

FIRST-ORDER SYSTEM LEAST-SQUARES METHOD FOR THE OPTIMAL CONTROL BY THE STOKES EQUATIONS

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ABSTRACT

We consider the approximate solution of optimal control problem governed by the Stokes equations. For the constrained optimization problem, Lagrange multiplier method is used. We employ a least-squares functional developed in [3] for the two- and three-dimensional first order optimality systems, by introducing a *velocity flux* variable and associated curl and *trace* equations. With the provided H^2 -regularity of that system, ellipticity in an H^1 product norm is established. This yields optimal discretization error estimates for finite element spaces in a product H^1 norm. The resulting matrix becomes symmetric and positive definite. For numerical tests, we apply V-cycle multigrid methods to the total discrete algebraic system.

COUPLED FIRST-ORDER SYSTEM FORMULATIONS

Let $\Omega \subset \mathbb{R}^n (n = 2 \text{ or } 3)$ be an open, connected and bounded domain with Lipschitz boundary $\partial\Omega$. We consider the control problem consisting of minimizing the quadratic functional defined by

$$\min \mathcal{J}(\mathbf{u}, \mathbf{f}) = \frac{1}{2} \int_{\Omega} |\mathbf{u} - \hat{\mathbf{u}}|^2 dx + \frac{\sigma}{2} \int_{\Omega} |\mathbf{f}|^2 dx \quad (1)$$

subject to the Stokes equation

$$\left\{ \begin{array}{ll} -\nu \Delta \mathbf{u} + \nabla p = \mathbf{f} & \text{in } \Omega \\ \nabla \cdot \mathbf{u} = 0 & \text{in } \Omega \\ \mathbf{u} = \mathbf{0} & \text{on } \partial\Omega \\ \int_{\Omega} p dx = 0, & \end{array} \right. \quad (2)$$

where σ is positive constant.

Then, the constrained optimization problem (1)-(2) is the equivalent to the unconstrained optimization problem of finding saddle points of the Lagrangian functional. These saddle points

may be found by solving the optimality system:

$$\begin{cases} -\nu\Delta\mathbf{u} + \nabla p = \mathbf{f} & \text{in } \Omega \\ \nu\Delta\mathbf{v} + \nabla q + \mathbf{u} = \hat{\mathbf{u}} & \text{in } \Omega \\ \nabla \cdot \mathbf{u} = 0 & \text{in } \Omega \\ \nabla \cdot \mathbf{v} = 0 & \text{in } \Omega \\ -\sigma\mathbf{f} = \mathbf{v} & \text{in } \Omega. \end{cases} \quad (3)$$

According to [3], This problem can be rewritten as the first order system of partial differential equations by introducing the velocity flux variable $\underline{\mathbf{U}} = \nabla\mathbf{u}^t = (\nabla u_1, \nabla u_2, \dots, \nabla u_n)$, and $\underline{\mathbf{V}} = \nabla\mathbf{v}^t = (\nabla v_1, \nabla v_2, \dots, \nabla v_n)$, (3) can be written as an extended first order system of partial differential equations:

$$\begin{cases} -\nu(\nabla \cdot \underline{\mathbf{U}})^t + \nabla p + \frac{\mathbf{v}}{\sigma} = \mathbf{0} & \text{in } \Omega \\ \underline{\mathbf{U}} - \nabla\mathbf{u}^t = \underline{\mathbf{0}} & \text{in } \Omega \\ \nabla \cdot \mathbf{u} = 0 & \text{in } \Omega \\ \nabla \times \underline{\mathbf{U}} = \mathbf{0} & \text{in } \Omega \\ \nabla(\text{tr}\underline{\mathbf{U}}) = \mathbf{0} & \text{in } \Omega \\ \mathbf{u} = \mathbf{0} & \text{on } \Gamma \\ \mathbf{n} \times \underline{\mathbf{U}} = \mathbf{0} & \text{on } \Gamma, \end{cases} \quad (4)$$

$$\begin{cases} \nu(\nabla \cdot \underline{\mathbf{V}})^t + \nabla q + \mathbf{u} = \hat{\mathbf{u}} & \text{in } \Omega \\ \underline{\mathbf{V}} - \nabla\mathbf{v}^t = \underline{\mathbf{0}} & \text{in } \Omega \\ \nabla \cdot \mathbf{v} = 0 & \text{in } \Omega \\ \nabla \times \underline{\mathbf{V}} = \mathbf{0} & \text{in } \Omega \\ \nabla(\text{tr}\underline{\mathbf{V}}) = \mathbf{0} & \text{in } \Omega \\ \mathbf{v} = \mathbf{0} & \text{on } \Gamma \\ \mathbf{n} \times \underline{\mathbf{V}} = \mathbf{0} & \text{on } \Gamma, \end{cases} \quad (5)$$

Then, we may make a least-squares functional as in [3], which are

$$\begin{aligned} G(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q; \hat{\mathbf{u}}) &= \|\nu(\nabla \cdot \underline{\mathbf{U}})^t - \nabla p - \frac{\mathbf{v}}{\sigma}\|^2 + \|\underline{\mathbf{U}} - \nabla\mathbf{u}^t\|^2 + \|\nabla \cdot \mathbf{u}\|^2 + \|\nabla \times \underline{\mathbf{U}}\|^2 \\ &+ \|\nabla\text{tr}(\underline{\mathbf{U}})\|^2 + \|\nu(\nabla \cdot \underline{\mathbf{V}})^t + \nabla q + \mathbf{u} - \hat{\mathbf{u}}\|^2 + \|\underline{\mathbf{V}} - \nabla\mathbf{v}^t\|^2 \\ &+ \|\nabla \cdot \mathbf{v}\|^2 + \|\nabla \times \underline{\mathbf{V}}\|^2 + \|\nabla\text{tr}(\underline{\mathbf{V}})\|^2. \end{aligned} \quad (6)$$

Hence we are going to minimize $G(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q; \hat{\mathbf{u}})$. Since the coercivity and continuity of the bilinear form corresponding to G are equivalent to stating that the functional G is norm-equivalent, the following theorems show the existence and uniqueness of the minimizer.

Theorem 1. Assume that the domain Ω is a bounded convex polyhedron or has $C^{1,1}$ boundary. Then there are two constants C_1 and C_2 dependent on σ and ν such that for any $(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q) \in \mathcal{W} \times \mathcal{W}$, we have

$$C_1 M(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q) \leq G(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q; \mathbf{0}), \quad (7)$$

and

$$G(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q; \mathbf{0}) \leq C_2 M(\underline{\mathbf{U}}, \mathbf{u}, p, \underline{\mathbf{V}}, \mathbf{v}, q), \quad (8)$$

where

$$M(\mathbf{U}, \mathbf{u}, p, \mathbf{V}, \mathbf{v}, q) = \|\mathbf{U}\|_1^2 + \|\mathbf{u}\|_1^2 + \|p\|_1^2 + \|\mathbf{V}\|_1^2 + \|\mathbf{v}\|_1^2 + \|q\|_1^2.$$

The discrete problem for minimizing (6) is equivalent to an linear algebraic system:

$$\begin{bmatrix} \mathbb{K}_1 & \mathbb{C}_1^t & \mathbb{C}_2^t & \mathbf{0} & \mathbb{C}_4^t & \mathbf{0} \\ \mathbb{C}_1 & \mathbb{K}_2 & \mathbf{0} & \mathbb{C}_3^t & \mathbf{0} & \mathbb{C}_6^t \\ \mathbb{C}_2 & \mathbf{0} & \mathbb{K}_3 & \mathbf{0} & \mathbb{C}_5^t & \mathbf{0} \\ \mathbf{0} & \mathbb{C}_3 & \mathbf{0} & \frac{1}{\sigma^2}\mathbb{K}_1 & \frac{1}{\sigma^2}\mathbb{C}_1^t & -\frac{1}{\sigma^2}\mathbb{C}_2^t \\ \mathbb{C}_4 & \mathbf{0} & \mathbb{C}_5 & \frac{1}{\sigma^2}\mathbb{C}_1 & \mathbb{K}_4 & \mathbf{0} \\ \mathbf{0} & \mathbb{C}_6 & \mathbf{0} & -\frac{1}{\sigma^2}\mathbb{C}_2 & \mathbf{0} & \frac{1}{\sigma^2}\mathbb{K}_3 \end{bmatrix} \begin{bmatrix} \vec{U} \\ \vec{u} \\ \vec{p} \\ \vec{V} \\ \vec{v} \\ \vec{q} \end{bmatrix} = \begin{bmatrix} \vec{0} \\ \vec{g}_1 \\ \vec{0} \\ \vec{g}_2 \\ \vec{0} \\ \vec{g}_3 \end{bmatrix}, \quad (9)$$

where $\vec{U} = (U_1, \dots, U_{4J})^t$, $\vec{u} = (u_1, \dots, u_{2J})^t$, $\vec{p} = (p_1, \dots, p_J)^t$, $\vec{V} = (V_1, \dots, V_{4J})^t$, $\vec{v} = (v_1, \dots, v_{2J})^t$, $\vec{q} = (q_1, \dots, q_J)^t$. The block matrices \mathbb{K}_1 and \mathbb{K}_4 are of size $4J \times 4J$, \mathbb{K}_2 is of size $2J \times 2J$, \mathbb{K}_3 is of size $J \times J$, \mathbb{C}_1 and \mathbb{C}_4 are of size $2J \times 4J$, \mathbb{C}_2 is of size $J \times 4J$, \mathbb{C}_3 is of size $4J \times 2J$, \mathbb{C}_5 is of size $2J \times J$ and \mathbb{C}_6 is of size $J \times 2J$.

NUMERICAL EXPERIMENTATIONS WITH MULTIGRID METHOD

Table 1 The L^2 -errors $\|\mathbf{u}^h - \hat{\mathbf{u}}\|$ between the controlled flow \mathbf{u}^h and the target flow $\hat{\mathbf{u}}$, the L^2 -norm of control $\|\mathbf{f}^h\|$ and the values of the cost functional for the different σ when $h = 1/32$.

ν	σ	$\ \mathbf{u}^h - \hat{\mathbf{u}}\ $	$\ \mathbf{f}^h\ $	$\mathcal{J}(\mathbf{u}^h, \mathbf{f}^h)$
1	1	1.273898834e-1	2.334794985e-3	8.116816833e-3
	10^{-1}	1.239166887e-1	1.890008890e-2	7.695533536e-3
	10^{-2}	3.700037256e-2	5.435863697e-2	6.992880920e-4
	10^{-3}	7.202576194e-4	8.281429727e-3	2.936765583e-7
	10^{-4}	6.576050304e-4	7.602196141e-2	5.051891188e-7
10^{-1}	1	1.210256866e-1	2.213912607e-2	7.568678854e-3
	10^{-1}	3.584283083e-2	6.274538301e-2	8.392034154e-4
	10^{-2}	1.078722145e-3	3.168275641e-2	5.600806001e-6
	10^{-3}	5.974841291e-4	4.485327213e-2	1.184401653e-6
	10^{-4}	6.145601493e-4	6.459243032e-1	2.104975236e-5
10^{-2}	1	2.323299951e-2	3.816631056e-2	9.982197639e-4
	10^{-1}	6.127488652e-4	2.493789498e-2	3.128266088e-5
	10^{-2}	5.433031180e-4	1.010494889e-2	6.581390993e-7
	10^{-3}	5.523628405e-4	5.717007883e-2	1.786761311e-6
	10^{-4}	5.532245028e-4	7.030575325e-1	2.486752338e-5

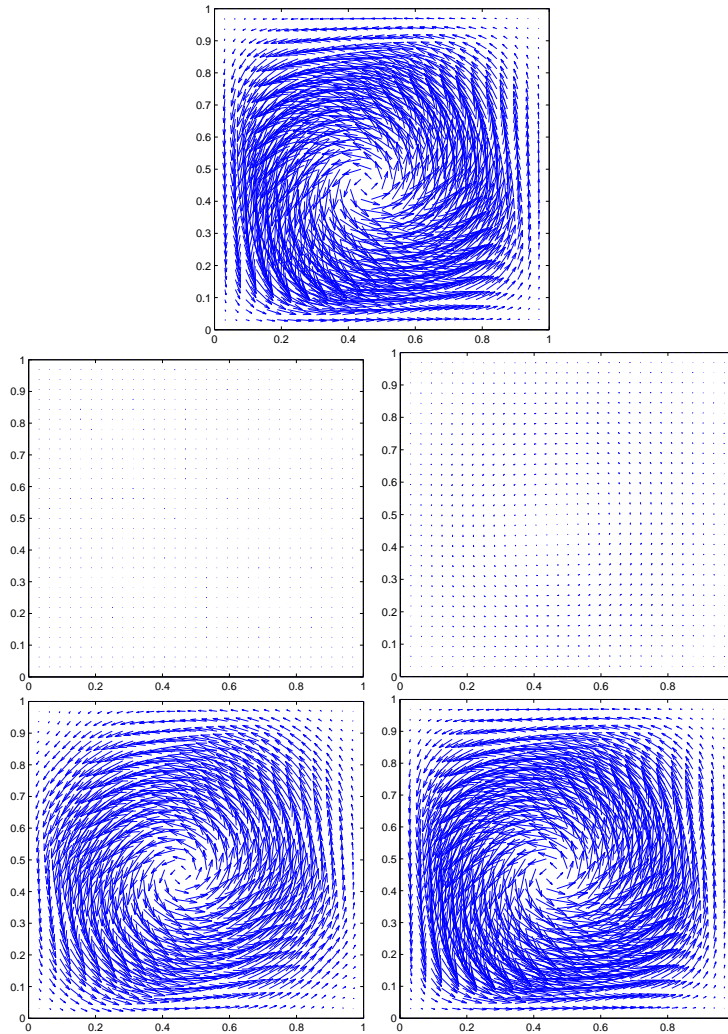


Figure 1. Target flow(top) and controlled flows at $\sigma = 1$ (middle left), 0.1 (middle right), 0.01 (bottom left) and $\sigma = 0.001$ (bottom right) for $\nu = 1$, when $h = 1/32$.

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