AN ALGORITHM FOR SOLVING NON-SYMMETRIC SYSTEMS ARISING FROM SMOOTHER FDM

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The contribution deals with a fast method for solving large scale algebraic saddle-point systems arising from a new fictitious domain formulation of elliptic boundary value problems.

Keywords: saddle-point system, fictitious domain method, Schur complement, orthogonal projectors, BiCGSTAB algorithm, multigrid

1 Introduction

The classical fictitious domain method (FDM) enforces boundary conditions in PDE's by Lagrange multipliers defined on the boundary γ of the original domain ω . Therefore the computed solution has a singularity on γ that can result in an intrinsic error. The basic idea of our modification consists in introducing new control variables (instead of Lagrange multipliers) defined on an auxiliary boundary Γ located outside of $\overline{\omega}$ [1]. In this approach, the singularity is moved away from $\overline{\omega}$ so that the computed solution is smoother in ω and the discretization error has a significantly higher rate of convergence in ω .

The respective finite element discretization leads typically to a non-symmetric saddle-point system

$$\begin{pmatrix} A & B_1^{\top} \\ B_2 & 0 \end{pmatrix} \begin{pmatrix} \hat{u} \\ \lambda \end{pmatrix} = \begin{pmatrix} f \\ g \end{pmatrix}, \tag{1}$$

where an $(n \times n)$ diagonal block A is possibly singular and $(m \times n)$ off-diagonal blocks B_1 , B_2 have full row-rank and they are highly sparse. Moreover, m is much smaller than n and the defect l of A, i.e., l = n - rank A, is much smaller than m. For solving such systems, it is convenient to use a method based on the Schur complement reduction. If A is singular, the reduced system has again a saddle-point structure. Fortunately after applying orthogonal projectors, we obtain an equation in terms of λ only that can be efficiently solved by the projected variant of the BiCGSTAB algorithm [1]. This procedure generalizes ideas used in FETI domain decomposition methods, in which A is symmetric, positive semidefinite and $B_1 = B_2$.

2 Fictitious domain method

Let us consider a non-homogeneous Dirichlet boundary value problem:

$$-\Delta u = f \quad \text{in } \omega, \qquad u = g \quad \text{on } \gamma, \tag{2}$$

where $\omega \subset \mathbb{R}^2$ is a bounded domain with the Lipschitz boundary γ , $f \in L^2_{loc}(\mathbb{R}^2)$ and $g \in H^{1/2}(\gamma)$ are given data.

Let $\Xi \supset \overline{\omega}$ be another domain with the Lipschitz boundary Γ , dist $(\Gamma, \gamma) = \delta$ for some $\delta > 0$ given. Finally, let $\Omega \supset \overline{\Xi}$ be a fictitious domain (a box, e.g.). We define a problem:

$$Find (\hat{u}, \lambda) \in H_0^1(\Omega) \times H^{-1/2}(\Gamma) \text{ such that} \int_{\Omega} \nabla \hat{u} \cdot \nabla v \, dx = \int_{\Omega} f v \, dx + \langle \lambda, v \rangle_{\Gamma} \quad \forall v \in H_0^1(\Omega), \langle \hat{u}, \mu \rangle_{\gamma} = \langle g, \mu \rangle_{\gamma} \quad \forall \mu \in H^{-1/2}(\gamma).$$

$$(3)$$

If γ and Γ are smooth enough then the first component of the solution to (3) satisfies $\hat{u} \in H^{3/2-\epsilon}(\Omega) \ \forall \epsilon > 0$, while its restrictions are smoother $\hat{u}_{|\Xi} \in H^2(\Xi)$, $\hat{u}_{|\Omega\setminus\overline{\Xi}} \in H^2(\Omega\setminus\overline{\Xi})$. It means that the singularity of \hat{u} is located on Γ , where generally a non-zero jump of the normal derivative $\frac{\partial \hat{u}}{\partial \nu}$ occurs. Since Γ has a positive distance from γ , one can expect that our variant of FDM will increase the convergence rate of computed solutions in ω .

3 Algorithm

Let us return to the system (1) resulting from a finite element discretization of (3), in which we use same notation for the discrete analogies of \hat{u} , λ , f and g. Our algorithm is based on the *generalized Schur complement* of A in (1) that is defined by

$$\mathcal{S} = \begin{pmatrix} -B_2 A^{\dagger} B_1^{\top} & B_2 N \\ M^{\top} B_1^{\top} & 0 \end{pmatrix},$$

where A^{\dagger} is a generalized inverse to A and columns of N and M span the null-spaces $\mathbb{N}(A)$ and $\mathbb{N}(A^{\top})$, respectively.

Theorem 1 [1] Let us assume that (1) has a unique solution. Its second component λ is the first component of a solution to

$$\begin{pmatrix} F & G_1^\top \\ G_2 & 0 \end{pmatrix} \begin{pmatrix} \lambda \\ \alpha \end{pmatrix} = \begin{pmatrix} d \\ e \end{pmatrix}, \tag{4}$$

where $F := B_2 A^{\dagger} B_1^{\top}, G_1 := -N^{\top} B_2^{\top}, G_2 := -M^{\top} B_1^{\top}, d := B_2 A^{\dagger} f - g$ and $e := -M^{\top} f$. The first component \hat{u} is given by the formulae

$$\hat{u} = A^{\dagger}(f - B_1^{\top}\lambda) + N\alpha.$$

Let us point out that (4) is formally the same saddle-point system as (1), but its size is considerably smaller. We will modify it by two orthogonal projectors

$$P_1 := I - G_1^{\top} (G_1 G_1^{\top})^{-1} G_1, \quad P_2 := I - G_2^{\top} (G_2 G_2^{\top})^{-1} G_2$$

on the null-spaces $\mathbb{N}(G_1)$, $\mathbb{N}(G_2)$, respectively. The following results are keys to the algorithm.

Lemma 1 [1] The linear operator $P_1F : \mathbb{N}(G_2) \mapsto \mathbb{N}(G_1)$ is invertible.

Theorem 2 [1] Let $\lambda_{\mathbb{N}} \in \mathbb{N}(G_2)$, $\lambda_{\mathbb{R}} \in \mathbb{R}(G_2^{\top})$. Then $\lambda = \lambda_{\mathbb{N}} + \lambda_{\mathbb{R}}$ is the first component of a solution to (4) iff

$$\lambda_{\mathbb{R}} = G_2^\top (G_2 G_2^\top)^{-1} e$$

and

$$P_1 F \lambda_{\mathbb{N}} = P_1 (d - F \lambda_{\mathbb{R}}).$$

The second component α is given by

$$\alpha = (G_1 G_1^\top)^{-1} G_1 (d - F\lambda).$$

Algorithm: Projected Schur Complement Method (PSCM)

Step 1.a: Assemble $G_1 = -N^{\top}B_2^{\top}, G_2 = -M^{\top}B_1^{\top}, d = B_2A^{\dagger}f - g$ and $e = -M^{\top}f$.Step 1.b: Assemble $H_1 = (G_1G_1^{\top})^{-1}$ and $H_2 = (G_2G_2^{\top})^{-1}$.Step 1.c: Assemble $\lambda_{\mathbb{R}} = G_2^{\top}H_2e$.Step 1.d: Assemble $\tilde{d} = P_1(d - F\lambda_{\mathbb{R}})$.Step 1.e: Solve the equation $P_1F\lambda_{\mathbb{N}} = \tilde{d}$ on $\mathbb{N}(G_2)$.Step 1.f: Compute $\lambda = \lambda_{\mathbb{N}} + \lambda_{\mathbb{R}}$.Step 2: Compute $\alpha = H_1G_1(d - F\lambda)$.Step 3: Compute $\hat{u} = A^{\dagger}(f - B_1^{\top}\lambda) + N\alpha$.

The heart of the algorithm consists in Step 1.e. Its solution can be computed by a *projected* Krylov subspace method. The projected BiCGSTAB algorithm [1] can be derived from the non-projected one by choosing an initial iterate $\lambda_{\mathbb{N}}^0$ on $\mathbb{N}(G_2)$, projecting the initial residual in $\mathbb{N}(G_2)$ and replacing the operator P_1F by its projected version P_2P_1F . Finally, let us point out that convergence of the projected BiCGSTAB algorithm can be accelerated by a multigrid technique.

4 Numerical experiments

Let ω be the ellipse, $\omega \equiv \{(x,y) \in \mathbb{R}^2 | (x-0.5)^2/0.4^2 + (y-0.5)^2/0.2^2 < 1\}$, and the fictitious domain $\Omega = (0,1) \times (0,1)$. We will assume that the right hand-sides f and g in (2) are chosen appropriately to the exact solution $\hat{u}_{ex}(x,y) = 100 \left((x-0.5)^3 - (y-0.5)^3 \right) - x^2$.

The space $H_0^1(\Omega)$ in (3) is replaced by $H_{per}^1(\Omega)$ enabling us to use the Fourier direct method [2] to compute actions of A^{\dagger} , where A is the positive semidefinite discrete Laplacian resulting from the discretization of $H_{per}^1(\Omega)$ by piecewise bilinear functions defined on a rectangulation of Ω with a stepsize h. The spaces $H^{-1/2}(\Gamma)$ and $H^{-1/2}(\gamma)$ are approximated by piecewise constant functions defined on partitions of polygonal approximations of Γ and γ , respectively.

In Table 1, we report the errors of the approximate solutions with respect to the stepsize h in the $H^1(\omega)$ -norm together with the number of BiCGSTAB iterations. We compare the classical FDM based on Lagrange multipliers and our modification (3), in which the auxiliary boundary Γ arises by shifting γ in the direction of the outward normal vector with $\delta = 8h$. From the computed errors, we determine the convergence rates (the last row of the table) that are considerably higher for modified case.

	Classical		Modified		Modified+Multigrid	
Step h	Iters.	$\operatorname{Err}_{H^1(\omega)}$	Iters.	$\operatorname{Err}_{H^1(\omega)}$	Iters.	$\operatorname{Err}_{H^1(\omega)}$
1/128	8	1.9647e + 0	13	1.6878e-2	11	1.8988e-2
1/256	9	1.2884e + 0	25	7.7891e-3	13	7.6303e-3
1/512	12	8.6517e-1	40	4.0160e-3	19	3.8638e-3
1/1024	18	6.0510e-1	58	1.9098e-3	21	1.7758e-3
1/2048	25	4.4015e-1	86	9.9299e-4	31	9.8213e-4
Conv. rates:		0.54		1.02		1.07

Table 1: Comparisons of the methods.

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