

First-Order System Least Squares *FOSLS*

DOE & NSF & IBM

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Chan	Jespersen	Povinelli	
Chang	Jiang	Shao	

HISTORICAL GAP

fast solvers

&

uniform performance

Reynolds number, Lamé constants,...

Outline

no frills

- I. Basic FOSLS
- II. FOSLS Tools
- III. Anatomy of FOSLS for C-D
- IV. FOSLL* & Results
- V. Concluding Remarks

- I. Basics**
- II. Tools**
- III. C-D**
- IV. LL***
- V. End**

FOSLoSophy

PHYSICAL PROCESS



CONTINUUM MODEL



ALGEBRAIC MODEL



ALGEBRAIC SOLVER



IMPLEMENTATION



design effort

My Matrix Grail

symmetry

$$Lx = b, \quad L \text{ general}$$

Normal Equations: $A = L^T L$

$$L^T L x = L^T b$$

Dual Equations: $A = L L^T$

$$L L^T y = b$$

Dual Matrix Equations

dual (ghost) variables

$$x = L^T y$$

$Lx = b$ solvable

solution x $(L) = \text{Range}(L^T)$

$$x = L^T y$$

$$LL^T y = b$$

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Matrix Minimization Principles

“energy” functionals

$$Lx = b, \quad L \text{ general}$$

Normal Equations:

$$\|Lx - b\|^2 = \langle L^T Lx, x \rangle - 2\langle x, L^T b \rangle + \langle b, b \rangle$$

Dual Equations:

$$\langle LL^T y, y \rangle - 2\langle y, b \rangle$$

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My REAL Matrix Grail

symmetry & positive definiteness

$$AX = b, \quad A \text{ spd}$$

$$\text{error} = e = x - X$$

Energy Functional: solving = minimizing energy error

$$\begin{aligned} F(x) &= \langle Ax, x \rangle - 2\langle x, b \rangle \\ &= \langle A(x-X), x-X \rangle - \langle AX, X \rangle \\ &= \langle Ae, e \rangle + \text{constant} \end{aligned}$$

?nice?

My ACTUAL Real Grail

spd & nice (elliptic perhaps)

Let $L = -\Delta_h$ be a discrete Laplacian.

$A = L^2$ $O(h^{-4})$ condition number!

TROUBLE!

But “no” trouble with $A = L$:

minimize $\langle Ae, e \rangle \sim \|e\|^2 = H^1$ semi-norm 2

H^1 -ellipticity multigrid

I. Basics
II. Tools
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Bottom Line

Objective:

$L \longrightarrow A$ spd & H^1 -elliptic.

Trouble:

$L = (\quad h)^2$ is spd & H^2 -elliptic!?

Cure:

Compute $A = L^{1/2}$?!

Solution:

Get access to the PDE !

- I. Basics
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FOSLS Illustration

$$p'' + ap' = f$$

↑ second order

ignore boundary

First-Order System: $u = p'$

$$Lu = u' + au = f$$

FOSLS Functional:

$$F(u; f) = \|Lu - f\|^2$$

$$= \int (u' + au - f)^2 dx$$

$$\sim \int (e')^2 dx \quad H^1\text{-ellipticity}$$

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Multigrid

$$\min F(q^h)$$

Relaxation: $S_i^h H^1$

$$F(q^h + s_i^h; f) = \min_{t \in \mathbf{R}} F(q^h + t_i^h; f)$$

Coarsening: $S^{2h} S^h$

$$F(q^h + p^{2h}; f) = \min_{q^{2h} \in S^{2h}} F(q^h + q^{2h}; f)$$

Optimality (Oswald, Zhang, Bramble, ...):

homogeneous/bilinear part

$$F(p + e; f) - F(p; f) = E(e) \sim \|e\|_1^2$$

Minimization Principle Advantages

$$\min F(q)$$

- sense of optimality
- error monitor $F(q)$
- monotone: $T \quad S \quad \min_T F \quad \min_S F$
- continuum functions & subspaces
- approximation theory analysis
- elliptic
 - standard FE & MG with theory
 - final H^1 accuracy

I. Basics
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FOSLS Essence

divide & conquer

Motivation: scalar (esp. elliptic) problems are “well” understood

scalar (elliptic) energy minimization principle

Plan: imbue PDE with “nice” energy principle

Objective: PDE loosely coupled scalar equations

Purpose: *standard* finite elements and multigrid methods

Approach: PDE First Order System Energy

I. Basics
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Issues

FOSLS

- New variables
- New equations
- New boundary conditions
- Scaling
- Other norms

I. Basics
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Error Measure

Objective: approximation + *reasonable* error measure

Ideal: optimal approximation in the measure

Reasonable:

- always accurate values: L^2
- often accurate derivatives: H^1

Conventional Measure ($L = \infty$):

- residual = $\|Lp - f\|_0 = \|L(p - p^*)\|_0 = \|Le\|_0 = \|e\|_0$
- oscillatory error overly large residuals

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FOSLS Error Measure

for *any* approximation \mathbf{w}
to the exact solution \mathbf{w}^*

Absolute :

$$F(\mathbf{w}; \mathbf{f}) = \|\mathbf{L}\mathbf{w} - \mathbf{f}\|_0^2 = \|\mathbf{L}(\mathbf{w} - \mathbf{w}^*)\|_0^2 \\ \sim \|\mathbf{w} - \mathbf{w}^*\|_1^2$$

Relative :

$$\frac{F(\mathbf{w}; \mathbf{f})}{F(\mathbf{0}; \mathbf{f})} = \frac{\|\mathbf{L}\mathbf{w} - \mathbf{f}\|_0^2}{\|\mathbf{L}\mathbf{w}^*\|_0^2} \sim \frac{\|\mathbf{w} - \mathbf{w}^*\|_1^2}{\|\mathbf{w}^*\|_1^2}$$

!!! powerful tool for adaptive refinement !!!

!avoids patch test for nonconforming elements!

#1: *New Variables*

reducing order

Poisson:

- $p = f$

$$\mathbf{u} = p$$

CFD/Elasticity:

- $\mathbf{u} + \text{Re}(\mathbf{u}^t)^t \mathbf{u} + p = f$

- $\mathbf{u} + p = g$

$$\mathbf{U} = \mathbf{u}^t$$

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#2: *New Equations*

good norm equivalence

Poisson :

$$\mathbf{u} - p = \mathbf{0}$$

$$\bullet \mathbf{u} + f = 0$$

$$\times \mathbf{u} = \mathbf{0}$$

CFD/Elasticity :

$$\mathbf{U} - \mathbf{u}^t = \mathbf{O}$$

$$- \bullet \mathbf{U} + \text{Re } \mathbf{U}^t \mathbf{u} + p = \mathbf{f}$$

$$\bullet \mathbf{u} + p = g$$

nonlinearity across
variables only

$$\times \mathbf{U} = \mathbf{O}$$

$$\& \text{trace } \mathbf{U} + p = 0$$

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Curl Option

Poisson system :

$$\mathbf{u} - p = \mathbf{0}$$

$$\nabla \cdot \mathbf{u} = f$$

$$\nabla \times \mathbf{u} = \mathbf{0}$$

Div - free error : $\nabla \cdot \mathbf{e} = 0$

$$\mathbf{u} + \mathbf{e} - p - \mathbf{0} = \mathbf{e}$$

$$\nabla \cdot (\mathbf{u} + \mathbf{e}) - f = 0$$

$$\nabla \times (\mathbf{u} + \mathbf{e}) = \nabla \times \mathbf{e}$$

#3: *New Boundary Conditions*

good norms

Poisson Case : $\mathbf{u} = p$

Neumann :

$$\mathbf{n} \cdot \nabla p = 0$$

$$\mathbf{n} \cdot \mathbf{u} = 0$$

Dirichlet :

$$p = 0$$

$$\mathbf{n} \times \mathbf{u} = \mathbf{0}$$

Robin/Radiation :

$$\mathbf{n} \cdot \nabla p + \alpha p = 0$$

$$\mathbf{n} \cdot \mathbf{u} + \beta p = 0$$

Tangential Boundary Condition

$y = \text{constant line}$

$$p = 0$$

$$p_x = 0$$

$$p \parallel \mathbf{n}$$

$$\mathbf{n} \times p = 0$$

#4: Preconditioners

nonsmooth data or L^2 measure

ILLUSTRATE WITH POISSON: $\mathbf{u} - p = \mathbf{0}$

$$\left(- \right)^{-\frac{1}{2}} \left(\bullet \mathbf{u} + f \right) = 0$$

$$\left(- \right)^{-\frac{1}{2}} \left(\times \mathbf{u} \right) = \mathbf{0}$$

$$F_{-1}(\mathbf{u}, p; f) =$$

$$\|\mathbf{u} - p\|_0^2 + \|\bullet \mathbf{u} - f\|_{-1}^2 + \|\times \mathbf{u}\|_{-1}^2$$

$$F_{-1}(\mathbf{u}, p; 0) \sim \|p\|_0^2 + \|\mathbf{u}\|_0^2$$

Inverse Norm

Precondition $Lw = f$ by $B = (-\Delta)^{-1/2}$:

$$BLw = Bf$$

$$\begin{aligned}\|q\|_{-1}^2 &= \sup_{r \in H^1} \frac{\langle q, r \rangle^2}{\|r\|_1^2} \\ &= \langle (-\Delta)^{-1} q, q \rangle \\ &= \|(-\Delta)^{-1/2} q\|_0^2\end{aligned}$$

direct computation by MG

#5: Scaling

balanced equivalence

Poisson Case: $\mathbf{u} = p$

$$F(\mathbf{u}, p; f) =$$

$$\|\mathbf{u} - p\|^2 + (1 - \alpha)(\|\nabla \cdot \mathbf{u} + f\|^2 + \|\beta \times \mathbf{u}\|^2)$$

Stage I:

$$F_0(\mathbf{u}; f) = \|\nabla \cdot \mathbf{u} + f\|^2 + \|\beta \times \mathbf{u}\|^2$$

$$F_0(\mathbf{u}; 0) \sim \|\mathbf{u}\|^2$$

Stage II:

$$F_1(p; \mathbf{u}) = \|\mathbf{u} - p\|^2$$

$$F_1(p; \mathbf{0}) = \|p\|^2$$

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III. FOSLS for Convection Diffusion

convection-diffusion

$$\begin{aligned} - \bullet \quad & p + ap_x = f, \quad 0 \leq x, y \leq 1 \\ & p(0, y) = p(1, y) = 0 \\ & p_y(x, 0) = p_y(x, 1) = 0 \end{aligned}$$

Guide

$$- \bullet \quad p + ap_x = -e^{ax} (e^{-ax} p_x)_x - p_{yy}$$

I. Basics
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Convection Diffusion 1-3

1 : New Variables

$$u = e^{-ax} p_x$$

$$v = p_y$$

flux

2 : New Equations

$$-e^{ax} u_x - v_y + f = 0 \quad \text{div}$$

$$e^{ax} u_y - v_x = 0 \quad \text{'curl'}$$

3 : New Boundary Conditions

$$p(0, y) = p(1, y) = 0$$

$$v(0, y) = v(1, y) = 0$$

$$v(x, 0) = v(x, 1) = 0$$

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Convection Diffusion 5

tool #5

$f = 0$ for simplicity

First Stage :

$$e^{-ax/2} \quad \times \quad \left\{ \begin{array}{l} -e^{ax} u_x - v_y = 0 \\ e^{ax} u_y - v_x = 0 \end{array} \right.$$

Troublesome Scaling :

$$\begin{aligned} & \left\| -e^{ax} u_x - v_y \right\|^2 + \left\| e^{ax} u_y - v_x \right\|^2 \\ &= \left\| e^{ax} u_x \right\|^2 + 2 \langle e^{ax} u_x, v_y \rangle + \left\| v_y \right\|^2 \\ &+ \left\| e^{ax} u_y \right\|^2 - 2 \langle e^{ax} u_y, v_x \rangle + \left\| v_x \right\|^2 \end{aligned}$$

decoupled u & v ?

Convection Diffusion 5

continued

Better Scaling : $s_{\pm}(x, y) = e^{\pm ax/2}$

$$\left\| -s_+ u_x - s_- v_y \right\|^2 + \left\| s_+ u_y - s_- v_x \right\|^2$$

$$= \left\| s_+ u_x \right\|^2 + 2 \langle u_x, v_y \rangle + \left\| s_- v_y \right\|^2$$

divided & conquered

$$+ \left\| s_+ u_y \right\|^2 - 2 \langle u_y, v_x \rangle + \left\| s_- v_x \right\|^2$$

- **uniform** scaled H^1 equivalence
- scaling damps boundary layer behavior
- valid for more general convection term
- Poisson equation for recovering p

More General Form

$$s_- \sim e^{-ax/2}$$

$$\mathbf{u} = p$$
$$-\bullet \mathbf{u} + au = f$$

$$F_0(\mathbf{u}; f) =$$
$$\| s_- [-\bullet \mathbf{u} + au - f] \|^2$$
$$+ \| s_- [\times \mathbf{u}] \|^2$$

$$F_0(\mathbf{u}; 0) \sim$$
$$\| s_- \mathbf{u} \|^2 + \| s_- \mathbf{v} \|^2$$

IV. FOSLL*

L^2 error minimization

Energy Functional: $LL^*V = f, U = L^*V$

$$\langle LL^*v, v \rangle - 2 \langle v, f \rangle$$

$$= \langle L^*v, L^*v \rangle - 2 \langle v, LL^*V \rangle \pm \langle V, LL^*V \rangle$$

$$= \langle L^*v, L^*v \rangle - 2 \langle L^*v, L^*V \rangle \pm \langle L^*V, L^*V \rangle$$

$$= \|L^*v - U\|^2 - \text{constant}$$

precise L^2 error minimization

*FOSLL**

theoretical issues

L is a first-order differential operator
with boundary conditions.

L^* is a first-order differential operator
with adjoint boundary conditions.

The solution of $Lu = f$ is in the range of L^* .

L^* is H^1 -elliptic: $\|L^*v\| \sim \|v\|$.

- I. Basics
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FOSLL*

sample application
2D

Problem:

$$\mathbf{a} \cdot \mathbf{p} + \mathbf{b} \cdot \mathbf{p} = f$$

First-Order System: $\mathbf{u} = a^{1/2} \mathbf{p}$

$$\begin{aligned} \mathbf{a} \cdot \mathbf{u} + \mathbf{b} \cdot \mathbf{u} &= f \\ \times a^{-1/2} \mathbf{u} &= 0 \end{aligned}$$

FOSLL*

formal dual LL*

2D

$$\begin{matrix} \bullet a^{1/2} + \mathbf{b} \bullet \\ \times a^{-1/2} \end{matrix} \begin{pmatrix} -a^{1/2} & + \mathbf{b} & a^{-1/2} & \times \end{pmatrix} \begin{matrix} w \\ z \end{matrix}$$

$$= \begin{matrix} (\bullet a^{1/2} + \mathbf{b} \bullet) (-a^{1/2} + \mathbf{b}) (\mathbf{b} \bullet a^{-1/2} \times)^* \\ \mathbf{b} \bullet a^{-1/2} \times \times a^{-1} \times \end{matrix} \begin{matrix} w \\ z \end{matrix}$$

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FOSLL* Numerical Example

two stage

problem:

$$\bullet a \quad p + \mathbf{b} \bullet \quad p = f, \quad a = \begin{cases} 1, & x \leq 0.5 \\ , & x > 0.5 \end{cases}$$

solution:

$$p = \begin{cases} ((2 - 4)x^2 + (4 -))\sin(\pi y), & x \leq 0.5 \\ (-6x^2 + 7x - 1)\sin(\pi y), & x > 0.5 \end{cases}$$

method:

FE uniform linears & AMG W(1,1) cycles

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Results

fine grids: $h=1/128$ $\mathbf{b}^t = (0,0)$ & $\mathbf{b}^t = (6,9)$

jump	$\ \text{MG}\ $	FE accuracy
1	0.08	$\left\{ \begin{array}{l} O(h^2) \text{ for } p \text{ in } L^2 \\ O(h) \text{ for } p \text{ in } H^1 \\ O(h) \text{ for } \mathbf{u} \text{ in } L^2 \end{array} \right.$
10	0.13	
100	0.25	

convection scales: $= 1$ $\mathbf{b}^t = (6,9)$

h	$=$	1/4	1/8	1/16	1/32	1/64	1/128
$\ \text{MG}\ $	$=$	0.34	0.64	0.51	0.31	0.15	0.08

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FOSLS/FOSLL Advantages*

- minimization principle advantages
- systematic once functional is properly posed
- ‘decoupled’ elliptic or “nice” equations in each variable
- avoids LBB/staggering/artificial diffusion/stabilization...
- final H^1 or L^2 accuracy for all variables
- F is its own practical performance measure
- generally excellent numerical performance

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FOSLS/FOSLL Disadvantages*

- possibly more variables
 - but possibly smaller grids
- H^1 equivalence for FOSLS needs smoothness/regularity
 - but FOSLL* does not
- conservation is not automatic
 - but can be done
- very tricky to get the functional & spaces just right
 - ✓ new variables & equations + careful scaling

I. Basics
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FOSLS Applications

various stages of development

- diffusion-convection
- neutron transport
- Stokes & Navier-Stokes
- elasticity
- electromagnetics
- eigenproblems
- coupled systems: flow/structure/em/egg
- elliptic grid generation
- boundary functionals
- discontinuities
- highly convective flows
- inverse problems
- time-dependent problems

FURTHER INFO

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<ftp://amath.colorado.edu/pub/fosls>

<http://amath.colorado.edu/appm/faculty/stevem>

University of Colorado

I. Basics
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V. End!