



# A superconvergent hybridizable discontinuous Galerkin method for Dirichlet boundary control of elliptic PDEs

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## Abstract

We begin an investigation of hybridizable discontinuous Galerkin (HDG) methods for approximating the solution of Dirichlet boundary control problems governed by elliptic PDEs. These problems can involve atypical variational formulations, and often have solutions with low regularity on polyhedral domains. These issues can provide challenges for numerical methods and the associated numerical analysis. We propose an HDG method for a Dirichlet boundary control problem for the Poisson equation, and obtain optimal a priori error estimates for the control. Specifically, under certain assumptions, for a 2D convex polygonal domain we show the control converges at a superlinear rate. We present 2D and 3D numerical experiments to demonstrate our theoretical results.

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## 1 Introduction

We consider the following elliptic Dirichlet boundary control problem on a Lipschitz polyhedral domain  $\Omega \subset \mathbb{R}^d$ ,  $d = 2, 3$ , with boundary  $\Gamma = \partial\Omega$ :

$$\min J(u), \quad J(u) = \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u\|_{L^2(\Gamma)}^2, \quad (1)$$

where  $\gamma > 0$  and  $y$  is the solution of the Poisson equation with nonhomogeneous Dirichlet boundary conditions

$$-\Delta y = f \quad \text{in } \Omega, \quad (2)$$

$$y = u \quad \text{on } \Gamma. \quad (3)$$

It is well known that the Dirichlet boundary control problem (1)–(3) is equivalent to the optimality system

$$-\Delta y = f \quad \text{in } \Omega, \quad (4a)$$

$$y = u \quad \text{on } \Gamma, \quad (4b)$$

$$-\Delta z = y - y_d \quad \text{in } \Omega, \quad (4c)$$

$$z = 0 \quad \text{on } \Gamma, \quad (4d)$$

$$u = \gamma^{-1} \frac{\partial z}{\partial \mathbf{n}} \quad \text{on } \Gamma, \quad (4e)$$

where  $\mathbf{n}$  is the unit outer normal to  $\Gamma$ .

Dirichlet boundary control has many applications in fluid flow problems and other fields, and therefore the mathematical study of these control problems has become an important area of research. Major theoretical and computational developments have been made in the recent past; see, e.g., [8, 19, 20, 24–26, 31–34, 53, 56, 58]. However, only in the last 10 years have researchers developed thorough well-posedness, regularity, and finite element error analysis results for elliptic PDEs; see [1, 6, 21, 42, 59] and the references therein. One difficulty of Dirichlet boundary control problems is that the Dirichlet boundary data does not directly enter a standard variational setting for the PDE; instead, the state equation is understood in a very weak sense. Also, solutions of the optimality system typically do not have high regularity on polyhedral domains; corners cause the normal derivative of the adjoint state  $\partial z / \partial \mathbf{n}$  in the optimality condition (4) to have limited smoothness. Solutions with limited regularity can lead to complications for numerical methods and numerical analysis.

To avoid the difficulties described above, researchers have considered other approaches including modified cost functionals [13, 30, 32, 48], approximating the Dirichlet boundary condition with a Robin boundary condition [2, 3, 5, 35, 53], and weak boundary penalization [9].

One way to approximate the solution of the original problem without penalization and also avoid the variational difficulty is to use a mixed finite element method. Recently, Gong and Yan [29] considered this approach and obtained

$$\|u - u_h\|_{0,\Gamma} = O(h^{1-1/s})$$

when the control belongs to  $H^{1-1/s}(\Gamma)$  and the lowest order Raviart–Thomas elements are used for the computation.

As researchers continue to investigate Dirichlet boundary control problems of increasingly complexity, it may become natural to utilize discontinuous Galerkin methods for the spatial discretization of problems involving strong convection and discontinuities. We have performed preliminary computations using an hybridizable discontinuous Galerkin (HDG) method for a similar Dirichlet boundary control problem for the Stokes equations. Our preliminary results for this problem indicate that the optimal control can indeed be discontinuous at the corners of the domain. Before we continue to investigate problems of such complexity, we begin this line of research by considering an HDG method to approximate the solution of the above Dirichlet boundary control problem.

HDG methods also utilize a mixed formulation and therefore avoid the variational difficulty of the Dirichlet control problem. Furthermore, the number of degrees of freedom for HDG methods are much less than standard mixed methods or other DG approaches. Moreover, the RT element is a special case of the HDG method. We provide more background about HDG methods in Sect. 3.

We propose an HDG method to approximate the control in Sect. 3, and in Sect. 4 we prove an optimal superlinear rate of convergence for the control in 2D using discontinuous linear elements for the control on a quasi-uniform mesh under certain assumptions on the domain and  $y_d$ . To give a specific example, for a rectangular 2D domain and  $y_d \in H^1(\Omega) \cap L^\infty(\Omega)$ , we obtain the following a priori error bounds for the state  $y$ , adjoint state  $z$ , their fluxes  $\mathbf{q} = -\nabla y$  and  $\mathbf{p} = -\nabla z$ , and the optimal control  $u$ :

$$\begin{aligned} \|y - y_h\|_{0,\Omega} &= O(h^{3/2-\varepsilon}), & \|z - z_h\|_{0,\Omega} &= O(h^{3/2-\varepsilon}), \\ \|\mathbf{q} - \mathbf{q}_h\|_{0,\Omega} &= O(h^{1-\varepsilon}), & \|\mathbf{p} - \mathbf{p}_h\|_{0,\Omega} &= O(h^{3/2-\varepsilon}), \end{aligned}$$

and

$$\|u - u_h\|_{0,\Gamma} = O(h^{3/2-\varepsilon}),$$

for any  $\varepsilon > 0$ . We demonstrate the performance of the HDG method with numerical experiments in 2D and 3D in Sect. 5.

We note that an optimal superlinear convergence rate for the control on polygonal domains has also been recently obtained in [1, Theorem 4.1 and Remark 4.8] for standard continuous finite elements using linear elements on a superconvergence mesh or quadratic elements on a quasi-uniform mesh. (Also see [21] for similar results on curved domains.) The number of degrees of freedom for these standard finite element methods is lower than the HDG method considered here with discontinuous linear elements for the control. However, we do not give a thorough comparison of the methods here since, as mentioned above, the primary goal of this work is to begin an

investigation of HDG methods for Dirichlet boundary control problems before moving to more complex problems, for which DG methods may be more appropriate.

## 2 Background: the optimality system and regularity

To begin, we review some fundamental results concerning the optimality system for the control problem in polyhedral domains and the regularity of the solution in 2D polygonal domains. We also provide a mixed formulation of the optimality system and prove it is well-posed.

Throughout the paper we adopt the standard notation  $W^{m,p}(\Omega)$  for Sobolev spaces on  $\Omega$  with norm  $\|\cdot\|_{m,p,\Omega}$  and seminorm  $|\cdot|_{m,p,\Omega}$ . We denote  $W^{m,2}(\Omega)$  by  $H^m(\Omega)$  with norm  $\|\cdot\|_{m,\Omega}$  and seminorm  $|\cdot|_{m,\Omega}$ . Also,  $H_0^1(\Omega) = \{v \in H^1(\Omega) : v = 0 \text{ on } \partial\Omega\}$ . We denote the  $L^2$ -inner products on  $L^2(\Omega)$  and  $L^2(\Gamma)$  by

$$(v, w) = \int_{\Omega} vw \quad \forall v, w \in L^2(\Omega),$$

$$\langle v, w \rangle = \int_{\Gamma} vw \quad \forall v, w \in L^2(\Gamma).$$

Define the space  $H(\text{div}; \Omega)$  as

$$H(\text{div}, \Omega) = \{\mathbf{v} \in [L^2(\Omega)]^d, \nabla \cdot \mathbf{v} \in L^2(\Omega)\}.$$

The inner product  $(\cdot, \cdot)$  is defined above only for scalar variables, but we use the same inner product notation for vector functions:  $(\mathbf{q}, \mathbf{p}) = \int_{\Omega} \mathbf{q} \cdot \mathbf{p}$  for all  $\mathbf{q}, \mathbf{p} \in [L^2(\Omega)]^d$ . Throughout this paper, the norm corresponding to a given inner product is defined in the standard way.

We use the bracket  $[\cdot, \cdot]_{\Gamma}$  to denote the  $H^{-1/2}(\Gamma) \times H^{1/2}(\Gamma)$  duality pairing. Recall for  $\mathbf{r} \in H(\text{div}, \Omega)$ , it is known that  $\mathbf{r} \cdot \mathbf{n} \in H^{-1/2}(\Gamma)$ . Also, for any  $v \in H^1(\Omega)$  with boundary trace  $v|_{\Gamma} \in H^{1/2}(\Gamma)$ , we have the integration by parts formula (see, e.g., [55, Theorem 6.1])

$$(v, \nabla \cdot \mathbf{r}) = [v|_{\Gamma}, \mathbf{r} \cdot \mathbf{n}]_{\Gamma} - (\nabla v, \mathbf{r}) \quad \text{for all } \mathbf{r} \in H(\text{div}, \Omega). \quad (5)$$

Furthermore, for any  $v_{\Gamma} \in H^{1/2}(\Gamma)$  and  $\mathbf{r} \in H(\text{div}, \Omega)$  satisfying  $\mathbf{r} \cdot \mathbf{n} \in L^2(\Gamma)$ , we have that the duality pairing reduces to the  $L^2(\Gamma)$  inner product (see, e.g., [55, Section 6.1]), i.e.,

$$[v|_{\Gamma}, \mathbf{r} \cdot \mathbf{n}]_{\Gamma} = \langle v|_{\Gamma}, \mathbf{r} \cdot \mathbf{n} \rangle.$$

### 2.1 The optimality system

For the remainder of this section, we assume  $f, y_d \in L^2(\Omega)$  and  $\Omega \subset \mathbb{R}^d$ ,  $d = 2, 3$ , is a bounded convex polyhedral domain with boundary  $\Gamma$ . Below, we recall a precise well-posedness result from [42] for the optimal control problem.

**Lemma 2.1** [42, Lemma 2.6] *The pair  $(u, y) \in L^2(\partial\Omega) \times L^2(\Omega)$  is a solution of the optimal control problem if and only if there exists an adjoint state  $z \in H^2(\Omega) \cap H_0^1(\Omega)$  such that  $(u, y, z)$  solves the optimality system*

$$-(y, \Delta\varphi) + \langle u, \partial\varphi/\partial\mathbf{n} \rangle = (f, \varphi) \quad \forall \varphi \in H^2 \cap H_0^1(\Omega), \quad (6a)$$

$$-(\psi, \Delta z) - (y, \psi) = -(y_d, \psi) \quad \forall \psi \in L^2(\Omega), \quad (6b)$$

$$\langle \gamma u - \partial z/\partial\mathbf{n}, \chi \rangle = 0 \quad \forall \chi \in L^2(\partial\Omega). \quad (6c)$$

**Remark 2.2** In [42], the authors only consider the case of a convex polygonal domain, but their proof also works for a convex polyhedral domain in  $\mathbb{R}^3$ .

As mentioned earlier, the state Eqs. (2), (3) in the optimal control problem is written formally and should be understood to hold in the very weak sense (6a). Similarly, the formal optimality system (4) should be understood in the weak sense (6).

The convexity of the domain gives additional regularity for  $u$  and  $y$ .

**Lemma 2.3** *The solution of the optimality system satisfies*

$$y \in H^1(\Omega), \quad \partial z/\partial\mathbf{n} \in H^{1/2}(\Gamma), \quad u \in H^{1/2}(\Gamma).$$

**Proof** Since  $y - y_d \in L^2(\Omega)$  and  $\Omega$  is a convex polyhedral domain, Eq. (6b) along with [5, Lemma A.2] for the 2D case and [28, Equation (3.5) with  $q = 2$ ] for the 3D case give  $\partial z/\partial\mathbf{n} \in H^{1/2}(\Gamma)$ . The desired regularity for  $u$  follows from (6c). Lastly, Eq. (6a),  $f \in L^2(\Omega)$ , and  $u \in H^{1/2}(\Gamma)$  imply  $y \in H^1(\Omega)$ .  $\square$

To avoid the variational difficulty we follow the strategy introduced by Wei Gong and Ningning Yan [29] and consider a mixed formulation of the optimality system. Introduce two flux variables  $\mathbf{q} = -\nabla y$  and  $\mathbf{p} = -\nabla z$ . We consider the following mixed weak form of (4a)-(4e): Find  $(\mathbf{q}, y, \mathbf{p}, z, u) \in H(\text{div}, \Omega) \times L^2(\Omega) \times H(\text{div}, \Omega) \times L^2(\Omega) \times H^{1/2}(\Gamma)$  such that

$$(\mathbf{q}, \mathbf{r}_1) - (y, \nabla \cdot \mathbf{r}_1) + [u, \mathbf{r}_1 \cdot \mathbf{n}]_\Gamma = 0, \quad (7a)$$

$$(\nabla \cdot \mathbf{q}, w_1) = (f, w_1), \quad (7b)$$

$$(\mathbf{p}, \mathbf{r}_2) - (z, \nabla \cdot \mathbf{r}_2) = 0, \quad (7c)$$

$$(\nabla \cdot \mathbf{p}, w_2) - (y, w_2) = -(y_d, w_2), \quad (7d)$$

$$[\gamma u + \mathbf{p} \cdot \mathbf{n}, \xi]_\Gamma = 0 \quad (7e)$$

for all  $(\mathbf{r}_1, w_1, \mathbf{r}_2, w_2, \xi) \in H(\text{div}, \Omega) \times L^2(\Omega) \times H(\text{div}, \Omega) \times L^2(\Omega) \times H^{1/2}(\Gamma)$ . This mixed formulation can be derived directly from the formally stated optimality system (4). Below, we work with the precise statement of the optimality system (6) and prove the well-posedness of this mixed problem.

Since we require  $u \in H^{1/2}(\Gamma)$ , all of the duality pairings in the above mixed formulation are well defined. Since the domain is convex, Lemma 2.3 implies this requirement for the optimal control  $u$  is not a restriction.

Next, we show the well-posedness of the mixed weak form of the optimality system.

**Theorem 2.4** Let  $(u, y, z)$  be the solution of the optimality system (6), and define  $\mathbf{q} = -\nabla y$  and  $\mathbf{p} = -\nabla z$ . Then  $(\mathbf{q}, y, \mathbf{p}, z, u) \in H(\text{div}, \Omega) \times L^2(\Omega) \times H(\text{div}, \Omega) \times L^2(\Omega) \times H^{1/2}(\Gamma)$  is the unique solution of the mixed optimality system (7).

**Proof** Let  $(u, y, z) \in L^2(\Gamma) \times L^2(\Omega) \times [H^2(\Omega) \cap H_0^1(\Omega)]$  be the unique solution of the optimality system (6). Lemma 2.3 gives the additional regularity  $u \in H^{1/2}(\Gamma)$ ,  $y \in H^1(\Omega)$ , and  $\partial z/\partial \mathbf{n} \in H^{1/2}(\Gamma)$ . Furthermore, this regularity along with (6a) gives that  $u$  is the boundary trace of  $y$ . Then  $(\mathbf{q}, y, \mathbf{p}, z, u)$  give a solution to the mixed formulation. To be complete, we give the details.

For  $\mathbf{r}_1, \mathbf{r}_2 \in H(\text{div}, \Omega)$ , Eqs. (7a) and (7c) follow directly from the integration by parts formula (5) applied to  $y \in H^1(\Omega)$  and  $z \in H_0^1(\Omega)$ , respectively. Also, since  $\partial z/\partial \mathbf{n} \in H^{1/2}(\Gamma)$ ,  $\mathbf{p} = -\nabla z$ , and  $u = -\gamma^{-1}\partial z/\partial \mathbf{n}$  holds in  $L^2(\Gamma)$ , it is clear that (7e) holds. Furthermore, since  $z \in H^2 \cap H_0^1(\Omega)$  satisfies the dual problem (6b) and  $\mathbf{p} = -\nabla z \in H(\text{div}, \Omega)$ , it is clear that (7d) holds.

It remains to show  $\mathbf{q} \in H(\text{div}, \Omega)$  and that (7b) holds. Since  $u \in H^{1/2}(\Gamma)$ ,  $y \in H^1(\Omega)$ ,  $u$  is the boundary trace of  $y$ , and  $\mathbf{q} = -\nabla y$ , the very weak form of the state Eq. (6a) gives that for any  $\varphi \in H^2 \cap H_0^1(\Omega)$  we have

$$(f, \varphi) = -(y, \Delta \varphi) + \langle u, \partial \varphi / \partial \mathbf{n} \rangle = -(\mathbf{q}, \nabla \varphi).$$

Restricting to  $\varphi \in C_0^\infty(\Omega)$  gives, by definition, that  $\nabla \cdot \mathbf{q} = f \in L^2(\Omega)$  in the distributional sense, i.e.,

$$(f, \varphi) = (\nabla \cdot \mathbf{q}, \varphi) \quad \text{for all } \varphi \in C_0^\infty(\Omega).$$

This implies  $\mathbf{q} \in H(\text{div}, \Omega)$ . Also, by density, the above equation holds for all  $\varphi \in L^2(\Omega)$ , and therefore (7b) holds.

Next, to prove the uniqueness we set  $f = y_d = 0$  and show zero is the only solution of the mixed formulation (7). Take  $(\mathbf{r}_1, w_1) = (\mathbf{p}, -z)$ ,  $(\mathbf{r}_2, w_2) = (-\mathbf{q}, y)$ , and  $\xi = -u$ , and then sum the equations to give

$$\gamma \|u\|_{L^2(\Gamma)}^2 + \|y\|_{L^2(\Omega)}^2 = 0.$$

Since  $\gamma > 0$ , this implies  $u = 0$  and  $y = 0$ . Next, since  $y = 0$  and  $u = 0$ , taking  $\mathbf{r}_1 = \mathbf{q}$  implies  $\mathbf{q} = \mathbf{0}$ . Also, taking  $\mathbf{r}_2 = \mathbf{p}$  and  $w_2 = z$  and summing these two equations implies  $\mathbf{p} = \mathbf{0}$ . This leaves

$$0 = -(z, \nabla \cdot \mathbf{r}_2) \quad \text{for all } \mathbf{r}_2 \in H(\text{div}, \Omega).$$

Let  $z_2$  be the solution of  $-\Delta z_2 = z$  with zero Dirichlet boundary conditions. Since the domain is convex, we have  $z_2 \in H^2(\Omega) \cap H_0^1(\Omega)$ . Set  $\mathbf{r}_2 = \nabla z_2 \in H(\text{div}, \Omega)$ . This gives  $0 = \|z\|_{L^2(\Omega)}^2$ , which implies  $z = 0$ . This proves the uniqueness.  $\square$

## 2.2 Regularity

One of the main reasons that the Dirichlet boundary control problem can be challenging numerically is that the solution can have very low regularity, and this restricts the convergence rates of finite element and DG methods. In order to prove a superlinear convergence rate for the optimal control for the HDG method in Sect. 4, we assume the solution has the following fractional Sobolev regularity:

$$u \in H^{r_u}(\Gamma), \quad y \in H^{r_y}(\Omega), \quad z \in H^{r_z}(\Omega), \quad \mathbf{q} \in H^{r_q}(\Omega), \quad \mathbf{p} \in H^{r_p}(\Omega), \quad (8)$$

with

$$r_u > 1, \quad r_y > 1, \quad r_z > 2, \quad r_q > 1/2, \quad r_p > 1. \quad (9)$$

We require  $r_q > 1/2$  in order to guarantee  $\mathbf{q}$  has a well-defined trace on the boundary  $\Gamma$ . We note that it may be possible to use the techniques in [38] to lower the regularity requirement on  $\mathbf{q}$ . We leave this to be considered elsewhere.

For a 2D convex polygonal domain and  $f = 0$ , we use a recent regularity result of Mateos and Neitzel [41] below to give conditions on the domain and  $y_d$  to guarantee the solution has the above regularity. For a higher dimensional convex polyhedral domain, the regularity theory is much more complicated, and we do not attempt to provide conditions to guarantee the above regularity in this work.

**Theorem 2.5** [41, Lemma 3 and Corollary 1] *Assume  $\Omega$  is a convex polygonal domain and  $f = 0$ . Let  $\omega \in [\pi/3, \pi)$  be the largest interior angle of  $\Gamma$ , and define  $p_\Omega, r_\Omega$  by*

$$p_\Omega = \frac{2}{2 - \pi / \max\{\omega, \pi/2\}} \in (2, \infty],$$

and

$$r_\Omega = 1 + \frac{\pi}{\omega} \in (2, 4].$$

If  $y_d \in L^p(\Omega) \cap H^{r-2}(\Omega)$  for all  $p < p_\Omega$  and  $r < r_\Omega$ , then the solution  $(u, y, z)$  satisfies

$$\begin{aligned} u &\in H^{r-3/2}(\Gamma) \cap W^{1-1/p, p}(\Gamma), \\ y &\in H^{r-1}(\Omega) \cap W^{1, p}(\Omega), \\ z &\in H_0^1(\Omega) \cap H^r(\Omega) \cap W^{2, p}(\Omega) \end{aligned}$$

for all

$$p < p_\Omega, \quad r < \min\{3, r_\Omega\}.$$

**Corollary 2.6** *Under the assumptions of Theorem 2.5, the flux variables  $\mathbf{q} = -\nabla y$  and  $\mathbf{p} = -\nabla z$  satisfy*

$$\mathbf{q} \in H^{r-2}(\Omega) \cap H(\text{div}, \Omega), \quad \mathbf{p} \in H^{r-1}(\Omega) \cap H(\text{div}, \Omega)$$

for all  $r < \min\{3, r_\Omega\}$ .

The regularity for the flux variable  $\mathbf{q} = -\nabla y$  is low; Corollary 2.6 only guarantees  $\mathbf{q} \in H^{r_q}$  for some  $0 < r_q < 1$ . For the HDG approximation theory, we need the regularity condition  $r_q > 1/2$ . We can guarantee this condition by restricting the maximum interior angle  $\omega$ . Specifically, if  $y_d$  has the required smoothness and  $\omega$  satisfies

$$\omega \in [\pi/3, 2\pi/3],$$

then  $r_\Omega \in (5/2, 4]$  and we are guaranteed  $\mathbf{q} \in H^{r_q}$  for some  $r_q > 1/2$ .

Also, when we restrict  $\omega \in [\pi/3, 2\pi/3]$  as above, this guarantees  $u \in H^{r_u}$  for some  $1 < r_u < 3/2$  and furthermore the regularity assumption (8), (9) is satisfied. For a rectangular domain, we have  $p_\Omega = \infty$  and  $r_\Omega = 3$ . Therefore if  $y_d \in H^1(\Omega) \cap L^\infty(\Omega)$  we are guaranteed the fractional Sobolev regularity

$$r_u = \frac{3}{2} - \varepsilon, \quad r_y = 2 - \varepsilon, \quad r_z = 3 - \varepsilon, \quad r_q = 1 - \varepsilon, \quad r_p = 2 - \varepsilon$$

for any  $\varepsilon > 0$ .

### 3 HDG formulation and implementation

A mixed method can avoid the variational difficulty by introducing flux variables  $\mathbf{q}$  and  $\mathbf{p}$  and the equation for the optimal control (7e). However, these two additional vector variables will increase the computational cost, even if the lowest order RT method is used.

We introduce an HDG method for the optimality system (4) to take advantage of the mixed formulation and also reduce the computational cost compared to standard mixed methods. Specifically, we introduce the flux variables but eliminate them before we solve the global equation; this significantly reduces the number of degrees of freedom.

HDG methods were proposed by Cockburn et al. in [15] as an improvement of tradition discontinuous Galerkin methods and have many applications; see, e.g., [7, 11, 16–18, 44–47, 57]. The approximate scalar variable and flux are expressed in an element-by-element fashion in terms of an approximate trace of the scalar variable along the element boundary. Then, a unique value for the trace at inter-element boundaries is obtained by enforcing flux continuity. This leads to a global equation system in terms of the approximate boundary traces only. The high number of globally coupled degrees of freedom is significantly reduced compared to other DG methods and standard mixed methods.

Before we introduce the HDG method, we first set some notation. Let  $\{\mathcal{T}_h\}$  be a conforming quasi-uniform polyhedral mesh of  $\Omega$ . We denote by  $\partial\mathcal{T}_h$  the set  $\{\partial K : K \in \mathcal{T}_h\}$ . For an element  $K$  of the collection  $\mathcal{T}_h$ , let  $e = \partial K \cap \Gamma$  denote the boundary face of  $K$  if the  $d - 1$  Lebesgue measure of  $e$  is non-zero. For two elements  $K^+$  and

$K^-$  of the collection  $\mathcal{T}_h$ , let  $e = \partial K^+ \cap \partial K^-$  denote the interior face between  $K^+$  and  $K^-$  if the  $d - 1$  Lebesgue measure of  $e$  is non-zero. Let  $\varepsilon_h^o$  and  $\varepsilon_h^\partial$  denote the sets of interior and boundary faces, respectively. We denote by  $\varepsilon_h$  the union of  $\varepsilon_h^o$  and  $\varepsilon_h^\partial$ . We finally introduce

$$(w, v)_{\mathcal{T}_h} = \sum_{K \in \mathcal{T}_h} (w, v)_K, \quad \langle \zeta, \rho \rangle_{\partial \mathcal{T}_h} = \sum_{K \in \mathcal{T}_h} \langle \zeta, \rho \rangle_{\partial K}.$$

Similar to the notation defined in Sect. 2, we also use the inner product notation  $(\cdot, \cdot)_{\mathcal{T}_h}$  for vector functions.

Let  $\mathcal{P}^k(D)$  denote the set of polynomials of degree at most  $k$  on a domain  $D$ . We introduce the discontinuous finite element spaces

$$\mathbf{V}_h := \{\mathbf{v} \in [L^2(\Omega)]^d : \mathbf{v}|_K \in [\mathcal{P}^k(K)]^d, \forall K \in \mathcal{T}_h\}, \quad (10)$$

$$W_h := \{w \in L^2(\Omega) : w|_K \in \mathcal{P}^{k+1}(K), \forall K \in \mathcal{T}_h\}, \quad (11)$$

$$M_h := \{\mu \in L^2(\varepsilon_h) : \mu|_e \in \mathcal{P}^k(e), \forall e \in \varepsilon_h\}. \quad (12)$$

The space  $W_h$  is for scalar variables, while  $\mathbf{V}_h$  is for flux variables and  $M_h$  is for boundary trace variables. Note that the polynomial degree for the scalar variables is one order higher than the polynomial degree for the other variables. Also, the boundary trace variables will be used to eliminate the state and flux variables from the coupled global equations, thus substantially reducing the number of degrees of freedom.

Let  $M_h(o)$  and  $M_h(\partial)$  denote the spaces defined in the same way as  $M_h$ , but with  $\varepsilon_h$  replaced by  $\varepsilon_h^o$  and  $\varepsilon_h^\partial$ , respectively. Note that  $M_h$  consists of functions which are continuous inside the faces (or edges)  $e \in \varepsilon_h$  and discontinuous at their borders. In addition, for any function  $w \in W_h$  we use  $\nabla w$  to denote the piecewise gradient on each element  $K \in \mathcal{T}_h$ . A similar convention applies to the divergence operator  $\nabla \cdot \mathbf{r}$  for all  $\mathbf{r} \in \mathbf{V}_h$ .

### 3.1 The HDG formulation

To approximate the solution of the mixed weak form (4a)–(4e) of the optimality system, the HDG method seeks approximate fluxes  $\mathbf{q}_h, \mathbf{p}_h \in \mathbf{V}_h$ , states  $y_h, z_h \in W_h$ , interior element boundary traces  $\hat{y}_h^o, \hat{z}_h^o \in M_h(o)$ , and boundary control  $u_h \in M_h(\partial)$  satisfying

$$(\mathbf{q}_h, \mathbf{r}_1)_{\mathcal{T}_h} - (y_h, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle \hat{y}_h^o, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle u_h, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} = 0, \quad (13a)$$

$$-(\mathbf{q}_h, \nabla w_1)_{\mathcal{T}_h} + \langle \hat{\mathbf{q}}_h \cdot \mathbf{n}, w_1 \rangle_{\partial \mathcal{T}_h} = (f, w_1)_{\mathcal{T}_h} \quad (13b)$$

for all  $(\mathbf{r}_1, w_1) \in \mathbf{V}_h \times W_h$ ,

$$(\mathbf{p}_h, \mathbf{r}_2)_{\mathcal{T}_h} - (z_h, \nabla \cdot \mathbf{r}_2)_{\mathcal{T}_h} + \langle \hat{z}_h^o, \mathbf{r}_2 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \quad (13c)$$

$$-(\mathbf{p}_h, \nabla w_2)_{\mathcal{T}_h} + \langle \widehat{\mathbf{p}}_h \cdot \mathbf{n}, w_2 \rangle_{\partial \mathcal{T}_h} - (y_h, w_2)_{\mathcal{T}_h} = -(y_d, w_2)_{\mathcal{T}_h} \quad (13d)$$

for all  $(\mathbf{r}_2, w_2) \in \mathbf{V}_h \times W_h$ ,

$$\langle \widehat{\mathbf{q}}_h \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0 \quad (13e)$$

for all  $\mu_1 \in M_h(o)$ ,

$$\langle \widehat{\mathbf{p}}_h \cdot \mathbf{n}, \mu_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0 \quad (13f)$$

for all  $\mu_2 \in M_h(o)$ , and

$$\langle u_h, \mu_3 \rangle_{\varepsilon_h^\partial} + \langle \gamma^{-1} \widehat{\mathbf{p}}_h \cdot \mathbf{n}, \mu_3 \rangle_{\varepsilon_h^\partial} = 0 \quad (13g)$$

for all  $\mu_3 \in M_h(\partial)$ .

The numerical traces on  $\partial \mathcal{T}_h$  are defined as

$$\widehat{\mathbf{q}}_h \cdot \mathbf{n} = \mathbf{q}_h \cdot \mathbf{n} + h^{-1} (P_M y_h - \widehat{y}_h^o) \quad \text{on } \partial \mathcal{T}_h \setminus \varepsilon_h^\partial, \quad (13h)$$

$$\widehat{\mathbf{q}}_h \cdot \mathbf{n} = \mathbf{q}_h \cdot \mathbf{n} + h^{-1} (P_M y_h - u_h) \quad \text{on } \varepsilon_h^\partial, \quad (13i)$$

$$\widehat{\mathbf{p}}_h \cdot \mathbf{n} = \mathbf{p}_h \cdot \mathbf{n} + h^{-1} (P_M z_h - \widehat{z}_h^o) \quad \text{on } \partial \mathcal{T}_h \setminus \varepsilon_h^\partial, \quad (13j)$$

$$\widehat{\mathbf{p}}_h \cdot \mathbf{n} = \mathbf{p}_h \cdot \mathbf{n} + h^{-1} P_M z_h \quad \text{on } \varepsilon_h^\partial, \quad (13k)$$

where  $P_M$  denotes the standard  $L^2$ -orthogonal projection from  $L^2(\varepsilon_h)$  onto  $M_h$ . This completes the formulation of the HDG method.

The HDG formulation with  $h^{-1}$  stabilization, polynomial degree  $k+1$  for the scalar unknown, and polynomial degree  $k$  for the other unknowns was originally introduced by Lehrenfeld in [36] and first analyzed by Oikawa in [49]. See [14, Sections 6.5-6.6] for more about the history of this method, and its relationship to other HDG and DG methods. This HDG method is widely considered to be a superconvergent method. Specifically, if the solution of the PDE is smooth enough, then in many cases  $O(h^{k+2})$  error estimates can be obtained for the state variable for all  $k \geq 0$ ; see [38, 49] for the Poisson equation, [50] for linear elasticity, [51] for linear convection diffusion, and [52] the Navier–Stokes equations. Hence, from the viewpoint of globally coupled degrees of freedom, this method achieves superconvergence for the scalar variable. However, this method loses the superconvergence when  $k=0$  for convection diffusion [51] and the Navier–Stokes equations [52]. A fix for linear convection diffusion is found in [10], but a fix for the Navier–Stokes equations is not known.

### 3.2 Implementation

To arrive at the HDG formulation we implement numerically, we insert (13h)–(13k) into (13a)–(13g), and find after some simple manipulations that

$$(\mathbf{q}_h, \mathbf{p}_h, y_h, z_h, \widehat{y}_h^o, \widehat{z}_h^o, u_h) \in \mathbf{V}_h \times \mathbf{V}_h \times W_h \times W_h \times M_h(o) \times M_h(o) \times M_h(\partial)$$

is the solution of the following weak formulation:

$$(\mathbf{q}_h, \mathbf{r}_1)_{\mathcal{T}_h} - (y_h, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle \hat{\mathbf{y}}_h^o, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle u_h, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} = 0, \quad (14a)$$

$$(\mathbf{p}_h, \mathbf{r}_2)_{\mathcal{T}_h} - (z_h, \nabla \cdot \mathbf{r}_2)_{\mathcal{T}_h} + \langle \hat{\mathbf{z}}_h^o, \mathbf{r}_2 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \quad (14b)$$

$$(\nabla \cdot \mathbf{q}_h, w_1)_{\mathcal{T}_h} + \langle h^{-1} P_M y_h, w_1 \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} \hat{\mathbf{y}}_h^o, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \quad (14c)$$

$$- \langle h^{-1} u_h, w_1 \rangle_{\varepsilon_h^\partial} = (f, w_1)_{\mathcal{T}_h}, \quad (14d)$$

$$(\nabla \cdot \mathbf{p}_h, w_2)_{\mathcal{T}_h} + \langle h^{-1} P_M z_h, w_2 \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} \hat{\mathbf{z}}_h^o, w_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \quad (14e)$$

$$-(y_h, w_2)_{\mathcal{T}_h} = -(y_d, w_2)_{\mathcal{T}_h}, \quad (14f)$$

$$\langle \mathbf{q}_h \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} y_h, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle h^{-1} \hat{\mathbf{y}}_h^o, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \quad (14g)$$

$$\langle \mathbf{p}_h \cdot \mathbf{n}, \mu_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} z_h, \mu_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle h^{-1} \hat{\mathbf{z}}_h^o, \mu_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \quad (14h)$$

$$\langle u_h, \mu_3 \rangle_{\varepsilon_h^\partial} + \langle \gamma^{-1} \mathbf{p}_h \cdot \mathbf{n}, \mu_3 \rangle_{\varepsilon_h^\partial} + \langle \gamma^{-1} h^{-1} z_h, \mu_3 \rangle_{\varepsilon_h^\partial} = 0 \quad (14i)$$

for all  $(\mathbf{r}_1, \mathbf{r}_2, w_1, w_2, \mu_1, \mu_2, \mu_3) \in V_h \times V_h \times W_h \times W_h \times M_h(o) \times M_h(o) \times M_h(\partial)$ .

### 3.2.1 Matrix equations

Assume  $V_h = \text{span}\{\varphi_i\}_{i=1}^{N_1}$ ,  $W_h = \text{span}\{\phi_i\}_{i=1}^{N_2}$ ,  $M_h(o) = \text{span}\{\psi_i\}_{i=1}^{N_3}$ , and  $M_h(\partial) = \text{span}\{\psi_i\}_{i=1+N_3}^{N_4}$ . Then

$$\begin{aligned} \mathbf{q}_h &= \sum_{j=1}^{N_1} q_j \varphi_j, & \mathbf{p}_h &= \sum_{j=1}^{N_1} p_j \varphi_j, & y_h &= \sum_{j=1}^{N_2} y_j \phi_j, & z_h &= \sum_{j=1}^{N_2} z_j \phi_j, \\ \hat{\mathbf{y}}_h^o &= \sum_{j=1}^{N_3} \alpha_j \psi_j, & \hat{\mathbf{z}}_h^o &= \sum_{j=1}^{N_3} \gamma_j \psi_j, & u_h &= \sum_{j=1+N_3}^{N_4} \beta_j \psi_j. \end{aligned} \quad (15)$$

Substitute (15) into (14a)–(14i) and use the corresponding test functions to test (14a)–(14i), respectively, to obtain the matrix equation

$$\begin{bmatrix} A_1 & 0 & -A_2 & 0 & A_8 & 0 & A_9 \\ 0 & A_1 & 0 & -A_2 & 0 & A_8 & 0 \\ A_2^T & 0 & A_5 & 0 & -A_{10} & 0 & -A_{11} \\ 0 & A_2^T & -A_4 & A_5 & 0 & -A_{10} & 0 \\ A_8^T & 0 & A_{10}^T & 0 & A_{11} & 0 & 0 \\ 0 & A_8^T & 0 & A_{10}^T & 0 & A_{11} & 0 \\ 0 & \gamma^{-1} A_{12} & 0 & \gamma^{-1} A_{13} & 0 & 0 & A_{14} \end{bmatrix} \begin{bmatrix} \mathbf{q} \\ \mathbf{p} \\ \mathbf{y} \\ \mathbf{z} \\ \hat{\mathbf{y}} \\ \hat{\mathbf{z}} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ b_1 \\ -b_2 \\ 0 \\ 0 \\ 0 \end{bmatrix}. \quad (16)$$

Here,  $\mathbf{q}, \mathbf{p}, \mathbf{y}, \mathbf{z}, \widehat{\mathbf{y}}, \widehat{\mathbf{z}}, \mathbf{u}$  are the coefficient vectors for  $\mathbf{q}_h, \mathbf{p}_h, \mathbf{y}_h, \mathbf{z}_h, \widehat{\mathbf{y}}_h^o, \widehat{\mathbf{z}}_h^o, \mathbf{u}_h$ , respectively, and

$$\begin{aligned} A_1 &= [(\boldsymbol{\varphi}_j, \boldsymbol{\varphi}_i)_{\mathcal{T}_h}], \quad A_2 = [(\phi_j, \nabla \cdot \boldsymbol{\varphi}_i)_{\mathcal{T}_h}], \quad A_3 = [(\psi_j, \boldsymbol{\varphi}_i \cdot \mathbf{n})_{\mathcal{T}_h}], \\ A_4 &= [(\phi_j, \phi_i)_{\mathcal{T}_h}], \quad A_5 = [\langle h^{-1} P_M \phi_j, \phi_i \rangle_{\partial \mathcal{T}_h}], \quad A_6 = [\langle h^{-1} \psi_j, \psi_i \rangle_{\partial \mathcal{T}_h}], \\ A_7 &= [\langle h^{-1} \psi_j, \varphi_i \rangle_{\partial \mathcal{T}_h}], \quad b_1 = [(f, \phi_i)_{\mathcal{T}_h}], \quad b_2 = [(y_d, \phi_i)_{\mathcal{T}_h}]. \end{aligned}$$

The remaining matrices  $A_8 - A_{14}$  are constructed by extracting the corresponding rows and columns from  $A_3, A_6$ , and  $A_7$ . In the actual computation, to save memory we do not assemble the large matrix in Eq. (16).

Equation (16) can be rewritten as

$$\begin{bmatrix} B_1 & B_2 & B_3 \\ -B_2^T & B_4 & B_5 \\ B_6 & B_7 & B_8 \end{bmatrix} \begin{bmatrix} \boldsymbol{\alpha} \\ \boldsymbol{\beta} \\ \boldsymbol{\gamma} \end{bmatrix} = \begin{bmatrix} 0 \\ \mathbf{b} \\ 0 \end{bmatrix}, \quad (17)$$

where  $\boldsymbol{\alpha} = [\mathbf{q}; \mathbf{p}], \boldsymbol{\beta} = [\mathbf{y}; \mathbf{z}], \boldsymbol{\gamma} = [\widehat{\mathbf{y}}; \widehat{\mathbf{z}}; \mathbf{u}]$ ,  $\mathbf{b} = [b_1; -b_2]$ , and  $\{B_i\}_{i=1}^8$  are the corresponding blocks of the coefficient matrix in (16).

Due to the discontinuous nature of the approximation spaces  $V_h$  and  $W_h$ , the first two equations of (17) can be used to eliminate both  $\boldsymbol{\alpha}$  and  $\boldsymbol{\beta}$  in an element-by-element fashion. As a consequence, we can write system (17) as

$$\begin{bmatrix} \boldsymbol{\alpha} \\ \boldsymbol{\beta} \end{bmatrix} = \begin{bmatrix} G_1 & H_1 \\ G_2 & H_2 \end{bmatrix} \begin{bmatrix} \boldsymbol{\gamma} \\ \mathbf{b} \end{bmatrix} \quad (18)$$

and

$$B_6 \boldsymbol{\alpha} + B_7 \boldsymbol{\beta} + B_8 \boldsymbol{\gamma} = 0. \quad (19)$$

We provide details on the element-by-element construction of  $G_1, G_2$  and  $H_1, H_2$  in the appendix. Next, we eliminate both  $\boldsymbol{\alpha}$  and  $\boldsymbol{\beta}$  to obtain a reduced globally coupled equation for  $\boldsymbol{\gamma}$  only:

$$\mathbb{K} \boldsymbol{\gamma} = -\mathbb{F} \mathbf{b}, \quad (20)$$

where

$$\mathbb{K} = B_6 G_1 + B_7 G_2 + B_8 \quad \text{and} \quad \mathbb{F} = B_6 H_1 + B_7 H_2.$$

Once  $\boldsymbol{\gamma}$  is available, both  $\boldsymbol{\alpha}$  and  $\boldsymbol{\beta}$  can be recovered from (18).

**Remark 3.1** For HDG methods, the standard approach is to first compute the local solver independently on each element and then assemble the global system. The process we follow here is to first assemble the global system and then reduce its dimension by simple block-diagonal algebraic operations. The two approaches are equivalent.

The introduction of the approximate boundary trace unknowns, the “hybridization” process, allows this reduction of the global system to take place; see [14] for more information.

Equation (18) says we can express the approximate scalar state variable and corresponding fluxes in terms of the approximate traces on the element boundaries. The global Eq. (20) only involves the approximate traces. Therefore, the high number of globally coupled degrees of freedom in the HDG method is significantly reduced. This is one excellent feature of HDG methods.

### 3.3 Discretize-then-optimize and optimize-then-discretize

To approximate the solution of the optimal control problem (1)–(3), in Sects. 3.1, 3.2 we first derived the first-order necessary optimality conditions at the PDE level and then used the HDG method to discretize the optimality system. This strategy is called the optimize-then-discretize (OD) approach.

Another strategy is the discretize-then-optimize (DO) approach. Here, one first discretizes the PDE optimization problem to obtain a finite dimensional optimization problem. An advantage of this approach is that existing optimization algorithms can be utilized. However, this approach can yield spurious numerical results if the DO and OD approaches do not yield equivalent results; see, e.g., [37,39].

Since the DO approach allows the use of existing optimization algorithms, it is important to devise algorithms for which discretization and optimization *commute*, i.e., the DO and OD approaches yield equivalent results. In this section, we prove the two HDG approaches are equivalent. We follow the technique used in [61], where we prove optimization and discretization commute for an embedded discontinuous Galerkin method for a convection diffusion distributed control problem.

For the DO approach, we consider the following HDG discrete cost functional

$$\min_{u_h \in M_h(\partial)} \frac{1}{2} \|y_h - y_d\|_{\mathcal{T}_h}^2 + \frac{\gamma}{2} \|u_h\|_{\varepsilon_h^\partial}^2, \quad \gamma > 0,$$

subject to the HDG discretized state equations for  $(\mathbf{q}_h, y_h, \widehat{y}_h^o) \in \mathbf{V}_h \times W_h \times M_h(o)$  given by

$$\begin{aligned} (\mathbf{q}_h, \mathbf{r}_1)_{\mathcal{T}_h} - (y_h, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle \widehat{y}_h^o, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle u_h, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} = 0, \\ (\nabla \cdot \mathbf{q}_h, w_1)_{\mathcal{T}_h} + \langle h^{-1} P_M y_h, w_1 \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} \widehat{y}_h^o, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ - \langle h^{-1} u_h, w_1 \rangle_{\varepsilon_h^\partial} = (f, w_1)_{\mathcal{T}_h}, \\ \langle \mathbf{q}_h \cdot \mathbf{n} + h^{-1} (P_M y_h - \widehat{y}_h^o), \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \end{aligned}$$

which hold for any  $(\mathbf{r}_1, w_1, \mu_1) \in \mathbf{V}_h \times W_h \times M_h(o)$ .

The Lagrangian functional defined for the adjoint variables  $(\mathbf{p}_h, z_h, \widehat{z}_h^o) \in \mathbf{V}_h \times W_h \times M_h(o)$  is given by

$$\begin{aligned}
\mathcal{L}_h(\mathbf{q}_h, y_h, \widehat{y}_h^o; \mathbf{p}_h, z_h, \widehat{z}_h^o) = & \frac{1}{2} \|y_h - y_d\|_{\mathcal{T}_h}^2 + \frac{\gamma}{2} \|u_h\|_{\varepsilon_h^\partial}^2 \\
& + (\mathbf{q}_h, \mathbf{p}_h)_{\mathcal{T}_h} - (y_h, \nabla \cdot \mathbf{p}_h)_{\mathcal{T}_h} + \langle \widehat{y}_h^o, \mathbf{p}_h \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle u_h, \mathbf{p}_h \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} \\
& - (\nabla \cdot \mathbf{q}_h, z_h)_{\mathcal{T}_h} - \langle h^{-1} P_M y_h, z_h \rangle_{\partial \mathcal{T}_h} + \langle h^{-1} \widehat{y}_h^o, z_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} u_h, z_h \rangle_{\varepsilon_h^\partial} \\
& + (f, z_h)_{\mathcal{T}_h} + \langle \mathbf{q}_h \cdot \mathbf{n} + h^{-1} (P_M y_h - \widehat{y}_h^o), \widehat{z}_h^o \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}. \tag{21}
\end{aligned}$$

Due to the linear-quadratic structure of the optimization problem, we obtain the optimality system by setting the partial Fréchet-derivatives of (21) with respect to the flux  $\mathbf{q}_h$ , state  $y_h$ , numerical trace  $\widehat{y}_h^o$  and control  $u_h$  equal to zero. This gives the adjoint equations and optimality condition

$$\begin{aligned}
\frac{\partial \mathcal{L}_h}{\partial \mathbf{q}_h} \mathbf{r}_2 &= (\mathbf{p}_h, \mathbf{r}_2)_{\mathcal{T}_h} - (z_h, \nabla \cdot \mathbf{r}_2)_{\mathcal{T}_h} + \langle \widehat{z}_h^o, \mathbf{r}_2 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \\
\frac{\partial \mathcal{L}_h}{\partial y_h} w_2 &= -(\nabla \cdot \mathbf{p}_h, w_2)_{\mathcal{T}_h} - \langle h^{-1} P_M z_h, w_2 \rangle_{\partial \mathcal{T}_h} + \langle h^{-1} \widehat{z}_h^o, w_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\
&\quad + (y_h - y_d, w_2)_{\mathcal{T}_h} = 0, \\
\frac{\partial \mathcal{L}_h}{\partial \widehat{y}_h^o} \mu_2 &= \langle \mathbf{p}_h \cdot \mathbf{n} + h^{-1} (z_h - \widehat{z}_h^o), \mu_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} = 0, \\
\frac{\partial \mathcal{L}_h}{\partial u_h} \mu_3 &= \langle \gamma u_h + \mathbf{p}_h \cdot \mathbf{n} + h^{-1} z_h, \mu_3 \rangle_{\varepsilon_h^\partial} = 0,
\end{aligned}$$

which hold for all  $(\mathbf{r}_2, w_2, \mu_2, \mu_3) \in \mathbf{V}_h \times W_h \times M_h(o) \times M_h(\partial)$ . Comparing the HDG discretized optimality system (14) with the above equations shows the two approaches are equivalent, i.e., OD = DO.

## 4 Error analysis

Next, we provide a convergence analysis of the above HDG method for the Dirichlet boundary control problem. Throughout this section, we assume  $\Omega \subset \mathbb{R}^d$ ,  $d = 2, 3$ , is a bounded convex polyhedral domain, the regularity condition (8), (9) is satisfied, and  $h$  is bounded above by some positive constant. For the 2D case, recall Sect. 2 provides conditions on  $\Omega$  and  $y_d$  guaranteeing the required regularity.

Through this section, we use  $A \lesssim B$  to indicate that there exists a positive constant  $C$  such that  $A \leq C B$ , where  $C$  only depends on the polynomial degree  $k$ , the domain  $\Omega$  and the shape regularity of the mesh.

### 4.1 Main result

First, we present the following main theoretical result of this work. Recall we assume the fractional Sobolev regularity exponents satisfy

$$r_u > 1, \quad r_y > 1, \quad r_z > 2, \quad r_q > 1/2, \quad r_p > 1.$$

**Theorem 4.1** For

$$s_y = \min\{r_y, k+2\}, s_z = \min\{r_z, k+2\}, s_q = \min\{r_q, k+1\}, s_p = \min\{r_p, k+1\},$$

we have

$$\begin{aligned} \|u - u_h\|_{\varepsilon_h^0} &\lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}, \\ \|y - y_h\|_{\mathcal{T}_h} &\lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}, \\ \|\mathbf{q} - \mathbf{q}_h\|_{\mathcal{T}_h} &\lesssim h^{s_p - 1} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - 2} \|z\|_{s_z, \Omega} + h^{s_q} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - 1} \|y\|_{s_y, \Omega}, \\ \|\mathbf{p} - \mathbf{p}_h\|_{\mathcal{T}_h} &\lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}, \\ \|z - z_h\|_{\mathcal{T}_h} &\lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

Using the regularity results for the 2D case presented in Sect. 2, we obtain the following result.

**Corollary 4.2** Suppose  $d = 2$  and  $f = 0$ . Let  $\omega \in [\pi/3, 2\pi/3)$  be the largest interior angle of  $\Gamma$ , and define  $p_\Omega, r_\Omega$  by

$$p_\Omega = \frac{2}{2 - \pi / \max\{\omega, \pi/2\}} \in (4, \infty], \quad r_\Omega = 1 + \frac{\pi}{\omega} \in (5/2, 4].$$

Assume  $y_d \in L^p(\Omega) \cap H^{r-2}(\Omega)$  for all  $p < p_\Omega$  and  $r < r_\Omega$ . If  $k = 1$ , then for any  $r$  satisfying  $5/2 < r < \min\{3, r_\Omega\}$  we have

$$\begin{aligned} \|u - u_h\|_{\varepsilon_h^0} &\lesssim h^{r - \frac{3}{2}} (\|\mathbf{p}\|_{H^{r-1}(\Omega)} + \|z\|_{H^r(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|y - y_h\|_{\mathcal{T}_h} &\lesssim h^{r - \frac{3}{2}} (\|\mathbf{p}\|_{H^{r-1}(\Omega)} + \|z\|_{H^r(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|\mathbf{q} - \mathbf{q}_h\|_{\mathcal{T}_h} &\lesssim h^{r - 2} (\|\mathbf{p}\|_{H^{r-1}(\Omega)} + \|z\|_{H^r(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|\mathbf{p} - \mathbf{p}_h\|_{\mathcal{T}_h} &\lesssim h^{r - \frac{3}{2}} (\|\mathbf{p}\|_{H^{r-1}(\Omega)} + \|z\|_{H^r(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|z - z_h\|_{\mathcal{T}_h} &\lesssim h^{r - \frac{3}{2}} (\|\mathbf{p}\|_{H^{r-1}(\Omega)} + \|z\|_{H^r(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}). \end{aligned}$$

Furthermore, if  $k = 0$ , then for any  $r$  as above we have

$$\begin{aligned} \|u - u_h\|_{\varepsilon_h^0} &\lesssim h^{1/2} (\|\mathbf{p}\|_{H^1(\Omega)} + \|z\|_{H^2(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|y - y_h\|_{\mathcal{T}_h} &\lesssim h^{1/2} (\|\mathbf{p}\|_{H^1(\Omega)} + \|z\|_{H^2(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|\mathbf{p} - \mathbf{p}_h\|_{\mathcal{T}_h} &\lesssim h^{1/2} (\|\mathbf{p}\|_{H^1(\Omega)} + \|z\|_{H^2(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}), \\ \|z - z_h\|_{\mathcal{T}_h} &\lesssim h^{1/2} (\|\mathbf{p}\|_{H^1(\Omega)} + \|z\|_{H^2(\Omega)} + \|\mathbf{q}\|_{H^{r-2}(\Omega)} + \|y\|_{H^{r-1}(\Omega)}). \end{aligned}$$

Note that  $\min\{3, r_\Omega\}$  is always greater than  $5/2$ , which guarantees a superlinear convergence rate for all variables except  $\mathbf{q}$  if  $k = 1$ . Also, if  $\Omega$  is a rectangle (i.e.,

$\omega = \pi/2$  and  $y_d \in H^1(\Omega) \cap L^\infty(\Omega)$ , then  $r_\Omega = 3$  and we obtain an  $O(h^{3/2-\varepsilon})$  convergence rate for  $u$ ,  $y$ ,  $z$ , and  $\mathbf{p}$ , and an  $O(h^{1-\varepsilon})$  convergence rate for  $\mathbf{q}$  for any  $\varepsilon > 0$ . For  $k = 1$ , the convergence rates are optimal for the control  $u$  and the flux  $\mathbf{q}$ , but suboptimal for the other variables. For  $k = 0$ , the convergence rates are suboptimal for all variables.

## 4.2 Preliminary material

Before we prove the main result, we discuss  $L^2$  projections, an HDG operator  $\mathcal{B}$ , and the well-posedness of the HDG equations.

For any element  $K$  and boundary face  $e$ , we define the standard  $L^2$  projections  $\Pi : [L^2(K)]^d \rightarrow [\mathcal{P}^k(K)]^d$ ,  $\Pi : L^2(K) \rightarrow \mathcal{P}^{k+1}(K)$ , and  $P_M : L^2(e) \rightarrow \mathcal{P}^k(e)$ , which satisfy

$$\begin{aligned} (\Pi \mathbf{q}, \mathbf{r})_K &= (\mathbf{q}, \mathbf{r})_K, & \forall \mathbf{r} \in [\mathcal{P}^k(K)]^d, \\ (\Pi u, w)_K &= (u, w)_K, & \forall w \in \mathcal{P}^{k+1}(K), \\ \langle P_M m, \mu \rangle_e &= \langle m, \mu \rangle_e, & \forall \mu \in \mathcal{P}^k(e). \end{aligned} \quad (22)$$

Note that  $\Pi : [L^2(\Omega)]^d \rightarrow V_h$ ,  $\Pi : L^2(\Omega) \rightarrow W_h$ , and  $P_M : L^2(\varepsilon_h) \rightarrow M_h$ .

In the analysis, we use the following classical results:

$$\|\mathbf{q} - \Pi \mathbf{q}\|_{\mathcal{T}_h} \lesssim h^{s_q} \|\mathbf{q}\|_{s_q, \Omega}, \quad \|y - \Pi y\|_{\mathcal{T}_h} \lesssim h^{s_y} \|y\|_{s_y, \Omega}, \quad (23a)$$

$$\|y - \Pi y\|_{\partial \mathcal{T}_h} \lesssim h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}, \quad \|\mathbf{q} - \Pi \mathbf{q}\|_{\partial \mathcal{T}_h} \lesssim h^{s_q - \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega}, \quad (23b)$$

$$\|w\|_{\partial \mathcal{T}_h} \lesssim h^{-\frac{1}{2}} \|w\|_{\mathcal{T}_h} \quad \forall w \in W_h, \quad (23c)$$

where  $s_q$  and  $s_y$  are defined in Theorem 4.1. We note that (23a) follows directly from [54, Theorem 2.6], and the inverse inequality (23c) can be found in [60]. Since  $y \in H^1(\Omega)$ , the inequality for  $y$  in (23b) follows from an approximation result [54, Theorem 2.6], a trace inequality [4, Theorem 1.6.6], and the stability of the  $L^2$  projection in the  $H^1$  norm [22, Lemma 1.131]. The same proof works for the inequality for  $\mathbf{q}$  in (23b) if  $\mathbf{q}$  is also in  $H^1(\Omega)$ . However, if  $q \in H^{r_q}$  for some  $r_q \in (1/2, 1)$ , then the inequality follows from an approximation result [27, Lemma A.5], a trace inequality [23, Lemma 7.2], and the stability of the  $L^2$  projection in the  $H^{r_q}$  norm (which can be shown using an interpolation argument). We have the same projection error bounds for  $\mathbf{p}$ ,  $z$ , and other variables.

To shorten lengthy equations, for any  $(\mathbf{q}_h, y_h, \widehat{y}_h^o; \mathbf{r}_1, w_1, \mu_1) \in [V_h \times W_h \times M_h(o)]^2$ , we define the HDG operator  $\mathcal{B}$  as follows:

$$\begin{aligned} \mathcal{B}(\mathbf{q}_h, y_h, \widehat{y}_h^o; \mathbf{r}_1, w_1, \mu_1) &= (\mathbf{q}_h, \mathbf{r}_1)_{\mathcal{T}_h} - (y_h, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle \widehat{y}_h^o, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad - (\mathbf{q}_h, \nabla w_1)_{\mathcal{T}_h} + \langle \mathbf{q}_h \cdot \mathbf{n} + h^{-1} P_M y_h, w_1 \rangle_{\partial \mathcal{T}_h} \\ &\quad - \langle h^{-1} \widehat{y}_h^o, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \mathbf{q}_h \cdot \mathbf{n} + h^{-1} (P_M y_h - \widehat{y}_h^o), \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}. \end{aligned} \quad (24)$$

By the definition of  $\mathcal{B}$ , we can rewrite the HDG formulation of the optimality system (13) as follows: find  $(\mathbf{q}_h, \mathbf{p}_h, y_h, z_h, \widehat{y}_h^o, \widehat{z}_h^o, u_h) \in \mathbf{V}_h \times \mathbf{V}_h \times W_h \times W_h \times M_h(o) \times M_h(o) \times M_h(\partial)$  such that

$$\mathcal{B}(\mathbf{q}_h, y_h, \widehat{y}_h^o; \mathbf{r}_1, w_1, \mu_1) = -\langle u_h, \mathbf{r}_1 \cdot \mathbf{n} - h^{-1}w_1 \rangle_{\varepsilon_h^\partial} + (f, w_1)_{\mathcal{T}_h}, \quad (25a)$$

$$\mathcal{B}(\mathbf{p}_h, z_h, \widehat{z}_h^o; \mathbf{r}_2, w_2, \mu_2) = (y_h - y_d, w_2)_{\mathcal{T}_h}, \quad (25b)$$

$$\gamma^{-1} \langle \mathbf{p}_h \cdot \mathbf{n} + h^{-1} P_M z_h, \mu_3 \rangle_{\varepsilon_h^\partial} = -\langle u_h, \mu_3 \rangle_{\varepsilon_h^\partial} \quad (25c)$$

for all  $(\mathbf{r}_1, \mathbf{r}_2, w_1, w_2, \mu_1, \mu_2, \mu_3) \in \mathbf{V}_h \times \mathbf{V}_h \times W_h \times W_h \times M_h(o) \times M_h(o) \times M_h(\partial)$ .

Next, we present a basic property of the operator  $\mathcal{B}$  and show the HDG Eq. (25) has a unique solution.

**Lemma 4.3** *For any  $(\mathbf{v}_h, w_h, \mu_h) \in \mathbf{V}_h \times W_h \times M_h(o)$ , we have*

$$\begin{aligned} \mathcal{B}(\mathbf{v}_h, w_h, \mu_h; \mathbf{v}_h, w_h, \mu_h) &= (\mathbf{v}_h, \mathbf{v}_h)_{\mathcal{T}_h} + \langle h^{-1}(P_M w_h - \mu_h), P_M w_h - \mu_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle h^{-1} P_M w_h, P_M w_h \rangle_{\varepsilon_h^\partial}. \end{aligned}$$

**Proof** By the definition of  $\mathcal{B}$  in (24) and integration by parts, we have

$$\begin{aligned} &\mathcal{B}(\mathbf{v}_h, w_h, \mu_h; \mathbf{v}_h, w_h, \mu_h) \\ &= (\mathbf{v}_h, \mathbf{v}_h)_{\mathcal{T}_h} - (w_h, \nabla \cdot \mathbf{v}_h)_{\mathcal{T}_h} + \langle \mu_h, \mathbf{v}_h \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - (\mathbf{v}_h, \nabla w_h)_{\mathcal{T}_h} \\ &\quad + \langle \mathbf{v}_h \cdot \mathbf{n} + h^{-1} P_M w_h, w_h \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} \mu_h, w_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad - \langle \mathbf{v}_h \cdot \mathbf{n} + h^{-1} (P_M w_h - \mu_h), \mu_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= (\mathbf{v}_h, \mathbf{v}_h)_{\mathcal{T}_h} + \langle h^{-1} P_M w_h, w_h \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} \mu_h, w_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad - \langle h^{-1} (P_M w_h - \mu_h), \mu_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= (\mathbf{v}_h, \mathbf{v}_h)_{\mathcal{T}_h} + \langle h^{-1} (P_M w_h - \mu_h), P_M w_h - \mu_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M w_h, P_M w_h \rangle_{\varepsilon_h^\partial}. \end{aligned}$$

□

**Proposition 4.4** *There exists a unique solution of the HDG Eq. (25).*

**Proof** Since the system (25) is finite dimensional, we only need to prove the uniqueness. Therefore, we assume  $y_d = f = 0$  and we show the system (25) only has the trivial solution.

First, by the definition of  $\mathcal{B}$ , we have

$$\begin{aligned} &\mathcal{B}(\mathbf{q}_h, y_h, \widehat{y}_h^o; \mathbf{p}_h, -z_h, -\widehat{z}_h^o) + \mathcal{B}(\mathbf{p}_h, z_h, \widehat{z}_h^o; -\mathbf{q}_h, y_h, \widehat{y}_h^o) \\ &= (\mathbf{q}_h, \mathbf{p}_h)_{\mathcal{T}_h} - (y_h, \nabla \cdot \mathbf{p}_h)_{\mathcal{T}_h} + \langle \widehat{y}_h^o, \mathbf{p}_h \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + (\mathbf{q}_h, \nabla z_h)_{\mathcal{T}_h} \\ &\quad - \langle \mathbf{q}_h \cdot \mathbf{n} + h^{-1} P_M y_h, z_h \rangle_{\partial \mathcal{T}_h} + \langle h^{-1} \widehat{y}_h^o, z_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \mathbf{q}_h \cdot \mathbf{n} + h^{-1} (P_M y_h - \widehat{y}_h^o), \widehat{z}_h^o \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - (\mathbf{p}_h, \mathbf{q}_h)_{\mathcal{T}_h} + (z_h, \nabla \cdot \mathbf{q}_h)_{\mathcal{T}_h} \end{aligned}$$

$$\begin{aligned}
& - \langle \widehat{z}_h^o, \mathbf{q}_h \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - (\mathbf{p}_h, \nabla y_h)_{\mathcal{T}_h} + \langle \mathbf{p}_h \cdot \mathbf{n} + h^{-1} P_M z_h, y_h \rangle_{\partial \mathcal{T}_h} \\
& - \langle h^{-1} \widehat{z}_h^o, y_h \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \mathbf{p}_h \cdot \mathbf{n} + h^{-1} (P_M z_h - \widehat{z}_h^o), \widehat{y}_h^o \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}.
\end{aligned}$$

Integrating by parts and using the properties of  $P_M$  in (22) gives

$$\mathcal{B}(\mathbf{q}_h, y_h, \widehat{y}_h^o; \mathbf{p}_h, -z_h, -\widehat{z}_h^o) + \mathcal{B}(\mathbf{p}_h, z_h, \widehat{z}_h^o; -\mathbf{q}_h, y_h, \widehat{y}_h^o) = 0.$$

Next, take  $(\mathbf{r}_1, w_1, \mu_1) = (\mathbf{p}_h, -z_h, -\widehat{z}_h^o)$ ,  $(\mathbf{r}_2, w_2, \mu_2) = (-\mathbf{q}_h, y_h, \widehat{y}_h^o)$ , and  $\mu_3 = -\gamma u_h$  in the HDG Eqs. (25a)–(25c), respectively, and sum to obtain

$$(y_h, y_h)_{\mathcal{T}_h} + \gamma \|u_h\|_{\varepsilon_h^\partial}^2 = 0.$$

This implies  $y_h = 0$  and  $u_h = 0$  since  $\gamma > 0$ .

Next, taking  $(\mathbf{r}_1, w_1, \mu_1) = (\mathbf{q}_h, y_h, \widehat{y}_h^o)$  and  $(\mathbf{r}_2, w_2, \mu_2) = (\mathbf{p}_h, z_h, \widehat{z}_h^o)$  in Lemma 4.3 gives  $\mathbf{q}_h = \mathbf{p}_h = \mathbf{0}$ ,  $\widehat{y}_h^o = 0$ ,  $P_M z_h = 0$  on  $\varepsilon_h^\partial$  and

$$P_M z_h - \widehat{z}_h^o = 0 \quad \text{on } \partial \mathcal{T}_h \setminus \varepsilon_h^\partial. \quad (26)$$

Substituting (26) into (13c), and remembering again  $P_M z_h = 0$  on  $\varepsilon_h^\partial$ , we get

$$-(z_h, \nabla \cdot \mathbf{r}_2)_{\mathcal{T}_h} + \langle P_M z_h, \mathbf{r}_2 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h} = 0.$$

Use the property of  $P_M$  in (22), integrate by parts, and take  $\mathbf{r}_2 = \nabla z_h$  to obtain

$$(\nabla z_h, \nabla z_h)_{\mathcal{T}_h} = 0.$$

Thus,  $z_h$  is constant on each  $K \in \mathcal{T}_h$ . Therefore,  $P_M z_h$  is constant on  $\partial K$  for each  $K \in \mathcal{T}_h$ . Since  $P_M z_h = 0$  on  $\varepsilon_h^\partial$  and  $P_M z_h = \widehat{z}_h^o = 0$  on  $\partial \mathcal{T}_h \setminus \varepsilon_h^\partial$ , this gives  $z_h = 0$  on  $\partial \mathcal{T}_h$ . Therefore  $z_h = 0$  since  $z_h$  is constant on each  $K \in \mathcal{T}_h$ .

#### 4.3 Proof of main result

To prove the main result, we follow a similar strategy taken by Gong and Yan [29], see also [12, 40, 43], and introduce an auxiliary problem with the approximate control  $u_h$  in (25a) replaced by a projection of the exact optimal control. We first bound the error between the solutions of the auxiliary problem and the mixed weak form (4a)–(4e) of the optimality system. Then we bound the error between the solutions of the auxiliary problem and the HDG problem (25). A simple application of the triangle inequality then gives a bound on the error between the solutions of the HDG problem and the mixed form of the optimality system.

The precise form of the auxiliary problem is given as follows: find  $(\mathbf{q}_h(u), \mathbf{p}_h(u), y_h(u), z_h(u), \widehat{y}_h^o(u), \widehat{z}_h^o(u)) \in \mathbf{V}_h \times \mathbf{V}_h \times W_h \times W_h \times M_h(o) \times M_h(o)$  such that

$$\mathcal{B}(\mathbf{q}_h(u), y_h(u), \widehat{y}_h^o(u); \mathbf{r}_1, w_1, \mu_1) = -\langle P_M u, \mathbf{r}_1 \cdot \mathbf{n} - h^{-1} w_1 \rangle_{\varepsilon_h^\partial} + (f, w_1)_{\mathcal{T}_h}, \quad (27a)$$

$$\mathcal{B}(\mathbf{p}_h(u), z_h(u), \widehat{z}_h^o(u); \mathbf{r}_2, w_2, \mu_2) = (y_h(u) - y_d, w_2)_{\mathcal{T}_h} \quad (27b)$$

for all  $(\mathbf{r}_1, \mathbf{r}_2, w_1, w_2, \mu_1, \mu_2) \in \mathbf{V}_h \times \mathbf{V}_h \times W_h \times W_h \times M_h(o) \times M_h(o)$ .

We split the proof of the main result, Theorem 4.1, in 7 steps. We begin by bounding the error between the solutions of the auxiliary problem and the mixed form (4a)–(4e) of the optimality system. We split the errors in the variables using the  $L^2$  projections. In steps 1–3, we focus on the primary variables, i.e., the state  $y$  and the flux  $\mathbf{q}$ , and we use the following notation:

$$\begin{aligned} \delta^{\mathbf{q}} &= \mathbf{q} - \Pi \mathbf{q}, & \varepsilon_h^{\mathbf{q}} &= \Pi \mathbf{q} - \mathbf{q}_h(u), \\ \delta^y &= y - \Pi y, & \varepsilon_h^y &= \Pi y - y_h(u), \\ \delta^{\widehat{y}} &= y - P_M y, & \varepsilon_h^{\widehat{y}} &= P_M y - \widehat{y}_h^o(u), \\ \widehat{\delta}_1 &= \delta^{\mathbf{q}} \cdot \mathbf{n} + h^{-1} P_M \delta^y, & \widehat{\varepsilon}_1 &= \varepsilon_h^{\mathbf{q}} \cdot \mathbf{n} + h^{-1} (P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}). \end{aligned} \quad (28)$$

#### 4.3.1 Step 1: The error equation for part 1 of the auxiliary problem (27a)

**Lemma 4.5** *For any  $(\mathbf{r}_1, w_1, \mu_1) \in \mathbf{V}_h \times W_h \times M_h(o)$ , we have*

$$\mathcal{B}(\varepsilon_h^{\mathbf{q}}, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; \mathbf{r}_1, w_1, \mu_1) = -\langle \widehat{\delta}_1, w_1 \rangle_{\partial \mathcal{T}_h} + \langle \widehat{\delta}_1, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}}. \quad (29)$$

**Proof** By the definition of the operator  $\mathcal{B}$  in (24), we have

$$\begin{aligned} &\mathcal{B}(\Pi \mathbf{q}, \Pi y, P_M y; \mathbf{r}_1, w_1, \mu_1) \\ &= (\Pi \mathbf{q}, \mathbf{r}_1)_{\mathcal{T}_h} - (\Pi y, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle P_M y, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} \\ &\quad - (\Pi \mathbf{q}, \nabla w_1)_{\mathcal{T}_h} + \langle \Pi \mathbf{q} \cdot \mathbf{n} + h^{-1} P_M \Pi y, w_1 \rangle_{\partial \mathcal{T}_h} \\ &\quad - \langle h^{-1} P_M y, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} - \langle \Pi \mathbf{q} \cdot \mathbf{n} - h^{-1} P_M \delta^y, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}}. \end{aligned}$$

By properties of the  $L^2$  projections (22), we have

$$\begin{aligned} \mathcal{B}(\Pi \mathbf{q}, \Pi y, P_M y; \mathbf{r}_1, w_1, \mu_1) &= (\mathbf{q}, \mathbf{r}_1)_{\mathcal{T}_h} - (y, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle y, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} \\ &\quad - (\mathbf{q}, \nabla w_1)_{\mathcal{T}_h} + \langle \mathbf{q} \cdot \mathbf{n}, w_1 \rangle_{\partial \mathcal{T}_h} - \langle \delta^{\mathbf{q}} \cdot \mathbf{n}, w_1 \rangle_{\partial \mathcal{T}_h} \\ &\quad + \langle h^{-1} P_M \Pi y, w_1 \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} P_M y, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} \\ &\quad - \langle \mathbf{q} \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} + \langle \delta^{\mathbf{q}} \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} \\ &\quad + \langle h^{-1} P_M \delta^y, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}}. \end{aligned}$$

Note that the exact state  $y$  and exact flux  $\mathbf{q}$  satisfy

$$\begin{aligned} (\mathbf{q}, \mathbf{r}_1)_{\mathcal{T}_h} - (y, \nabla \cdot \mathbf{r}_1)_{\mathcal{T}_h} + \langle y, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} &= - \langle u, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^{\mathbf{q}}}, \\ -(\mathbf{q}, \nabla w_1)_{\mathcal{T}_h} + \langle \mathbf{q} \cdot \mathbf{n}, w_1 \rangle_{\partial \mathcal{T}_h} &= (f, w_1)_{\mathcal{T}_h}, \\ \langle \mathbf{q} \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\mathbf{q}}} &= 0 \end{aligned}$$

for all  $(\mathbf{r}_1, w_1, \mu_1) \in \mathbf{V}_h \times W_h \times M_h(o)$  and  $(\mathbf{r}_1, w_1, \mu_1) \in \mathbf{V}_h \times W_h \times M_h$ . Here, the last equation holds since  $\mathbf{q} \in H(\text{div}, \Omega)$  and  $\mu_1$  is single-valued on each interior face. Then we have

$$\begin{aligned}\mathcal{B}(\Pi \mathbf{q}, \Pi y, P_M y; \mathbf{r}_1, w_1, \mu_1) &= -\langle u, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} + (f, w_1)_{\mathcal{T}_h} - \langle \delta^{\mathbf{q}} \cdot \mathbf{n}, w_1 \rangle_{\partial \mathcal{T}_h} \\ &\quad + \langle h^{-1} P_M \Pi y, w_1 \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} P_M y, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \delta^{\mathbf{q}} \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \delta^y, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}.\end{aligned}$$

Subtract part 1 of the auxiliary problem (27a) from the above equality, use  $y = u$  on  $\varepsilon_h^\partial$ , and use  $\langle P_M u, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} = \langle u, \mathbf{r}_1 \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial}$  to obtain the result:

$$\begin{aligned}\mathcal{B}(\varepsilon_h^{\mathbf{q}}, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; \mathbf{r}_1, w_1, \mu_1) &= -\langle P_M u, h^{-1} w_1 \rangle_{\varepsilon_h^\partial} - \langle \delta^{\mathbf{q}} \cdot \mathbf{n}, w_1 \rangle_{\partial \mathcal{T}_h} \\ &\quad + \langle h^{-1} P_M \Pi y, w_1 \rangle_{\partial \mathcal{T}_h} - \langle h^{-1} P_M y, w_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \delta^{\mathbf{q}} \cdot \mathbf{n}, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \delta^y, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= -\langle \widehat{\delta}_1, w_1 \rangle_{\partial \mathcal{T}_h} + \langle \widehat{\delta}_1, \mu_1 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}.\end{aligned}$$

□

### 4.3.2 Step 2: Estimate for $\varepsilon_h^{\mathbf{q}}$

We first provide a key inequality which was proven in Lemma 3.2 in [51]. In order to obtain the estimate for  $\varepsilon_h^{\mathbf{q}}$ , we first provide a key inequality which was proven in [51].

**Lemma 4.6** *We have*

$$\begin{aligned}\|\nabla \varepsilon_h^y\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\varepsilon_h^y\|_{\varepsilon_h^\partial} \\ \lesssim \|\varepsilon_h^{\mathbf{q}}\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial}.\end{aligned}$$

**Lemma 4.7** *We have*

$$\begin{aligned}\|\varepsilon_h^{\mathbf{q}}\|_{\mathcal{T}_h}^2 + h^{-1} \|P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial}^2 \\ \lesssim h^{2s_q} \|\mathbf{q}\|_{s_q, \Omega}^2 + h^{2s_y - 2} \|y\|_{s_y, \Omega}^2.\end{aligned}\tag{30}$$

**Proof** First, the basic property of  $\mathcal{B}$  in Lemma 4.3 gives

$$\mathcal{B}(\varepsilon_h^{\mathbf{q}}, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; \varepsilon_h^{\mathbf{q}}, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}) = (\varepsilon_h^{\mathbf{q}}, \varepsilon_h^{\mathbf{q}})_{\mathcal{T}_h} + h^{-1} \|P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial}^2.$$

Then, taking  $(\mathbf{r}_1, w_1, \mu_1) = (\varepsilon_h^{\mathbf{q}}, \varepsilon_h^y, \varepsilon_h^{\widehat{y}})$  in (29) in Lemma 4.5 gives

$$\begin{aligned}(\varepsilon_h^{\mathbf{q}}, \varepsilon_h^{\mathbf{q}})_{\mathcal{T}_h} + h^{-1} \|P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial}^2 \\ = -\langle \widehat{\delta}_1, \varepsilon_h^y - \varepsilon_h^{\widehat{y}} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \widehat{\delta}_1, \varepsilon_h^y \rangle_{\varepsilon_h^\partial}\end{aligned}$$

$$\begin{aligned}
&= -\langle \delta^q \cdot \mathbf{n}, \varepsilon_h^y - \widehat{\varepsilon}_h^y \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - h^{-1} \langle \delta^y, P_M \varepsilon_h^y - \widehat{\varepsilon}_h^y \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\
&\quad - \langle \delta^q \cdot \mathbf{n}, \varepsilon_h^y \rangle_{\varepsilon_h^\partial} - h^{-1} \langle \delta^y, P_M \varepsilon_h^y \rangle_{\varepsilon_h^\partial} \\
&\leq \|\delta^q\|_{\partial \mathcal{T}_h} \|\varepsilon_h^y - \widehat{\varepsilon}_h^y\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-1} \|\delta^y\|_{\partial \mathcal{T}_h} \|P_M \varepsilon_h^y - \widehat{\varepsilon}_h^y\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\
&\quad + \|\delta^q\|_{\partial \mathcal{T}_h} \|\varepsilon_h^y\|_{\varepsilon_h^\partial} + h^{-1} \|\delta^y\|_{\partial \mathcal{T}_h} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial} \\
&\leq h^{1/2} \|\delta^q\|_{\partial \mathcal{T}_h} h^{-1/2} \|\varepsilon_h^y - \widehat{\varepsilon}_h^y\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\
&\quad + h^{-1/2} \|\delta^y\|_{\partial \mathcal{T}_h} h^{-1/2} \|P_M \varepsilon_h^y - \widehat{\varepsilon}_h^y\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\
&\quad + h^{1/2} \|\delta^q\|_{\partial \mathcal{T}_h} h^{-1/2} \|\varepsilon_h^y\|_{\varepsilon_h^\partial} + h^{-1/2} \|\delta^y\|_{\partial \mathcal{T}_h} h^{-1/2} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial}.
\end{aligned}$$

By Young's inequality, Lemma 4.6, and the approximation properties of the  $L^2$  projections in (23) we obtain

$$\begin{aligned}
&\|\varepsilon_h^q\|_{\mathcal{T}_h}^2 + h^{-1} \|P_M \varepsilon_h^y - \widehat{\varepsilon}_h^y\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial}^2 \\
&\lesssim h \|\delta^q\|_{\partial \mathcal{T}_h}^2 + h^{-1} \|\delta^y\|_{\partial \mathcal{T}_h}^2 \\
&\lesssim h^{2s_q} \|\mathbf{q}\|_{s_q, \Omega}^2 + h^{2s_y-2} \|y\|_{s_y, \Omega}^2.
\end{aligned}$$

□

#### 4.3.3 Step 3: Estimate for $\varepsilon_h^y$ by a duality argument

Next, we introduce the dual problem for any given  $\Theta$  in  $L^2(\Omega)$ :

$$\begin{aligned}
\Phi + \nabla \Psi &= 0 && \text{in } \Omega, \\
\nabla \cdot \Phi &= \Theta && \text{in } \Omega, \\
\Psi &= 0 && \text{on } \Gamma.
\end{aligned} \tag{31}$$

Since the domain  $\Omega$  is convex, we have the following regularity estimate

$$\|\Phi\|_{H^1(\Omega)} + \|\Psi\|_{H^2(\Omega)} \lesssim \|\Theta\|_{\mathcal{T}_h}. \tag{32}$$

Before we estimate  $\varepsilon_h^y$  we introduce the following notation, which is similar to the earlier notation in (28):

$$\delta^\Phi = \Phi - \Pi \Phi, \quad \delta^\Psi = \Psi - \Pi \Psi, \quad \delta^{\widehat{\Psi}} = \Psi - P_M \Psi. \tag{33}$$

By the regularity estimate (32) and the approximation properties of  $L^2$  projections, we have the following bounds:

$$\|\delta^\Phi\|_{\partial \mathcal{T}_h} \lesssim h^{\frac{1}{2}} \|\Theta\|_{\mathcal{T}_h}, \quad \|\delta^\Psi\|_{\partial \mathcal{T}_h} \lesssim h^{\frac{3}{2}} \|\Theta\|_{\mathcal{T}_h}. \tag{34}$$

**Lemma 4.8** *We have*

$$\|\varepsilon_h^y\|_{\mathcal{T}_h} \lesssim h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}. \tag{35}$$

**Proof** Consider the dual problem (31) and let  $\Theta = \varepsilon_h^y$ . In the definition (24) of  $\mathcal{B}$ , take  $(\mathbf{r}_1, w_1, \mu_1)$  to be  $(-\boldsymbol{\Pi}\Phi, \Pi\Psi, P_M\Psi)$  and use  $\Psi = 0$  on  $\varepsilon_h^\partial$  to obtain

$$\begin{aligned} & \mathcal{B}(\varepsilon_h^q, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; -\boldsymbol{\Pi}\Phi, \Pi\Psi, P_M\Psi) \\ &= -(\varepsilon_h^q, \boldsymbol{\Pi}\Phi)_{\mathcal{T}_h} + (\varepsilon_h^y, \nabla \cdot \boldsymbol{\Pi}\Phi)_{\mathcal{T}_h} - \langle \varepsilon_h^{\widehat{y}}, \boldsymbol{\Pi}\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad - (\varepsilon_h^q, \nabla \Pi\Psi)_{\mathcal{T}_h} + \langle \widehat{\boldsymbol{\varepsilon}}_1, \Pi\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \widehat{\boldsymbol{\varepsilon}}_1, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \varepsilon_h^q \cdot \mathbf{n} + h^{-1}P_M\varepsilon_h^y, \Pi\Psi \rangle_{\varepsilon_h^\partial}. \end{aligned} \quad (36)$$

Next, integrating by parts, using properties of  $L^2$  projections, and using  $\nabla \cdot \boldsymbol{\Phi} = \varepsilon_h^y$  gives

$$\begin{aligned} (\varepsilon_h^y, \nabla \cdot \boldsymbol{\Pi}\Phi)_{\mathcal{T}_h} &= \langle \varepsilon_h^y, \boldsymbol{\Pi}\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h} - (\nabla \varepsilon_h^y, \boldsymbol{\Pi}\Phi)_{\mathcal{T}_h} \\ &= \langle \varepsilon_h^y, \boldsymbol{\Pi}\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h} - (\nabla \varepsilon_h^y, \boldsymbol{\Phi})_{\mathcal{T}_h} \\ &= -\langle \varepsilon_h^y, \delta^\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h} + (\varepsilon_h^y, \nabla \cdot \boldsymbol{\Phi})_{\mathcal{T}_h} \\ &= -\langle \varepsilon_h^y, \delta^\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h} + \|\varepsilon_h^y\|_{\mathcal{T}_h}^2. \end{aligned}$$

Similarly, using  $\boldsymbol{\Phi} + \nabla\Psi = \mathbf{0}$  gives

$$\begin{aligned} -(\varepsilon_h^q, \nabla \Pi\Psi)_{\mathcal{T}_h} &= -\langle \varepsilon_h^q \cdot \mathbf{n}, \Pi\Psi \rangle_{\partial\mathcal{T}_h} + (\nabla \cdot \varepsilon_h^q, \Pi\Psi)_{\mathcal{T}_h} \\ &= -\langle \varepsilon_h^q \cdot \mathbf{n}, \Pi\Psi \rangle_{\partial\mathcal{T}_h} + (\nabla \cdot \varepsilon_h^q, \Psi)_{\mathcal{T}_h} \\ &= -\langle \varepsilon_h^q \cdot \mathbf{n}, \Pi\Psi \rangle_{\partial\mathcal{T}_h} + \langle \varepsilon_h^q \cdot \mathbf{n}, \Psi \rangle_{\partial\mathcal{T}_h} - \langle \varepsilon_h^q, \nabla\Psi \rangle_{\mathcal{T}_h} \\ &= \langle \varepsilon_h^q \cdot \mathbf{n}, (P_M\Psi - \Pi\Psi) \rangle_{\partial\mathcal{T}_h} + \langle \varepsilon_h^q, \boldsymbol{\Phi} \rangle_{\mathcal{T}_h}. \end{aligned}$$

Then Eq. (36) becomes

$$\begin{aligned} & \mathcal{B}(\varepsilon_h^q, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; -\boldsymbol{\Pi}\Phi, \Pi\Psi, P_M\Psi) \\ &= -(\varepsilon_h^q, \boldsymbol{\Phi})_{\mathcal{T}_h} - \langle \varepsilon_h^y, \delta^\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h} + \|\varepsilon_h^y\|_{\mathcal{T}_h}^2 - \langle \varepsilon_h^{\widehat{y}}, \boldsymbol{\Pi}\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \varepsilon_h^q \cdot \mathbf{n}, P_M\Psi - \Pi\Psi \rangle_{\partial\mathcal{T}_h} + \langle \varepsilon_h^q, \boldsymbol{\Phi} \rangle_{\mathcal{T}_h} + \langle \widehat{\boldsymbol{\varepsilon}}_1, \Pi\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \widehat{\boldsymbol{\varepsilon}}_1, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle h^{-1}P_M\varepsilon_h^y, \Pi\Psi \rangle_{\varepsilon_h^\partial}. \end{aligned}$$

Since  $\boldsymbol{\Phi}$  is single-valued on element faces, we have  $\langle \varepsilon_h^{\widehat{y}}, \boldsymbol{\Phi} \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} = 0$ . This gives  $-\langle \varepsilon_h^{\widehat{y}}, \boldsymbol{\Pi}\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} = \langle \varepsilon_h^{\widehat{y}}, \delta^\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial}$ . Also, since  $\langle \varepsilon_h^q \cdot \mathbf{n}, P_M\Psi \rangle_{\partial\mathcal{T}_h} = \langle \varepsilon_h^q \cdot \mathbf{n}, \Psi \rangle_{\partial\mathcal{T}_h}$  and  $\langle \widehat{\boldsymbol{\varepsilon}}_1, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} = \langle \widehat{\boldsymbol{\varepsilon}}_1, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial}$ , we have

$$\begin{aligned} & \mathcal{B}(\varepsilon_h^q, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; -\boldsymbol{\Pi}\Phi, \Pi\Psi, P_M\Psi) \\ &= -\langle \varepsilon_h^y - \varepsilon_h^{\widehat{y}}, \delta^\Phi \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} + \|\varepsilon_h^y\|_{\mathcal{T}_h}^2 - h^{-1} \langle P_M\varepsilon_h^y - \varepsilon_h^{\widehat{y}}, \delta^\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad - \langle \varepsilon_h^y, \delta^\Phi \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} - h^{-1} \langle P_M\varepsilon_h^y, \delta^\Psi \rangle_{\varepsilon_h^\partial}. \end{aligned}$$

On the other hand, Eq. (29) in Lemma 4.5 gives

$$\mathcal{B}(\varepsilon_h^q, \varepsilon_h^y, \varepsilon_h^{\widehat{y}}; -\boldsymbol{\Pi}\boldsymbol{\Phi}, \Pi\Psi, P_M\Psi) = -\langle \widehat{\delta}_1, \Pi\Psi \rangle_{\partial\mathcal{T}_h} + \langle \widehat{\delta}_1, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial}.$$

Moreover,

$$\begin{aligned} & \langle \widehat{\delta}_1, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= \langle \delta^q \cdot \mathbf{n} + h^{-1} P_M \delta^y, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= \langle \boldsymbol{q} \cdot \mathbf{n}, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \boldsymbol{\Pi}\boldsymbol{q} \cdot \mathbf{n}, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \delta^y, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= -\langle \boldsymbol{\Pi}\boldsymbol{q} \cdot \mathbf{n}, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \delta^y, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= \langle \boldsymbol{q} \cdot \mathbf{n}, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \boldsymbol{\Pi}\boldsymbol{q} \cdot \mathbf{n}, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \delta^y, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= \langle \widehat{\delta}_1, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &= \langle \widehat{\delta}_1, \Psi \rangle_{\partial\mathcal{T}_h}, \end{aligned}$$

where we used  $\langle \boldsymbol{q} \cdot \mathbf{n}, P_M\Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} = 0$  and  $\langle \boldsymbol{q} \cdot \mathbf{n}, \Psi \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} = 0$ , which hold since  $\boldsymbol{q} \in H(\text{div}, \Omega)$ , and we also used  $\Psi = 0$  on  $\varepsilon_h^\partial$ .

Comparing the above two equalities gives

$$\begin{aligned} \|\varepsilon_h^y\|_{\mathcal{T}_h}^2 &= \langle \varepsilon_h^y - \varepsilon_h^{\widehat{y}}, \delta^{\boldsymbol{\Phi}} \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-1} \langle P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}, \delta^{\Psi} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \varepsilon_h^y, \delta^{\boldsymbol{\Phi}} \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} + h^{-1} \langle P_M \varepsilon_h^y, \delta^{\Psi} \rangle_{\varepsilon_h^\partial} + \langle \widehat{\delta}_1, \delta^{\Psi} \rangle_{\partial\mathcal{T}_h} \\ &= \langle \varepsilon_h^y - \varepsilon_h^{\widehat{y}}, \delta^{\boldsymbol{\Phi}} \cdot \mathbf{n} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-1} \langle P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}, \delta^{\Psi} \rangle_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \langle \varepsilon_h^y, \delta^{\boldsymbol{\Phi}} \cdot \mathbf{n} \rangle_{\varepsilon_h^\partial} + h^{-1} \langle P_M \varepsilon_h^y, \delta^{\Psi} \rangle_{\varepsilon_h^\partial} + \langle \delta^q \cdot \mathbf{n} + h^{-1} P_M \delta^y, \delta^{\Psi} \rangle_{\partial\mathcal{T}_h} \\ &\lesssim h^{-\frac{1}{2}} \|\varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \cdot h^{\frac{1}{2}} \|\delta^{\boldsymbol{\Phi}}\|_{\partial\mathcal{T}_h} \\ &\quad + h^{-\frac{1}{2}} \|P_M \varepsilon_h^y - \varepsilon_h^{\widehat{y}}\|_{\partial\mathcal{T}_h \setminus \varepsilon_h^\partial} \cdot h^{-\frac{1}{2}} \|\delta^{\Psi}\|_{\partial\mathcal{T}_h} \\ &\quad + h^{-\frac{1}{2}} \|\varepsilon_h^y\|_{\varepsilon_h^\partial} \cdot h^{\frac{1}{2}} \|\delta^{\boldsymbol{\Phi}}\|_{\partial\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^y\|_{\varepsilon_h^\partial} \cdot h^{-\frac{1}{2}} \|\delta^{\Psi}\|_{\partial\mathcal{T}_h} \\ &\quad + \|\delta^q\|_{\partial\mathcal{T}_h} \cdot \|\delta^{\Psi}\|_{\partial\mathcal{T}_h} + h^{-1} \|\delta^y\|_{\partial\mathcal{T}_h} \cdot \|\delta^{\Psi}\|_{\partial\mathcal{T}_h} \\ &\lesssim (h^{s_q+1} \|\boldsymbol{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}) \|\varepsilon_h^y\|_{\mathcal{T}_h}, \end{aligned}$$

where we used Lemmas 4.6, 4.7, and the bounds (34) to obtain the final inequality.  $\square$

As a consequence of Lemmas 4.7, 4.8, a simple application of the triangle inequality gives optimal convergence rates for  $\|\boldsymbol{q} - \boldsymbol{q}_h(u)\|_{\mathcal{T}_h}$  and  $\|y - y_h(u)\|_{\mathcal{T}_h}$ :

#### Lemma 4.9

$$\|\boldsymbol{q} - \boldsymbol{q}_h(u)\|_{\mathcal{T}_h} \leq \|\delta^q\|_{\mathcal{T}_h} + \|\varepsilon_h^q\|_{\mathcal{T}_h} \lesssim h^{s_q} \|\boldsymbol{q}\|_{s_q, \Omega} + h^{s_y-1} \|y\|_{s_y, \Omega}, \quad (37a)$$

$$\|y - y_h(u)\|_{\mathcal{T}_h} \leq \|\delta^y\|_{\mathcal{T}_h} + \|\varepsilon_h^y\|_{\mathcal{T}_h} \lesssim h^{s_q+1} \|\boldsymbol{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}. \quad (37b)$$

#### 4.3.4 Step 4: The error equation for part 2 of the auxiliary problem (27b)

We continue to bound the error between the solutions of the auxiliary problem and the mixed form (4a)–(4e) of the optimality system. In steps 4–5, we focus on the dual variables, i.e., the state  $z$  and the flux  $\mathbf{p}$ . We split the errors in the variables using the  $L^2$  projections, and we use the following notation.

$$\begin{aligned}\delta^{\mathbf{p}} &= \mathbf{p} - \Pi \mathbf{p}, & \varepsilon_h^{\mathbf{p}} &= \Pi \mathbf{p} - \mathbf{p}_h(u), \\ \delta^z &= z - \Pi z, & \varepsilon_h^z &= \Pi z - z_h(u), \\ \delta^{\hat{z}} &= z - P_M z, & \varepsilon_h^{\hat{z}} &= P_M z - \hat{z}_h^o(u), \\ \hat{\delta}_2 &= \delta^{\mathbf{p}} \cdot \mathbf{n} + h^{-1} P_M \delta^z.\end{aligned}\quad (38)$$

The derivation of the error equation for part 2 of the auxiliary problem (27b) is similar to the analysis for part 1 of the auxiliary problem in step 1 in Sect. 4.3.1; the only difference is there is one more term  $(y - y_h(u), w_2)_{\mathcal{T}_h}$  in the right hand side. Therefore, we state the result and omit the proof.

**Lemma 4.10** *For any  $(\mathbf{r}_2, w_2, \mu_2) \in V_h \times W_h \times M_h(o)$ , we have*

$$\begin{aligned}\mathcal{B}(\varepsilon_h^{\mathbf{p}}, \varepsilon_h^z, \varepsilon_h^{\hat{z}}, \mathbf{r}_2, w_2, \mu_2) \\ = -\langle \hat{\delta}_2, w_2 \rangle_{\partial \mathcal{T}_h} + \langle \hat{\delta}_2, \mu_2 \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\partial}} + (y - y_h(u), w_2)_{\mathcal{T}_h}.\end{aligned}\quad (39)$$

#### 4.3.5 Step 5: Estimate for $\varepsilon_h^{\mathbf{p}}$ and $\varepsilon_h^z$

Before we estimate  $\varepsilon_h^{\mathbf{p}}$ , we give the following discrete Poincaré inequality from [51, p. 354].

**Lemma 4.11** *We have*

$$\|\varepsilon_h^z\|_{\mathcal{T}_h} \lesssim \|\nabla \varepsilon_h^z\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\varepsilon_h^z - \varepsilon_h^{\hat{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\partial}} + h^{-\frac{1}{2}} \|\varepsilon_h^z\|_{\varepsilon_h^{\partial}}. \quad (40)$$

**Lemma 4.12** *We have*

$$\begin{aligned}\|\varepsilon_h^{\mathbf{p}}\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\hat{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\partial}} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^{\partial}} \\ \lesssim h^{s_p} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-1} \|z\|_{s_z, \Omega} + h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}, \\ \|\varepsilon_h^z\|_{\mathcal{T}_h} \lesssim h^{s_p} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-1} \|z\|_{s_z, \Omega} + h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}.\end{aligned}$$

**Proof** First, we note the key inequality in Lemma 4.6 is valid with  $(z, \mathbf{p}, \hat{z})$  in place of  $(y, \mathbf{q}, \hat{y})$ . This gives

$$\begin{aligned}\|\nabla \varepsilon_h^z\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\varepsilon_h^z - \varepsilon_h^{\hat{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\partial}} + h^{-\frac{1}{2}} \|\varepsilon_h^z\|_{\varepsilon_h^{\partial}} \\ \lesssim \|\varepsilon_h^{\mathbf{p}}\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\hat{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^{\partial}} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^{\partial}},\end{aligned}\quad (41)$$

which we use below. Next, the basic property of  $\mathcal{B}$  in Lemma 4.3 gives

$$\mathcal{B}(\varepsilon_h^p, \varepsilon_h^z, \varepsilon_h^{\tilde{z}}, \varepsilon_h^p, \varepsilon_h^z, \varepsilon_h^{\tilde{z}}) = (\varepsilon_h^p, \varepsilon_h^p)_{\mathcal{T}_h} + h^{-1} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}^2.$$

Then taking  $(r_2, w_2, \mu_2) = (\varepsilon_h^p, \varepsilon_h^z, \varepsilon_h^{\tilde{z}})$  in (39) in Lemmas 4.10 and 4.11 give

$$\begin{aligned} & (\varepsilon_h^p, \varepsilon_h^p)_{\mathcal{T}_h} + h^{-1} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}^2 \\ &= -\langle \widehat{\delta}_2, \varepsilon_h^z - \varepsilon_h^{\tilde{z}} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - \langle \widehat{\delta}_2, \varepsilon_h^z \rangle_{\varepsilon_h^\partial} + (y - y_h(u), \varepsilon_h^z)_{\mathcal{T}_h} \\ &= -\langle \delta^p \cdot \mathbf{n}, \varepsilon_h^z - \varepsilon_h^{\tilde{z}} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} - h^{-1} \langle \delta^z, P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad - \langle \delta^p \cdot \mathbf{n}, \varepsilon_h^z \rangle_{\varepsilon_h^\partial} - h^{-1} \langle \delta^z, P_M \varepsilon_h^z \rangle_{\varepsilon_h^\partial} + (y - y_h(u), \varepsilon_h^z)_{\mathcal{T}_h} \\ &\leq \|\delta^p\|_{\partial \mathcal{T}_h} \|\varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-1} \|\delta^z\|_{\partial \mathcal{T}_h} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + \|\delta^p\|_{\partial \mathcal{T}_h} \|\varepsilon_h^z\|_{\varepsilon_h^\partial} + h^{-1} \|\delta^z\|_{\partial \mathcal{T}_h} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} \\ &\quad + \|y - y_h(u)\|_{\mathcal{T}_h} \|\varepsilon_h^z\|_{\mathcal{T}_h} \\ &\leq h^{1/2} \|\delta^p\|_{\partial \mathcal{T}_h} h^{-1/2} \|\varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\delta^z\|_{\partial \mathcal{T}_h} h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} \\ &\quad + h^{1/2} \|\delta^p\|_{\partial \mathcal{T}_h} h^{-1/2} \|\varepsilon_h^z\|_{\varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\delta^z\|_{\partial \mathcal{T}_h} h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} \\ &\quad + \|y - y_h(u)\|_{\mathcal{T}_h} \|\varepsilon_h^z\|_{\mathcal{T}_h} \\ &\lesssim h^{1/2} \|\delta^p\|_{\partial \mathcal{T}_h} (\|\varepsilon_h^p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}) \\ &\quad + h^{-\frac{1}{2}} \|\delta^z\|_{\partial \mathcal{T}_h} h^{-\frac{1}{2}} (\|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}) \\ &\quad + \|y - y_h(u)\|_{\mathcal{T}_h} (\|\nabla \varepsilon_h^z\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\varepsilon_h^z\|_{\varepsilon_h^\partial}) \\ &\lesssim h^{1/2} \|\delta^p\|_{\partial \mathcal{T}_h} (\|\varepsilon_h^p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}) \\ &\quad + h^{-\frac{1}{2}} \|\delta^z\|_{\partial \mathcal{T}_h} h^{-\frac{1}{2}} (\|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}) \\ &\quad + \|y - y_h(u)\|_{\mathcal{T}_h} (\|\varepsilon_h^p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}). \end{aligned}$$

Applying Young's inequality and Lemma 4.9 gives

$$\begin{aligned} & (\varepsilon_h^p, \varepsilon_h^p)_{\mathcal{T}_h} + h^{-1} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial}^2 + h^{-1} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial}^2 \\ &\lesssim h \|\delta^p\|_{\partial \mathcal{T}_h}^2 + h^{-1} \|\delta^z\|_{\partial \mathcal{T}_h}^2 + \|y_h(u) - y\|_{\mathcal{T}_h}^2 \\ &\lesssim h^{2s_p} \|\mathbf{p}\|_{s_p, \Omega}^2 + h^{2s_z-2} \|z\|_{s_z, \Omega}^2 + h^{2s_q+2} \|\mathbf{q}\|_{s_q, \Omega}^2 + h^{2s_y} \|y\|_{s_y, \Omega}^2. \end{aligned}$$

This gives

$$\begin{aligned} & \|\varepsilon_h^p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \varepsilon_h^{\tilde{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} \\ &\lesssim h^{s_p} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-1} \|z\|_{s_z, \Omega} + h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}, \end{aligned}$$

$$\begin{aligned}
\|\varepsilon_h^z\|_{\mathcal{T}_h} &\lesssim \|\nabla \varepsilon_h^z\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \widehat{\varepsilon}_h^z\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} \\
&\lesssim \|\varepsilon_h^p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z - \widehat{\varepsilon}_h^z\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} \\
&\lesssim h^{s_p} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-1} \|z\|_{s_z, \Omega} + h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}.
\end{aligned}$$

□

As a consequence, a simple application of the triangle inequality gives optimal convergence rates for  $\|\mathbf{p} - \mathbf{p}_h(u)\|_{\mathcal{T}_h}$  and  $\|z - z_h(u)\|_{\mathcal{T}_h}$ :

### Lemma 4.13

$$\begin{aligned}
\|\mathbf{p} - \mathbf{p}_h(u)\|_{\mathcal{T}_h} &\leq \|\delta^p\|_{\mathcal{T}_h} + \|\varepsilon_h^p\|_{\mathcal{T}_h} \\
&\lesssim h^{s_p} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-1} \|z\|_{s_z, \Omega} + h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega},
\end{aligned} \tag{42a}$$

$$\begin{aligned}
\|z - z_h(u)\|_{\mathcal{T}_h} &\leq \|\delta^z\|_{\mathcal{T}_h} + \|\varepsilon_h^z\|_{\mathcal{T}_h} \\
&\lesssim h^{s_p} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-1} \|z\|_{s_z, \Omega} + h^{s_q+1} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y} \|y\|_{s_y, \Omega}.
\end{aligned} \tag{42b}$$

### 4.3.6 Step 6: Estimate for $\|u - u_h\|_{\varepsilon_h^\partial}$ and $\|y - y_h\|_{\mathcal{T}_h}$

Next, we bound the error between the solutions of the auxiliary problem and the HDG problem (25). We use these error bounds and the error bounds in Lemma 4.9, 4.12, 4.13 to obtain the main result.

For the remaining steps, we denote

$$\begin{aligned}
\zeta_q &= \mathbf{q}_h(u) - \mathbf{q}_h, \quad \zeta_y = y_h(u) - y_h, \quad \zeta_{\widehat{y}} = \widehat{y}_h^o(u) - \widehat{y}_h^o, \\
\zeta_p &= \mathbf{p}_h(u) - \mathbf{p}_h, \quad \zeta_z = z_h(u) - z_h, \quad \zeta_{\widehat{z}} = \widehat{z}_h^o(u) - \widehat{z}_h^o.
\end{aligned}$$

Subtracting the auxiliary problem and the HDG problem gives the following error equations

$$\mathcal{B}(\zeta_q, \zeta_y, \zeta_{\widehat{y}}; \mathbf{r}_1, w_1, \mu_1) = -\langle P_M u - u_h, \mathbf{r}_1 \cdot \mathbf{n} - h^{-1} w_1 \rangle_{\varepsilon_h^\partial} \tag{43a}$$

$$\mathcal{B}(\zeta_p, \zeta_z, \zeta_{\widehat{z}}; \mathbf{r}_2, w_2, \mu_2) = (\zeta_y, w_2)_{\mathcal{T}_h} \tag{43b}$$

for all  $(\mathbf{r}_1, \mathbf{r}_2, w_1, w_2, \mu_1, \mu_2) \in V_h \times V_h \times W_h \times W_h \times M_h(o) \times M_h(o)$ .

### Lemma 4.14 We have

$$\begin{aligned}
&\|u - u_h\|_{\varepsilon_h^\partial}^2 + \gamma^{-1} \|\zeta_y\|_{\mathcal{T}_h}^2 \\
&= \langle u + \gamma^{-1} \mathbf{p}_h(u) \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h(u), u - u_h \rangle_{\varepsilon_h^\partial} \\
&\quad - \langle u_h + \gamma^{-1} \mathbf{p}_h \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h, u - u_h \rangle_{\varepsilon_h^\partial}.
\end{aligned}$$

**Proof** First, we have

$$\begin{aligned} & \langle u + \gamma^{-1} \mathbf{p}_h(u) \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h(u), u - u_h \rangle_{\varepsilon_h^\partial} \\ & - \langle u_h + \gamma^{-1} \mathbf{p}_h \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h, u - u_h \rangle_{\varepsilon_h^\partial} \\ & = \|u - u_h\|_{\varepsilon_h^\partial}^2 + \gamma^{-1} \langle \zeta_p \cdot \mathbf{n} + h^{-1} P_M \zeta_z, u - u_h \rangle_{\varepsilon_h^\partial}. \end{aligned}$$

As in the proof of Proposition 4.4, it can be shown that

$$\mathcal{B}(\zeta_q, \zeta_y, \zeta_{\bar{y}}; \zeta_p, -\zeta_z, -\zeta_{\bar{z}}) + \mathcal{B}(\zeta_p, \zeta_z, \zeta_{\bar{z}}; -\zeta_q, \zeta_y, \zeta_{\bar{y}}) = 0.$$

On the other hand, use the error equations in (43) to obtain

$$\begin{aligned} & \mathcal{B}(\zeta_q, \zeta_y, \zeta_{\bar{y}}; \zeta_p, -\zeta_z, -\zeta_{\bar{z}}) + \mathcal{B}(\zeta_p, \zeta_z, \zeta_{\bar{z}}; -\zeta_q, \zeta_y, \zeta_{\bar{y}}) \\ & = (\zeta_y, \zeta_y)_{\mathcal{T}_h} - \langle P_M u - u_h, \zeta_p \cdot \mathbf{n} + h^{-1} \zeta_z \rangle_{\varepsilon_h^\partial} \\ & = (\zeta_y, \zeta_y)_{\mathcal{T}_h} - \langle u - u_h, \zeta_p \cdot \mathbf{n} + h^{-1} P_M \zeta_z \rangle_{\varepsilon_h^\partial}. \end{aligned}$$

Comparing the above two equalities gives

$$(\zeta_y, \zeta_y)_{\mathcal{T}_h} = \langle u - u_h, \zeta_p \cdot \mathbf{n} + h^{-1} P_M \zeta_z \rangle_{\varepsilon_h^\partial}.$$

**Theorem 4.15** *We have*

$$\begin{aligned} \|u - u_h\|_{\varepsilon_h^\partial} & \lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}, \\ \|y - y_h\|_{\mathcal{T}_h} & \lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

**Proof** Since  $u + \gamma^{-1} \mathbf{p} \cdot \mathbf{n} = 0$  on  $\varepsilon_h^\partial$  by (7e) and  $u_h + \gamma^{-1} \mathbf{p}_h \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h = 0$  on  $\varepsilon_h^\partial$  by (13g) and (13k) we have

$$\begin{aligned} \|u - u_h\|_{\varepsilon_h^\partial}^2 + \gamma^{-1} \|\zeta_y\|_{\mathcal{T}_h}^2 & = \langle u + \gamma^{-1} \mathbf{p}_h(u) \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h(u), u - u_h \rangle_{\varepsilon_h^\partial} \\ & = \langle \gamma^{-1} (\mathbf{p}_h(u) - \mathbf{p}) \cdot \mathbf{n} + \gamma^{-1} h^{-1} P_M z_h(u), u - u_h \rangle_{\varepsilon_h^\partial} \\ & \lesssim (\|\mathbf{p}_h(u) - \mathbf{p}\|_{\partial \mathcal{T}_h} + h^{-1} \|P_M z_h(u)\|_{\varepsilon_h^\partial}) \|u - u_h\|_{\varepsilon_h^\partial}. \end{aligned}$$

Next, use an inverse inequality and Lemma 4.12 to get

$$\begin{aligned} \|\mathbf{p}_h(u) - \mathbf{p}\|_{\partial \mathcal{T}_h} & \leq \|\mathbf{p}_h(u) - \Pi \mathbf{p}\|_{\partial \mathcal{T}_h} + \|\Pi \mathbf{p} - \mathbf{p}\|_{\partial \mathcal{T}_h} \\ & \lesssim h^{-\frac{1}{2}} \|\mathbf{p}_h(u) - \Pi \mathbf{p}\|_{\mathcal{T}_h} + h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} \\ & \lesssim h^{s_p - \frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z - \frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q + \frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} \\ & \quad + h^{s_y - \frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

Also, use  $z = 0$  on  $\varepsilon_h^\partial$ , Lemma 4.12, and properties of the  $L^2$  projection to obtain

$$\begin{aligned} h^{-1} \|P_M z_h(u)\|_{\varepsilon_h^\partial} &= h^{-1} \|P_M z_h(u) - P_M \Pi z + P_M \Pi z - P_M z\|_{\varepsilon_h^\partial} \\ &\leq h^{-1} (\|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} + \|\Pi z - z\|_{\varepsilon_h^\partial}) \\ &\leq h^{-1} (\|P_M \varepsilon_h^z\|_{\varepsilon_h^\partial} + \|\Pi z - z\|_{\partial \mathcal{T}_h}) \\ &\lesssim h^{sp-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{sq+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} \\ &\quad + h^{sy-\frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

This gives

$$\|u - u_h\|_{\varepsilon_h^\partial} \lesssim h^{sp-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{sq+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{sy-\frac{1}{2}} \|y\|_{s_y, \Omega}.$$

Moreover, we have

$$\|\zeta_y\|_{\mathcal{T}_h} \lesssim h^{sp-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{sq+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{sy-\frac{1}{2}} \|y\|_{s_y, \Omega}.$$

Since  $y - y_h = y - y_h(u) + \zeta_y$ , by the triangle inequality and Lemma 4.9 we obtain

$$\|y - y_h\|_{\mathcal{T}_h} \lesssim h^{sp-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{sq+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{sy-\frac{1}{2}} \|y\|_{s_y, \Omega}.$$

□

Note that in the final estimate for  $\|y - y_h\|_{\mathcal{T}_h}$  in the above proof, the lower convergence rate for  $\|\zeta_y\|_{\mathcal{T}_h}$  dominates the higher convergence rate for  $\|y - y_h(u)\|_{\mathcal{T}_h}$  from Lemma 4.9. In Step 7 below, the convergence rates for  $\|\zeta_p\|_{\mathcal{T}_h}$  and  $\|\zeta_z\|_{\mathcal{T}_h}$  will also dominate the error bounds for the variables  $\mathbf{p}$  and  $z$ .

#### 4.3.7 Step 7: Estimates for $\|\mathbf{q} - \mathbf{q}_h\|_{\mathcal{T}_h}$ , $\|\mathbf{p} - \mathbf{p}_h\|_{\mathcal{T}_h}$ and $\|z - z_h\|_{\mathcal{T}_h}$

**Lemma 4.16** *We have*

$$\begin{aligned} \|\zeta_{\mathbf{q}}\|_{\mathcal{T}_h} &\lesssim h^{sp-1} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-2} \|z\|_{s_z, \Omega} + h^{sq} \|\mathbf{q}\|_{s_q, \Omega} + h^{sy-1} \|y\|_{s_y, \Omega}, \\ \|\zeta_{\mathbf{p}}\|_{\mathcal{T}_h} &\lesssim h^{sp-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{sq+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{sy-\frac{1}{2}} \|y\|_{s_y, \Omega}, \\ \|\zeta_z\|_{\mathcal{T}_h} &\lesssim h^{sp-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{sz-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{sq+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{sy-\frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

**Proof** By Lemma 4.3 and the error Eq. (43a), we have

$$\begin{aligned} &\mathcal{B}(\zeta_{\mathbf{q}}, \zeta_y, \zeta_{\hat{y}}; \zeta_{\mathbf{q}}, \zeta_y, \zeta_{\hat{y}}) \\ &= (\zeta_{\mathbf{q}}, \zeta_{\mathbf{q}})_{\mathcal{T}_h} + \langle h^{-1}(P_M \zeta_y - \zeta_{\hat{y}}), P_M \zeta_y - \zeta_{\hat{y}} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \zeta_y, P_M \zeta_y \rangle_{\varepsilon_h^\partial} \\ &= -\langle P_M u - u_h, \zeta_{\mathbf{q}} \cdot \mathbf{n} - h^{-1} \zeta_y \rangle_{\varepsilon_h^\partial} = -\langle u - u_h, \zeta_{\mathbf{q}} \cdot \mathbf{n} - h^{-1} P_M \zeta_y \rangle_{\varepsilon_h^\partial} \end{aligned}$$

$$\begin{aligned} &\lesssim \|u - u_h\|_{\varepsilon_h^\partial} (\|\zeta_q\|_{\varepsilon_h^\partial} + h^{-1} \|P_M \zeta_y\|_{\varepsilon_h^\partial}) \\ &\lesssim h^{-\frac{1}{2}} \|u - u_h\|_{\varepsilon_h^\partial} (\|\zeta_q\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \zeta_y\|_{\varepsilon_h^\partial}), \end{aligned}$$

which gives

$$\begin{aligned} \|\zeta_q\|_{\mathcal{T}_h} &\lesssim h^{-\frac{1}{2}} \|u - u_h\|_{\varepsilon_h^\partial} \\ &\lesssim h^{s_p-1} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-2} \|z\|_{s_z, \Omega} + h^{s_q} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y-1} \|y\|_{s_y, \Omega}. \end{aligned}$$

Next, we estimate  $\zeta_p$ . By Lemma 4.3, the error Eq. (43b), we have

$$\begin{aligned} \mathcal{B}(\zeta_p, \zeta_z, \zeta_{\bar{z}}; \zeta_p, \zeta_z, \zeta_{\bar{z}}) &= (\zeta_p, \zeta_p)_{\mathcal{T}_h} + \langle h^{-1}(P_M \zeta_z - \zeta_{\bar{z}}), P_M \zeta_z - \zeta_{\bar{z}} \rangle_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + \langle h^{-1} P_M \zeta_z, P_M \zeta_{\bar{z}} \rangle_{\varepsilon_h^\partial} \\ &= (\zeta_y, \zeta_z)_{\mathcal{T}_h} \\ &\leq \|\zeta_y\|_{\mathcal{T}_h} \|\zeta_z\|_{\mathcal{T}_h} \\ &\lesssim \|\zeta_y\|_{\mathcal{T}_h} (\|\nabla \zeta_z\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\zeta_z - \zeta_{\bar{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\zeta_z\|_{\varepsilon_h^\partial}) \\ &\lesssim \|\zeta_y\|_{\mathcal{T}_h} (\|\zeta_p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\zeta_z - \zeta_{\bar{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\zeta_z\|_{\varepsilon_h^\partial}), \end{aligned}$$

where we used the discrete Poincaré inequality in Lemma 4.11 and also Lemma 4.6. This implies

$$\begin{aligned} &\|\zeta_p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \zeta_z - \zeta_{\bar{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \zeta_z\|_{\varepsilon_h^\partial} \\ &\lesssim h^{s_p-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y-\frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

The discrete Poincaré inequality in Lemma 4.11 also gives

$$\begin{aligned} \|\zeta_z\|_{\mathcal{T}_h} &\lesssim \|\nabla \zeta_z\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|\zeta_z - \zeta_{\bar{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|\zeta_z\|_{\varepsilon_h^\partial} \\ &\lesssim \|\zeta_p\|_{\mathcal{T}_h} + h^{-\frac{1}{2}} \|P_M \zeta_z - \zeta_{\bar{z}}\|_{\partial \mathcal{T}_h \setminus \varepsilon_h^\partial} + h^{-\frac{1}{2}} \|P_M \zeta_z\|_{\varepsilon_h^\partial} \\ &\lesssim h^{s_p-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y-\frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

□

The above lemma along with the triangle inequality, Lemmas 4.9, 4.13 complete the proof of the main result:

**Theorem 4.17** *We have*

$$\begin{aligned} \|\mathbf{q} - \mathbf{q}_h\|_{\mathcal{T}_h} &\lesssim h^{s_p-1} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-2} \|z\|_{s_z, \Omega} + h^{s_q} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y-1} \|y\|_{s_y, \Omega}, \\ \|\mathbf{p} - \mathbf{p}_h\|_{\mathcal{T}_h} &\lesssim h^{s_p-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y-\frac{1}{2}} \|y\|_{s_y, \Omega}, \\ \|z - z_h\|_{\mathcal{T}_h} &\lesssim h^{s_p-\frac{1}{2}} \|\mathbf{p}\|_{s_p, \Omega} + h^{s_z-\frac{3}{2}} \|z\|_{s_z, \Omega} + h^{s_q+\frac{1}{2}} \|\mathbf{q}\|_{s_q, \Omega} + h^{s_y-\frac{1}{2}} \|y\|_{s_y, \Omega}. \end{aligned}$$

## 5 Numerical experiments

For our numerical experiments, we test problems similar to the examples considered in [29]; see also [6, 42, 48].

We begin with a 2D example on a square domain  $\Omega = [0, 1/4] \times [0, 1/4] \subset \mathbb{R}^2$ . The largest interior angle is  $\omega = \pi/2$ , and so  $r_\Omega = 3$  and  $p_\Omega = \infty$ . The data is chosen as

$$f = 0, \quad y_d = (x^2 + y^2)^s \quad \text{and} \quad \gamma = 1,$$

where  $s = 10^{-5}$ . Then  $y_d \in H^1(\Omega) \cap L^\infty(\Omega)$ . For the case  $k = 1$ , i.e., quadratic polynomials are used for the scalar variables, and linear polynomials are used for the flux variables and the boundary trace variables, Corollary 4.2 in Sect. 4 gives the convergence rates

$$\begin{aligned} \|y - y_h\|_{0,\Omega} &= O(h^{3/2-\varepsilon}), & \|z - z_h\|_{0,\Omega} &= O(h^{3/2-\varepsilon}), \\ \|\mathbf{q} - \mathbf{q}_h\|_{0,\Omega} &= O(h^{1-\varepsilon}), & \|\mathbf{p} - \mathbf{p}_h\|_{0,\Omega} &= O(h^{3/2-\varepsilon}), \end{aligned}$$

and

$$\|u - u_h\|_{0,\Gamma} = O(h^{3/2-\varepsilon}).$$

For  $k = 0$ , i.e., linear polynomials are used for the scalar variables, and piecewise constant functions are used for the other variables, Corollary 4.2 gives convergence rates of  $O(h^{1/2})$  for all variables.

Since we do not have an explicit expression for the exact solution, we solved the problem numerically for a triangulation with 262,144 elements, i.e.,  $h = 2^{-12}\sqrt{2}$  and compared this reference solution against other solutions computed on meshes with larger  $h$ .

The numerical results for  $k = 1$  are shown in Table 1. The convergence rates observed for  $\|\mathbf{q} - \mathbf{q}_h\|_{0,\Omega}$  and  $\|u - u_h\|_{0,\Gamma}$  are in agreement with our theoretical results, while the convergence rates for  $\|\mathbf{p} - \mathbf{p}_h\|_{0,\Omega}$ ,  $\|y - y_h\|_{0,\Omega}$ , and  $\|z - z_h\|_{0,\Omega}$  are higher than our theoretical results. A similar phenomena can be observed in [29, 48]. Only one work explained the above phenomena: May, Rannacher, and Vexler in [42] used a duality argument to obtain improved convergence rates for the state and dual state with the standard finite element method. To the best of our knowledge, it is not clear how to apply this technique to standard mixed finite element methods or the HDG method.

The numerical results for  $k = 0$  are shown in Table 2. The convergence rates for all variables are higher than the  $O(h^{1/2})$  rate from Corollary 4.2. As indicated in Sect. 4, the  $O(h^{1/2})$  rate from Corollary 4.2 for  $k = 0$  is suboptimal for all variables. Improving the numerical analysis for both the  $k = 1$  and  $k = 0$  cases is an interesting problem to be considered in the future.

For illustration, in Fig. 1 we plot the state  $y$ , adjoint state  $z$ , and their fluxes  $\mathbf{q}$  and  $\mathbf{p}$  computed using  $k = 1$ . The 2D regularity result in Sect. 2 indicates that the

**Table 1** 2D example,  $k = 1$ : error of control  $u$ , state  $y$ , adjoint state  $z$ , and their fluxes  $\mathbf{q}$  and  $\mathbf{p}$ 

$h/\sqrt{2}$	$2^{-4}$	$1/2^{-5}$	$2^{-6}$	$2^{-7}$	$2^{-8}$
$\ \mathbf{q} - \mathbf{q}_h\ _{0,\Omega}$	4.1343e-02	2.1025e-02	1.0677e-02	5.3865e-03	2.6959e-03
Order	–	0.9756	0.9776	0.9871	0.9986
$\ \mathbf{p} - \mathbf{p}_h\ _{0,\Omega}$	1.3463e-03	3.8638e-04	1.0849e-04	2.9862e-05	8.0969e-06
Order	–	1.8009	1.8325	1.8612	1.8828
$\ y - y_h\ _{0,\Omega}$	5.4609e-04	1.3647e-04	3.4763e-05	8.8037e-06	2.2236e-06
Order	–	2.0005	1.9730	1.9814	1.9852
$\ z - z_h\ _{0,\Omega}$	1.9671e-05	2.6887e-06	3.7026e-07	5.0372e-08	6.7776e-09
Order	–	2.8711	2.8603	2.8778	2.8940
$\ u - u_h\ _{0,\Gamma}$	7.3053e-03	2.6902e-03	9.7764e-04	3.5178e-04	1.2569e-04
Order	–	1.4412	1.4603	1.4746	1.4849

**Table 2** 2D example,  $k = 0$ : error of control  $u$ , state  $y$ , adjoint state  $z$ , and their fluxes  $\mathbf{q}$  and  $\mathbf{p}$ 

$h/\sqrt{2}$	$2^{-4}$	$1/2^{-5}$	$2^{-6}$	$2^{-7}$	$2^{-8}$
$\ \mathbf{q} - \mathbf{q}_h\ _{0,\Omega}$	4.7552e-02	3.4107e-02	2.1082e-02	1.2281e-02	6.9039e-03
Order	–	0.47942	0.69409	0.77961	0.83090
$\ \mathbf{p} - \mathbf{p}_h\ _{0,\Omega}$	1.6793e-03	9.8644e-04	5.2097e-04	2.6498e-04	1.3302e-04
Order	–	0.76759	0.92104	0.97531	0.99429
$\ y - y_h\ _{0,\Omega}$	7.3260e-04	3.2546e-04	1.0577e-04	3.1075e-05	8.7640e-06
Order	–	1.1706	1.6215	1.7671	1.8261
$\ z - z_h\ _{0,\Omega}$	7.3656e-05	2.0645e-05	5.4062e-06	1.3718e-06	3.4375e-07
Order	–	1.8350	1.9331	1.9785	1.9967
$\ u - u_h\ _{0,\Gamma}$	7.4915e-03	4.6700e-03	2.5730e-03	1.3539e-03	6.9528e-04
Order	–	0.68183	0.85996	0.92630	0.96148

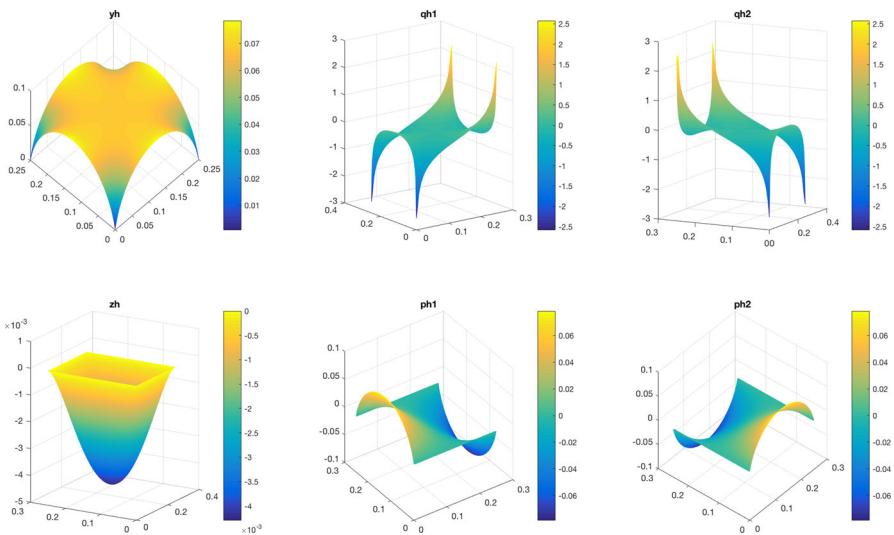
primary flux  $\mathbf{q}$  can have low regularity. In this example, it does indeed appear that  $\mathbf{q}$  has singularities at the corners of the domain. We plot the computed control in Fig. 2. These figures can be compared to similar plots in [6,48].

Next, we consider a 3D extension of the 2D example above. The domain is a cube  $\Omega = [0, 1/32] \times [0, 1/32] \times [0, 1/32]$ , and the data is chosen as

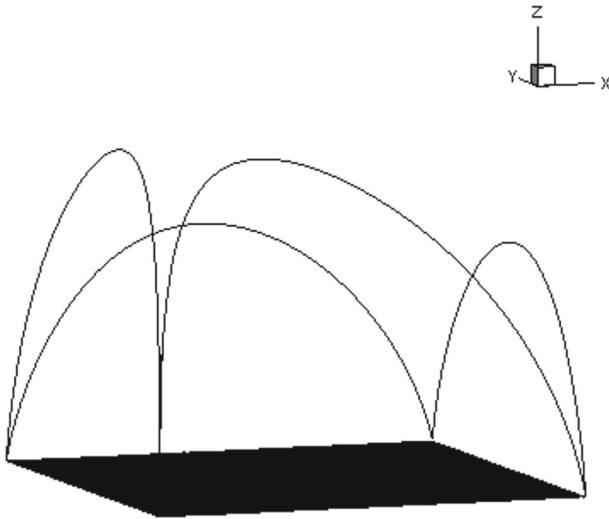
$$f = 0, \quad y_d = (x^2 + y^2 + z^2)^s \quad \text{and} \quad \gamma = 1,$$

where  $s = -1/4 + 10^{-5}$ , so that  $y_d \in H^1(\Omega)$ . In this case, we did not attempt to determine the regularity of the control and other variables; we simply present the numerical results here.

As in the 2D example above, we do not have an explicit expression for the exact solution. Therefore, we solved the problem numerically using  $k = 1$  for a triangulation with 196,608 tetrahedrons, i.e.,  $h = 2^{-12}\sqrt{3}$  and compared this reference solution against other solutions computed on meshes with larger  $h$ . The numerical results are



**Fig. 1** 2D example,  $k = 1$ : The computed primary state  $y_h$ , the primary flux  $q_h$ , the dual state  $z_h$ , and the dual flux  $p_h$



**Fig. 2** The optimal control  $u_h$  for the 2D example

shown in Table 3. The observed convergence rates for all variables are similar to the results for the 2D example above.

## 6 Conclusions

We proposed an HDG method to approximate the solution of an optimal Dirichlet boundary control problem for the Poisson equation. We obtained an optimal superlin-

**Table 3** 3D example,  $k = 1$ : error of control  $u$ , state  $y$ , adjoint state  $z$ , and their fluxes  $\mathbf{q}$  and  $\mathbf{p}$ 

$h/\sqrt{3}$	$2^{-6}$	$2^{-7}$	$2^{-8}$	$2^{-9}$
$\ \mathbf{q} - \mathbf{q}_h\ _{0,\Omega}$	9.2640e-03	5.2580e-03	2.7462e-03	1.2475e-03
Order	–	0.81712	0.93706	1.1384
$\ \mathbf{p} - \mathbf{p}_h\ _{0,\Omega}$	3.5425e-05	1.2283e-05	3.8463e-06	1.1022e-06
Order	–	1.5281	1.6751	1.8032
$\ y - y_h\ _{0,\Omega}$	1.6040e-05	4.5070e-06	1.2191e-06	2.9781e-07
Order	–	1.8314	1.8864	2.0333
$\ z - z_h\ _{0,\Omega}$	7.8545e-08	1.3058e-08	2.0042e-09	2.8775e-10
Order	–	2.5886	2.7039	2.8001
$\ u - u_h\ _{0,\Gamma}$	4.5932e-04	1.8934e-04	7.1955e-05	2.4123e-05
Order	–	1.2785	1.3958	1.5767

ear rate of convergence for the control in 2D under certain assumptions on the domain, the target state  $y_d$ , and the polynomial degree of the HDG method. Numerical experiments confirmed our theoretical results for this superlinear convergence rate when piecewise linear polynomials were used to approximate the control (Table 1).

Our results indicate HDG methods have potential for solving more complex Dirichlet boundary control problems. We plan to investigate HDG methods for Dirichlet boundary control of other PDEs, including convection dominated diffusion problems and fluid flows. These problems may involve solutions with large gradients or shocks, and it is natural to consider HDG methods for such problems.

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## A Local solver

By simple algebraic operations in Eq. (17), we obtain the following formulas for the matrices  $G_1$ ,  $G_2$ ,  $H_1$ , and  $H_2$  in (18):

$$\begin{aligned} G_1 &= B_1^{-1} B_2 (B_4 + B_2^T B_1^{-1} B_2)^{-1} (B_5 + B_2^T B_1^{-1} B_3) - B_1^{-1} B_3, \\ G_2 &= -(B_4 + B_2^T B_1^{-1} B_2)^{-1} (B_5 + B_2^T B_1^{-1} B_3), \\ H_1 &= -B_1^{-1} B_2 (B_4 + B_2^T B_1^{-1} B_2)^{-1}, \\ H_2 &= (B_4 + B_2^T B_1^{-1} B_2)^{-1}. \end{aligned}$$

In general, this process is impractical; however, for the HDG method described in this work, these matrices can be easily computed. This is one of the advantages of the HDG method. We briefly describe this process below.

Since the spaces  $V_h$  and  $W_h$  consist of discontinuous polynomials, some of the system matrices are block diagonal and each block is small and symmetric positive definite. Let us call a matrix of this form a SSPD block diagonal matrix. The inverse of a SSPD block diagonal matrix is another SSPD block diagonal matrix, and the inverse can be easily constructed by computing the inverse of each small block. Furthermore, the inverse of each small block can be computed independently; and therefore computing the inverse can be easily done in parallel.

It can be checked that  $B_1$  is a SSPD block diagonal matrix, and therefore  $B_1^{-1}$  is easily computed and is also a SSPD block diagonal matrix. Therefore, the matrices  $G_1$ ,  $G_2$ ,  $H_1$ , and  $H_2$  are easily computed if  $B_4 + B_2^T B_1^{-1} B_2$  is also easily inverted. We show below that this is the case.

First, it can be checked that  $B_2$  is block diagonal with small blocks, but the blocks are not symmetric or definite. This implies  $B_2^T B_1^{-1} B_2$  is block diagonal with small nonnegative definite blocks. Next,  $B_4 = \begin{bmatrix} A_5 & 0 \\ -A_4 & A_5 \end{bmatrix}$ , where  $A_4$  and  $A_5$  are both SSPD block diagonal. Due to the structure of  $B_1$  and  $B_2$ , the matrix  $B_2^T B_1^{-1} B_2 + B_4$  has the form  $\begin{bmatrix} C_1 & 0 \\ -A_4 & C_2 \end{bmatrix}$ , where  $C_1$  and  $C_2$  are SSPD block diagonal. The inverse can be easily computed using the formula

$$\begin{bmatrix} C_1 & 0 \\ -A_4 & C_2 \end{bmatrix}^{-1} = \begin{bmatrix} C_1^{-1} & 0 \\ C_2^{-1} A_4 C_1^{-1} & C_2^{-1} \end{bmatrix}.$$

Furthermore,  $C_1^{-1}$ ,  $C_2^{-1}$  and  $C_2^{-1} A_4 C_1^{-1}$  are both SSPD block diagonal.

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