A GENERALIZED MAC SCHEME ON CURVILINEAR DOMAINS

YIN-LIANG HUANG†, JIAN-GUO LIU‡, AND WEI-CHENG WANG§

Abstract. We propose a simple finite difference scheme for Navier–Stokes equations in primitive formulation on curvilinear domains. With proper boundary treatment and interplay between covariant and contravariant components, the spatial discretization admits exact Hodge decomposition and energy identity. As a result, the pressure can be decoupled from the momentum equation with explicit time stepping. No artificial pressure boundary condition is needed. In addition, it can be shown that this spatially compatible discretization leads to uniform inf-sup condition, which plays a crucial role in the pressure approximation of both dynamic and steady state calculations. Numerical experiments demonstrate the robustness and efficiency of our scheme.

Key words. Navier–Stokes equations, inf-sup condition, mark-and-cell scheme, incompressible flow, pressure boundary condition, pressure Poisson equation

AMS subject classifications. 35Q30, 65M06, 65M12, 76D05

DOI. 10.1137/120875508

1. Introduction. In the numerical computation of the Navier–Stokes equation

\[
\begin{align*}
\mathbf{u}_t + (\mathbf{u} \cdot \nabla) \mathbf{u} + \nabla p &= \nu \nabla^2 \mathbf{u} + \mathbf{f} \quad \text{in } \Omega, \\
\nabla \cdot \mathbf{u} &= 0 \quad \text{in } \Omega, \\
\mathbf{u} &= 0 \quad \text{on } \Gamma,
\end{align*}
\]

one of the key issues is the proper implementation of boundary conditions. Since the pressure \( p \) is not described by an evolutionary equation, one could instead treat it as a Lagrangian multiplier to enforce the discrete divergence-free constraint. To realize it in a discrete setting, the discrete gradient, curl, and divergence operators are required to satisfy certain compatibility conditions so that the discrete Hodge decomposition can be performed exactly and efficiently. This is key to decoupling the pressure from the momentum equation and efficiency of the scheme for time-dependent problems. In this approach, the pressure is no longer solved via an elliptic PDE and there is no boundary condition involved for the pressure. The classical mark-and-cell (MAC) scheme [Le, DHSW, HW] can be interpreted as a typical example of this approach [An].

Inspired by the classical MAC scheme, we propose here a generalized MAC (GMAC) scheme on a curvilinear coordinate that preserves the desired properties.

*Submitted to the journal’s Computational Methods in Science and Engineering section April 30, 2012; accepted for publication (in revised form) May 14, 2013; published electronically September 4, 2013. This work was partially supported by the National Science Foundation, the Center of Mathematical Sciences at NTHU, the National Science Council, and the National Center for Theoretical Sciences in Taiwan.

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This is done by careful construction of appropriate second order finite difference operators such that the differential identities such as
\begin{equation}
\text{curl} \circ \text{grad} \equiv 0, \quad \text{div} \circ \text{curl} \equiv 0
\end{equation}
remain valid in the discrete setting. In particular, the discrete Hodge decomposition for vector fields can be performed exactly and the corresponding linear system for the pressure is symmetric and semidefinite. No pressure boundary condition is needed. The scheme is finite difference in nature and easy to implement on curvilinear domains with simple geometry. Overall, the resulting scheme is robust and efficient, as demonstrated by our numerical examples.

The exact Hodge decomposition also leads to a very simple error analysis for the velocity field [HLW]. Rigorous error analysis for the classical MAC scheme was first obtained in [HoW] and for MAC-like schemes on Cartesian grids in [We]. The proof in [HoW, We] is based on high order Strang’s expansion. In contrast, the argument in [HLW] explores the special structure of the spatial discretization and makes use of both the stream function and the discrete analogue of the differential identities (1.4). As a result, an optimal $O(h^2)$ error estimate is obtained provided the exact velocity is in $C^4$ and pressure in $C^3$. This may be the minimal regularity requirement in finite difference setting.

In addition to exact discrete Hodge decomposition, compatibility among spatial discretizations is closely related to the pressure error. The heart of this matter is widely known as the inf-sup condition or Ladyzhenskaya–Babuška–Brezzi (LBB) condition (see (3.74) below). Spatial discretizations that do not satisfy the inf-sup condition usually result in degradation of the accuracy in pressure. This is well documented for steady state [BS] and dynamic calculations (see, for example, [We]). A well-known example is the $Q_1-P_0$ element with LBB constant $\beta_h = O(h)$ [BN]. In contrast, GMAC is staggered and supported the same way as the $Q_1-P_0$ element (and a few other finite difference schemes that fail to satisfy the inf-sup condition), except GMAC results in a different nine-point differencing formula for the viscous term. Nevertheless, it can be shown [HLW] that GMAC admits a uniform LBB estimate with an $O(1)$ lower bound for smooth grids in two dimensions. We also give strong numerical evidence for this assertion. See sections 3.5 and 4 for details.

Related works on finite difference and finite volume methods on mapped grids include those designed for the Navier–Stokes equation in primitive variables on nonstaggered or fully staggered curvilinear grids [ZSK, GeS, LB, TC]. See also [NC, CDHM, Ar, ArL, LWa, HS] for numerical methods in curvilinear coordinates on the Navier–Stokes equation with the vorticity-stream formulation, convection-diffusion equation, and other linear equations. In addition, a vast amount of research works have contributed to the development of computational incompressible flows in many aspects, for example, parallel implementation on large-scale simulation [OOB, KK], adaptive refinement techniques [ABCHW, RBLCB], fast iterative methods [ESW], multigrid methods [BHM, TOS], and domain decomposition [TW]. For an overview of recent developments in CFD, see, for example, [DFM, GS, KS, FP] and the references therein.

The rest of the paper is organized as follows. In section 2, we review the classical MAC scheme, the boundary treatment, and discrete Hodge decomposition. In section 3, we describe our GMAC scheme in curvilinear coordinates. The velocity components are at the same location for convenience in both programming and applications. Representation of the Navier–Stokes equation in a skewed local coordinate leads to a natural discretization that gives rise to desired crucial discrete identities. One of the key issues here is to incorporate the boundary conditions into the finite difference operators so that exact summation by parts identity holds. The key
properties of the MAC scheme and the exact discrete Hodge decomposition mentioned above are retained here even for the nonhomogeneous boundary conditions. The three-dimensional (3D) version of our scheme, as well as possible variants, are documented in the appendix. Finally, we perform a systematic numerical test and report the results in section 4.

2. Classical MAC scheme, energy identity, and Hodge decomposition.
In this section, we review the classical MAC scheme and fundamental discrete identities associated with it. We first recall the essential ingredients that lead to the energy estimate and the well-posedness of the Navier–Stokes equation (1.1)–(1.3). Take the inner product with $u$ on both sides of (1.1) and then integrate over $\Omega$ using the following facts:

\begin{align}
(2.1) \quad & \langle u, \nabla p \rangle = -\langle \nabla \cdot u, p \rangle, \\
(2.2) \quad & \text{the vector Laplacian } \nabla^2 \text{ is symmetric and nonpositive,} \\
(2.3) \quad & \langle N(u), u \rangle = 0,
\end{align}

where $N(u) = u \cdot \nabla u$ and $\langle u, v \rangle = \int_{\Omega} u \cdot v \, dx$. It is easy to obtain the basic energy estimate (for $f \equiv 0$):

\begin{equation}
(2.4) \quad \frac{1}{2} \frac{d}{dt} \| u \|^2 + \nu \| \nabla u \|^2 = 0.
\end{equation}

It is therefore desirable if a numerical scheme can preserve the discrete analogue of (2.1)–(2.3) and therefore guarantee the stability of the scheme. A well-known example satisfying (2.1), (2.2) is the classical MAC scheme [Le, DHSW, HW], where the pressure and the components of the velocity field are placed on staggered grids in such a way that second order centered difference, divergence-free constraint, and no-slip, no-penetration boundary conditions all fit naturally and elegantly with the placement of the variables (Figure 2.1).

The fully discrete MAC scheme with explicit treatment for the viscous term is given by

\begin{equation}
\left\{ \begin{array}{l}
\frac{u^{n+1} - u^n}{\Delta t} + N_h(u^n) + \nabla_h p^{n+1} = \nu \nabla_h^2 u^n + f^n \quad \text{on } \rightarrow \text{ and } \uparrow, \\
\nabla_h \cdot u^{n+1} = 0 \quad \text{on } \bullet, \\
u^{n+1} = 0 \quad \text{on } \Gamma.
\end{array} \right.
\end{equation}

Fig. 2.1. The MAC grid.
The MAC scheme (2.5) is to be completed with extensions on the ghost points outside the computational domain (for example, on the left of \( i = 0 \)),

\[
(2.6) \quad v_{-\frac{1}{2},j}, \quad p_{-\frac{1}{2},j-\frac{1}{2}}.
\]

The ghost values in (2.6) are needed to enforce the tangential component of the no-slip boundary condition, to evaluate the viscous term and nonlinear term near the boundary, and to impose an artificial pressure boundary condition, which is the central issue in the pressure Poisson formulation and the key to decoupling the pressure from the momentum equation.

Alternatively, the staggered placement of the variables offers an equivalent interpretation of (2.5) without resorting to the ghost values and artificial pressure boundary condition. This is realized by introducing the reduced operator (cf. Anderson [An]) that retains a fraction of the original finite difference operators and incorporates the no-slip condition on the boundary grids.

Using the reduced operators (denoted as primed operators), one can recast (2.5) as

\[
(2.7) \quad \begin{cases}
\frac{u^{n+1} - u^n}{\Delta t} + N_h(u^n) + \nabla_h p^{n+1} = \nu \nabla^2_h u^n + f^n & \text{on } \to \text{ and } \uparrow, \\
\nabla_h \cdot u^{n+1} = 0 & \text{on } \bullet,
\end{cases}
\]

Note that in (2.7) the system in self-contained. The ghost values (2.6) are no longer needed. One can therefore identify the pressure gradient as discrete Hodge projection of the acceleration terms onto the orthogonal complement of the divergence-free subspace,

\[
(2.8) \quad \frac{u^{n+1}}{\Delta t} + \nabla_h p^{n+1} = \frac{u^n}{\Delta t} - N_h(u^n) + \nu \nabla^2_h u^n + f^n, \quad \nabla_h \cdot u^{n+1} = 0.
\]

Moreover, the pressure can be easily decoupled from (2.8). No artificial pressure boundary condition is needed. (The curvilinear analogue of the reduced operators and aforementioned properties will be explained in section 3.2 and therefore are not detailed here.) The resulting scheme is robust and efficient. However, the restriction of Cartesian grids has limited the applicability and popularity of the MAC scheme. In addition, issues of high order time discretization and cell Reynolds number constraint have raised controversy and were not fully understood until the 1990s [EL]. In (2.5), we have illustrated these issues using first order forward Euler discretization. Proper high order time discretization and its connection with the cell Reynolds number can be found in [EL].

Motivated by the success of the MAC scheme, we propose in this work a second order finite difference scheme for the Navier–Stokes equation on curvilinear domains. The velocity field and pressure are placed on cell centers and grid points, respectively. With a set of skewed coordinates, the Navier–Stokes equation can be discretized naturally in such a way that the discrete analogue of (2.1)–(2.3) remain valid. As a consequence, the resulting scheme admits a discrete energy estimate. In addition, the scheme preserves the vector identities (1.4) in discrete settings. This is key to discrete Hodge decomposition and plays an essential role in the efficiency as well as rigorous error analysis of our scheme.

Another appealing feature of the MAC scheme is about the following generalized Stokes system \((\alpha \geq 0)\):

\[
(2.9) \quad \begin{cases}
(\alpha - \nabla^2_h) u + \nabla_h p = f & \text{on } \to \text{ and } \uparrow, \\
\nabla_h \cdot u = 0 & \text{on } \bullet.
\end{cases}
\]
Equation (2.9) arises naturally from steady state calculation or partially implicit time stepping for the Navier–Stokes equation on a MAC grid. The solvability of (2.9) and uniform bound of the solution operator are direct consequences of the LBB condition. The verification of the LBB condition is vital to pressure error estimates for both static and dynamic problems and is a long-standing open problem for the MAC scheme. In [HWu], the authors constructed a new finite element method for the Stokes system based on three sets of tessellation of rectangular cells. The resulting scheme combined with quadrature formulas corresponds to the classical MAC scheme (2.9) equipped with the Orszag–Israel vorticity boundary condition [OI] (see also [EL, Table I]),

\begin{equation}
\begin{align*}
    u_{0,j+\frac{1}{2}} = 0, \\
    v_{-\frac{1}{2},j} = -2v_{-\frac{1}{2},j} + \frac{1}{3}v_{-\frac{1}{2},j},
\end{align*}
\end{equation}

and satisfies the uniform LBB estimate. This method is later extended to Navier–Stokes equation in [HY]. The issue of a uniform LBB estimate for a MAC scheme with the original reflection boundary condition, \( v_{-\frac{1}{2},j} = -v_{-\frac{1}{2},j} \), will be addressed in a forthcoming paper. An alternative approach for the pressure error estimate on the MAC scheme using high order Strang’s expansion can be found in [HW].

3. Generalized MAC scheme on curvilinear domains. The generalized MAC scheme is based on discretizing the Navier–Stokes equation in rotational form:

\begin{equation}
\begin{align*}
    u_t + \omega \times u + \nabla p &= -\nu \nabla \times \omega + f \\
    \omega &= \nabla \times u \\
    \nabla \cdot u &= 0 \\
    u &= 0
\end{align*}
\end{equation}

in \( \Omega \),

\begin{align*}
    \nabla \cdot u &= 0 \\
    u &= 0
\end{align*}

on \( \Gamma \).

On a curvilinear domain, we place all three components of the velocity on cell centers \( x(\xi_{1i}, \xi_{2j}, \xi_{3k}) \), while the pressure and the vorticity components are placed on the grid points \( x(\xi_{1i}, \xi_{2j}, \xi_{3k}) \). Here \( (\xi_{1i}, \xi_{2j}, \xi_{3k}) \) is the coordinate in the computational domain and \( x \) is the position vector in physical domain \( \Omega \).

In addition to generalization to curvilinear coordinates, GMAC differs from the classical MAC scheme in the placement of the velocity components. One advantage of placing all three components of velocity at the same place is that (2.3) can be naturally realized in the discrete setting. Together with other vector identities, this ensures the stability of GMAC. More importantly, the resulting discrete Laplacian for \( p \) is self-adjoint and nonpositive definite as long as the cells are nonsingular, regardless of the regularity of the grids. See the discussion in section 3.4. For the fully staggered case, where different components of \( u \) are placed on different positions as in the classical MAC case, positivity of the pressure equation on curvilinear domains may require extra assumption on smoothness of the grids [BD, pp. 147–150].

For simplicity of presentation, we start with the two-dimensional (2D) case:

\begin{equation}
\begin{align*}
    u_t + \omega u^\perp + \nabla p &= \nu \nabla^\perp \omega + f \\
    \omega &= \nabla^\perp \times u \\
    \nabla \cdot u &= 0 \\
    u &= 0
\end{align*}
\end{equation}

in \( \Omega \),

\begin{align*}
    \nabla \cdot u &= 0 \\
    u &= 0
\end{align*}

on \( \Gamma \).

The discretization of (3.2) and boundary treatment in the curvilinear coordinate will be explained in detail in following subsections.
3.1. Differential operators in curvilinear coordinate. In the 2D case, \( x = (x, y) \) is the position vector in the physical domain and \((\xi^1, \xi^2)\) the coordinate in the computational domain with mesh size \( \Delta \xi^1 = h_1 \) and \( \Delta \xi^2 = h_2 \). We further introduce a new set of coordinates in the skewed direction by

\[
\xi^1 := \frac{h_2 \xi^1 + h_1 \xi^2}{\sqrt{h_1^2 + h_2^2}}, \quad \xi^2 := \frac{-h_2 \xi^1 + h_1 \xi^2}{\sqrt{h_1^2 + h_2^2}},
\]

as illustrated in Figure 3.1.

Once the a local coordinate is chosen, the intrinsic differential operators can be determined following standard procedure. Throughout this paper, we use \( \times \) to emphasis that the corresponding quantities are computed in the skewed variables \( \times \xi^\alpha \).

Denote by

\[
\times e_1 = \frac{\partial x}{\partial \xi^1}, \quad \times e_2 = \frac{\partial x}{\partial \xi^2},
\]

\[
\times e^1 = \nabla \xi^1, \quad \times e^2 = \nabla \xi^2;
\]

the metric tensors with respect to the skewed coordinate \((\xi^1, \xi^2)\) are then given by

\[
\times g^{\mu\nu} = \times e^\mu \cdot \times e^\nu, \quad \times g_{\mu\nu} = \times e_\mu \cdot \times e_\nu, \quad \mu, \nu = 1, 2,
\]

and

\[
\times g := \text{det}(\times g_{\mu\nu}).
\]

The following identities follow immediately from the definition:

\[
\sqrt{\times g} = \text{det} \left( \frac{\partial x}{\partial \xi} \right),
\]

\[
\sum_{\gamma=1}^{2} \times g^{\mu\gamma} \times g_{\gamma\nu} = \delta^\mu_\nu.
\]

We use \((\tilde{u}^1, \tilde{u}^2)\) and \((\tilde{u}^1, \tilde{u}^2)\) to denote the covariant and contravariant components of a vector field \( u \) in the \( \times \xi^\alpha \) coordinate:

\[
u = \times u^1 \times e_1 + \times u^2 \times e_2 = \times u^1 \times e^1 + \times u^2 \times e^2.
\]
The transformation between covariant and contravariant components is given by

\begin{equation}
\tilde{u}^\mu = \sum_{\gamma=1}^{2} g^{\mu\gamma} \tilde{u}_\gamma, \quad \tilde{u}_\nu = \sum_{\gamma=1}^{2} g_{\gamma\nu} \tilde{u}^\gamma, \quad \mu, \nu = 1, 2.
\end{equation}

We summarize relevant formulas for (3.2) as follows:

\begin{align*}
\nabla p &= \frac{\partial p}{\partial \xi^1} e_1 + \frac{\partial p}{\partial \xi^2} e_2, \\
\nabla^\perp \omega &= \frac{1}{\sqrt{g}} \left( -\frac{\partial \omega}{\partial \xi^1} e_1 + \frac{\partial \omega}{\partial \xi^2} e_2 \right), \\
\nabla \cdot \mathbf{u} &= \frac{1}{\sqrt{g}} \left( \frac{\partial}{\partial \xi^1} \left( \sqrt{g} u^1 \right) + \frac{\partial}{\partial \xi^2} \left( \sqrt{g} u^2 \right) \right), \\
\nabla^2 p &= \nabla \cdot \nabla p = \frac{1}{\sqrt{g}} \sum_{\mu, \nu = 1}^{2} \frac{\partial}{\partial \xi^\mu} \left( \sqrt{g} g^{\mu\nu} \frac{\partial p}{\partial \xi^\nu} \right), \\
\mathbf{u}^\perp &= \sqrt{g} (-\tilde{u}^2 e_1 + \tilde{u}^1 e_2) = \frac{1}{\sqrt{g}} (-\tilde{u}_2 e_1 + \tilde{u}_1 e_2), \\
\omega &= \nabla^\perp \cdot \mathbf{u} = \frac{1}{\sqrt{g}} \left( \frac{\partial \tilde{u}_2}{\partial \xi^1} - \frac{\partial \tilde{u}_1}{\partial \xi^2} \right).
\end{align*}

Whenever necessary, the covariant and contravariant components can be converted to each other using (3.9). For example,

\begin{equation}
\nabla p = \left( \frac{\partial p}{\partial \xi^1} \tilde{g}^{11} + \frac{\partial p}{\partial \xi^2} \tilde{g}^{21} \right) \tilde{e}_1 + \left( \frac{\partial p}{\partial \xi^1} \tilde{g}^{12} + \frac{\partial p}{\partial \xi^2} \tilde{g}^{22} \right) \tilde{e}_2.
\end{equation}

### 3.2. Spatial discretization

The choice of the skewed coordinate is motivated by the jump condition capturing scheme developed for elliptic interface problems [Wa]. It was shown that using the skewed coordinate as independent variables, the interface jump conditions can be naturally incorporated into the finite difference operator. The resulting scheme is symmetric, definite, and second order accurate even when the diffusion coefficient has a jump continuity across the material interface.

We now give a detailed description of the spatial discretization of our scheme, starting with the metric tensor. In the case that the coordinate mapping \((\xi^1, \xi^2) \mapsto (x, y)\) is explicitly given, one can compute the metric tensors from (3.4)–(3.8) and use it in the discretizations (3.30)–(3.36) below. Alternatively, one can compute the numerical metric tensors \(\tilde{g}_{\alpha\beta}^h\) from straightforward centered difference:

\begin{align*}
(g_{11})_{i+\frac{1}{2},j+\frac{1}{2}} &:= \frac{x_{i+1,j+1} - x_{i,j}}{h}, \quad (g_{12})_{i+\frac{1}{2},j+\frac{1}{2}} := \frac{x_{i+1,j+1} - x_{i+1,j}}{h}, \\
(g_{21})_{i+\frac{1}{2},j+\frac{1}{2}} &:= \frac{x_{i+1,j+1} - x_{i,j}}{h}, \quad (g_{22})_{i+\frac{1}{2},j+\frac{1}{2}} := \frac{x_{i+1,j+1} - x_{i+1,j}}{h}.
\end{align*}
where
\[ \tilde{h} := \frac{2h_1 h_2}{\sqrt{h_1^2 + h_2^2}} = \Delta \hat{\xi} = \Delta \hat{\xi}^2 \]
is the mesh size in the skewed directions \( \hat{\xi} \) and \( \hat{\xi}^2 \). Note that the indices \( i, j \) refer to the \( \xi \) variables, not the \( \hat{\xi} \) ones.

The numerical contravariant components (3.8) are defined through analogue of (3.8):
\[ (\tilde{g}_{\hat{\alpha \hat{\beta}}}) := (g_{\mu \nu})^{-1}, \]
that is,
\[ \sum_{\lambda=1}^{2} \tilde{g}^{\mu \lambda} \tilde{g}_{h_{\lambda \nu}} = \delta^\mu_\nu. \]
The numerical Jacobian on cell centers and grids is given by
\[ g_{h_{i,j}} := \det \begin{pmatrix} \tilde{g}_{h_{11}} & \tilde{g}_{h_{12}} \\ \tilde{g}_{h_{21}} & \tilde{g}_{h_{22}} \end{pmatrix} \]
and
\[ \sqrt{g_{h_{i,j}}} := \frac{1}{4} \left( \sqrt{g_{h_{i+\frac{1}{2},j+\frac{1}{2}}} + \sqrt{g_{h_{i+\frac{1}{2},j-\frac{1}{2}}} + \sqrt{g_{h_{i-\frac{1}{2},j+\frac{1}{2}}}} + \sqrt{g_{h_{i-\frac{1}{2},j-\frac{1}{2}}}} \right). \]

Next, we introduce the discrete grad, div, curl, and Laplacian. With a semi-staggered placement of the variables \( u, p, \) and \( \omega \) (see Figure 3.4), it is straightforward to discretize (3.2) using centered difference. The crucial issue here is the boundary treatment of these operators, which plays an essential role in the stability of our scheme. We define the one-dimensional reduced finite difference operator by
\[ (D' b)_i := \begin{cases} \frac{b_{i-\frac{1}{2}}}{h} & i = 0, \\ \frac{b_{i+\frac{1}{2}} - b_{i-\frac{1}{2}}}{h} & 1 \leq i \leq M - 1, \\ \frac{0 - b_{M+\frac{1}{2}}}{h} & i = M. \end{cases} \]

A direct consequence of the reduced differencing is the exact summation by parts identity as observed in [An].

**Proposition 1.**
\[ h \sum_{i=1}^{M} b_{i-\frac{1}{2}} (D a)_{i-\frac{1}{2}} = -h \sum_{i=0}^{M} a_i (D' b)_i, \]
where the primed sum denotes half weight on the boundary:
\[ \sum_{i=0}^{M} a_i = \frac{1}{2} a_0 + \sum_{i=1}^{M-1} a_i + \frac{1}{2} a_M. \]
and denote by $L^2(\bar{\Omega}_g, \mathbb{R})$ the collection of real valued functions on $\bar{\Omega}_g$ and $L^2(\Omega_c, \mathbb{R}^2)$ the vector fields on $\Omega_c$:

\begin{align}
(3.21) & \quad L^2(\bar{\Omega}_g, \mathbb{R}) := \{ \omega : \bar{\Omega}_g \to \mathbb{R} \}, \\
(3.22) & \quad L^2(\Omega_c, \mathbb{R}) := \{ u : \Omega_c \to \mathbb{R} \}, \\
(3.23) & \quad L^2(\Omega_c, \mathbb{R}^2) := \{ u : \Omega_c \to \mathbb{R}^2 \},
\end{align}

and

\begin{align}
(3.24) & \quad L^2(\bar{\Omega}_g, \mathbb{R})/\mathbb{R}^2 := \left\{ p \in L^2(\bar{\Omega}_g, \mathbb{R}) \left| \sum_{\bar{\Omega}_{go}} \{ (\sqrt{g_{ij}}) \}_{i,j} = 0 = \sum_{\bar{\Omega}_{go}} \{ (\sqrt{g_{ij}}) \}_{i,j} \right. \right\}, \\
(3.25) & \quad L^2(\bar{\Omega}_g, \mathbb{R}) := \{ \psi \in L^2(\bar{\Omega}_g, \mathbb{R}) \left| \psi = \text{constant on each connected component of } \Gamma \right. \}.
\end{align}

In view of Proposition 1, it is natural to define the reduced difference operator in the skewed variables as follows.

**Definition 1.**

1. For $a \in L^2(\bar{\Omega}_g, \mathbb{R})$,

\begin{align}
(3.26) & \quad (\tilde{D}_1 a)_{i-\frac{1}{2},j-\frac{1}{2}} := \frac{a_{i,j} - a_{i-1,j-1}}{\hat{h}}, \quad (\tilde{D}_2 a)_{i-\frac{1}{2},j-\frac{1}{2}} := \frac{a_{i-1,j} - a_{i,j-1}}{\hat{h}},
\end{align}

2. For $b \in L^2(\Omega_c, \mathbb{R})$,

\begin{align}
(3.27) & \quad (\tilde{D}_1 b)_{i,j} := \begin{cases} 
\frac{b_{i+\frac{1}{2},j+\frac{1}{2}} - b_{i-\frac{1}{2},j-\frac{1}{2}}}{h}, & 0 < i < M, \ 0 < j < N; \\
\frac{b_{i,j+\frac{1}{2}}}{\frac{1}{2}h}, & i = 0, \ 0 < j < N; \\
\frac{b_{i+\frac{1}{2},j}}{\frac{1}{2}h}, & (i,j) = (0,0); \\
1, & (i,j) = (0,N);
\end{cases}
\end{align}
is given by

\[
\begin{align*}
(\hat{D}_2 b)_{i,j} := \begin{cases} 
\frac{b_{i+\frac{1}{2},j+\frac{1}{2}} - b_{i+\frac{1}{2},j-\frac{1}{2}}}{\frac{1}{2}h}, & 0 < i < M, \ 0 < j < N; \\
-\frac{b_{i-\frac{1}{2},j+\frac{1}{2}}}{\frac{1}{2}h}, & i = 0, \ 0 < j < N; \\
0, & (i,j) = (0,0); \\
\frac{b_{i+\frac{1}{2},N-\frac{1}{2}}}{\frac{1}{2}h}, & (i,j) = (0, N).
\end{cases}
\end{align*}
\]

In Definition 1 and in the rest of the paper, we use the prime to indicate that the reduced difference is applied on the boundary when the standard finite difference requires grid points outside the computational domain.

The (reduced) finite difference in the $\hat{x}$ variables extends naturally to the discrete grad, div, curl, and Laplacian operators as follows.

**Definition 2.** For $a \in L^2(\Omega_g, \mathbb{R})$, we define

\[
\tilde{\nabla}_h : L^2(\Omega_g, \mathbb{R}) \to L^2(\Omega_c, \mathbb{R}^2), \quad \tilde{\nabla}_h a := (D_1 a) \hat{e}_1 + (D_2 a) \hat{e}_2,
\]

and

\[
\tilde{\nabla}_h^\perp : L^2(\Omega_g, \mathbb{R}) \to L^2(\Omega_c, \mathbb{R}^2), \quad \tilde{\nabla}_h^\perp a := -\frac{D_2 a}{\sqrt{g_h}} \hat{e}_1 + \frac{D_1 a}{\sqrt{g_h}} \hat{e}_2.
\]

**Definition 3.** Let $u = \hat{u}^1 \hat{e}_1 + \hat{u}^2 \hat{e}_2 \in L^2(\Omega_c, \mathbb{R}^2)$. We define

\[
\hat{\nabla}_h' : L^2(\Omega_c, \mathbb{R}^2) \to L^2(\Omega_g, \mathbb{R}), \quad \hat{\nabla}_h' u := \frac{1}{\sqrt{g_h}} \left( \hat{D}_1' \left( \sqrt{g_h} \hat{u}^1 \right) + \hat{D}_2' \left( \sqrt{g_h} \hat{u}^2 \right) \right),
\]

and

\[
\hat{\nabla}_h'^\perp : L^2(\Omega_c, \mathbb{R}^2) \to L^2(\Omega_g, \mathbb{R}), \quad \hat{\nabla}_h'^\perp u := \frac{1}{\sqrt{g_h}} (\hat{D}_1' \hat{u}_2 - \hat{D}_2' \hat{u}_1).
\]

The realization of (3.31) and (3.32) on typical interior, boundary, and corner points is given by

\[
(\hat{\nabla}_h' u)_{i,j} = \begin{cases} 
\frac{1}{\sqrt{g_h}} \left( \sqrt{\frac{1}{g_h}} \hat{u}^1_{i+\frac{1}{2},j+\frac{1}{2}} - \sqrt{\frac{1}{g_h}} \hat{u}^1_{i+\frac{1}{2},j-\frac{1}{2}} \right), & 0 < i < M, \ 0 < j < N; \\
\frac{2}{\sqrt{g_h}} \left( \sqrt{\frac{1}{g_h}} \hat{u}^1_{\frac{i+\frac{1}{2},j+\frac{1}{2}} - \sqrt{\frac{1}{g_h}} \hat{u}^1_{\frac{i+\frac{1}{2},j-\frac{1}{2}}} \right), & i = 0, \ 0 < j < N; \\
\frac{4}{\sqrt{g_h}} \left( \sqrt{\frac{1}{g_h}} \hat{u}^1_{\frac{i+\frac{1}{2}}{j+\frac{1}{2}} \right), & (i,j) = (0,0);
\end{cases}
\]

\[
(\hat{\nabla}_h'^\perp u)_{i,j} = \begin{cases} 
\frac{1}{\sqrt{g_h}} \left( \sqrt{\frac{1}{g_h}} \hat{u}^1_{i+\frac{1}{2},j+\frac{1}{2}} \right), & 0 < i < M, \ 0 < j < N; \\
\frac{2}{\sqrt{g_h}} \left( \sqrt{\frac{1}{g_h}} \hat{u}^1_{\frac{i+\frac{1}{2},j+\frac{1}{2}} - \sqrt{\frac{1}{g_h}} \hat{u}^1_{\frac{i+\frac{1}{2},j-\frac{1}{2}}} \right), & i = 0, \ 0 < j < N; \\
\frac{4}{\sqrt{g_h}} \left( \sqrt{\frac{1}{g_h}} \hat{u}^1_{\frac{i+\frac{1}{2}}{j+\frac{1}{2}} \right), & (i,j) = (0,0);
\end{cases}
\]

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where
\( (3.34) \)
\[
(\nabla_h^L \cdot u)_{i,j} = \begin{cases} 
\frac{1}{\sqrt{g_{h,j}}} \left( \frac{(\hat{u}_2)_{i+\frac{1}{2},j+\frac{1}{2}} - (\hat{u}_1)_{i+\frac{1}{2},j}}{h} \right), & 0 < i < M, \ 0 < j < N; \\
\frac{2}{\sqrt{g_{h,j}}} \left( \frac{(\hat{u}_2)_{i+\frac{1}{2},j+\frac{1}{2}} + (\hat{u}_1)_{i+\frac{1}{2},j-\frac{1}{2}}}{h} \right), & i = 0, \ 0 < j < N; \\
\frac{4}{\sqrt{g_{h,j}}} \left( \frac{(\hat{u}_2)_{i+\frac{1}{2},j+\frac{1}{2}}}{h} \right), & (i,j) = (0,0).
\end{cases}
\]

See also Figures 3.2 and 3.3.

Finally, the discrete Laplacian is defined in a similar way. (See also [Wa] for the reduced Laplacian in the skewed variables for elliptic interface problems.)

**DEFINITION 4.** For \( a \in L^2(\Omega_h, \mathbb{R}) \), define
\[
(3.35) \quad \triangle_h^L : L^2(\Omega_h, \mathbb{R}) \rightarrow L^2(\Omega_h, \mathbb{R}), \quad \triangle_h^L a = \frac{1}{\sqrt{g_h}} \sum_{\mu,\nu=1}^2 D_{\mu} \left( \sqrt{g_h} \tilde{a}_{\mu} D_{\nu} a \right).
\]

That is,
\[
(3.36) \quad (\triangle_h^L a)_{i,j} = \begin{cases} 
\frac{1}{\sqrt{g_{h,i}}} \left( \frac{(\hat{q}_h^1 D_{1a} + \hat{q}_h^2 D_{2a})_{i+\frac{1}{2},j}}{h} + \frac{(\hat{q}_h^1 D_{1a} + \hat{q}_h^2 D_{2a})_{i-\frac{1}{2},j}}{h} \right), & 0 < i < M ; \\
\frac{2}{\sqrt{g_{h,i}}} \left( \frac{(\hat{q}_h^1 D_{1a} + \hat{q}_h^2 D_{2a})_{i+\frac{1}{2},j}}{h} - \frac{(\hat{q}_h^1 D_{1a} + \hat{q}_h^2 D_{2a})_{i-\frac{1}{2},j}}{h} \right), & i = 0 ; \\
\frac{4}{\sqrt{g_{h,i}}} \left( \frac{(\hat{q}_h^1 D_{1a} + \hat{q}_h^2 D_{2a})_{i+\frac{1}{2},j}}{h} \right), & (i,j) = (0,0),
\end{cases}
\]

where \( q_h^\alpha = \sqrt{g_h g_h^\alpha} \).
We are now ready to state the key lemma associated with the reduced difference operators. Define the discrete inner products

\begin{equation}
\langle \mathbf{u}, \mathbf{v} \rangle_{\Omega_h} = h_1 h_2 \sum_{i=0}^{M-1} \sum_{j=0}^{N-1} \left( \mathbf{u} \cdot \mathbf{v} \right) \sqrt{g_h} \right)_{i+\frac{1}{2}, j+\frac{1}{2}} = h_1 h_2 \sum_{i=0}^{M-1} \sum_{j=0}^{N-1} \left( {\hat{u}^1 v^1 + \hat{u}^2 v^2} \right) \sqrt{g_h} \right)_{i+\frac{1}{2}, j+\frac{1}{2}} \quad \mathbf{u}, \mathbf{v} \in L^2(\Omega_h, \mathbb{R}^2),
\end{equation}

and the corresponding norms

\[ \| \mathbf{u} \|^2_{\Omega_h} = \langle \mathbf{u}, \mathbf{u} \rangle_{\Omega_h}, \quad \| a \|^2_{\Omega_h} = \langle a, a \rangle_{\Omega_h}, \]

where \( \sqrt{g_h} := \frac{h_1}{h_1^2 + h_2^2} \sqrt{g_h} \) is the numerical Jacobian with respect to the default coordinate \((\xi^1, \xi^2)\). Applying Proposition 1 in the skewed directions \( \xi^1 \) and \( \xi^2 \), it is easy to derive the following discrete identities.

**Lemma 3.1.** Let \( \mathbf{u} \in L^2(\Omega_h, \mathbb{R}^2) \) and \( a \in L^2(\Omega_h, \mathbb{R}) \). We have

1. \begin{equation}
\langle \mathbf{u},  \hat{\nabla}_h a \rangle_{\Omega_h} = -\langle \hat{\nabla}_h^t \cdot \mathbf{u}, a \rangle_{\Omega_h}, \end{equation}

2. \begin{equation}
\langle \mathbf{u},  \hat{\nabla}_h^t a \rangle_{\Omega_h} = -\langle \hat{\nabla}_h^t \cdot \mathbf{u}, a \rangle_{\Omega_h}, \end{equation}

3. \begin{equation}
\hat{\nabla}_h^t \cdot \hat{\nabla}_h a = \hat{\nabla}_h \cdot \hat{\nabla}_h a = \hat{\Delta}_h a \quad \text{on} \ \bar{\Omega}_g.
\end{equation}

4. If \( a \in L^2(\Omega_g, \mathbb{R}) \), then

\begin{equation}
\hat{\nabla}_h \cdot \hat{\nabla}_h a = \hat{\nabla}_h \cdot \hat{\nabla}_h a = 0 \quad \text{on} \ \bar{\Omega}_g.
\end{equation}

In addition, if \( a \in L^2(\Omega_g, \mathbb{R}) \), then

\begin{equation}
\hat{\nabla}_h \cdot \hat{\nabla}_h a = \hat{\nabla}_h \cdot \hat{\nabla}_h a = 0 \quad \text{on} \ \bar{\Omega}_g.
\end{equation}

**3.3. Generalized MAC scheme.** On a curvilinear domain, we place both components of the velocity on cell centers \( x(\xi^1_i + \frac{1}{2}, \xi^2_j + \frac{1}{2}) \), while the pressure and the vorticity are both placed on the grid points \( x(\xi^1_i, \xi^2_j) \), as shown in Figure 3.4.

The generalized MAC scheme with \( \mathbf{u} \) defined on cell centers (GMACc) can be summarized as follows.

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The GMACc scheme. Solve for $u \in C^1([0,T]; L^2(\Omega_c, \mathbb{R}^2))$ and $p \in C^0([0,T]; L^2(\Omega_g, \mathbb{R}))$ such that

$$
\begin{align*}
\dot{u}_t + \bar{\omega} u + \bar{\nabla} h p &= \nu \bar{\nabla}^2 \omega + f \quad \text{on } \Omega_c, \\
\omega &= \bar{\nabla}^2 \cdot u \quad \text{on } \Omega_g, \\
\bar{\nabla}^2 \cdot u &= 0 \quad \text{on } \Omega_g,
\end{align*}
$$

(3.44)

where $\bar{\omega}_{i-\frac{1}{2},j-\frac{1}{2}} = \frac{1}{4}(\omega_{i,j} + \omega_{i-1,j} + \omega_{i,j-1} + \omega_{i-1,j-1})$.

Another version of the generalized MAC scheme, GMACg, with $u$ defined on grids and $p$ on cell centers, is detailed in Appendix A.

As a direct consequence of (3.39)–(3.40), we have the following discrete energy estimate for (3.44).

**Lemma 3.2.** Let $(u, p, \omega)$ be a solution to (3.44) with $f = 0$; then

$$
\frac{1}{2} \frac{d}{dt} \|u\|_{\Omega_c}^2 + \nu \|\omega\|_{\Omega_g}^2 = 0.
$$

It is worth noting that the reduced divergence operator in the third equation of (3.44) has implicitly incorporated the no-penetration condition $u \cdot n = 0$ in a natural way. On the other hand, the reduced curl operator in the second equation of (3.44) has implicitly incorporated the no-slip condition $u \times n = 0$ on $\Gamma_g$. This can be interpreted as a local vorticity boundary condition.

### 3.3.1. Nonhomogeneous boundary velocity.

In case of inhomogeneous boundary velocity $u = u_b$ on $\Gamma$, such as the driven cavity flow, the corresponding reduced operators require proper modification.

Let $F \in L^2(\Omega_c, \mathbb{R})$ and $f \in L^2(\Gamma_c, \mathbb{R})$. With slight abuse of notation, we define the extended reduced operators, still denoted by $\tilde{D}_1', \tilde{D}_2'$, as follows:

$$
\begin{align*}
\tilde{D}_1' : L^2(\Omega_c, \mathbb{R}) \times L^2(\Gamma_c, \mathbb{R}) &\rightarrow L^2(\Omega_g, \mathbb{R}), \\
\tilde{D}_2' : L^2(\Omega_c, \mathbb{R}) \times L^2(\Gamma_c, \mathbb{R}) &\rightarrow L^2(\Omega_g, \mathbb{R}),
\end{align*}
$$

(3.45)

$$
\begin{align*}
\tilde{D}_1'(F \oplus f)_{i,j} := & \begin{cases} \\
F_{i+\frac{1}{2},j+\frac{1}{2}} - F_{i-\frac{1}{2},j-\frac{1}{2}} & \text{on } \Omega_g; \\
\frac{1}{h} f_{i+\frac{1}{2},j} & \text{for } (i,j) = (0,0); \\
F_{i+\frac{1}{2},\frac{1}{2}} - \frac{1}{2} (f_{0+\frac{1}{2}} + f_{i-\frac{1}{2},0}) & \quad 0 < i < M, \ j = 0;
\end{cases}
\end{align*}
$$

(3.46)
One can define 

\begin{equation}
\begin{cases}
F_i - \frac{j}{2} + \frac{1}{2} - F_i + \frac{j}{2} - \frac{1}{2} & \text{on } \hat{\Omega}_h; \\
\frac{F_i - \frac{j}{2} + \frac{1}{2} - \frac{1}{2}(f_i + \frac{1}{2}), 0) & 0 < i < M, j = 0; \\
f_0 - f_0 & (i, j) = (0, 0).
\end{cases}
\end{equation}

One can define \( \hat{\nabla}_h \cdot \) and \( \hat{\nabla}_h^{1/2} : L^2(\Omega_c, \mathbb{R}^2) \times L^2(\Gamma_c, \mathbb{R}^2) \mapsto L^2(\hat{\Omega}_h, \mathbb{R}) \) in a natural way.

When the boundary data is identically zero, (3.46)–(3.47) reduce to the original reduced operators,

\begin{equation}
\hat{D}_\alpha' (F \oplus 0) = \hat{D}_\alpha' F,
\end{equation}

\begin{equation}
\hat{\nabla}_h' \cdot (u \oplus 0) = \hat{\nabla}_h \cdot u, \quad \hat{\nabla}_h^{1/2} \cdot (u \oplus 0) = \hat{\nabla}_h^{1/2} \cdot u.
\end{equation}

The corresponding scheme for inhomogeneous boundary velocity is given by

\begin{equation}
\begin{split}
\hat{\nabla}_h' \cdot (u \oplus b) &= \nu \hat{\nabla}_h' \cdot \omega + \nabla_h' \cdot f \\
\omega &= \hat{\nabla}_h^{1/2} \cdot (u \oplus u_b) \\
\hat{\nabla}_h' \cdot (u \oplus u_b) &= 0
\end{split}
\end{equation}

subject to the following compatibility condition for the boundary velocity \( u_b \):

\begin{equation}
\langle 1_{\hat{\Omega}_h}, \hat{\nabla}_h' \cdot (0 \oplus u_b) \rangle_{\hat{\Omega}_h} = 0 = \langle 1_{\hat{\Omega}_h}, \hat{\nabla}_h' \cdot (0 \oplus u_b) \rangle_{\hat{\Omega}_h}.
\end{equation}

To see this, we first note that the kernel of \( \hat{\nabla}_h \cdot \) consists of linear combinations of indicator functions of \( \Omega_{ge} \) and \( \hat{\Omega}_{ge} \):

\begin{equation}
\ker(\hat{\nabla}_h) = \text{span}\{1_{\hat{\Omega}_{ge}}, 1_{\hat{\Omega}_{ge}}\}.
\end{equation}

From (3.48), (3.39), and (3.52), it follows that for all \( c_1, c_2 \in \mathbb{R} \),

\begin{equation}
\langle c_1 1_{\hat{\Omega}_{ge}}, c_2 1_{\hat{\Omega}_{ge}}, \hat{\nabla}_h' \cdot (u \oplus u_b) \rangle_{\hat{\Omega}_h} = \langle c_1 1_{\hat{\Omega}_{ge}}, c_2 1_{\hat{\Omega}_{ge}}, \hat{\nabla}_h' \cdot (u \oplus 0) + \nabla_h' \cdot (0 \oplus u_b) \rangle_{\hat{\Omega}_h} = \langle c_1 1_{\hat{\Omega}_{ge}}, c_2 1_{\hat{\Omega}_{ge}}, \hat{\nabla}_h' \cdot (0 \oplus u_b) \rangle_{\hat{\Omega}_h}.
\end{equation}

In view of (3.53) and the third equation of (3.50), the compatibility condition (3.51) follows. It is easy to see that both conditions in (3.51) can be interpreted naturally as a discrete analogue of \( \int_{\Gamma} u_b \cdot n = 0 \) since

\begin{equation}
\langle 1_{\hat{\Omega}_{ge}}, \hat{\nabla}_h' \cdot (0 \oplus u_b) \rangle_{\hat{\Omega}_h} = \langle 1_{\hat{\Omega}_{ge}}, \hat{\nabla}_h' \cdot (0 \oplus u_b) \rangle_{\hat{\Omega}_h} = \frac{1}{2} \sum_{i=1}^{M} u_b \cdot n \Delta \ell_h
\end{equation}

\begin{equation}
= \frac{1}{2} \left( \sum_{i=1}^{M} \left( u_b \cdot n \Delta \ell_h, 0 \right)_{i, 0, 0} + u_b \cdot n \Delta \ell_h, 0, 0 \right) + \sum_{j=1}^{N} \left( u_b \cdot n \Delta \ell_h, 0, 0 \right)_{0, j, 0} + u_b \cdot n \Delta \ell_h, M, j, 0 \right).
\end{equation}
where, for example,
\begin{equation}
(u_h \cdot n \Delta \ell_h)_{i-\frac{1}{2},0} = \frac{h_1 h_2}{h} \left( \sqrt{g_h} (-\frac{x_6^2}{\ell_h} - \frac{x_6^1}{\ell_h}) \right)_{i-\frac{1}{2},0},
\end{equation}

(3.55) \[ (u_h \cdot n \Delta \ell_h)_{0,j-\frac{1}{2}} = \frac{h_1 h_2}{h} \left( \sqrt{g_h} \left( \frac{x_6^2}{\ell_h} - \frac{x_6^1}{\ell_h} \right) \right)_{0,j-\frac{1}{2}}, \]

with
\begin{equation}
(u_h \cdot n)_{i-\frac{1}{2},0} = \left( \frac{-u_6^2}{\sqrt{g^{22}}} \right)_{i-\frac{1}{2},0} = \left( \frac{(-\frac{x_6^2}{\ell_h} - \frac{x_6^1}{\ell_h})h_2}{h} \right)_{i-\frac{1}{2},0},
\end{equation}

(3.56) \[ (u_h \cdot n)_{0,j-\frac{1}{2}} = \left( \frac{-u_6^1}{\sqrt{g^{11}}} \right)_{0,j-\frac{1}{2}} = \left( \frac{(\frac{x_6^2}{\ell_h} - \frac{x_6^1}{\ell_h})h_1}{h} \right)_{0,j-\frac{1}{2}}, \]

(3.57)
\begin{align*}
(\Delta \ell_h)_{i-\frac{1}{2},0} &:= (h_1 \sqrt{g_h} \sqrt{g^{22}})_{i-\frac{1}{2},0} = \left( h_1 \sqrt{g_h} \sqrt{g^{11}} \right)_{i-\frac{1}{2},0} \approx \Delta \xi^1 \left| \frac{\partial x}{\partial \xi^1} \right|_{i-\frac{1}{2},0}, \\
(\Delta \ell_h)_{0,j-\frac{1}{2}} &:= (h_2 \sqrt{g_h} \sqrt{g^{11}})_{0,j-\frac{1}{2}} = \left( h_2 \sqrt{g_h} \sqrt{g^{22}} \right)_{0,j-\frac{1}{2}} \approx \Delta \xi^2 \left| \frac{\partial x}{\partial \xi^2} \right|_{0,j-\frac{1}{2}},
\end{align*}

and similarly on \((i - \frac{1}{2}, N)\) and \((M, j - \frac{1}{2})\).

Remark. In case of complex geometry, it may be necessary to patch the domain with several nonoverlapping coordinate charts, (see Figure 4.1). In this case, the metric tensor has a jump discontinuity across coordinate boundaries. We define the discrete divergence operator at a grid point \(P\) on the coordinate interface by

\begin{equation}
(\hat{\nabla}_h \cdot u)_P = \left( \frac{1}{2} \sqrt{g_h} \hat{\nabla}_h \cdot u \right)_{P^{+}} + \left( \frac{1}{2} \sqrt{g_h} \hat{\nabla}_h \cdot u \right)_{P^{-}}.
\end{equation}

(3.58) \[ (\hat{\nabla}_h \cdot u)_P = \left( \sum_{Q \in \text{chart}} \left( \sqrt{g_h} \right)_{Q(i)} \right)^{-1} \sum_{Q \in \text{chart}} \left( \frac{1}{4} \sqrt{g_h} \hat{\nabla}_h \cdot u \right)_{Q(i)}. \]

At a multiple coordinate junction \(Q\), such as the one given in Figure 3.5, the formula becomes

\begin{equation}
(\hat{\nabla}_h \cdot u)_Q = \left( \sum_{Q \in \text{chart}} \left( \sqrt{g_h} \right)_{Q(i)} \right)^{-1} \sum_{Q \in \text{chart}} \left( \frac{1}{4} \sqrt{g_h} \hat{\nabla}_h \cdot u \right)_{Q(i)}.
\end{equation}

(3.59) \[ A \text{ similar formula applies to discrete curl and discrete Laplacian. In this way, it is easy to see that the summation by parts identities in Lemma 3.1 remain valid. The same treatment for the discrete Laplacian on material interface has been proposed for elliptic interface problems in [Wa, HWA].} \]

3.4. Explicit and implicit time stepping.

3.4.1. Explicit time stepping and Hodge decomposition. As in section 2, we illustrate time stepping by the forward Euler method:

\begin{align*}
\frac{u^{n+1} - u^n}{\Delta t} + (\omega u^n) + \hat{\nabla}_h p^{n+1} = \nu \hat{\nabla}_h^2 \omega^n + f^n & \quad \text{on } \Omega_c, \\
\omega^n = \hat{\nabla}_h \cdot (u^n + u_b^n) = 0 & \quad \text{on } \Omega_g, \\
\hat{\nabla}_h \cdot (u^n + u_b^n) = 0 & \quad \text{on } \tilde{\Omega}_g.
\end{align*}

(3.60)
In high Reynolds number calculations, a high order Runge–Kutta method such as RK4 is needed for stability consideration [EL].

Given \((u^n, p^n)\) with \(\nabla_h' \cdot (u^n \oplus u^n_b) = 0\), (3.60) is solved via the following steps:

Step 1. Evaluate \(\omega^n\) up to the boundary

\[
\omega^n = \nabla_h' \cdot (u^n \oplus u^n_b) \quad \text{on } \bar{\Omega}_g. 
\]

Step 2. Evaluate \(u^*\) on cell centers

\[
\frac{u^{n+1} - u^n}{\Delta t} + (\hat{\omega}^\perp) = \nu \nabla_h \omega^n + f^n \quad \text{on } \Omega_c. 
\]

Step 3. Solve for \((u^{n+1}, p^{n+1})\) such that

\[
\frac{u^{n+1} - u^n}{\Delta t} + \hat{\nabla}_h p^{n+1} = 0 \quad \text{on } \Omega_c, \\
\hat{\nabla}_h' \cdot (u^{n+1} \oplus u^{n+1}_b) = 0 \quad \text{on } \bar{\Omega}_g. 
\]

This is the (inhomogeneous) Hodge decomposition for \(u^*\) and can be performed as follows:

Step 3-1. Solve for \(p^{n+1}\) up to the boundary from

\[
\hat{\Delta}_h' p^{n+1} = \frac{1}{\Delta t} \hat{\nabla}_h' \cdot (u^* \oplus u^{n+1}_b) \quad \text{on } \bar{\Omega}_g. 
\]

Step 3-2. Update \(u^{n+1}\) from

\[
\frac{u^{n+1} - u^n}{\Delta t} + \hat{\nabla}_h p^{n+1} = 0 \quad \text{on } \Omega_c. 
\]

It follows from (3.41), (3.49), (3.64), (3.65) that

\[
\hat{\nabla}_h' \cdot (u^{n+1} \oplus u^{n+1}_b) = \hat{\nabla}_h' \cdot (u^* \oplus u^{n+1}_b) + \hat{\nabla}_h' \cdot ((u^{n+1} - u^*) \oplus 0) \\
= \hat{\nabla}_h' \cdot (u^* \oplus u^{n+1}_b) + \hat{\nabla}_h' \cdot (u^{n+1} - u^*) \\
= 0. 
\]

Step 1 can be viewed as a vorticity boundary condition that incorporates the tangential component of \(u_b\). On the other hand, Step 3 depends on \(u_b\) only through
its normal component. In Step 3-1, the pressure is solved as a Lagrangian multiplier without introducing the ghost value and artificial pressure boundary conditions. From (3.39) and (3.41), it is easy to see that

\[ (q, \tilde{\Delta}_h p)_{\Omega_h} = - (\nabla_h q, \tilde{\Delta}_h p)_{\Omega_h} = - h_1 h_2 \sum_{i=1}^{M} \sum_{j=1}^{N} \left( \tilde{g}_{\mu}^{\alpha} \sum_{\nu=1}^{2} \tilde{g}_{\mu}^{\nu} (D_{\mu} p)(D_{\nu} q) \right)_{i - \frac{1}{2} j - \frac{1}{2}}. \]

In other words, \( \tilde{\Delta}_h \) is self-adjoint with respect to the inner product (3.38) and non-positive definite as long as each of the \( 2 \times 2 \) matrices \( (\tilde{g}_{\mu}^{\nu})_{\nu=1}^{2} \) is positive definite on the cell centers. This only requires the cells to be nonsingular. No further grid regularity is needed for stability concerns. In addition, (3.65)

\[ \ker(\tilde{\Delta}_h^{+}) = \ker(\tilde{\nabla}_h) = \text{span}\{1_{\Omega_h}, 1_{\Omega_h^c}\}; \]

thus (3.64) is solvable if and only if

\[ (c_1 1_{\Omega_h} + c_2 1_{\Omega_h^c}, \tilde{\nabla}_h \cdot (u^{n+1} \oplus u_h^{+}) )_{\Omega_h} = 0 \quad \text{for all } c_1, c_2 \in \mathbb{R}. \]

In view of (3.53), it follows that the solvability condition for (3.64) is exactly the compatibility condition (3.51). In addition, it is worth noting that (3.64) can be decoupled and solved on \( \Omega_h \) and \( \Omega_h^c \) separately.

### 3.4.2. Implicit time stepping and nonlinear stability

Here we list a few partially and fully implicit second order time discretizations of our scheme.

For low Reynolds number flows, explicit time stepping is subject to the parabolic time stepping constraint [EL]. As an alternative, the Stokes-based partially implicit time stepping is given by

\[
\begin{aligned}
\frac{u^{n+1} - u^n}{2\Delta t} + N_h(u^{n+1/2}) &+ \frac{1}{2} (\tilde{\nabla}_h p^{n+1} + \tilde{\nabla}_h p^n) = \frac{1}{2} \nu(\tilde{\nabla}_h \omega^{n+1} + \tilde{\nabla}_h \omega^n) + f^{n+1/2} & \quad \text{on } \Omega_c, \\
\tilde{\nabla}_h \cdot (u^{n+1} \oplus u_h^{+}) &+ \tilde{\nabla}_h \cdot (u^{n+1} \oplus u_h^{+}) = 0 & \quad \text{on } \Omega_g,
\end{aligned}
\]  

(3.69)

where \( N_h(u^{n+1/2}) = \frac{3}{2} \omega^n (u^n) - \frac{3}{2} \omega^{n-1} (u^{n-1}) + \frac{3}{2} \omega^{n+1} (u^{n+1}) \). It is not clear what the time stepping constraint for (3.69) is. Alternatively, one can discretize the nonlinear term semi-implicitly or fully implicitly to allow larger time steps. The special structure of our discretization of the nonlinear term guarantees unconditional stability with second order backward difference time stepping (assuming \( f = 0 \) and \( u_h = 0 \)):

\[
\begin{aligned}
\frac{3u^{n+1} - 4u^n + u^{n-1}}{2\Delta t} + N_h(u^{n+1}) + \frac{1}{2} (\tilde{\nabla}_h p^{n+1} + \tilde{\nabla}_h p^n) = \nu(\tilde{\nabla}_h \omega^{n+1} + \tilde{\nabla}_h \omega^n) & \quad \text{on } \Omega_c, \\
\tilde{\nabla}_h \cdot (u^{n+1} \oplus u_h^{+}) &+ \tilde{\nabla}_h \cdot (u^{n+1} \oplus u_h^{+}) = 0 & \quad \text{on } \Omega_g,
\end{aligned}
\]  

(3.70)

where \( N_h(u^{n+1}) = (2\omega^n - \omega^{n-1}) (u^{n+1}) \) for the semi-implicit scheme or \( N_h(u^{n+1}) = \omega^{n+1} (u^{n+1}) \) for the fully implicit scheme.
To see the stability of (3.70), we take the inner product of (3.70) with $u^{n+1}$ to get

$$
\left\langle u^{n+1}, \frac{3u^{n+1} - 4u^n + u^{n-1}}{2\Delta t} \right\rangle_{\Omega_c} + \nu \|\omega^{n+1}\|_{\Omega_g}^2 = 0.
$$

This gives

$$
\frac{3\|u^{n+1}\|_{\Omega_c}^2}{4\Delta t} - \frac{4\|u^n\|_{\Omega_c}^2}{4\Delta t} + \frac{\|u^{n-1}\|_{\Omega_c}^2}{4\Delta t} + \frac{4\|u^{n+1} - u^n\|_{\Omega_c}^2}{4\Delta t} + \frac{\|u^{n+1} - u^{n-1}\|_{\Omega_c}^2}{4\Delta t} + \nu \|\omega^{n+1}\|_{\Omega_g}^2 = 0.
$$

With the inequality $\|u^{n+1} - u^n\|_{\Omega_c}^2 \leq 2(\|u^{n+1} - u^n\|_{\Omega_c}^2 + \|u^n - u^{n-1}\|_{\Omega_c}^2)$, one obtains

$$
3\|u^{n+1}\|_{\Omega_c}^2 - \|u^n\|_{\Omega_c}^2 - 3\|u^{n-1}\|_{\Omega_c}^2 + \|u^0\|_{\Omega_c}^2 + 2\|u^{n+1} - u^n\|_{\Omega_c}^2 + 4\nu \Delta t \sum_{k=2}^{n+1} \|\omega^k\|_{\Omega_g}^2 \leq 0
$$

or

$$
\|u^{n+1}\|_{\Omega_c}^2 + 2\|u^{n+1} - u^n\|_{\Omega_c}^2 + 4\nu \Delta t \sum_{k=2}^{n+1} \|\omega^k\|_{\Omega_g}^2 \leq \|u^n\|_{\Omega_c}^2 + 2\|u^n - u^{n-1}\|_{\Omega_c}^2 + \|u^0\|_{\Omega_c}^2.
$$

This gives unconditional stability of (3.70).

Similarly, for steady state calculations with pseudo-time stepping where $\Delta t$ may vary spatially, it is not difficult to see that the first order backward differentiation formula

$$
\frac{u^{n+1} - u^n}{\Delta t} + N_h(u^{n+1}) + \hat{\nabla}_h p^{n+1} = \nu \hat{\nabla}_h \omega^{n+1} \quad \text{on } \Omega_c,
$$

$$
\omega^{n+1} = \hat{\nabla}'_h u^{n+1} \quad \text{on } \Omega_g,
$$

$$
\hat{\nabla}'_h \cdot u^{n+1} = 0 \quad \text{on } \Omega_g
$$

with $N_h(u^{n+1}) = \tilde{\omega}^n (u^{n+1})^\perp$ or $N_h(u^{n+1}) = \tilde{\omega}^{n+1} (u^{n+1})^\perp$ also leads to unconditional stability:

$$
\frac{1}{2} \left\| \frac{u^n}{\sqrt{\Delta t}} \right\|_{\Omega_c}^2 + \frac{1}{2} \sum_{k=1}^n \left\| \frac{u^k - u^{k-1}}{\sqrt{\Delta t}} \right\|_{\Omega_c}^2 + \nu \sum_{k=1}^n \|\omega^k\|_{\Omega_g}^2 \leq \frac{1}{2} \left\| \frac{u^0}{\sqrt{\Delta t}} \right\|_{\Omega_c}^2.
$$

### 3.5. Comparison with similarly staggered schemes.

It can be shown that GMAC is second order accurate on smooth grids in two dimensions for both velocity and pressure [HLW]. A key ingredient in the pressure error estimate for (3.44) is the inf-sup (LBB) condition which states that there exists a constant $\beta$, independent of the grid size, such that

$$
\inf_{p \in L^2(\Omega_g, \mathbb{R})} \sup_{u \in L^2(\Omega_c, \mathbb{R}^2)} \frac{\langle p, \hat{\nabla}'_h \cdot u \rangle_{\Omega_g}}{|p|_{\Omega_g} (\|u^2||_{\Omega_c} + \|\hat{\nabla}'_h \cdot u||_{\Omega_g} + \|\hat{\nabla}'_h \cdot u||_{\Omega_g})^2} := \beta_h \geq \beta.
$$
Roughly speaking, the higher the ratio $\text{DOF}(u)/\text{DOF}(p)$, the more likely the inf-sup condition (3.74) is to hold true and be verified. On 2D quadrilateral meshes, some numerical schemes are known to satisfy the inf-sup condition (3.74), including the $Q_1-P_{\ell-1}$ elements, $\ell \geq 2$ [GR] with $\text{DOF}(u)/\text{DOF}(p) \geq 8/3$, the $\mathcal{L}_1-P_0$ element [GR] with $\text{DOF}(u)/\text{DOF}(p) = 3$, and the schemes proposed in [Cho], [Han], and [RaTu] with $\text{DOF}(u)/\text{DOF}(p) \geq 4$. Here $\text{DOF}(u)$ is the degree of freedom for the velocity field, counting both components.

When the ratio of degrees of freedom becomes marginal, the compatibility of spatial discretization plays the key role in establishing the inf-sup condition. This is the case for the classical MAC scheme [HW] and the GMAC scheme. To demonstrate this point more clearly, we compare GMAC ((3.44) and (A.17)) against similarly staggered schemes with different (and noncompatible) spatial discretizations, including the BCG scheme (Bell, Colella, and Glaz [BCG]), the KF scheme (Kuznetsov, Fortin, and others [FPT, Ku, PT]), and the $Q_1-P_0$ element. For simplicity of presentation, all the discretization formulas throughout this subsection refer to interior nodes only. Proper reductions are required on the boundary for all these schemes. We omit the details.

Recall that the spatial discretization of the classical MAC scheme on 2D uniform Cartesian grids are given by

$$
\begin{align*}
\Delta_h u_{i,j-\frac{1}{4}} &= \frac{1}{h^2} (u_{i+1,j+\frac{1}{2}} + u_{i-1,j+\frac{1}{2}} + u_{i,j+\frac{1}{2}} + u_{i,j-\frac{1}{2}} - 4u_{i,j-\frac{1}{4}}), \\
D_x p_{i,j-\frac{1}{4}} &= \frac{1}{h} (p_{i+\frac{1}{4},j-\frac{1}{2}} - p_{i-\frac{1}{4},j-\frac{1}{2}}), \\
\Delta_h v_{i-\frac{1}{2},j} &= \frac{1}{h^2} (v_{i-\frac{1}{2},j+1} + v_{i-\frac{1}{2},j-1} + v_{i+\frac{1}{2},j} + v_{i-\frac{1}{2},j} - 4v_{i-\frac{1}{2},j}), \\
D_y p_{i-\frac{1}{2},j} &= \frac{1}{h} (p_{i-\frac{1}{2},j+\frac{1}{4}} - p_{i-\frac{1}{2},j-\frac{1}{4}}).
\end{align*}
$$

Like the MAC scheme, the discretization of the viscous term for BCG and KF use the standard five-point formula. Take the KF scheme, for example:

$$
\begin{align*}
\Delta_h u_{i,j} &= \frac{1}{h^2} (u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j}), \\
\Delta_h v_{i,j} &= \frac{1}{h^2} (v_{i+1,j} + v_{i-1,j} + v_{i,j+1} + v_{i,j-1} - 4v_{i,j}).
\end{align*}
$$

On the other hand, the pressure gradient for KF takes standard centered differencing, averaged in the transversal direction in order to maintain the second order local
truncation error,

\[
(D_x p)_{i,j} = \frac{1}{2h} \left( (p_{i+\frac{1}{2},j+\frac{1}{2}} - p_{i-\frac{1}{2},j+\frac{1}{2}}) + (p_{i+\frac{1}{2},j-\frac{1}{2}} - p_{i-\frac{1}{2},j-\frac{1}{2}}) \right),
\]

(3.77)

\[
(D_y p)_{i,j} = \frac{1}{2h} \left( (p_{i+\frac{1}{2},j+\frac{1}{2}} - p_{i+\frac{1}{2},j-\frac{1}{2}}) + (p_{i-\frac{1}{2},j+\frac{1}{2}} - p_{i-\frac{1}{2},j-\frac{1}{2}}) \right).
\]

(3.78)

Similar averaging is also needed for the divergence-free constraint.

In contrast, the viscous term in GMACc and GMACg is based on the skewed Laplacian. For example, in GMACg we have

\[
\Delta_h u_{i,j} = \frac{1}{2h^2} (u_{i+1,j+1} + u_{i-1,j+1} + u_{i+1,j-1} + u_{i-1,j-1} - 4u_{i,j}),
\]

(3.79)

\[
\Delta_h v_{i,j} = \frac{1}{2h^2} (v_{i+1,j+1} + v_{i-1,j+1} + v_{i+1,j-1} + v_{i-1,j-1} - 4v_{i,j}),
\]

and the same pressure gradient (3.77) as KF. Here we have adopted conventional vector notation in terms of the components $u, v$ in the default coordinate $(\xi^1, \xi^2) = (x, y)$. That is, we write $\mathbf{u} = (u, v)$ if $\mathbf{u} = u\mathbf{e}_x + v\mathbf{e}_y$.

The well-known div-unstable $Q_1$-$P_0$ element corresponds to a mixture of standard and skewed Laplacian,

\[
\Delta_h = \frac{1}{3} \Delta_h + \frac{2}{3} \overset{\wedge}{\Delta}_h,
\]

(3.80)

\[
\nabla^2_{\text{BCG}} u_{i+,j+} = (\Delta_h u_{i+\frac{1}{2},j+\frac{1}{2}}, \Delta_h v_{i+\frac{1}{2},j+\frac{1}{2}}),
\]

(3.81)

\[

abla^2_{\text{GMACg}} u_{i+,j+} = (\overset{\wedge}{\Delta}_h u_{i+\frac{1}{2},j+\frac{1}{2}}, \overset{\wedge}{\Delta}_h v_{i+\frac{1}{2},j+\frac{1}{2}}),
\]

(3.82)

\[
\nabla^2_{KF} u_{i,j} = (\Delta_h u_{i,j}, \Delta_h v_{i,j}),
\]

(3.83)

\[
\nabla^2_{\text{GMACc}} u_{i,j} = (\overset{\wedge}{\Delta}_h u_{i,j}, \overset{\wedge}{\Delta}_h v_{i,j}),
\]

(3.84)

\[
\nabla^2_{Q_1-P_0} u_{i,j} = (\overset{\wedge}{\Delta}_h u_{i,j}, \overset{\wedge}{\Delta}_h v_{i,j}).
\]

Note that in GMACc (3.44) and GMACg (A.17), the viscous term was originally proposed as $\overset{\wedge}{\nabla}^\perp \omega = \overset{\wedge}{\nabla}^\perp (\overset{\wedge}{\nabla} \cdot \mathbf{u})$, while in (3.81) and (3.83), it is recast in terms of the full vector Laplacian

\[
\nabla^2_{\text{GMAC}} \mathbf{u} = \overset{\wedge}{\nabla}^\perp \mathbf{u} := \overset{\wedge}{\nabla}^\perp (\overset{\wedge}{\nabla} \cdot \mathbf{u}) + \overset{\wedge}{\nabla}^\perp (\overset{\wedge}{\nabla}^\perp \cdot \mathbf{u})
\]

(3.85)

to make the comparison more comprehensible.

Equation (3.85) is the 2D discrete analogue of the identity

\[
\nabla^2 \mathbf{u} = \nabla (\nabla \cdot \mathbf{u}) - \nabla \times \nabla \times \mathbf{u}
\]

(3.86)

and is algebraically identical to $\overset{\wedge}{\nabla}^\perp (\overset{\wedge}{\nabla}^\perp \cdot \mathbf{u})$ under the incompressibility constraint $\overset{\wedge}{\nabla} \cdot \mathbf{u} = 0$.  

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All these schemes are semistaggered with \( \text{DOF}(u)/\text{DOF}(p) = 2 \) in two dimensions (lowest among discretizations based on quadrilateral meshes). They result in identical discretizations for the pressure gradient and incompressibility constraint, namely, the skewed gradient \( \nabla_h \cdot \hat{\nabla}_h \) and skewed divergence \( \hat{\nabla}_h \cdot \hat{\nabla}_h \). We believe that the spatial compatibility among the discrete gradient, divergence, and curl (as part of the vector Laplacian) operators

\[
(3.77) \quad \nabla_h \cdot \hat{\nabla}_h = \hat{\nabla}_h \cdot \nabla_h = 0
\]

plays an important role in the uniform LBB estimate for GMAC. We will demonstrate this result by numerically computing the LBB constant in section 4. A rigorous proof of the inf-sup condition (3.74) for GMAC can be found in [HLW]. We remark here that the classical MAC scheme is fully staggered with \( \text{DOF}(u)/\text{DOF}(p) = 2 \) in two dimensions. It is shown in [HWu] that the classical MAC scheme with a different boundary condition also satisfies the inf-sup condition.

4. Numerical result. In this section, we report several numerical test results. We start with the numerical computation of the LBB constants for the schemes mentioned in section 3.5. The result gives a very strong indication of the uniform LBB estimate for GMAC. We then proceed with a standard convergence test and benchmark problems. We have observed clean second order accuracy in both velocity and pressure in the convergence test. The benchmark simulation also shows good agreement with the results reported in the literature.

**Example 1. The LBB constant \( \beta_h \).** It is well known [RaTu, BF] that the LBB constant can be obtained through a generalized singular value decomposition. Take GMACc, for example; denote the matrix representations of the operators in (3.74) by

\[
A \sim \sqrt{g_{h,c}} \left( 1 - \nabla_h \hat{\nabla}_h \cdot \hat{\nabla}_h \right), \quad B \sim \sqrt{g_{h,g}} \hat{\nabla}_h \cdot \hat{\nabla}_h, \quad B^T \sim - \sqrt{g_{h,c}} \hat{\nabla}_h, \quad M \sim \sqrt{g_{h,g}},
\]

where the \( \sqrt{g_{h,c}} \) in \( B \) is defined on cell centers and \( \sqrt{g_{h,g}} \) defined on grid points with standard weighting (\( 1/2 \) on the edges and \( 1 \) on corners). The LBB constant in (3.74) can be characterized as

\[
(4.1) \quad \beta_h = \inf_{p \in \text{DOF}(p)} \sup_{\lambda \in \text{kernel}(B^T)} \frac{\lambda (p^T M^{-\frac{1}{2}} B A^{-\frac{1}{2}} \hat{u})}{\lambda (p^T)}
\]

In other words, \( \beta_h \) can be obtained numerically by computing the smallest nonzero eigenvalue of

\[
(4.2) \quad BA^{-1} B^T p = \mu^2 M p \quad \text{or equivalently} \quad - \hat{\nabla}_h \cdot (1 - \nabla_h \hat{\nabla}_h \cdot \hat{\nabla}_h \hat{\nabla}_h \cdot) \hat{\nabla}_h p = \mu^2 p
\]
and \( \beta_h = \sqrt{\min\{|\mu_i^2| \mid \mu_i^2 > 0\}} \) from (4.1). The computed \( \beta_h \) for GMAC and the schemes mentioned in section 3.5 are summarized in Table 4.1. It is clear that \( \beta_h = O(1) \) for both GMACc and GMACg. In contrast, KF, BCG, and \( Q_1-P_0 \) all result in \( \beta_h = O(h) \).

**Example 2. Convergence test on a patched domain.** We proceed with a standard accuracy check on a circular domain patched by five nonoverlapping coordinate charts as shown in Figure 4.1.

We take the exact solution to be

\[
\begin{align*}
\mathbf{u}^e(t, x, y) &= \cos(t) \cdot (-y(1-x^2-y^2), x(1-x^2-y^2)), \\
p^e(t, x, y) &= \sin(t) \cdot (1-x^2-y^2)
\end{align*}
\]

on the unit disk \( \Omega = \{x^2 + y^2 \leq 1\} \) and generate the corresponding forcing term \( \mathbf{f} \) with \( \nu = 1 \).

The result at \( T = 5.0 \) is summarized in Table 4.2. We used semi-implicit time stepping (3.70) with \( \Delta t = 5\Delta x \), where \( \Delta x = L/N \) is the mesh size in the inner square. Here \( L = 1/(1+\sqrt{2}/2) \) and each coordinate chart is divided into \( N \times N \) cells. The result shows clean second order accuracy.

**Example 3. Benchmark test: Lid-driven cavity flow.** We continue with the benchmark problem of lid-driven cavity flow [Bu, GGS, BP]. Here the domain \( \Omega = [0, 1]^2 \). The flow is initially at rest and driven by a slip velocity \( \mathbf{u} = (1, 0) \) on top. The grids are adapted near the boundary to better resolve the flow pattern there. A 32 \( \times \) 32 adaptive grid partitioning is shown in Figure 4.2 for illustration. The results for \( \text{Re} = \frac{1}{\nu} = 1,000 \) with 128 \( \times \) 128 adaptive grids and \( \text{Re} = 10,000 \) with 180 \( \times \) 180 adaptive grids are presented in Figure 4.3 and Figure 4.4, respectively. Both are simulated till \( T \approx 1.5\sqrt{\text{Re}}: T_{1000} = 48.0, T_{10000} = 255.0 \) using fourth order explicit

---

**Table 4.1** \( \beta_h \) for GMAC and similarly staggered schemes.

<table>
<thead>
<tr>
<th># cells</th>
<th>GMACc</th>
<th>BCG</th>
<th>GMACg</th>
<th>KF</th>
<th>( Q_1-P_0 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>64 ( \times ) 64</td>
<td>4.774E-1</td>
<td>1.736E-2</td>
<td>4.405E-1</td>
<td>1.726E-2</td>
<td>2.976E-2</td>
</tr>
<tr>
<td>128 ( \times ) 128</td>
<td>4.669E-1</td>
<td>8.678E-3</td>
<td>4.374E-1</td>
<td>8.654E-3</td>
<td>1.496E-2</td>
</tr>
<tr>
<td>256 ( \times ) 256</td>
<td>4.593E-1</td>
<td>4.339E-3</td>
<td>4.351E-1</td>
<td>4.333E-3</td>
<td>7.497E-3</td>
</tr>
<tr>
<td>512 ( \times ) 512</td>
<td>4.536E-1</td>
<td>2.169E-3</td>
<td>4.335E-1</td>
<td>2.168E-3</td>
<td>3.753E-3</td>
</tr>
</tbody>
</table>

**Fig. 4.1.** Domain patched by nonoverlapping coordinate charts.

---

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Table 4.2
Absolute error and rate of convergence for Example 2.

<table>
<thead>
<tr>
<th># cells</th>
<th>$|u^e - u_h|_2$</th>
<th>Order</th>
<th>$|\omega^e - \omega_h|_2$</th>
<th>Order</th>
<th>$|p^e - p_h|_2$</th>
<th>Order</th>
</tr>
</thead>
<tbody>
<tr>
<td>5 × 16$^2$</td>
<td>3.48E − 4</td>
<td></td>
<td>1.64E − 3</td>
<td></td>
<td>8.33E − 4</td>
<td></td>
</tr>
<tr>
<td>5 × 32$^2$</td>
<td>8.51E − 5</td>
<td>2.03</td>
<td>4.01E − 4</td>
<td>2.03</td>
<td>2.04E − 4</td>
<td>2.03</td>
</tr>
<tr>
<td>5 × 64$^2$</td>
<td>2.14E − 5</td>
<td>1.99</td>
<td>1.01E − 4</td>
<td>1.99</td>
<td>5.04E − 5</td>
<td>2.02</td>
</tr>
<tr>
<td>5 × 128$^2$</td>
<td>5.32E − 6</td>
<td>2.00</td>
<td>2.51E − 5</td>
<td>2.00</td>
<td>1.26E − 5</td>
<td>2.01</td>
</tr>
</tbody>
</table>

Fig. 4.2. Illustration of the adaptive grid used for the lid-driven cavity flow (Example 3).

Fig. 4.3. Vorticity contour plot of the lid-driven cavity flow (Example 3) at Re = 1,000.
Fig. 4.4. Vorticity contour plot of the lid-driven cavity flow (Example 3) at Re = 10,000.

Fig. 4.5. 3D cavity with bottom topography.
Runge–Kutta time stepping. The Re = 1,000 case agrees well with the benchmark result in [BP].

Example 4. Convergence test for 3D cavity flow with bottom topography. In this example, we perform an accuracy check for GMAC3De-iso on a 3D cavity flow with bottom topography at Re = 2,000 with explicit fourth order Runge–Kutta time stepping.

The domain is \([0, 1] \times [0, 1] \times [g(x, y), 1]\), where \(g(x, y) = 0.15 \sin(2\pi x) \sin(2\pi y)\) is the bottom topography as shown in Figure 4.5.

We take the exact solution to be \(u^e = \nabla \psi^e, p^e(t, x, y, z) = \cos(t) \cos(x) \sin(y) e^z\), where \(\psi^e(t, x, y, z) = \cos(t)(\psi_1^e, \psi_2^e, \psi_3^e)\) and

\[
\psi_1^e(x, y, z) = \sin^2(\pi x) \sin^2(\pi y) (z - 1)^2(z - g(x, y))^2,
\psi_2^e(x, y, z) = \sin^2(\pi x) \sin^2(2\pi y) (z - 1)^2(z - g(x, y))^2,
\psi_3^e(x, y, z) = \sin^2(2\pi x) \sin^2(\pi y) (z - 1)^2(z - g(x, y))^2.
\]

The right-hand side \(f(t, x, y, z)\) is generated accordingly with \(\nu = 1/2000\).

The result at \(T = 5.0\) is summarized in Table 4.3 and shows second order accuracy.

5. Conclusion. In this paper, we have proposed a novel finite difference scheme for incompressible the Navier–Stokes equation. The main features of the new scheme are as follows:

1. It partially resembles the classical MAC scheme in the sense that velocity and pressure are defined on different locations.
2. It is applicable on general curvilinear domains and easy to implement.
3. The spatial discretization is based on intrinsic differential operators div, grad, curl and is mutually compatible. It naturally incorporates no-slip conditions and is endowed with desirable properties such as summation by parts identities and exact discrete Hodge decomposition. As a result, pressure can be decoupled from the momentum equation without imposing artificial boundary conditions.
4. The pressure Poisson equation is symmetric and semidefinite. Standard fast solvers for large linear systems are applicable.
5. More important, the spatial discretization satisfies the inf-sup condition, giving full second order accuracy for velocity and pressure in both dynamic and steady state calculations.

Systematic numerical experiments were conducted to verify the accuracy and robustness of the proposed scheme.

Appendix A. GMACg, another version of the generalized MAC scheme.

An alternative placement of the unknown variables

\[
(A.1) \quad u \in L^2(\bar{\Omega}_c, \mathbb{R}^2), \quad p, \omega \in L^2(\Omega_c, \mathbb{R})
\]
leads to a variant of the scheme (3.44). In this setting, we place both components of the velocity on grid points, pressure and vorticity on cell centers. Note that the boundary velocity $u_b \in L^2(\Gamma_g, \mathbb{R}^2)$ is prescribed data and not listed in (A.1) as part of active variables.

Associated with this new placements are the function spaces

\begin{align}
L^2(\Omega_c, \mathbb{R}) &:= \{ a : \Omega_c \to \mathbb{R} \}, \\
L^2(\Omega_c, \mathbb{R}) &:= \{ u_b : \Gamma_g \to \mathbb{R}^2 \}, \\
L^2(\Omega_c, \mathbb{R}) &:= \{ a : \Omega_c \to \mathbb{R} \}
\end{align}

and finite difference operators

\begin{align}
\nabla_h^\omega : L^2(\Omega_c, \mathbb{R}) &\to L^2(\Omega_g, \mathbb{R}), \\
\nabla_h^p : L^2(\Omega_c, \mathbb{R}) &\to L^2(\Omega_g, \mathbb{R})
\end{align}

Note that the operators $\nabla_h^\omega$ and $\nabla_h^p$ are exactly the standard finite difference operators as (3.29) and (3.30), except (A.3) and (A.4) are defined for vector fields on $\Omega_c$. We use the superscript $\circ$ to emphasize that $\nabla_h^\omega$ and $\nabla_h^p$ are interior gradient operators whose range is vector fields defined only on the interior grids.

For inhomogeneous boundary velocity, the operators $\nabla_h^\omega$ and $\nabla_h^p$ involve both the active variable $u \in L^2(\Omega_g, \mathbb{R}^2)$ and the data $u_b \in L^2(\Gamma_g, \mathbb{R}^2)$. Denoting by $U = u \oplus u_b$, that is,

\begin{align}
\nabla_h^\omega : L^2(\Omega_g, \mathbb{R}^2) &\to L^2(\Omega_g, \mathbb{R}^2), \\
\nabla_h^p : L^2(\Omega_g, \mathbb{R}^2) &\to L^2(\Omega_g, \mathbb{R}^2)
\end{align}

we define

\begin{align}
\nabla_h^\omega : L^2(\Omega_g, \mathbb{R}^2) &\to L^2(\Omega_g, \mathbb{R}^2), \\
\nabla_h^p : L^2(\Omega_g, \mathbb{R}^2) &\to L^2(\Omega_g, \mathbb{R}^2)
\end{align}
Lemma A.1. If \( u \in L^2(\Omega_g, \mathbb{R}^2) \), then we have

\begin{align}
(A.10) \quad \langle u, v \rangle_{\Omega_g} &= h_1 h_2 \sum_{i=1}^{M-1} \sum_{j=1}^{N-1} ((u \cdot v) \sqrt{g_h})_{i,j}, \quad u, v \in L^2(\Omega_g, \mathbb{R}^2), \\
(A.11) \quad \langle a, b \rangle_{\Omega_c} &= h_1 h_2 \sum_{i=1}^{M} \sum_{j=1}^{N} (a \cdot b \sqrt{g_h})_{i-\frac{1}{2}, j-\frac{1}{2}} \quad a, b \in L^2(\Omega_c, \mathbb{R}).
\end{align}

The counterpart of Lemma 3.1 is also valid.

**Lemma A.1.** If \( u \in L^2(\Omega_g, \mathbb{R}^2) \) and \( a \in L^2(\Omega_c, \mathbb{R}) \), then we have

1. \( \langle u, \nabla_h^0 a \rangle_{\Omega_g} = -\langle \nabla_h^0 u, a \rangle_{\Omega_c} \).
2. \( \langle u, \nabla_h^0 a \rangle_{\Omega_g} = -\langle \nabla_h^0 u, a \rangle_{\Omega_c} \).
3. \( \nabla_h^0 \cdot (\nabla_h^0 a) = \nabla_h^0 \cdot (\nabla_h^0 a) = \Delta_h^0 a \quad \text{on } \Omega_c. \)
4. If \( a \in L^2(\Omega_c, \mathbb{R}) \), then

\( \nabla_h^0 \cdot (\nabla_h^0 a) = \nabla_h^0 \cdot (\nabla_h^0 a) = 0 \quad \text{on } \Omega_c, \)

where \( \Omega_c = \{ x(\xi_i, \xi_j) \in \Omega_c : 2 \leq i \leq M-1, 2 \leq j \leq N-1 \} \). If in addition \( a \) is constant on \( \Omega_c \), then

\( \nabla_h^0 \cdot (\nabla_h^0 a) = \nabla_h^0 \cdot (\nabla_h^0 a) = 0 \quad \text{on } \Omega_c. \)

The generalized MAC scheme with \( u \) defined on grids (GMACg) is as follows.

**The GMACg scheme.** Solve for \( u \in C^1([0, T]; L^2(\Omega_g, \mathbb{R}^2)) \) and \( p \in C^0([0, T]; L^2(\Omega_c, \mathbb{R})) \) such that

\begin{align}
(A.17) \quad u_t + \nu u + \nabla_h^0 p &= \nu \nabla_h^0 \omega + f \quad \text{on } \Omega_g, \\
\omega &= \nabla_h^0 \cdot (u \oplus u_h) \quad \text{on } \Omega_c, \\
\nabla_h^0 \cdot (u \oplus u_h) &= 0 \quad \text{on } \Omega_c.
\end{align}
The compatibility condition for the boundary velocity \( \mathbf{u}_b \) is

\begin{equation}
\langle \mathbf{1}_{\Omega_c}, \nabla \mathbf{u}_b \cdot (\mathbf{0} \oplus \mathbf{u}_b) \rangle_{\Omega_c} = 0 = \langle \mathbf{1}_{\Omega_c}, \nabla \mathbf{u}_b \cdot (\mathbf{0} \oplus \mathbf{u}_b) \rangle_{\Omega_c},
\end{equation}

where

\begin{equation}
\Omega_c := \{ \mathbf{x}(\xi_{i-\frac{1}{2}}, \xi_{j-\frac{1}{2}}) \in \Omega_c, \ i + j \text{ is even} \},
\end{equation}

\begin{equation}
\Omega_c := \{ \mathbf{x}(\xi_{i-\frac{1}{2}}, \xi_{j-\frac{1}{2}}) \in \Omega_c, \ i + j \text{ is odd} \}.
\end{equation}

Furthermore,

\begin{equation}
\ker(\Delta \mathbf{u}) = \ker(\nabla \mathbf{u}) = \text{span}\{\mathbf{1}_{\Omega_c}, \mathbf{1}_{\Omega_c}\}.
\end{equation}

**Appendix B. The GMAC3D scheme.** Generalization of GMAC to 3D with the semistaggered variable placement as either

\begin{equation}
\text{GMAC3Dc : } \mathbf{u} \in L^2(\Omega_c, \mathbb{R}^3), \quad p \in L^2(\Omega_g, \mathbb{R}), \quad \omega \in L^2(\Omega_g, \mathbb{R}^3),
\end{equation}

or

\begin{equation}
\text{GMAC3Dg : } \mathbf{u} \in L^2(\Omega_g, \mathbb{R}^3), \quad p \in L^2(\Omega_c, \mathbb{R}), \quad \omega \in L^2(\Omega_c, \mathbb{R}^3),
\end{equation}

is straightforward once a skewed coordinate system is chosen. We only elaborate on GMAC3Dc.

Take, for example, \((\xi^1, \xi^2, \xi^3)\) as shown in left top corner of Figure B.1:

\begin{align}
\xi^1 &= \frac{h_3 \xi^2 + h_2 \xi^3}{\sqrt{h_1^2 + h_2^2}}, & h_1 &= \Delta \xi^1 = \frac{2h_2 h_3}{\sqrt{h_1^2 + h_2^2}}, \\
\xi^2 &= \frac{h_1 \xi^3 + h_3 \xi^1}{\sqrt{h_1^2 + h_2^2}}, & h_2 &= \Delta \xi^2 = \frac{2h_3 h_1}{\sqrt{h_1^2 + h_2^2}}, \\
\xi^3 &= \frac{h_2 \xi^1 + h_1 \xi^2}{\sqrt{h_1^2 + h_2^2}}, & h_3 &= \Delta \xi^3 = \frac{2h_1 h_2}{\sqrt{h_1^2 + h_2^2}}.
\end{align}

The discrete metric tensors \(\{g_{\alpha \beta}^h\}_{\alpha, \beta = 1}^3\) and \(\sqrt{g_h}\) with respect to \((\xi^1, \xi^2, \xi^3)\), as well as the reduced difference operators \(\hat{D}_\alpha\) and \(\hat{D}_\beta\), can be defined in a similar way as in the 2D case.

We therefore arrive at the first version of our scheme for 3D Navier–Stokes equation (3.1).

**The GMAC3Dc-skew scheme.** Solve for \(\mathbf{u} \in C^1([0,T]; L^2(\Omega_c, \mathbb{R}^3))\) and \(p \in C^0([0,T]; L^2(\Omega_g, \mathbb{R}))\) such that

\begin{align}
\mathbf{u}_t + \nabla \times \mathbf{u} + \nabla p &= \nu \nabla \times \omega + \mathbf{f} & \text{on } \Omega_c, \\
\omega &= \nabla' \times \mathbf{u} & \text{on } \Omega_g, \\
\nabla' \cdot \mathbf{u} &= 0 & \text{on } \Omega_g,
\end{align}
where

\[(B.5) \quad \hat{\nabla}_h : L^2(\bar{\Omega}_g, \mathbb{R}) \mapsto L^2(\Omega_c, \mathbb{R}^3), \quad \hat{\nabla}_h p := (\hat{D}_1 p) \hat{e}_1 + (\hat{D}_2 p) \hat{e}_2 + (\hat{D}_3 p) \hat{e}_3,\]

\[(B.6) \quad \hat{\nabla}'_h : L^2(\Omega_c, \mathbb{R}^3) \mapsto L^2(\bar{\Omega}_g, \mathbb{R}^3), \quad \hat{\nabla}'_h \cdot u := \frac{1}{\sqrt{\hat{g}_h}} \left( \sum_{\alpha=1}^{3} D'_\alpha \left( \sqrt{\hat{g}_h} \hat{g}_\alpha \right) \right),\]

\[(B.7) \quad \hat{\nabla}'_h \times u := \frac{1}{\sqrt{\hat{g}_h}} \left( \sum_{\gamma=1}^{3} \pi'_\gamma \left( \sqrt{\hat{g}_h} \hat{g}_\gamma \right) \right),\]

\[(B.8) \quad \hat{\nabla}'_h \times \omega := \frac{1}{\sqrt{\hat{g}_h}} \left( \sum_{\gamma=1}^{3} \pi'_\gamma \left( \sqrt{\hat{g}_h} \hat{g}_\gamma \right) \right),\]

and

\[(B.9) \quad \hat{\triangle}'_h : L^2(\bar{\Omega}_g, \mathbb{R}) \mapsto L^2(\bar{\Omega}_g, \mathbb{R}), \quad \hat{\triangle}'_h p = \frac{1}{\sqrt{\hat{g}_h}} \sum_{\mu,\nu=1}^{3} \pi'_{\mu,\nu} \left( \sqrt{\hat{g}_h} \hat{g}_{\mu,\nu} \hat{D}_\nu p \right).\]

Alternatively, one could have chosen \(\{\hat{\xi}_\alpha\}_{\alpha=1}^{3}, \{\hat{\xi}'_\alpha\}_{\alpha=1}^{3}, \) or \(\{\hat{\xi}'_\alpha\}_{\alpha=1}^{3}\) as coordinates and discretized accordingly (Figure B.1).

The resulting schemes differ from each other slightly. It should be noted that the averaged version of the four skew operators, for example,

\[(B.10) \quad \nabla_h p := \frac{1}{4} (\hat{\nabla}_h p + \hat{\nabla}'_h p + \hat{\nabla}_h p + \hat{\nabla}'_h p),\]
is isotropic or, more precisely, invariant under the coordinate transformation \((\xi^1, \xi^2, \xi^3) \mapsto (\xi^2, \xi^3, \xi^1)\). We thus also propose the following isotropic version of the 3D scheme and elaborate with more details below.

We first give an alternative expression for the isotropic gradient operator. To this end, we define the following operators with respect to the default coordinate \(\{\xi^1, \xi^2, \xi^3\}\):

\[
\begin{align*}
(D)_{11} &:= D_1 A_2 A_3 : L^2(\bar{\Omega}_g, \mathbb{R}) \rightarrow L^2(\bar{\Omega}_c, \mathbb{R}), \\
(D)_{12} &:= D_2 A_3 A_1 : L^2(\bar{\Omega}_g, \mathbb{R}) \rightarrow L^2(\bar{\Omega}_c, \mathbb{R}), \\
(D)_{13} &:= D_3 A_1 A_2 : L^2(\bar{\Omega}_g, \mathbb{R}) \rightarrow L^2(\bar{\Omega}_c, \mathbb{R}), \\
(D)_{21} &:= D'_1 A'_2 A'_3 : L^2(\bar{\Omega}_c, \mathbb{R}) \rightarrow L^2(\bar{\Omega}_g, \mathbb{R}), \\
(D)_{22} &:= D'_2 A'_3 A'_1 : L^2(\bar{\Omega}_c, \mathbb{R}) \rightarrow L^2(\bar{\Omega}_g, \mathbb{R}), \\
(D)_{23} &:= D'_3 A'_1 A'_2 : L^2(\bar{\Omega}_c, \mathbb{R}) \rightarrow L^2(\bar{\Omega}_g, \mathbb{R}),
\end{align*}
\]

where \(D\) is the standard short-stencil finite difference operator, \(D'\) is the reduced finite difference operator as defined in (3.11), and the averaging operators are defined by

\[
(Ag)_{i-\frac{1}{2}} = \frac{1}{2}(g_{i+1} + g_{i-1}), \quad 1 \leq i \leq N, \quad (A' f)_i = \begin{cases} f_i & i = 0, \\ \frac{1}{2}(f_{i+1} + f_{i-1}) & 1 \leq i \leq N - 1, \\ f_{i-N} & i = N. \end{cases}
\]

With lengthy but elementary calculations, it can be shown that the isotropic gradient defined in (B.10) can be recast in terms of the operators in the default coordinate

\[
\nabla_h \omega = (\bar{D}_1 p)e^1 + (\bar{D}_2 p)e^2 + (\bar{D}_3 p)e^3 \in L^2(\bar{\Omega}_c, \mathbb{R}^3),
\]

where \(e^\alpha = \nabla \xi^\alpha\), \(e_\alpha = \partial x / \partial \xi^\alpha\). We therefore recast the isotropic 3D GMAC scheme as follows.

**The GMAC3Dc-iso scheme.** Solve for \(u \in C^1([0,T]; L^2(\bar{\Omega}_c, \mathbb{R}^3))\) and \(p \in C^0([0,T]; L^2(\bar{\Omega}_g, \mathbb{R}))\) such that

\[
\begin{align*}
\nabla_h \omega + \bar{\omega} \times u + \nabla_h p &= -\nu \nabla_h \omega + f \quad \text{on } \bar{\Omega}_c, \\
\omega &= \nabla'_h \times u \quad \text{on } \bar{\Omega}_c, \\
\nabla'_h \cdot u &= 0 \quad \text{on } \bar{\Omega}_g,
\end{align*}
\]

with \(\omega = (\bar{\omega}^1, \bar{\omega}^2, \bar{\omega}^3)\) and \(\bar{\omega} \times u \in L^2(\bar{\Omega}_c, \mathbb{R}^3)\) defined by

\[
\bar{\omega} \times u := \sqrt{gh} \begin{vmatrix} e^1 & e^2 & e^3 \\ \bar{\omega}^1 & \bar{\omega}^2 & \bar{\omega}^3 \\ u^1 & u^2 & u^3 \end{vmatrix}, \quad \bar{\omega}^\alpha := A_1 A_2 A_3 \omega^\alpha \in L^2(\bar{\Omega}_c, \mathbb{R}),
\]

and

\[
\begin{align*}
\nabla'_h \cdot : L^2(\bar{\Omega}_c, \mathbb{R}^3) &\rightarrow L^2(\bar{\Omega}_g, \mathbb{R}), \\
\nabla'_h \cdot u &= \frac{1}{\sqrt{gh}} \left( \sum_{\alpha=1}^{3} D'_\alpha (\sqrt{g_h} \omega^\alpha) \right), \\
\nabla'_h \times : L^2(\bar{\Omega}_c, \mathbb{R}^3) &\rightarrow L^2(\bar{\Omega}_g, \mathbb{R}^3), \\
\nabla'_h \times u &= \frac{1}{\sqrt{gh}} \begin{vmatrix} e_1 & e_2 & e_3 \\ D'_1 & D'_2 & D'_3 \\ u_1 & u_2 & u_3 \end{vmatrix}, \\
\nabla_h \times : L^2(\bar{\Omega}_c, \mathbb{R}^3) &\rightarrow L^2(\bar{\Omega}_g, \mathbb{R}^3), \\
\nabla_h \times \omega &= \frac{1}{\sqrt{gh}} \begin{vmatrix} e_1 & e_2 & e_3 \\ D_1 & D_2 & D_3 \\ \omega_1 & \omega_2 & \omega_3 \end{vmatrix}.
\end{align*}
\]

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It can be shown that the analogue of (B.10) remains valid for the discrete operators \( \bar{\nabla}' \cdot, \bar{\nabla}' \times, \) and \( \bar{\nabla}' h \times \) as well. In (B.20)–(B.23), the covariant components and contravariant components of a vector field \( \mathbf{v} \) in \( L^2(\Omega_c, \mathbb{R}^3) \) or \( L^2(\bar{\Omega}_g, \mathbb{R}^3) \) are transformed to each other through

\[
\sum_{\nu=1}^{3} g^h_{\mu\nu} v_{\nu} \quad \text{and} \quad \sum_{\nu=1}^{3} g^h_{\mu\nu} v^{\nu}
\]

whenever necessary. Here \( g^h_{\mu\nu} \) and \( g^h_{\mu\nu} \) are (numerical) metric tensors with respect to the default coordinate \((\xi^1, \xi^2, \xi^3)\) defined on cell centers and satisfy

\[
\sum_{\gamma=1}^{3} g^h_{\mu\gamma} g^h_{\nu\gamma} = \delta^\mu_\nu.
\]

The corresponding isotropic Laplacian is given by

\[
\bar{\Delta}'_h : L^2(\bar{\Omega}_g, \mathbb{R}) \mapsto L^2(\bar{\Omega}_g, \mathbb{R}), \quad \bar{\Delta}'_h p = \bar{\nabla}'_h \cdot \bar{\nabla}'_h p = \frac{1}{\sqrt{g^h}} \sum_{\mu,\nu=1}^{3} \bar{D}'_\mu (\sqrt{g^h} g^h_{\mu\nu} \bar{D}'_\nu p).
\]

The 3D analogue of Lemma 3.1 remains valid for the isotropic difference operators.

**Lemma B.1.** For \( u \in L^2(\Omega_c, \mathbb{R}^3) \), \( v \in L^2(\bar{\Omega}_g, \mathbb{R}^3) \), and \( p \in L^2(\bar{\Omega}_g, \mathbb{R}) \), we have

1. \( \langle u, \bar{\nabla}'_h p \rangle_{\Omega_c} = -\langle \bar{\nabla}'_h \cdot u, p \rangle_{\bar{\Omega}_g}. \)
2. \( \langle u, \bar{\nabla}'_h \times \omega \rangle_{\Omega_c} = \langle \bar{\nabla}'_h \times u, \omega \rangle_{\bar{\Omega}_g}. \)
3. \( \bar{\Delta}'_h p = \bar{\nabla}'_h \cdot \bar{\nabla}'_h p. \)
4. If \( \psi \in L^2(\bar{\Omega}_g, \mathbb{R}^3) \), then

\[
\bar{\nabla}'_h \cdot \bar{\nabla}'_h \times \psi \equiv 0 \quad \text{on} \quad \bar{\Omega}_g.
\]

If in addition, \( \psi \times n = 0 \) on \( \Gamma_g \), then

\[
\bar{\nabla}'_h \cdot \bar{\nabla}'_h \times \psi \equiv 0 \quad \text{on} \quad \bar{\Omega}_g.
\]

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