

Finite element methods in scientific computing

Wolfgang Bangerth, Colorado State University

Lecture 21:

Linear solvers for problems with more than one solution variable

Linear systems for vector-valued problems

Recall the mixed form of the Laplace equation:

$$\begin{aligned} K^{-1}\mathbf{u} + \nabla p &= \mathbf{0} \\ -\nabla \cdot \mathbf{u} &= -f \end{aligned}$$

The associated weak form is

$$(\boldsymbol{\phi}_u, K^{-1}\mathbf{u}) - (\nabla \cdot \boldsymbol{\phi}_u, p) - (\phi_p, \nabla \cdot \mathbf{u}) = (\phi_p, -f)$$

Expanding the solution and testing with discrete test functions yields this matrix:

$$A_{ij} = (\boldsymbol{\phi}_{i,u}, K^{-1}\boldsymbol{\phi}_{j,u}) - (\nabla \cdot \boldsymbol{\phi}_{i,u}, \phi_{j,p}) - (\phi_{i,p}, \nabla \cdot \boldsymbol{\phi}_{j,u})$$

Linear systems for vector-valued problems

Let us look at this more closely:

$$A_{ij} = (\phi_{i,u}, K^{-1} \phi_{j,u}) - (\nabla \cdot \phi_{i,u}, \phi_{j,p}) - (\phi_{i,p}, \nabla \cdot \phi_{j,u})$$

- We build the bilinear form like this in code because it's convenient
- However, in reality, shape functions are nonzero only in either the \mathbf{u} or p components!
- I.e., they are either

$$\Phi_i = \begin{pmatrix} \phi_{i,u} \\ 0 \end{pmatrix} \quad \text{or} \quad \Phi_i = \begin{pmatrix} \mathbf{0} \\ \phi_{i,p} \end{pmatrix}$$

Note: See step-8/the "vector-valued module" on this.

Linear systems for vector-valued problems

Let us look at this more closely:

$$A_{ij} = (\phi_{i,u}, K^{-1} \phi_{j,u}) - (\nabla \cdot \phi_{i,u}, \phi_{j,p}) - (\phi_{i,p}, \nabla \cdot \phi_{j,u})$$

- If index i is so that the shape function is of \mathbf{u} -type, then we get a matrix row that is in fact

$$A_{ij} = (\phi_{i,u}, K^{-1} \phi_{j,u}) - (\nabla \cdot \phi_{i,u}, \phi_{j,p})$$

- If index i is of p -type, then we get

$$A_{ij} = -(\phi_{i,p}, \nabla \cdot \phi_{j,u})$$

- We can do the same for index j .

Linear systems for vector-valued problems

Let us look at this more closely:

$$A_{ij} = (\phi_{i,u}, K^{-1} \phi_{j,u}) - (\nabla \cdot \phi_{i,u}, \phi_{j,p}) - (\phi_{i,p}, \nabla \cdot \phi_{j,u})$$

- If we enumerate degrees of freedom so that all the \mathbf{u} -types come first and then the p -types, then this leads to a *block matrix*
- Here, we get:

$$A = \begin{pmatrix} M & B \\ B^T & 0 \end{pmatrix}$$

- How does one approach solving linear systems with this matrix?

Linear systems for vector-valued problems

What is difficult about this?

- The matrix

$$A = \begin{pmatrix} M & B \\ B^T & 0 \end{pmatrix}$$

is symmetric but indefinite (i.e., has positive & negative eigenvalues)

- Conjugate Gradients (CG) requires the matrix to be symmetric & positive definite → won't work
- Preconditioners such as Jacobi or SSOR need to divide by diagonal elements → won't work

Linear systems for vector-valued problems

What should we do?

- Could use a direct solver that computes a sparse LU factorization of A
- Could use iterative solver alternatives to CG:
 - MINRES
 - SymmLQ
 - GMRES
- Would need to think more about preconditioners

More on this: Lectures 34-38.

For now: Exploit the block structure $A = \begin{pmatrix} M & B \\ B^T & 0 \end{pmatrix}$

Linear systems for vector-valued problems

Block structures in deal.II:

- Block structured matrices like

$$A = \begin{pmatrix} M & B \\ B^T & 0 \end{pmatrix}$$

are represented by classes *BlockSparsityPattern/BlockSparseMatrix*

- Block structured vectors like

$$U = \begin{pmatrix} u \\ p \end{pmatrix}$$

are represented by class *BlockVector*.

Note: From here on, u, p are *vectors*, not solution variables.

Solvers

Let us consider the linear system:

$$\begin{pmatrix} M & B \\ B^T & 0 \end{pmatrix} \begin{pmatrix} u \\ p \end{pmatrix} = \begin{pmatrix} F \\ G \end{pmatrix}$$

- This can equivalently be written as

$$Mu + Bp = F$$

$$B^T u = G$$

- We can eliminate this variable by variable:

$$u = -M^{-1}Bp + M^{-1}F$$

$$B^T M^{-1}Bp = -G + B^T M^{-1}F$$

Solvers

Let us consider the linear system:

$$\begin{pmatrix} M & B \\ B^T & 0 \end{pmatrix} \begin{pmatrix} u \\ p \end{pmatrix} = \begin{pmatrix} F \\ G \end{pmatrix}$$

- We have transformed this into a *decoupled system*:

$$B^T M^{-1} B p = -G + B^T M^{-1} F$$

$$M u = -B p + F$$

- We call $S = B^T M^{-1} B$ the *Schur complement* of the matrix
- S is a symmetric, positive definite matrix
- We can solve the equation for p with CG

Solvers

In order to solve the system

$$B^T M^{-1} B p = -G + B^T M^{-1} F$$

$$M u = -B p + F$$

we first need to solve with $S = B^T M^{-1} B$ and then with M .

Advantages:

- Both S and M are symmetric and positive definite
- We can use Conjugate Gradients and our usual preconditioners

Solvers

In order to solve the system

$$B^T M^{-1} B p = -G + B^T M^{-1} F$$

$$M u = -B p + F$$

we first need to solve with $S = B^T M^{-1} B$ and then with M .

Problem:

- S is not known element by element
- We need to represent it in code
- We need to access individual blocks of A

But: The deal.II solvers do not need to know matrix elements. They only need operator actions!

step-20

Representing solvers for

$$S p = -G + B^T M^{-1} F$$

$$M u = F - B p$$

in deal.II requires us to **represent** S, M and present it to the CG solver.

Note: We also need a preconditioner. See the tutorial program for more information!

Solvers in deal.II

In deal.II, solvers are templated on both the vector and *operator* class:

```
template <class Vector>
class SolverCG {
    ...
    public:
        template <class Matrix, class Preconditioner>
        void
        solve (const Matrix      &A,
              Vector            &x,
              const Vector      &b,
              const Preconditioner &precondition);
};
```

Requirements: (i) Operator and vector need to be compatible; (ii) operator needs to have a function `vmult (Vector &out, const Vector &in);`

Example

Here: Schur complement operator for mixed Laplace:

```
class SchurComplement {
public:
    SchurComplement (const BlockSparseMatrix &A);

    void vmult (Vector<double> &dst,
               const Vector<double> &src) const
    {
        Vector<double> tmp1(src.size());
        Vector<double> tmp2(src.size());

        A.block(0,1).vmult (tmp1, src);
        some_inverse_of(A.block(0,0))->vmult (tmp2, tmp1);
        A.block(1,0).vmult (dst, tmp2);
    }
};
```

Example

Also necessary:

- Need to represent the inverse of M
- Need to come up with a way to precondition the Schur complement

Let's just take a look at the code!

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