
A_{11}^H = P_{11}^T A_{11} P_{11}, \quad A_{22}^H = P_{22}^T A_{22} P_{22}
\]
where \( A_{11} \) and \( A_{22} \) are the (1,1) and (2,2) block of (3.6), \( A_{11}^H \) and \( A_{22}^H \) refer to their projections on a coarse mesh, respectively.

**4.2. Relaxation.** As before, we consider the following hybrid scheme. Suppose that the conjugate gradient iteration is actually applied to (3.1) and that (3.6) is only used within the preconditioner. To do this, it is necessary to convert residuals of (3.1) to right hand sides of (3.6) within the preconditioner. This is done by applying \([I \quad D_1]^T\) to the residual. Approximate solutions to (3.6) are then converted back to a form suitable for (3.1) via \( D_1 a^1 + a^2 \).

Algorithm 2 illustrates such a smoother proposed by Hiptmair that combines standard smoothing of the original equations with standard smoothing of the equations projected to the null-space [11].

**Algorithm 2**: \( \tilde{u} = \text{FineRelaxation}(A, D_1, \tilde{u}, b) \)

1. \( \tilde{u} \leftarrow \text{StandardRelaxation}(A, \tilde{u}, b) \)
2. \( c \leftarrow \text{StandardRelaxation}(D_1^T A D_1, 0, D_1^T (b - A \tilde{u}) ) \)
3. \( \tilde{u} \leftarrow \tilde{u} + D_1 c \)
4. \( \tilde{u} \leftarrow \text{StandardRelaxation}(A, \tilde{u}, b) \)

The key is that the error is smooth after this initial relaxation. Since the error is smooth, fine grid relaxation may be omitted from the AMG V-cycles in \( \text{Solve()} \), as (3.1) and (1.1) are equivalent.

It is important to realize that this special smoother is only needed on the finest level. A standard smoother can be used on coarse levels within the AMG procedures for the (1,1) and (2,2) blocks. Finally, an additive version of the Hiptmair smoother may also be considered for \( \text{FineRelaxation()} \).

**4.3. AMG algorithm preconditioner.** We now give the entire AMG-based preconditioner for the block Jacobi version in Algorithm 3. \( \text{PreFineRelaxation()} \) is identical to Algorithm 2 except step one is omitted. This also avoids the residual calculation in step two as the initial guess to a preconditioner is always zero. \( \text{PostFineRelaxation()} \) is identical to Algorithm 2 except step four is omitted to keep
the preconditioner symmetric when StandardRelaxation() employs a symmetric algorithm. Of course, residual calculations can also be avoided using additive forms of this smoother.

The algorithm essentially involves two AMG solves for nodal vector Laplacians: \( A_{11}^H \) corresponding to the (1,1) block and \( A_{22}^H \) corresponding to the (2,2) block. In addition, some relaxation must be performed on the original fine mesh system. Specifically, there are three major components of the preconditioner.

1. Hiptmair smoother for \( H(\text{div}) \) (see also Hiptmair [11]).
2. AMG for \( P_{11}^T A_{11} P_{11} \) within the (1,1)-block.
3. AMG for \( P_{22}^T A_{22} P_{22} \) within the (2,2)-block.

The detailed algorithm is listed as follows:

**Algorithm 3**: \( \tilde{u} = \text{Block Preconditioner}(r) \)

% Setup Phase

Form \( A_{11}^H \leftarrow P_{11}^T A_{11} P_{11} \) efficiently;
Standard_AMG_Setup(\( A_{11}^H \));
Form \( A_{22}^H \leftarrow P_{22}^T A_{22} P_{22} \) efficiently;
Standard_AMG_Setup(\( A_{22}^H \));

% Solve Phase

\( \tilde{u} \leftarrow \text{PreFineRelaxation}(D_2^T M_3 D_2 + M_2, D_1, 0, r) \);
\( \tilde{r} \leftarrow r - (D_2^T M_3 D_2 + M_2) \tilde{u} \);
% Perform V-cycles on \( A_{11}^H \) and \( A_{22}^H \)
\( a \leftarrow \text{Standard_AMG_Vcycle}(A_{11}^H, 0, P_{11}^T \tilde{r}) ; \)
\( p \leftarrow \text{Standard_AMG_Vcycle}(A_{22}^H, 0, P_{22}^T \tilde{r}) ; \)
\( \tilde{u} \leftarrow \tilde{u} + P_{11} a + D_1 P_{22} p ; \)
\( \tilde{u} \leftarrow \text{PostFineRelaxation}(D_2^T M_3 D_2 + M_2, D_1, \tilde{u}, r) ; \)

5. **Numerical results.** All the numerical experiments are conducted in a three-dimensional unit cube domain \( \Omega = \{(x, y, z) \in \mathbb{R}^3 : 0 \leq x, y, z \leq 1\} \) with homogeneous Neumann boundary condition. The domain is meshed by uniform cubes, and each cube is divided into 6 tetrahedra.

The proposed solver was implemented using CG in MATLAB. The first level and the first grid transfer of Algorithm 3 is also implemented in MATLAB. The ML’s smoothed aggregation solver is used for \( A_{11}^H \) and \( A_{22}^H \), through the mlmex MATLAB interface [9]. A single V-cycle of AMG is used for both the (1,1) and (2,2) block, using the efficient variant of Algorithm 2 (smoother). Unless otherwise stated, we use two steps of symmetric Gauss-Seidel sub-smoothing on both faces and edges. For all experiments the CG tolerance is \( 1 \times 10^{-10} \).

5.1. **Constant coefficients.** As the first experiment, we consider the constant coefficients case. We assume that \( \lambda = \mu = 1 \) in \( \Omega \). Table 5.1 reports the number of iterations with different meshsize. We note that the number of iterations are almost identical whether we include the gauge terms in the (1,1) and (2,2)-block or not. By this reason, we will omit the gauge term in the following numerical experiments.
Compatible Gauge Approaches for $H(\text{div})$

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Table 5.1

Number of iterations for CG-accelerated AMG on the 3D tetrahedral mesh problem with constant coefficients, using Algorithm 3. The size of the problem and the number of SGS smoothing steps are varied.

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Table 5.2

Number of iterations for CG-accelerated AMG on the 3D tetrahedral mesh problem with jump coefficients, using Algorithm 3. $\mu_0$ varies inside $[1/3, 2/3]$, and 1 elsewhere, and $\lambda \equiv 1$.

5.2. Variable $\mu$. We experiment with jumps in $\mu$ by considering two regions with constant values of $\mu$. Specifically, define

$$\Omega_0 = \left\{ (x, y, z) : \frac{1}{3} \leq x, y, z \leq \frac{2}{3} \right\}, \ \Omega_1 = \Omega \setminus \Omega_0;$$

let $\mu \equiv 1$ in $\Omega_1$ and choose $\mu = \mu_0$ to be a constant inside $\Omega_0$. $\lambda$ is fixed to be 1 throughout the whole domain $\Omega$. Table 5.2 reports the number of iterations on different meshsize. Note that the number of iterations are quite robust with respect to the variation of the coefficient $\mu$.

5.3. Variable $\lambda$. We now consider the jump on $\lambda$. Same as before, we choose $\lambda = \lambda_0$ to be a constant which varies from $10^{-4}$ to $10^4$ inside the domain $\Omega_0$, and $\lambda = 1$ elsewhere. This time, we fix $\mu$ to be 1 in the whole domain $\Omega$. Table 5.3 reports the number of iterations on different meshsize. Again, the number of iterations remains fairly constant.

6. Conclusions. In this paper, we proposed an AMG based preconditioner for the $H(\text{div})$ equation. We reformulated the equation by using the compatible gauge approaches, and formed a $2 \times 2$ system which is equivalent to the original discrete linear equations. Then we combined the AMG solvers for the (1,1) and (2,2) blocks of this system in certain way, and used it as the preconditioner of the original linear system. We also presented some numerical experiments to show the robustness of this algorithm. These experiments showed that the algorithm is very robust even with the presence of large jump coefficients.

REFERENCES
Table 5.3
Number of iterations for CG-accelerated AMG on the 3D tetrahedral mesh problem with jump coefficients, using Algorithm 3. $\lambda_0$ varies inside $[1/3,2/3]$, and 1 elsewhere, and $\mu \equiv 1$.

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[12] R. Hiptmair and J. Xu, Nodal auxiliary space preconditioning in $H$($\nabla$) and $H$($\nabla$)$\nabla$ spaces, tech. report, 2006.