FAST AUXILIARY SPACE PRECONDITIONER FOR LINEAR ELASTICITY IN MIXED FORM

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Abstract. A block diagonal preconditioner with the minimal residual method and a block triangular preconditioner with the generalized minimal residual method are developed for Hu-Zhang mixed finite element methods of linear elasticity. They are based on a new stability result of the saddle point system in mesh-dependent norms. The mesh-dependent norm for the stress corresponds to the mass matrix which is easy to invert while the displacement it is spectral equivalent to Schur complement. A fast auxiliary space preconditioner based on the $H^1$ conforming linear element of the linear elasticity problem is then designed for solving the Schur complement. For both diagonal and triangular preconditioners, it is proved that the conditioning numbers of the preconditioned systems are bounded above by a constant independent of both the crucial Lamé constant and the mesh-size. Numerical examples are presented to support theoretical results. As byproducts, a new stabilized low order mixed finite element method is proposed and analyzed and superconvergence results of Hu-Zhang element are obtained.

1. Introduction

We consider fast solvers for the Hu-Zhang mixed finite element methods [28, 29, 30] for linear elasticity, namely fast solvers for inverting the following saddle point system

$$(1.1) \quad \begin{pmatrix} M^\lambda_h & B^T_h \\ B_h & 0 \end{pmatrix},$$

where $M^\lambda_h$ is the mass matrix weighted by the compliance tensor and $B_h$ is the discretization of the div operator. The subscript $h$ is the mesh size of a underlying triangulation and the superscript $\lambda$ is the Lamé number which could be very large for nearly incompressible material. We aim to develop preconditioners robust to both $h$ and $\lambda$.

In [28, 29, 30], a stability result is established in the $H(\text{div}; \Omega) \times L^2(\Omega)$ norm whose matrix form is $(M_h + B^T_h M^{-1} w_h B_h) \times M_{u,h}$, where $M_{u,h}$ is the mass matrix for the displacement and $M_h$ is the abbreviation of $M_{h}^0$. By the theory developed

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by Mardal and Winther [41], the following block diagonal preconditioner leads to a parameter independent condition number of the preconditioned system

$$(1.2) \begin{pmatrix} M_h + B_h^T M^{-1}_{u,h} B_h & O \\ O & M^{-1}_{u,h} \end{pmatrix}.$$

To compute the first block of (1.2), however, a non-trivial solver should be designed to account for the discrete div operator.

Motivated by our recent work [21], we shall establish another stability result of (1.1) in mesh dependent norms $\| \cdot \|_{0,h} \times | \cdot |_{1,h}$ whose equivalent matrix form is $M_h \times B_h M^{-1}_h B_h^T$. Therefore we can use the block diagonal preconditioner

$$(1.3) \begin{pmatrix} M^{-1}_h & O \\ O & (B_h M^{-1}_h B_h^T)^{-1} \end{pmatrix}$$

together with the MINRES method to solve (1.1). The mass matrix $M_h$ can be further replaced by its diagonal matrix and thus a spectral equivalent approximation of $M^{-1}_h$ is easy to construct. The difficulty is the inverse of the Schur complement.

We shall develop a fast auxiliary space preconditioner for the Schur complement. The auxiliary space preconditioner was initially designed by Xu [52] to avoid the difficulty in creating a sequence of nonnested grids or nonnested finite element spaces. As a two level method, the auxiliary space preconditioner involves smoothing on the fine level space which is usually the to-be-solved finite element space, and a coarse grid correction on an auxiliary space which is much more flexible to choose.

It has been successfully applied to many finite element methods for partial differential equations, including conforming and nonconforming finite element method for the second order or fourth order problem [52, 53], $H$(curl) and $H$(div) problems [26, 33, 34, 47, 35], DG type discretizations [17, 23, 20, 39, 54], and general symmetric positive definite problems [36] etc.

We use the $H^1$ conforming linear finite element discretization on the same mesh for the linear elasticity equation with parameter $\lambda = 0$ as the auxiliary problem to preconditioning the Schur complement. Since $\lambda = 0$, we can solve the auxiliary problem by geometric multigrid methods for structured meshes and algebraic multigrid methods in general. Using the Korn’s inequality, we can further adopt the $H^1$ conforming linear finite element discretization for vector-type Poisson equation as the auxiliary problem.

Our stability result is robust to the parameter $\lambda$ and $h$, therefore the condition number of the preconditioned system is uniformly bounded with respect to both the size of the problem and the parameter $\lambda$. The later is notoriously difficult to construct for linear elasticity. Furthermore our results hold without the full regularity assumption.

We now give a brief literature review on robust multigrid methods for the linear elasticity problem. Discretization of the linear elasticity equations can be classified into three categories: displacement primary formulation, displacement-pressure mixed formulation and stress-displacement mixed formulation. Robust conforming and nonconforming multigrid methods for the primary formulation have been discussed in [44, 50, 38], and discontinuous Galerkin $H$(div)-conforming method in [27]. The W-cycle multigrid methods are the most studied multigrid methods for the displacement-pressure mixed formulation, which can be found in [37, 14] for conforming discretization and [11, 12] for nonconforming discretization. A V-cycle multigrid method for the finite difference discretization was developed in [55]. In
[6], the Taylor-Hood element method was reduced to the pressure Schur complement equation, based on which an inner/outer iteration scheme was set up. So far the solvers for the stress-displacement mixed formulation are mainly concentrated on the block diagonal preconditioned MINRES method, see [32, 49, 42]. In [32], the multigrid preconditioner was advanced for the PEERS element method with weakly symmetric stress. As for the Arnold-Winther element discretization, the overlapping Schwarz preconditioner was exploited in [49], and the variable V-cycle multigrid preconditioner was developed in [42]. The majority of existing works is to deal with the discrete null space ker(div) by using either block-wise Gauss-Seidels smoother or overlapping Schwarz smoothers. And only works in [38, 27, 14, 32, 42] do not rely on the $H^2$ regularity assumption. As we mentioned early our approach do not require a prior knowledge of the discrete ker(div). We transfer this difficulty to solve the Schur complement problem but with $\lambda = 0$, which only involves standard Poisson-type solvers. So it is much easier to implement and analyze.

To further improve the performance, we propose the following block-triangular preconditioner

$$\begin{pmatrix} I & D_h^{-1}B_h^T \\ 0 & 0 \end{pmatrix}^{-1} \begin{pmatrix} D_h & 0 \\ B_h & \tilde{S}_h \end{pmatrix}^{-1},$$

where $D_h$ is the diagonal of $M_h$ and $\tilde{S}_h = B_hD_h^{-1}B_h^T$ will be further preconditioned by the auxiliary space preconditioner we mentioned before. Numerical results in Section 6 show that the preconditioned GMRES converges around 40 steps to push the relative tolerance below $10^{-8}$.

Results in this paper can be also applied to other $H(\text{div})$ conforming and symmetric stress elements developed in [5, 2, 1, 3]. Indeed we present our results for both the original Hu-Zhang element $k \geq n + 1$ and a new stabilized version for $1 \leq k \leq n$.

The rest of this article is organized as follows. In Section 2, we present the mixed finite element methods for linear elasticity. In Section 3, we establish the stability based on the mesh dependent norms. Then we describe the block diagonal and triangular preconditioners in Section 4 and construct an auxiliary space preconditioner in Section 5. In Section 6, we give some numerical experiments to demonstrate the efficiency and robustness of our preconditioners. Throughout this paper, we use “$\lesssim \cdots$” to mean that “$\leq C \cdots$”, where $C$ is a generic positive constant independent of $h$ and the Lamé constant $\lambda$, which may take different values at different appearances.

2. Mixed Finite Element Methods

Assume that $\Omega \subset \mathbb{R}^n$ is a bounded polytope. Denote by $\mathbb{S}$ the space of all symmetric $n \times n$ tensors. Given a bounded domain $G \subset \mathbb{R}^n$ and a non-negative integer $m$, let $H^m(G)$ be the usual Sobolev space of functions on $G$, and $\mathbf{H}^m(G; \mathbb{X})$ be the usual Sobolev space of functions taking values in the finite-dimensional vector space $\mathbb{X}$ for $\mathbb{X} = \mathbb{S}$ or $\mathbb{R}^n$. The corresponding norm and semi-norm are denoted respectively by $\| \cdot \|_{m,G}$ and $| \cdot |_{m,G}$. Let $(\cdot, \cdot)_G$ be the standard inner product on $L^2(G)$ or $L^2(\mathbb{X}; G)$. If $G$ is $\Omega$, we abbreviate $\| \cdot \|_{m,G}$, $| \cdot |_{m,G}$ and $(\cdot, \cdot)_G$ by $\| \cdot \|_m$, $| \cdot |_m$ and $(\cdot, \cdot)$, respectively. Let $\mathbf{H}^m_0(G; \mathbb{R}^n)$ be the closure of $C_0^\infty(G; \mathbb{R}^n)$ with respect to the norm $\| \cdot \|_{m,G}$. Denote by $\mathbf{H}(\text{div}, G; \mathbb{S})$ the Sobolev space of square-integrable symmetric tensor fields with square-integrable divergence. For
any $\tau \in H(\text{div}, \Omega; \mathbb{S})$, we equip the following norm

$$
\|\tau\|_{H(\text{div})} := \left(\|\tau\|_0^2 + \|\text{div}\tau\|_0^2\right)^{1/2}.
$$

The Hellinger-Reissner mixed formulation of the linear elasticity under the load $f \in L^2(\Omega; \mathbb{R}^n)$ is given as follows: Find $(\sigma, u) \in \Sigma \times V := H(\text{div}, \Omega; \mathbb{S}) \times L^2(\Omega; \mathbb{R}^n)$ such that

\begin{align}
(2.1) & \quad a(\sigma, \tau) + b(\tau, u) = 0 \quad \forall \tau \in \Sigma, \\
(2.2) & \quad b(\sigma, v) = -(f, v) \quad \forall v \in V;
\end{align}

where

$$
a(\sigma, \tau) := (\mathfrak{A}\sigma, \tau), \quad b(\tau, v) := (\text{div}\tau, v)
$$

with $\mathfrak{A}$ being the compliance tensor of fourth order defined by

$$
\mathfrak{A}\sigma := \frac{1}{2\mu} \left(\sigma - \frac{\lambda}{n\lambda + 2\mu} (\text{tr}\sigma) \delta\right).
$$

Here $\delta := (\delta_{ij})_{n \times n}$ is the Kronecker tensor, tr is the trace operator, and positive constants $\lambda$ and $\mu$ are the Lamé constants.

Suppose the domain $\Omega$ is subdivided by a family of shape regular simplicial grids $T_h$ (cf. [15, 22]) with $h := \max_{K \in T_h} h_K$ and $h_K := \text{diam}(K)$. Let $F_h$ be the union of all $n - 1$ dimensional faces of $T_h$ and $F_h^i$ be the union of all $n - 1$ dimensional interior faces. For any $F \in F_h$, denote by $h_F$ its diameter and fix a unit normal vector $\nu_F$. Let $P_m(G)$ stand for the set of all polynomials in $G$ with the total degree no more than $m$, and $P_m(G; \mathbb{X})$ denote the tensor or vector version of $P_m(G)$ for $\mathbb{X}$ being $\mathbb{S}$ or $\mathbb{R}^n$, respectively.

Consider two adjacent simplices $K^+$ and $K^-$ sharing an interior face $F$. Denote by $\nu^+$ and $\nu^-$ the unit outward normals to the common face $F$ of the simplices $K^+$ and $K^-$, respectively. For a vector-valued function $w$, write $w^+ := w|_{K^+}$ and $w^- := w|_{K^-}$. Then define a jump as

$$
[w] := \begin{cases} 
  w^+(\nu^+ \cdot \nu_F) + w^-(\nu^- \cdot \nu_F), & \text{if } F \in F_h^i, \\
  w, & \text{if } F \in F_h \setminus F_h^i.
\end{cases}
$$

For each $K \in T_h$, define an $H(\text{div}, K; \mathbb{S})$ bubble function space of polynomials of degree $k$ as

$$
B_{K,k} := \{ \tau \in P_k(K; \mathbb{S}) : \tau \nu|_{\partial K} = 0 \}.
$$

It is easy to check that $B_{K,1}$ is merely the zero space. Denote the vertices of simplex $K$ by $x_{K,0}, \cdots, x_{K,n}$. If not causing confusion, we will abbreviate $x_{K,i}$ as $x_i$ for $i = 0, \cdots, n$. For any edge $x_i x_j (i \neq j)$ of element $K$, let $t_{i,j}$ be the associated unit tangent vectors and

$$
T_{i,j} := t_{i,j}t_{i,j}^T, \quad 0 \leq i < j \leq n.
$$

It has been proved in [28] that the $(n + 1)n/2$ symmetric tensors $T_{i,j}$ form a basis of $\mathbb{S}$, and for $k \geq 2$,

$$
B_{K,k} = \sum_{0 \leq i < j \leq n} \lambda_i \lambda_j P_{k-2}(K) T_{i,j},
$$

where $P_k(K)$ is the $k$-th order polynomial on $K$. For $k = 1$, we define $B_{K,1} := \{ 0 \}$. The following lemma provides an essential estimate for the bubble function spaces.

Lemma 2.1. There exists a constant $C > 0$ such that

$$
\|w\|_{H(\text{div}, K; \mathbb{S})} \leq C \|w\|_{H(\text{div}, \Omega; \mathbb{S})}.
$$

The proof of the above lemma is given in the appendix.
where $\lambda_i$ is the associated barycentric coordinates corresponding to $x_i$ for $i = 0, \cdots, n$. Some global finite element spaces are given by

$$B_{k,h} := \{ \tau \in H(\text{div}; \Omega; \mathbb{S}) : \tau|_K \in B_{K,k} \quad \forall K \in \mathcal{T}_h \},$$

$$\tilde{\Sigma}_{k,h} := \{ \tau \in H^1(\Omega; \mathbb{S}) : \tau|_K \in P_{k}(K; \mathbb{S}) \quad \forall K \in \mathcal{T}_h \},$$

$$\Sigma_h := \tilde{\Sigma}_{k,h} + B_{k,h},$$

$$V_h := \{ v \in L^2(\Omega; \mathbb{R}^n) : v|_K \in P_{k-1}(K; \mathbb{R}^n) \quad \forall K \in \mathcal{T}_h \},$$

with $\lambda_0$ with a slight abuse of notation, we use the same notation $\| \cdot \|_a$ in [28]:

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with $\epsilon(v) := (\nabla v + (\nabla v)^T) / 2$ being the linearized strain tensor.

With previous preparation, the mixed finite element method for linear elasticity proposed in [28, 29, 30, 19] is defined as follows: Find $(\sigma_h, u_h) \in \Sigma_h \times V_h$ such that

$$\begin{align}
(2.3) & \quad a(\sigma_h, \tau_h) + b(\tau_h, u_h) = 0 \quad \forall \tau_h \in \Sigma_h, \\
(2.4) & \quad b(\sigma_h, v_h) - c(u_h, v_h) = -(f, v_h) \quad \forall v_h \in V_h,
\end{align}$$

where

$$c(u_h, v_h) := \eta \sum_{F \in \mathcal{F}_h} h_F^{-1} \int_F [u_h] \cdot [v_h] \, ds,$$

$$\eta := \begin{cases} 0, & \text{if } k \geq n + 1, \\ 1, & \text{if } 1 \leq k \leq n. \end{cases}$$

The bilinear form $c(\cdot, \cdot)$ involving the jump of displacement is introduced to stabilize the discretization which is only necessary for low order polynomials, i.e., $1 \leq k \leq n$. Note that the scaling $h_F^{-1}$ is different with the one in [19].

Choosing appropriate bases of $\Sigma_h$ and $V_h$, we can write the matrix form of (2.3)-(2.4) as

$$\begin{pmatrix}
M^h & B^T_h \\
B_h & -C_h
\end{pmatrix}
\begin{pmatrix}
\sigma_h \\
u_h
\end{pmatrix}
= \begin{pmatrix}
f
\end{pmatrix}.$$
section we shall establish another stability result of \((2.5)\) in mesh dependent norms which leads to a new block-diagonal preconditioner.

### 3. Stability Based On Mesh Dependent Norms

To construct a new block diagonal preconditioner, we will show that the bilinear form \(A(\cdot, \cdot, \cdot)\) is stable on \(\Sigma_h \times V_h\) with mesh dependent norms.

For each \(K \in T_h\), denote by \(\nu_i\) the unit outward normal vector of the \(i\)-th face of element \(K\). For any \(\tau_h \in \Sigma_h\) and \(v_h \in V_h\), define

\[
\|\tau_h\|_{0,h}^2 := \|\tau_h\|^2 + \sum_{F \in F_h} h_F \|\tau_h \nu_F\|^2_{0,F},
\]

\[
|v_h|_{1,h}^2 := \sup_{\tau_h \in \Sigma_h} \frac{b(\tau_h,v_h)}{\|\tau_h\|_{0,h}}, \quad \forall v_h \in V_h.
\]

Here \(\varepsilon_h\) is element-wise symmetric gradient. We shall prove the stability of \((2.5)\) in the mesh dependent norms \(\|\cdot\|_{0,h} \times |\cdot|_{1,h}\). The key is the following inf-sup condition: for \(k \geq n + 1\)

\[
|v_h|_{1,h} \lesssim \sup_{\tau_h \in \Sigma_h} \frac{b(\tau_h,v_h)}{\|\tau_h\|_{0,h}}, \quad \forall v_h \in V_h.
\]

For low order cases \(1 \leq k \leq n\), in addition to a variant of the inf-sup condition, we also need a coercivity result in the null space of the div operator.

#### 3.1. Properties on mesh dependent norms

We first present a different basis of the symmetric tensor space \(S\). Inside a simplex formed by vertices \(x_0, \ldots, x_n\), we label the face opposite to \(x_i\) as the \(i\)-th face \(F_i\). For the edge \(x_ix_j\), \(i \neq j\), define

\[
N_{i,j} := \frac{1}{2(\nu_i^T t_{i,j}) (\nu_j^T t_{i,j})} (\nu_i \nu_j^T + \nu_j \nu_i^T), \quad 0 \leq i < j \leq n.
\]

Here recall that \(t_{i,j}\) is an unit tangent vector of edge \(x_ix_j\) and \(\nu_i\) is the unit outwards normal vector of face \(F_i\). Due to the shape regularity of the triangulation, it holds

\[
\nu_i^T t_{i,j} \approx 1, \quad 0 \leq i < j \leq n.
\]

By direct manipulation, we have the following results about \(T_{i,j}\) and \(N_{i,j}\):

\[
T_{i,j} : N_{k,l} = \varepsilon_{ik} \varepsilon_{jl}, \quad 0 \leq i < j \leq n, \quad 0 \leq k < l \leq n,
\]

\[
T_{i,j} : T_{i,j} = 1, \quad N_{i,j} : N_{i,j} \approx 1, \quad 0 \leq i < j \leq n.
\]

Thus the \((n+1)n/2\) symmetric tensors \(\{N_{i,j}\}\) also form a basis of \(S\) which is the dual to \(\{T_{i,j}\}\).

**Lemma 3.1.** For any \(q_{ij} \in L^2(K)\), \(0 \leq i < j \leq n\), let \(\tau_1 = \sum_{0 \leq i < j \leq n} q_{ij} T_{i,j}\) and \(\tau_2 = \sum_{0 \leq i < j \leq n} q_{ij} N_{i,j}\), then it holds

\[
\|\tau_1\|^2_{0,K} \approx \|\tau_2\|^2_{0,K} \approx \sum_{0 \leq i < j \leq n} \|q_{ij}\|^2_{0,K}.
\]
Proof. Using the Cauchy-Schwarz inequality and (3.3), we have
\[ \|\tau_1\|_{0,K}^2 \leq \frac{(n+1)n}{2} \sum_{0 \leq i < j \leq n} \|q_{ij}T_{i,j}\|_{0,K}^2 = \frac{(n+1)n}{2} \sum_{0 \leq i < j \leq n} \|q_{ij}\|_{0,K}^2, \]
\[ \|\tau_2\|_{0,K}^2 \leq \frac{(n+1)n}{2} \sum_{0 \leq i < j \leq n} \|q_{ij}N_{i,j}\|_{0,K}^2 \lesssim \frac{(n+1)n}{2} \sum_{0 \leq i < j \leq n} \|q_{ij}\|_{0,K}^2. \]

On the other side, it follows from Cauchy-Schwarz inequality and (3.2),
\[ \sum_{0 \leq i < j \leq n} \|q_{ij}\|_{0,K}^2 = \sum_{0 \leq i < j \leq n} \int_K q_{ij}^2 \, dx = \sum_{0 \leq i < j \leq n} \sum_{0 \leq k < l \leq n} \int_K q_{ij} q_{kl} \delta_{ik} \delta_{jl} \, dx 
= \sum_{0 \leq i < j \leq n} \sum_{0 \leq k < l \leq n} \int_K q_{ij} T_{i,j} : q_{kl} N_{k,l} \, dx 
= \int_K \tau_1 : \tau_2 \, dx \leq \|\tau_1\|_{0,K} \|\tau_2\|_{0,K}. \]

Hence we conclude the result by combining the last three inequalities. \( \square \)

We then embed \( \varepsilon_h(\mathbf{V}_h) \) into the \( \mathbf{H}^1(\mathbf{div} \times K; \mathbb{S}) \) bubble function space. For each element \( K \in \mathcal{T}_h \), introduce a bijective connection operator \( E_K : P_{k-2}(K; \mathbb{S}) \rightarrow B_{K,k} \) with \( k \geq 2 \) as follows: for any \( \tau = \sum_{0 \leq i < j \leq n} q_{ij} N_{i,j} \) with \( q_{ij} \in P_{k-2}(K) \), \( 0 \leq i < j \leq n \), define
\[ E_K \tau := \sum_{0 \leq i < j \leq n} \lambda_i \lambda_j q_{ij} T_{i,j}. \]

Applying Lemma 3.1 and the scaling argument, we get for any \( \tau \in P_{k-2}(K; \mathbb{S}) \)
\[ \|E_K \tau\|_{0,K}^2 \approx \sum_{0 \leq i < j \leq n} \|\lambda_i \lambda_j q_{ij}\|_{0,K}^2 \approx \sum_{0 \leq i < j \leq n} \|q_{ij}\|_{0,K}^2 \approx \|\tau\|_{0,K}^2. \]

Denote by \( E \) the elementwise global version of \( E_K \), i.e. \( E|_K := E_K \) for each \( K \in \mathcal{T}_h \).

Third, we give an equivalent formulation of the mesh dependent norm \( | \cdot |_{1,h} \). For each \( F \in \mathcal{F}_h \), denote by \( \pi_F \) the orthogonal projection operator from \( L^2(F; \mathbb{R}^n) \) onto \( P_1(F; \mathbb{R}^n) \). Define the broken \( H^1 \) space as
\[ H^1(\mathcal{T}_h; \mathbb{R}^n) := \{ v \in L^2(\Omega; \mathbb{R}^n) : v|_K \in H^1(K; \mathbb{R}^n) \quad \forall K \in \mathcal{T}_h \}. \]

The domain of mesh dependent norm \( | \cdot |_{1,h} \) can be extended from \( \mathbf{V}_h \) to \( H^1(\mathcal{T}_h; \mathbb{R}^n) \).

Lemma 3.2. We have the norm equivalence:
\[ |v|_{1,h}^2 \approx \|\varepsilon_h(v)\|_0^2 + \sum_{F \in \mathcal{F}_h} h_F^{-1}\|\pi_F[v]\|_{0,F}^2 \quad \forall v \in H^1(\mathcal{T}_h; \mathbb{R}^n). \]
For any element $K \in \mathcal{T}_h$, let $\pi_K$ be an interpolation operator from $H^1(K; \mathbb{R}^n)$ onto $R(K)$ defined by (3.1)-(3.2) in [13]. And let $\pi$ be the elementwise global version of $\pi_K$, i.e. $\pi|_K = \pi_K$ for each $K \in \mathcal{T}_h$. It follows from (3.3)-(3.4) in [13] that for any $v \in H^1(T_h; \mathbb{R}^n)$,

$$
\sum_{F \in \mathcal{F}_h} h_F^{-1} \| \nabla v - \pi F[v] \|_{0,F}^2 = \sum_{F \in \mathcal{F}_h} h_F^{-1} \| v - \pi v - \pi F[v] - \pi F[v - \pi v] \|_{0,F}^2 
\leq \sum_{F \in \mathcal{F}_h} h_F^{-1} \| v - \pi v \|_{0,F}^2 \lesssim \| \varepsilon_h(v) \|_{0}^2.
$$

Then the equivalence (3.6) follows from the triangle inequality.

We shall also use the following discrete Korn’s inequality (cf. (1.22) in [13] and (34) in [4])

$$
\| \nabla h v \|_0^2 + \| v \|_0^2 \lesssim \| \varepsilon_h(v) \|_0^2 + \sum_{F \in \mathcal{F}_h} h_F^{-1} \| \pi F[v] \|_{0,F}^2 \quad \forall v \in H^1(T_h; \mathbb{R}^n).
$$

Together with (3.6), we conclude $| \cdot |_{1,h}$ defines a norm on $V_h$.

### 3.2. inf-sup condition in mesh dependent norms

The inf-sup condition we need is actually for the subspace $\hat{\Sigma}_h$ with vanished mean trace, c.f., (3.9) below. It is obvious that inf-sup condition (3.9) implies inf-sup condition (3.1). On the other hand, if inf-sup condition (3.1) is true, then (3.9) holds by taking $\hat{\tau}_h = \tau_h - \left( \frac{1}{|\Omega|} \int_{\Omega} \text{tr} \tau_h \, dx \right) \delta$. Therefore inf-sup conditions (3.1) and (3.9) are equivalent.

**Lemma 3.3.** For $k \geq n + 1$, we have the following inf-sup condition

$$
|v_h|_{1,h} \lesssim \sup_{\tau_h \in \Sigma_h} \frac{b(\tau_h, v_h)}{\| \tau_h \|_{0,h}}
$$

for any $v_h \in V_h$.

**Proof.** Given a $v_h \in V_h$, we shall construct a $\hat{\tau}_h \in \hat{\Sigma}_h$ to verify (3.9).

We first control the norm $|\varepsilon_h(v_h)|_0$. For any $v_h \in V_h$, take $\tau_1 = E \varepsilon_h(v_h)$. It follows from (3.4)

$$
\| \tau_1 \|_0 \propto \| \varepsilon_h(v_h) \|_0.
$$

According to integration by parts and (3.5), there exists a constant $C_1 > 0$ such that

$$
b(\tau_1, v_h) = \int_{\Omega} \tau_1 : \varepsilon_h(v_h) \, dx \geq C_1 \| \varepsilon_h(v_h) \|_0^2.
$$

Next we control the jump term. Choose $\tau_2 \in \Sigma_h$ such that all the degrees of freedom (cf. Lemma 2.1 in [19]) for $\tau_2$ vanish except the following one:

$$
\int_F (\tau_2 v_F) \cdot w \, ds = h_F^{-1} \int_F [v_h] \cdot w \, ds \quad \forall w \in P_1(F; \mathbb{R}^n) \quad \text{on each face } F.
$$

Then we have

$$
\int_K \tau_2 : \varepsilon_h(v_h) \, dx = 0, \quad \int_F (\tau_2 v_F) \cdot \pi F[v_h] \, ds = h_F^{-1} \| \pi F[v_h] \|_{0,F}^2,
$$

$$
\| \tau_2 \|_0^2 \lesssim \sum_{F \in T_h} h_F^{-1} \| \pi F[v_h] \|_{0,F}^2.
$$
Thus by (3.7) and (3.12), there exists a constant $C_2 > 0$ such that

$$
b(\tau_2, v_h) = \sum_{F \in \mathcal{F}_h} \int_F (\tau_2 v_F) \cdot [v_h] \, ds
$$

$$
= \sum_{F \in \mathcal{F}_h} \int_F (\tau_2 v_F) \cdot ([v_h] - \pi_F[v_h]) \, ds + \sum_{F \in \mathcal{F}_h} h_F^{-1} \|\pi_F[v_h]\|_{0,F}^2
$$


(3.13) \geq -C_2 \|v_h\|_0^2 + \frac{1}{2} \sum_{F \in \mathcal{F}_h} h_F^{-1} \|\pi_F[v_h]\|_{0,F}^2.

Now taking $\tau_h = \tau_1 + \frac{C_1}{2C_2} \tau_2$, it holds from (3.11) and (3.13)

$$
b(\tau_h, v_h) = b(\tau_1, v_h) + \frac{C_1}{2C_2} b(\tau_2, v_h)
$$

$$
\geq \frac{C_1}{2} \|v_h\|_0^2 + \frac{C_1}{4C_2} \sum_{F \in \mathcal{F}_h} h_F^{-1} \|\pi_F[v_h]\|_{0,F}^2.
$$

Thanks to (3.6), we get

$$
|v_h|_{1,h}^2 \lesssim b(\tau_h, v_h).
$$

On the other hand, it follows from the inverse inequality, (3.10) and (3.12)

$$
\|\tau_h\|_{0,h} \lesssim \|\tau_h\|_0 \lesssim |v_h|_{1,h}.
$$

Finally the inf-sup condition (3.1) is the result of the last two inequalities and consequently (3.9) holds by taking $\tau_h = \tau_h - \left(\frac{1}{n} \int_\Omega \text{tr} \tau_h \, dx\right) \delta$.

3.3. Coercivity in the null space of the div operator. Besides the inf-sup condition, another issue of the linear elasticity in the mixed form is the coercivity of bilinear form $a(\cdot, \cdot)$. On the whole space: for all $\sigma \in \Sigma$,

$$
a(\sigma, \sigma) \geq \frac{1}{n\lambda + 2\mu} \|\sigma\|_0^2.
$$

The coercivity constant, unfortunately, is in the order of $O(1/\lambda)$ as $\lambda \to +\infty$. Namely it is not robust to $\lambda$. To obtain a robust coercivity, we first recall the following inequality which implies the coercivity in the null space of the div operator.

Lemma 3.4 (Proposition 9.1.1 in [8]). For $\tau \in H(\text{div}, \Omega; \mathbb{S})$ satisfying $\int_\Omega \text{tr} \tau \, dx = 0$, we have

$$
\|\tau\|_0 \lesssim \|\tau\|_a + \|\text{div} \tau\|_{-1},
$$

where $\|\tau\|_a^2 := a(\tau, \tau)$ and $\|\text{div} \tau\|_{-1} = \sup_{v \in H^1_0(\Omega; \mathbb{R}^n)} b(\tau, v)/|v|_1$.

We then move to the discrete case. Define discrete norms

$$
\|\text{div} \tau\|_{-1,h} := \sup_{v_h \in V_h} \frac{b(\tau, v_h)}{|v_h|_{1,h}},
$$

$$
\|h \text{div} \tau\|_2^2 := \sum_{K \in \mathcal{T}_h} h_K^2 \|\text{div} \tau\|_{0,K}^2.
$$

Let $Q_h$ be the $L^2$ orthogonal projection from $L^2(\Omega; \mathbb{R}^n)$ onto $V_h$, which will be abbreviated as $Q_h$. It holds the following error estimate (cf. [22, 15])

$$
(3.15) \|v - Q_h v\|_{0,K} + h_K^{1/2} \|v - Q_h v\|_{0,\partial K} \lesssim h_K^\text{min\{k,m\}} |v|_{m,K} \quad \forall \ v \in H^m(\Omega; \mathbb{R}^n)
$$

with integer $m \geq 1.$
Lemma 3.5. For any $\tau \in H(\text{div}, \Omega; S)$ satisfying $\int_{\Omega} \text{tr}\tau \, dx = 0$, we have
\[ \|\tau\|_0 \lesssim \|\tau\|_a + \|h \text{div} \tau\| + \|\text{div} \tau\|_{-1,h}. \]

Proof. It is sufficient to prove the case $k = 1$. Let $v \in H^1_0(\Omega; \mathbb{R}^n)$, then it follows from the Cauchy-Schwarz inequality and (3.15)
\[ b(\tau, v) = b(\tau, v - Q_h v) + b(\tau, Q_h v) \lesssim \|h \text{div} \tau\|_1 + b(\tau, Q_h v). \]
Again by (3.15), it holds
\[ |Q_h v|_{1,h}^2 = \sum_{F \in \mathcal{F}_h} h_F^{-1} \|[Q_h v]\|_{0,F}^2 = \sum_{F \in \mathcal{F}_h} h_F^{-1} \|[Q_h v - v]\|_{0,F}^2 \lesssim |v|_{1}^2. \]

Hence we get from the last two inequalities
\[ \|\text{div} \tau\|_{-1} = \sup_{v \in H^1_0(\Omega; \mathbb{R}^n)} \frac{b(\tau, v)}{|v|_1} \lesssim \|h \text{div} \tau\| + \sup_{v_h \in V_h} \frac{b(\tau, v_h)}{|v_h|_{1,h}}. \]

Therefore we can end the proof by using Lemma 3.4.

3.4. Stability in mesh dependent norms. We now present stability in mesh dependent norms. For $k \geq n + 1$, since there is no stabilization term and $\text{div} \Sigma_h \subset V_h$, then $\ker(\text{div}) \cap \Sigma_h \subset \ker(\text{div}) \cap \Sigma$. The stability follows from Lemma 3.4 and inf-sup condition (3.9).

Theorem 3.6. For $k \geq n + 1$, it follows for any $(\tilde{\sigma}_h, \tilde{u}_h) \in \tilde{\Sigma}_h \times V_h$,
\[ \|\tilde{\sigma}_h\|_{0,h} + |\tilde{u}_h|_{1,h} \lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h} \frac{A(\tilde{\sigma}_h, \tilde{u}_h; \tau_h, v_h)}{\|\tau_h\|_{0,h} + |v_h|_{1,h}}. \]

Corollary 3.7. Let $k \geq n + 1$. Assume that $\sigma \in H^{k+1}(\Omega; S)$ and $u \in H^k(\Omega; \mathbb{R}^n)$, then
\[ \|\sigma - \sigma_h\|_{0,h} + |Q_h u - u_h|_{1,h} \lesssim h^{k+1} \|\sigma\|_{k+1}, \]
\[ |u - u_h|_{1,h} \lesssim h^{k-1}(\|\sigma\|_{k+1} + \|u\|_{k+1}). \]
Moreover, when $\Omega$ is convex, we have
\[ \|Q_h u - u_h\|_0 \lesssim h^{k+2} \|\sigma\|_{k+1}. \]

Proof. Subtracting (2.3)-(2.4) from (2.1)-(2.2), we get the error equation
\[ a(\sigma - \sigma_h, \tau_h) + b(\tau_h, u - u_h) = 0 \quad \forall \tau_h \in \Sigma_h, \]
\[ b(\sigma - \sigma_h, v_h) = 0 \quad \forall v_h \in V_h. \]
Let $I^H_Z$ be the standard interpolation from $H^1(\Omega; S)$ to $\Sigma_h$ defined in [28, Remark 3.1], and it holds
\[ \text{div}(I^H_Z \sigma) = Q_h(\text{div} \sigma). \]

Thus we have from (3.22)
\[ b(I^H_Z \sigma - \sigma_h, v_h) = b(\sigma - \sigma_h, v_h) = 0. \]

By the definition of $Q_h$ and (3.21),
\[ b(\tau_h, Q_h u - u_h) = b(\tau_h, u - u_h) = -a(\sigma - \sigma_h, \tau_h). \]
Combining the last two equalities, it holds
\[ a(I_h^{HZ} \sigma - \sigma_h, Q_h u - u_h; r_h, v_h) \]
\[ = a(I_h^{HZ} \sigma - \sigma_h, r_h) + b(r_h, Q_h u - u_h) + b(I_h^{HZ} \sigma - \sigma_h, v_h) \]
\[ = a(I_h^{HZ} \sigma - \sigma, r_h), \]
which together with (3.17) implies
\[ \| I_h^{HZ} \sigma - \sigma_h \|_{0,h} + |Q_h u - u_h|_{1,h} \lesssim \| I_h^{HZ} \sigma - \sigma \|_{0,h}. \]
Therefore we will achieve (3.18)-(3.19) by using the last inequality, and the error estimate of \( I_h^{HZ} \) and \( Q_h \). The error estimate (3.20) can be derived by using the duality argument as in [24, 46].

**Remark 3.8.** The optimal convergence rate of \( \| \sigma - \sigma_h \|_{0,h} \) has been mentioned in [28, Remarks 3.1-3.2] and [30, Remarks 3.6], but the 2-order higher superconvergent rates of \( |Q_h u - u_h|_{1,h} \) and \( \|Q_h u - u_h\|_0 \) are new which can be used to reconstruct a better approximation of displacement. The convergence rate of \( |u - u_h|_{1,h} \) is also optimal.

Due to the stabilization term (for inf-sup condition), our proof of the stability is more complicated for the low order case \( 1 \leq k \leq n \).

**Theorem 3.9.** For \( 1 \leq k \leq n \), it holds for any \((\tilde{\sigma}_h, \tilde{u}_h) \in \tilde{\Sigma}_h \times V_h\) that
\[ \| \tilde{\sigma}_h \|_{0,h} + |\tilde{u}_h|_{1,h} \lesssim \sup_{(r_h, v_h) \in \Sigma_h \times V_h} \lambda(\tilde{\sigma}_h, \tilde{u}_h; r_h, v_h). \]

**Proof.** As demonstrated in Lemma 3.3, it is equivalent to prove
\[ \| \tilde{\sigma}_h \|_{0,h} + |\tilde{u}_h|_{1,h} \lesssim \sup_{(r_h, v_h) \in \Sigma_h \times V_h} \lambda(\tilde{\sigma}_h, \tilde{u}_h; r_h, v_h) := \beta. \]

The notation \( \beta \) is introduced just for ease of presentation. Let \( \tau_1 = E \varepsilon_h(\tilde{u}_h) \) for \( k \geq 2 \) and \( \tau_1 = 0 \) for \( k = 1 \), then it holds from Cauchy-Schwarz inequality
\[ \lambda(\tilde{\sigma}_h, \tilde{u}_h; \tau_1, 0) = a(\tilde{\sigma}_h, \tau_1) + b(\tau_1, \tilde{u}_h) \geq -\| \tilde{\sigma}_h \|_a \| \tau_1 \|_a + b(\tau_1, \tilde{u}_h). \]

Using (3.10)-(3.11), there exists a constant \( C_3 > 0 \) such that
\[ \lambda(\tilde{\sigma}_h, \tilde{u}_h; \tau_1, 0) \geq C_1 \| \varepsilon_h(\tilde{u}_h) \|_0^2 - C_3 \| \tilde{\sigma}_h \|_a \| \varepsilon_h(\tilde{u}_h) \|_0 \]
\[ \geq \frac{C_1}{2} \| \varepsilon_h(\tilde{u}_h) \|_0^2 - \frac{C_3^2}{2C_1} \| \tilde{\sigma}_h \|_a^2. \]
Let \( v_1 \in V_h \) such that \( v_1|_K = h^\frac{1}{2} \text{div} \tilde{\sigma}_h \) for each \( K \in T_h \). Applying inverse inequality, we have
\[ |v_1|_{1,h} \lesssim \| h \text{div} \tilde{\sigma}_h \| \lesssim \| \tilde{\sigma}_h \|_0. \]
Thus there exists a constant \( C_4 > 0 \) such that
\[ \lambda(\tilde{\sigma}_h, \tilde{u}_h; 0, v_1) = b(\tilde{\sigma}_h, v_1) - c(\tilde{u}_h, v_1) \geq \| h \text{div} \tilde{\sigma}_h \|^2 - \| \tilde{u}_h \|_c \| v_1 \|_c \]
\[ \geq \| h \text{div} \tilde{\sigma}_h \|^2 - C_4 \| \tilde{u}_h \|_c \| h \text{div} \tilde{\sigma}_h \| \]
\[ \geq \frac{1}{2} \| h \text{div} \tilde{\sigma}_h \|^2 - \frac{C_4^2}{2} \| \tilde{u}_h \|_c^2. \]
Thus we obtain from the last two inequalities
\[ \eta \|
\bar{\sigma} - \sigma_h \|_a + h \div \bar{\sigma} - \sigma_h \| + \sup_{v_h \in V_h} \frac{b(\bar{\sigma}, v_h)}{|v_h|_{1,h}} \]
\[ = \|\bar{\sigma} - \sigma_h \| + h \div \bar{\sigma} - \sigma_h \| + \sup_{v_h \in V_h} \frac{A(\bar{\sigma}, \bar{u}_h; 0, v_h) + c(\bar{u}_h, v_h)}{|v_h|_{1,h}} \]
\[ \lesssim \|\bar{\sigma} - \sigma_h \| + h \div \bar{\sigma} - \sigma_h \| + |\bar{u}_h|_{1,h} + \beta. \]
which together with (3.10) and (3.27) indicates
\[ \|\bar{\sigma} - \sigma_h\|^2_a + \|h \div (\bar{\sigma} - \sigma_h)\|^2 + \|\bar{u}_h\|^2_{1,h} \lesssim A(\bar{\sigma}, \bar{u}_h; \tau_h, v_h) \lesssim \beta \|\bar{\sigma} - \sigma_h\|_0 + |\bar{u}_h|_{1,h}. \]
According to Lemma 3.5 and the definition of \( \beta \), it holds
\[ \|\sigma - \sigma_h\|_0 \lesssim \|\bar{\sigma} - \sigma_h\| + \|h \div \bar{\sigma} - \sigma_h\| + \sup_{v_h \in V_h} \frac{b(\bar{\sigma}, v_h)}{|v_h|_{1,h}} \]
\[ = \|\bar{\sigma} - \sigma_h\| + h \div \bar{\sigma} - \sigma_h \| + \sup_{v_h \in V_h} \frac{A(\bar{\sigma}, \bar{u}_h; 0, v_h) + c(\bar{u}_h, v_h)}{|v_h|_{1,h}} \]
\[ \lesssim \|\bar{\sigma} - \sigma_h\| + h \div \bar{\sigma} - \sigma_h \| + |\bar{u}_h|_{1,h} + \beta. \]
Thus we obtain from the last two inequalities
\[ \|\bar{\sigma} - \sigma_h\|^2_0 + |\bar{u}_h|_{1,h}^2 \lesssim \|\bar{\sigma} - \sigma_h\|^2_a + \|h \div \bar{\sigma} - \sigma_h\|^2 + |\bar{u}_h|_{1,h}^2 + \beta^2 \]
\[ \lesssim \beta (\|\bar{\sigma} - \sigma_h\|_0 + |\bar{u}_h|_{1,h}) + \beta^2, \]
which implies inf-sup condition (3.25). \( \square \)

**Corollary 3.10.** Let \( 1 \leq k \leq n \). Assume that \( \sigma \in H^{k+1}(\Omega; \mathbb{S}) \) and \( u \in H^k(\Omega; \mathbb{R}^n) \), then
\[ \|\sigma - \sigma_h\|_0 + |u - u_h|_{1,h} \lesssim h^{k-1} (\|\sigma\|_{k+1} + \|u\|_k). \]
The convergence of rate of \( |u - u_h|_{1,h} \) is optimal. But the \( L^2 \)-type error of \( \|\sigma - \sigma_h\|_0 \) is two order less.

**Remark 3.11.** Using the stability in mesh dependent norms established in [40, 9], the MINRES method with additive Schwarz preconditioner was developed for the mixed finite element methods of the Poisson problem in [43], and the CG method with auxiliary space preconditioner for the corresponding Schur complement problem was designed in [25]. Similar stability in mesh dependent norm for the mixed finite macroelement methods of the linear elasticity can be found in [45], hence the fast auxiliary space preconditioner constructed in this paper can be easily extended to these mixed methods. For example, the block-triangular preconditioner similar to (4.5) for the mixed Poisson problem has been included in iFEM [18] since 2012.

### 3.5 Postprocessing
Based on the superconvergent results of the displacement in (3.18) and (3.20), we will construct a superconvergent postprocessed displacement from \( (\sigma_h, u_h) \) for the higher order case \( k \geq n + 1 \) in this subsection.

To this end, let
\[ V_h^* := \left\{ v \in L^2(\Omega; \mathbb{R}^n) : v|_K \in P_{k+1}(K; \mathbb{R}^n), \quad \forall K \in T_h \right\}. \]
Then a postprocessed displacement can be defined as follows: Find \( u_h^* \in V_h^* \) such that
\[ (3.29) \quad Q_h u_h^* = u_h, \]
(3.30) \( (\varepsilon(u_h^*), \varepsilon(v))_K = (\mathfrak{A}\sigma_h, \varepsilon(v))_K \quad \forall \ v \in (I - Q_h)V_h|_K \),

for any \( K \in \mathcal{T}_h \). To derive the error estimate for the postprocessed displacement \( u_h^* \), we will merge the mixed finite element method (2.3)-(2.4) and the postprocessing (3.29)-(3.30) into one method as in [40]. To be specific, find \((\sigma_h, u_h^*) \in \Sigma_h \times V_h^*\) such that 

\[
(A_h(\sigma_h, u_h^*; \tau_h, v_h^*) = -(Q_h f, v_h^*) \quad \forall \ (\tau_h, v_h^*) \in \Sigma_h \times V_h^*,
\]

where 

\[
A_h(\sigma_h, u_h^*; \tau_h, v_h^*) := A(\sigma_h, u_h^*; \tau_h, v_h^*) + (\varepsilon_h(u_h^*) - \mathfrak{A}\sigma_h, \varepsilon_h(v_h^* - Q_h v_h^*)),
\]

**Lemma 3.12.** The mixed finite element method (2.3)-(2.4) and the problem (3.31) are equivalent in the following sense: if \((\sigma_h, u_h^*) \in \Sigma_h \times V_h^*\) is the solution of the problem (3.31) and let \( u_h = Q_h u_h^* \), then \((\sigma_h, u_h) \in \Sigma_h \times V_h\) solves the mixed finite element method (2.3)-(2.4); Conversely, if \((\sigma_h, u_h) \in \Sigma_h \times V_h\) is the solution of the mixed finite element method (2.3)-(2.4) and \( u_h^* \in V_h^* \) is the postprocessed displacement defined by (3.29)-(3.30), then \((\sigma_h, u_h^*) \in \Sigma_h \times V_h^*\) solves the problem (3.31).

**Proof.** Taking any \((\tau_h, v_h) \in \Sigma_h \times V_h\), and noting the fact that \( v_h = Q_h v_h^* \) and \( \text{div}\Sigma_h \subset V_h \), we have 

\[
A_h(\sigma_h, u_h^*; \tau_h, v_h^*) = A(\sigma_h, u_h^*; \tau_h, v_h) = a(\sigma_h, \tau_h) + b(\tau_h, u_h^*) + b(\sigma_h, v_h)
\]

(3.32)

\[
a(\sigma_h, \tau_h) + b(\tau_h, Q_h u_h^*) + b(\sigma_h, v_h) = A(\sigma_h, Q_h u_h^*; \tau_h, v_h).
\]

Hence we can see from (3.32) that \((\sigma_h, u_h)\) solves the mixed finite element method (2.3)-(2.4) if \((\sigma_h, u_h^*)\) is the solution of the problem (3.31).

Conversely, since \( \text{div}\Sigma_h \subset V_h \) and \((I - Q_h)^2 = I - Q_h\), it follows from (3.32) and (3.29) 

\[
A_h(\sigma_h, u_h^*; \tau_h, v_h^*) = A_h(\sigma_h, u_h^*; \tau_h, Q_h v_h^* + A_h(\sigma_h, u_h^*; 0, v_h^* - Q_h v_h^*)
\]

(3.33)

\[
= A_h(\sigma_h, u_h^*; \tau_h, Q_h v_h^*) + (\varepsilon_h(u_h^*) - \mathfrak{A}\sigma_h, \varepsilon_h(v_h^* - Q_h v_h^*))
\]

which together with (3.30) means that \((\sigma_h, u_h^*)\) solves the problem (3.31). \qed

**Lemma 3.13.** For any \( v \in H^1(\mathcal{T}_h; \mathbb{R}^n) \), it holds that 

\[
|v - Q_h v|_{1,h} \approx \|\varepsilon_h(v - Q_h v)\|_0.
\]

**Proof.** It is sufficient to prove 

\[
\sum_{F \in \mathcal{F}_h} h_F^{-1} \|v - Q_h v\|_{0,F}^2 \lesssim \|\varepsilon_h(v - Q_h v)\|_0^2.
\]

Let \( \pi \) be defined as in Lemma 3.2 and \( w = v - Q_h v \). It follows from (3.3) in [13] 

\[
\sum_{F \in \mathcal{F}_h} h_F^{-1} \|w - Q_h w\|_{0,F}^2 = \sum_{F \in \mathcal{F}_h} h_F^{-1} \|(w - \pi w) - Q_h (w - \pi w)\|_{0,F}^2 
\]

(3.36)

\[
\leq \sum_{K \in \mathcal{T}_h} |w - \pi w|_{1,K}^2 \lesssim \|\varepsilon_h(w)\|_0^2.
\]

On the other hand, 

\[
\sum_{F \in \mathcal{F}_h} h_F^{-1} \|v - Q_h v\|_{0,F}^2 = \sum_{F \in \mathcal{F}_h} h_F^{-1} \|w - Q_h w\|_{0,F}^2.
\]
Therefore (3.35) follows from (3.36).

**Theorem 3.14.** For any \((\tilde{\sigma}_h, \tilde{u}_h^*) \in \tilde{\Sigma}_h \times V_h^*, h > 0\), it follows

\begin{equation}
\|\tilde{\sigma}_h\|_{0,h} + |\tilde{u}_h|_{1,h} \lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h^*|_{1,h}}.
\end{equation}

**Proof.** For any \(v_h \in V_h\), we have from (3.32)

\[ A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h) = A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*). \]

Since \((\tilde{\sigma}_h, Q_h \tilde{u}_h^*) \in \tilde{\Sigma}_h \times V_h\), it holds from (3.17)

\[ \|\tilde{\sigma}_h\|_{0,h} + |Q_h \tilde{u}_h|_{1,h} \lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h} \frac{A_h(\tilde{\sigma}_h, Q_h \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h|_{1,h}} = \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h)}{\|\tau_h\|_{0,h} + |v_h|_{1,h}} \]

\begin{equation}
\lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h^*|_{1,h}}.
\end{equation}

Similarly as in (3.33), we get

\[ A_h(\tilde{\sigma}_h, \tilde{u}_h^*; 0, \tilde{u}_h^* - Q_h \tilde{u}_h^*) = (\varepsilon_h(\tilde{u}_h^*) - A\tilde{\sigma}_h, \varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*)). \]

Then we rewrite it as

\[ \|\varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*)\|_0 = A(\tilde{\sigma}_h - \varepsilon_h(Q_h \tilde{u}_h^*), \varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*) + A_h(\tilde{\sigma}_h, \tilde{u}_h^*; 0, \tilde{u}_h^* - Q_h \tilde{u}_h^*). \]

According to the triangle inequality and (3.38), it holds

\[ \|A\tilde{\sigma}_h - \varepsilon_h(Q_h \tilde{u}_h^*)\|_0 \lesssim \|\tilde{\sigma}_h\|_0 + \|\varepsilon_h(Q_h \tilde{u}_h^*)\|_0 \lesssim \|\tilde{\sigma}_h\|_0 + |Q_h \tilde{u}_h|_{1,h}\]

\[ \lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h^*|_{1,h}}. \]

Due to (3.34), we have

\[ A_h(\tilde{\sigma}_h, \tilde{u}_h^*; 0, \tilde{u}_h^* - Q_h \tilde{u}_h^*) \]

\[ \lesssim \|\varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*)\|_0 \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; 0, v_h^* - Q_h v_h^*)}{\|\varepsilon_h(v_h^* - Q_h v_h^*)\|_0} \]

\[ \lesssim \|\varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*)\|_0 \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; 0, v_h^* - Q_h v_h^*)}{\|v_h^* - Q_h v_h^*\|_{1,h}} \]

\[ \lesssim \|\varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*)\|_0 \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h^*|_{1,h}}. \]

Using the last two inequalities and Cauchy-Schwarz inequality, we get from (3.39)

\[ \|\varepsilon_h(\tilde{u}_h^* - Q_h \tilde{u}_h^*)\|_0 \lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h^*|_{1,h}}, \]

c which together with (3.34) implies

\begin{equation}
|\tilde{u}_h^* - Q_h \tilde{u}_h^*|_{1,h} \lesssim \sup_{(\tau_h, v_h) \in \Sigma_h \times V_h^*} \frac{A_h(\tilde{\sigma}_h, \tilde{u}_h^*; \tau_h, v_h^*)}{\|\tau_h\|_{0,h} + |v_h^*|_{1,h}}.
\end{equation}
Finally we can finish the proof by combining (3.38) and (3.40).

\begin{proof}
By direct computation, we have 
\[ A_h(\sigma - \sigma_h, u - Q_h^* u; \tau_h, v_h^*) = -(f, v_h^*) \quad \forall (\tau_h, v_h^*) \in \Sigma_h \times V_h^*.\]
Combining with (3.31), we get the error equation
\[ A_h(\sigma - \sigma_h, u - u_h^*; \tau_h, v_h^*) = (Q_h f - f, v_h^*) \quad \forall (\tau_h, v_h^*) \in \Sigma_h \times V_h^*.\]
Let \( Q_h^* \) be the \( L^2 \) orthogonal projection from \( L^2(\Omega; \mathbb{R}^n) \) onto \( V_h^* \). It holds from (3.23) that 
\[ A_h(\sigma - I_h^{HZ} \sigma, u - Q_h^* u; \tau_h, v_h^*) = a(\sigma - I_h^{HZ} \sigma, v_h^*) + \langle (\varepsilon_h(u - Q_h^* u) - \delta(\sigma - I_h^{HZ} \sigma), \varepsilon_h(v_h^*) \rangle \]
\[ = a(I_h^{HZ} \sigma, \tau_h) + (Q_h f - f, v_h^*) + \langle (\varepsilon_h(u - Q_h^* u) - \delta(\sigma - I_h^{HZ} \sigma), \varepsilon_h(v_h^*) \rangle.\]
Then we obtain from (3.43), Cauchy-Schwarz inequality and the error estimates of \( I_h^{HZ}, Q_h^* \) and \( Q_h \)
\[ A_h(I_h^{HZ} \sigma - \sigma_h, Q_h^* u - u_h^*; \tau_h, v_h^*) = a(I_h^{HZ} \sigma, \tau_h) - (\varepsilon_h(u - Q_h^* u) - \delta(\sigma - I_h^{HZ} \sigma), \varepsilon_h(v_h^*) \rangle \]
\[ \lesssim h^{k+1} (||\sigma||_{k+1} + ||u||_{k+2}) (||\tau_h||_0 + ||\varepsilon_h(v_h^*)||_0).\]
Applying the inf-sup condition (3.37), it follows
\[ ||I_h^{HZ} \sigma - \sigma_h||_{0,h} + ||Q_h^* u - u_h^*||_{1,h} \lesssim h^{k+1} (||\sigma||_{k+1} + ||u||_{k+2}).\]
Hence we will achieve (3.41) by using the triangle inequality, and the error estimates of \( I_h^{HZ} \) and \( Q_h^* \).

When \( \Omega \) is convex, we have from the triangle inequality, the error estimate of \( Q_h \) and (3.29)
\[ ||u - u_h^*||_0 \leq ||(I - Q_h)(u - u_h^*)||_0 + ||Q_h u - Q_h u_h^*||_0 \]
\[ \lesssim h||u - u_h^*||_{1,h} + ||Q_h u - u_h^*||_0.\]
Finally (3.42) is achieved by using (3.41) and (3.20).
\end{proof}

4. Block Diagonal and Triangular Preconditioners

Direct use of the mesh dependent norm \( ||\cdot||_{0,h} \times ||\cdot||_{1,h} \) would require the additional assembling of the jump term. In this section, we first derive equivalent matrix forms for these mesh dependent norms and then construct block-diagonal and block-triangular preconditioners.
4.1. Equivalent matrix forms of the mesh dependent norms. By the trace theorem and the inverse inequality, it is easy to see that

\[ \| \tau_h \|_{0,h} \simeq \| \tau_h \|_0 \quad \forall \tau_h \in \Sigma_h, \]

which implies that we can use the weighted mass matrix \( M_h^{\lambda} \) with \( \lambda = 0 \), i.e., \( M_h \).

For each \( v_h \in V_h \), denote by \( v_h \) the matrix representation of \( v_h \) based on the basis of \( V_h \) used to form the mass matrix \( M_{u,h} \) (cf. [51, Subsection 4.4]). For the mesh dependent norm \( | \cdot |_{1,h} \) of displacement, we can use the Schur complement of \((1,1)\) block, i.e., \( S_h := B_h M_h^{-1} B_h^T + C_h \). It is easy to see \( S_h \) is SPD and induce a norm \( \| \cdot \|_{S_h} \) on \( V_h \), i.e.

\[ \| v_h \|_{S_h}^2 := v_h^T S_h v_h, \quad \forall v_h \in V_h. \]

Lemma 4.1. We have the norm equivalence:

\[ | v_h |_{1,h} \simeq \| v_h \|_{S_h} \quad \forall v_h \in V_h. \]

Proof. We focus on the case \( k \geq n + 1 \) first. The low order case \( 1 \leq k \leq n \) can be proved similarly by adding the stabilization term.

The inf-sup condition (3.1) implies \( B_h^T \) is injective and thus \( S_h \) is SPD and defines an inner product on \( V_h \). The identity

\[ (v_h^T S_h v_h)^{1/2} = \| M_h^{-1/2} B_h^T v_h \| = \sup_{\tau_h \in \Sigma_h} \frac{b(\tau_h, v_h)}{\| \tau_h \|_0}, \quad \forall v_h \in V_h \]

follows from the Riesz representation. Here \( \| \cdot \| \) denotes the Euclidean norm of a vector. The inequality \( | v_h |_{1,h} \lesssim \| v_h \|_{S_h} \) is a combination of (3.1), (4.1), and (4.2).

From integration by parts, we can easily get \( b(\tau_h, v_h) \lesssim \| \tau_h \|_{0,h} | v_h |_{1,h} \). Then the inequality \( \| v_h \|_{S_h} \lesssim | v_h |_{1,h} \) follows from (4.1) and (4.2). \( \square \)

We define the operator \( P_h : \Sigma_h \times V_h \rightarrow \Sigma_h \times V_h \) with the matrix representation

\[ P_h = \begin{pmatrix} M_h^{\lambda} & 0 \\ 0 & S_h^{-1} \end{pmatrix}, \]

and denoted by

\[ L_h^{\lambda} = \begin{pmatrix} M_h^{\lambda} & B_h^T \\ B_h & -C_h \end{pmatrix}. \]

Theorem 4.2. The \( P_h \) is a uniform preconditioner for \( L_h^{\lambda} \), i.e., the corresponding operator norms

\[ \| P_h L_h^{\lambda} \|_{\Sigma_h \times V_h \rightarrow \Sigma_h \times V_h}, \| (P_h L_h^{\lambda})^{-1} \|_{\Sigma_h \times V_h \rightarrow \Sigma_h \times V_h} \]

are bounded and independent of parameters \( h \) and \( \lambda \).

The mass matrix \( M_h^{-1} \) can be further replaced by the inverse of the diagonal matrix or symmetric Gauss-Seidel iteration and thus the computation of \( M_h^{-1} \) is not a problem. The difficulty is the inverse of the Schur complement which will be further preconditioned by an auxiliary space preconditioner in the next section.
4.2. **Triangular Preconditioner.** When the diagonal of the mass matrix $D_h$ is used, we can make use of the block decomposition

$$
\begin{pmatrix}
D_h & B_h^T \\
B_h & -C_h
\end{pmatrix}
\begin{pmatrix}
I & D_h^{-1}B_h^T \\
0 & -I
\end{pmatrix}
= 
\begin{pmatrix}
D_h & 0 \\
B_h & \tilde{S}_h
\end{pmatrix},
$$

where $\tilde{S}_h = B_hD_h^{-1}B_h^T + C_h$ to obtain a triangular preconditioner.

We define the operator $G_h : \Sigma_h' \times V_h' \rightarrow \Sigma_h \times V_h$

$$(4.5) \quad G_h = \left( I - D_h B_h^T \right) \left( D_h 0 \right)^{-1};$$

If we denote by $\tilde{L}_h = \left( D_h B_h^T \right)$, it is trivial to verify that $G_h = \tilde{L}_h^{-1}$. For mass matrix $M_h$, by standard scaling argument, we have $D_h$ is spectrally equivalent to $M_h$ and so $\tilde{L}_h^{-1}$ is also stable in the mesh dependent norm. We thus obtain the following result. Detailed eigenvalue analysis of the preconditioned system can be found in [7].

**Theorem 4.3.** The $G_h$ is a uniform preconditioner for $L_h^\lambda$ i.e., the corresponding operator norms

$$
\|G_hL_h^\lambda\|_{\Sigma_h \times V_h \rightarrow \Sigma_h \times V_h}, \|G_hL_h^\lambda\|^{-1}_{\Sigma_h \times V_h \rightarrow \Sigma_h \times V_h}
$$

are bounded and independent of parameters $h$ and $\lambda$.

In both diagonal and triangular preconditioners, to be practical, we do not compute $S_h^{-1}$ or $\tilde{S}_h^{-1}$. Instead we shall apply the fast auxiliary space preconditioner to be developed in the next section.

5. **Auxiliary Space Preconditioner**

In this section we first review the framework on auxiliary space preconditioners developed by Xu [52] and then construct one for the linear elasticity problem in mixed forms. We use $H^1$ conforming linear element and primary formulation of linear elasticity with $\lambda = 0$ as the auxiliary space preconditioner and verify all assumptions needed in the framework.

5.1. **Framework.** Let

$$
\mathcal{V}_h := \left\{ v \in H^1_0(\Omega; \mathbb{R}^n) : v|_K \in P_1(K; \mathbb{R}^n) \quad \forall K \in T_h \right\}.
$$

Then $\mathcal{V}_h \subset V_h$ for $k \geq 2$, and

$$
|v_h|_{1,h} = \|\varepsilon(v_h)\|_0 \approx |v_h|_1 \quad \forall v_h \in \mathcal{V}_h.
$$

The conforming linear finite element method for the linear elasticity with $\lambda = 0$ is defined as follows: Find $u_h \in \mathcal{V}_h$ such that

$$
2\mu(\varepsilon(u_h), \varepsilon(v_h)) = (f, v_h) \quad \forall v_h \in \mathcal{V}_h.
$$

Denote $\mathcal{A} : \mathcal{V}_h \rightarrow \mathcal{V}_h$ by

$$
(\mathcal{A}w_h, v_h) := 2\mu(\varepsilon(w_h), \varepsilon(v_h)) \quad \forall w_h, v_h \in \mathcal{V}_h.
$$

It is apparent that the operator $\mathcal{A}$ is SPD.
In what follows we assume $T_h$ is quasi-uniform. Based on the norm equivalence (5.1), we can easily derive the estimate of spectral radius and condition number of the Schur complement operator $S$

\begin{equation}
\rho_S = \lambda_{\max}(S) \approx h^{-2}, \quad \kappa(S) = \frac{\lambda_{\max}(S)}{\lambda_{\min}(S)} \approx h^{-2}.
\end{equation}

The relation between $S$ and $S_h$ is given by

\[ S_h = M_{u,h}S \]

with $S$ being the matrix representation of $S$.

We introduce the auxiliary space preconditioner for the Schur complement. The idea is to construct a multigrid method using $V_h$ as the “fine” space and $V_h$ as the “coarse” space. Denote $B : V_h \to V_h$ to be such a “coarse” solver. It can be either an exact solver or an approximate solver that satisfies certain conditions, which will be given later. Next, on the fine space, we need a smoother $R : V_h \to V_h$, which is symmetric and positive definite. For example, $R$ can be a Jacobi or symmetric Gauss-Seidel smoother. Finally, to connect the “coarse” space with the “fine” space, we need a “prolongation” operator $\Pi : V_h \to V_h$. A “restriction” operator $\Pi^t : V_h \to V_h$ is consequently defined by

\[ (\Pi^t v, w) = (v, \Pi w) \quad \text{for} \quad v \in V_h \quad \text{and} \quad w \in V_h. \]

It is also well-known that the matrix representation of the restriction operator $\Pi^t$ is just the transpose of the matrix representation of the prolongation operator $\Pi$. Then, the auxiliary space preconditioner $X : V_h \to V_h$, following the definition in [52], is given by

\begin{align}
(5.3) \quad \text{Additive} & \quad X = R + \Pi B \Pi^t, \\
(5.4) \quad \text{Multiplicative} & \quad I - XS = (I - R^t S)(I - \Pi B \Pi^t S)(I - RS).
\end{align}

According to [52], the following theorem holds.

**Theorem 5.1** (Xu [52]). Assume that for all $v \in V_h$, $w \in V_h$,

\begin{align}
(5.5) \quad (Sv, v) \lesssim (R^{-1} v, v) \lesssim \rho_S(v, v), \\
(5.6) \quad (Aw, w) \lesssim (B A w, A w) \lesssim (Aw, w), \\
(5.7) \quad |\Pi w|_{1,h} \lesssim |w|_1 \quad \text{(stability of $\Pi$)},
\end{align}

and furthermore, assume that there exists a linear operator $P : V_h \to V_h$ such that

\begin{align}
(5.8) \quad |Pv|_1 \lesssim |v|_{1,h} \quad \text{(stability of $P$)}, \\
(5.9) \quad \|v - \Pi P v\|_0^2 \lesssim \rho_S^{-1} |v|_{1,h}^2 \quad \text{(approximability)}.
\end{align}

Then the preconditioner $X$ defined in (5.3) or (5.4) satisfies

\[ \kappa(X S) \lesssim 1. \]

**5.2. Construction.** Now we construct an auxiliary space preconditioner which satisfies all conditions in Theorem 5.1, namely, inequalities (5.5)-(5.9). It is straightforward to pick $B$ that satisfies condition (5.6). For example, $B$ can be either the direct solver, for which $B \sim A^{-1}$, or one step of classical multigrid iteration which satisfies condition (5.6).

The smoother $R$ is also easy to define. A Jacobi or a symmetric Gauss-Seidel smoother [10] will satisfy condition (5.5). The operator $\Pi$ is the natural inclusion
for \( k \geq 2 \) and the \( L^2 \) projection \( Q_h \) for \( k = 1 \), i.e., taking the averaging of nodal values inside each simplex. Then the condition (5.7) follows from (5.1) and (3.16) immediately.

The technical part is to define an operator \( P : V_h \to \mathcal{V}_h \) that satisfies the conditions (5.8)-(5.9). Note that operator \( P \) is needed only in the theoretical analysis. In the implementation, one needs \( \mathcal{B}, \mathcal{R} \) and \( \Pi \) only.

Construction of \( P \) is equivalent to specify the function values at each vertex. For an interior vertex \( x_i \) of \( T_h \), denoted by \( \Omega_i \) the vertex patch of \( x_i \), we will simply choose \((Pv)(x_i) := |\Omega_i|^{-1} \int_{\Omega_i} v \, dx\), i.e., the average of a discontinuous polynomial \( v \) in the vertex patch. For boundary vertex \( x_i \in \partial \Omega \), we set \((Pv)(x_i) := 0\).

For any \( K \in T_h \), let \( Q_K^0 v := (Q_K^0 v)|_K = |K|^{-1} \int_K v \, dx \). Define
\[ T_{h,i} := \{ K \in T_h : K \subset \Omega_i \}, \quad \mathcal{F}_{h,i} := \{ F \in \mathcal{F}_h : x_i \in F \}. \]

Obviously for interior nodes we have
\[(Pv)(x_i) = \sum_{K \in T_{h,i}} \frac{|K|}{|\Omega_i|} Q_K^0 v. \tag{5.10}\]

The error estimate of the operator \( P \) can be derived by standard argument used in [48, 13, 16, 31]. For completeness, we show it in details as follows.

**Lemma 5.2.** The operator \( P \) satisfies
\[ \|v - (Pv)\|_0 + h|Pv|_1 \lesssim h|v|_{1,h} \quad \forall \ v \in V_h. \]

**Proof.** According to (5.10), it holds for each interior node \( x_i \)
\[ |Q_K^0 v - (Pv)(x_i)|^2 \lesssim \sum_{K' \in T_{h,i}} |Q_{K'}^0 v - Q_K^0 v|^2 \lesssim \sum_{F \in \mathcal{F}_{h, i}} \|Q_h^0 v\|^2. \]

For each boundary node \( x_i \), we obtain by similar technique and the definition of jump on the boundary
\[ |Q_K^0 v - (Pv)(x_i)|^2 = |Q_K^0 v|^2 \lesssim \sum_{F \in \mathcal{F}_{h, i}} \|Q_h^0 v\|^2. \]

Then using the scaling argument, we have
\[ \sum_{K \in T_h} h_K^{-2} \|Q_K^0 v - P v\|_{0,K}^2 = \sum_{K \in T_h} \sum_{i=0}^n h_K^{n-2} |Q_K^0 v - (Pv)(x_{K,i})|^2 \lesssim \sum_{F \in \mathcal{F}_h} h_F^{-1} \|Q_h^0 v\|_{0,F}^2. \]

From the \( L^2 \) error estimate (3.15), discrete Korns inequality (3.8), and the norm equivalence (3.6), we get
\[ \sum_{K \in T_h} h_K^{-2} \|v - P v\|_{0,K}^2 \lesssim \sum_{K \in T_h} h_K^{-2} \|v - Q_K^0 v\|_{0,K}^2 + \sum_{K \in T_h} h_K^{-2} \|Q_K^0 v - P v\|_{0,K}^2 \lesssim \|v\|_{1,h}^2 + \sum_{F \in \mathcal{F}_h} h_F^{-1} \|Q_h^0 v - v\|_{0,F}^2 \lesssim \|v\|_{1,h}^2. \tag{5.11} \]
It follows from (5.11) and (3.15)
\[ |Pv|^2 = \sum_{K \in \mathcal{T}_h} |Pv - Q^0_K v|^2_{1,K} \lesssim \sum_{K \in \mathcal{T}_h} h_K^{-2} ||Pv - Q^0_K v||^2_{0,K} \]
\[ \lesssim \sum_{K \in \mathcal{T}_h} h_K^{-2} ||v - P v||^2_{0,K} + \sum_{K \in \mathcal{T}_h} h_K^{-2} ||v - Q^0_K v||^2_{0,K} \lesssim |v|^2_{1,h}. \]

Therefore we can finish the proof by combining the last two inequalities. \(\square\)

**Lemma 5.3.** For any \(v \in V_h\), it holds
\[ \| v - \Pi P v \|_0^2 \lesssim \rho_s^{-1} |v|^2_{1,h}. \]

**Proof.** For \(k \geq 2\), (5.12) is the result of Lemma 5.2 and (5.2). For \(k = 1\), we obtain from the triangle inequality, (3.15), Lemma 5.2 and (5.2)
\[ \| v - \Pi P v \|_0^2 = \| v - Q_h P v \|_0^2 \lesssim \| v - P v \|_0^2 + \| P v - Q_h P v \|_0^2 \]
\[ \lesssim \| v - P v \|_0^2 + h^2 \| P v \|_1^2 \lesssim h^2 | v |_{1,h}^2 \lesssim \rho_s^{-1} |v|^2_{1,h}, \]
as required. \(\square\)

Combining Lemma 4.1, Theorem 5.1, and Lemmas 5.2-5.3, we have the following estimate of the condition number of \(XS\).

**Theorem 5.4.** Let \(R\) be a Jacobi or a symmetric Gauss-Seidel smoother, \(B\) be one step of classical multigrid iteration, and \(\Pi\) be \(Q_h\). Then the preconditioner \(X\) defined in (5.3) or (5.4) satisfies
\[ \kappa(XS) \lesssim 1. \]

6. Numerical Results

In this section, we will report some numerical results to testify the efficiency and robustness of the auxiliary space preconditioners developed in Sections 4-5 for the mixed finite element method (2.3)-(2.4). Let \(\Omega = (-1,1)^2\), \(\mu = 0.5\) and the load \(f = 1\). We use the uniform triangulation \(\mathcal{T}_h\) of \(\Omega\). The stopping criteria of our iterative methods is the relative residual is less than \(10^{-8}\), and the initial guess is zero. We run the code on the laptop with Intel Core i5 CPU (1.7 GHz) and 4GB RAM.

6.1. Block Diagonal Preconditioner. First we use the minimal residual (MINRES) method with the block diagonal preconditioner
\[ \left( \begin{array}{cc} D_h^{-1} & 0 \\ 0 & (B_h D_h^{-1} B_h^{T} + C_h)^{-1} \end{array} \right) \]
to solve the mixed finite element method (2.3)-(2.4), where \(D_h\) is the diagonal matrix of \(M_h\). To solve the Schur complement \(B_h D_h^{-1} B_h^{T} + C_h\), we apply the multiplicative auxiliary space preconditioner (5.4), in which we employ three steps of the Gauss-Seidel smoother for \(R\) and one step of V-cycle multigrid method with one pre-smoothing and one post-smoothing for \(B\).

The iteration numbers and CPU time for the block diagonal preconditioned MINRES method are shown in Tables 1-3 for \(k = 1,2,3\), from which we can see that the iteration steps are uniform with respect to the meshsize \(h\) and the Lamé constant \(\lambda\).
Remark 6.1. The iteration steps can be further reduced by introduce a scaling $\ast \cdot B_h D_h^{-1} B_h^T$. 

### Table 1. The iteration steps and CPU time (in seconds) of block diagonal preconditioned MINRES method for $k = 1$

<table>
<thead>
<tr>
<th>#dofs</th>
<th>$\lambda = 0$ steps</th>
<th>$\lambda = 10$ steps</th>
<th>$\lambda = 100$ steps</th>
<th>$\lambda = 1000$ steps</th>
<th>$\lambda = +\infty$ steps</th>
</tr>
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<td>time</td>
<td>time</td>
<td>time</td>
</tr>
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<td>98 54.3</td>
</tr>
</tbody>
</table>

### Table 2. The iteration steps and CPU time (in seconds) of block diagonal preconditioned MINRES method for $k = 2$

<table>
<thead>
<tr>
<th>#dofs</th>
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<th>$\lambda = 100$ steps</th>
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</tr>
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<td>106 87.7</td>
</tr>
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</table>

### Table 3. The iteration steps and CPU time (in seconds) of block diagonal preconditioned MINRES method for $k = 3$

<table>
<thead>
<tr>
<th>#dofs</th>
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<td>90 62.8</td>
<td>96 66.3</td>
<td>98 67.4</td>
<td>98 67.4</td>
</tr>
</tbody>
</table>

6.2. **Block Triangular Preconditioner.** Next we examine the generalized minimal residual (GMRES) method with the block triangular preconditioner

$$
\left( \begin{array}{cc}
D_h & B_h^T \\
B_h & -C_h
\end{array} \right)^{-1} = 
\left( \begin{array}{cc}
I & D_h^{-1}B_h^T \\
0 & -I
\end{array} \right)
\left( \begin{array}{cc}
D_h & 0 \\
B_h & B_h D_h^{-1} B_h^T + C_h
\end{array} \right)^{-1}.
$$

Set restart=20 in the GMRES method. We still exploit the same multiplicative auxiliary space preconditioner as in the block diagonal preconditioner to solve the Schur complement.
The iteration numbers and CPU time for the block triangular preconditioned GMRES method are shown in Tables 4-6 for $k = 1, 2, 3$. Again the iteration steps are uniform with respect to the meshsize $h$ and the Lamé constant $\lambda$. The performance of the block triangular preconditioned GMRES method is better than the block diagonal preconditioned MINRES method. The iteration steps and CPU time are almost halved comparing with the block diagonal preconditioner.

Table 4. The iteration steps and CPU time (in seconds) of block triangular preconditioned GMRES method for $k = 1$

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<td>46 0.42</td>
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Table 5. The iteration steps and CPU time (in seconds) of block triangular preconditioned GMRES method for $k = 2$

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Table 6. The iteration steps and CPU time (in seconds) of block triangular preconditioned GMRES method for $k = 3$

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References


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