

## Storing of sparse matrices

- variables:
  - `npoin` = number of matrix rows (and columns)
  - `nzero` = number of nonzero matrix entries
- arrays storing the matrix:
  - `grid%irow(1:grid%npoin+1)` – integer array, storing the starts of row in the sequence storing of entries,
  - `grid%icol(1:grid%nzero)` – integer array, storing the column indexes
  - `grid%sparse(1:grid%nzero)` – real array, storing the values of matrix entries
- matrix-vector multiplication

```
npoin = grid%npoin
ip => grid%ip(1:npoin, 1:2)
v2(:) = 0.
do is = 1, npoin
  if(ip(is, 1) >=0 ) then
    i = ip(is, 2)

    do ks = grid%irow(is), grid%irow(is+1) -1
      js = grid%icol(ks)
      if(ip(js, 1) >=0) then
        j = ip(js, 2)
        v2(i) = v2(i) + grid%sparse(ks) * v1(j)
      endif
    enddo
  endif
enddo
```

Write a simple code for the solution of problem: find  $u : \Omega \rightarrow \mathbb{R}$  such that

$$-\Delta u(x) = g(x), \quad x \in \Omega, \quad (0.1)$$

$$u(x) = u_D(x), \quad x \in \partial\Omega_D, \quad (0.2)$$

$$\nabla u(x) \cdot \mathbf{n} = g_N(x), \quad x \in \partial\Omega_N, \quad (0.3)$$

where  $\Omega$  is a domain in  $\mathbb{R}^2$  with a boundary  $\partial\Omega$  consisting of two disjoint parts  $\partial\Omega_D$  and  $\partial\Omega_N$ ,  $g \in L^2(\partial\Omega_D)$ ,  $g_N \in L^2(\partial\Omega)$  and  $u_D$  is a trace of some  $u^* \in H^1(\Omega)$ .

- use  $P_1$ -conforming FE
- the arising linear system solve by the **Jacobi** and **BiCG** method
- the stiffness matrix **is computed and stored as sparse**
- compare with the dense variant of implementation for various sizes of matrices (see tutorial7)

Use the code: [http://msekc.e.karlin.mff.cuni.cz/~dolejsi/Vyuka/NS\\_source/FEM/FEM-code3.tgz](http://msekc.e.karlin.mff.cuni.cz/~dolejsi/Vyuka/NS_source/FEM/FEM-code3.tgz) [link](#)

- **mesh.f90** – reading the mesh from the file **triang**
  - **matrix.f90** – create the stiffness matrix, solution of  $\mathbb{A}x = b$
  - **sol.f90** – setting of RHS and BC (**input of data**)
  - **femP1.f90** – main code
- 
- type of boundary set in **subroutine Read\_mesh**, file **mesh.f90**
  - array **ip(:, 1)** – type of mesh vertices:  $> 0$  – interior,  $= 0$  – Neumann,  $< 0$  – Dirichlet,
  - array **ip(:, 2)** – index of vertex after removing Dirichlet nodes

## Syntax of the code:

```
> ./femP1 <J/B> <S/D>
<J> .. Jacobi
<B> .. BiCG

<S> .. sparse variant
<D> .. dense variant
```

## Basic tasks

1. study the code line by line, if something is unclear ask the teacher
2. compare the **dense** and **sparse** variants of the codes:  
[http://mseke.karlin.mff.cuni.cz/~dolejsi/Vyuka/NS\\_source/FEM/FEM-code.tgz](http://mseke.karlin.mff.cuni.cz/~dolejsi/Vyuka/NS_source/FEM/FEM-code.tgz) **dense**  
[http://mseke.karlin.mff.cuni.cz/~dolejsi/Vyuka/NS\\_source/FEM/FEM-code3.tgz](http://mseke.karlin.mff.cuni.cz/~dolejsi/Vyuka/NS_source/FEM/FEM-code3.tgz) **sparse**
3. use Linux codes:
  - `diff file1 file1`
  - `meld file1 file1 &` – has to be installed
4. perform numerical experiments for increasing number of elements using
  - (a) both solvers (**Jacobi**, **BiCG**)
  - (b) **dense** and **sparse** variants

## Further tasks

1. write a subroutine computing the error in the  $L^2$ -norm and  $H^1$ -seminorm (provided that the exact solution is known)
2. using computation on a sequence of meshes set the *experimental order of convergence*

## More advanced tasks

1. develop an algorithm for the sparse multiplication of the **transpose** matrix in the BiCG method (note that the actual implementation of BiCG works only for symmetric matrices!!!!)
2. test this subroutine with the code
3. in order to have a real nonsymmetric problem, solve the convection-diffusion equation

$$\begin{aligned} -\Delta u + \mathbf{b} \cdot \nabla u &= f && \text{in } \Omega, \\ u_h &= 0 && \text{on } \partial\Omega, \end{aligned} \tag{0.4}$$

where, e.g.,  $\mathbf{b} = (1, 0)^\top$ .

The approximate solution is given by

$$\int_{\Omega} \nabla u_h \cdot \nabla \varphi_h \, dx + \int_{\Omega} (\mathbf{b} \cdot \nabla u_h) \varphi_h \, dx = \int_{\Omega} f \varphi_h \, dx. \tag{0.5}$$

Let the approximate solution is  $u_h = \sum_{j=1}^N u_j \varphi_j$  then we have

$$\sum_{j=1}^N u_j \int_{\Omega} \nabla \varphi_j \cdot \nabla \varphi_i \, dx + \sum_{j=1}^N u_j \int_{\Omega} (\mathbf{b} \cdot \nabla \varphi_j) \varphi_i \, dx = \int_{\Omega} f \varphi_i \, dx, \quad i = 1, \dots, N. \tag{0.6}$$

If  $\mathbf{b} = (1, 0)^\top$  then  $(\mathbf{b} \cdot \nabla \varphi_j) = \frac{\partial \varphi_j}{\partial x_1}$ .

BiCG algorithms for the solution of

$$\mathbb{A}\mathbf{x} = \mathbf{b}, \quad \mathbb{A}^\top \mathbf{y} = \mathbf{c},$$

where  $\mathbb{A}$  is given matrix,  $\mathbf{b}$  and  $\mathbf{c}$  are the given-right-hand sides,  $\mathbf{x}_0$  and  $\mathbf{y}_0$  are initial guess.

- One can put  $\mathbf{c} := \mathbf{b}$  if any other better choice does not exist.
- If  $\mathbb{A}$  is symmetric and  $\mathbf{c} := \mathbf{b}$  then BiCG is equivalent to CG, only BiCG requires two times larger number of arithmetic operations.
- $\mathbb{P}$  is the preconditioned matrix in the following algorithm.

```

1: input  $\mathbb{A}$ ,  $\mathbf{x}_0$ ,  $\mathbf{y}_0$ ,  $\mathbb{P}$ 
2:  $\mathbf{r}_0 = \mathbf{b} - \mathbb{A}\mathbf{x}_0$ ,  $\mathbf{s}_0 = \mathbf{c} - \mathbb{A}^\top \mathbf{y}_0$ ,
3:  $\mathbf{p}_0 = \mathbb{P}^{-1}\mathbf{r}_0$ 
4:  $\mathbf{q}_0 = \mathbb{P}^{-\top}\mathbf{s}_0$ 
5:  $\tilde{\mathbf{r}}_0 = \mathbf{p}_0$ 
6: for  $k = 0, 1, \dots$  do
7:    $\alpha_k = \frac{\mathbf{s}_k^\top \tilde{\mathbf{r}}_k}{\mathbf{q}_k^\top \mathbb{A} \mathbf{p}_k}$ 
8:    $\mathbf{x}_{k+1} = \mathbf{x}_k + \alpha_k \mathbf{p}_k$ ,            $\mathbf{y}_{k+1} = \mathbf{y}_k + \alpha_k \mathbf{q}_k$ 
9:    $\mathbf{r}_{k+1} = \mathbf{r}_k - \alpha_k \mathbb{A} \mathbf{p}_k$ ,        $\mathbf{s}_{k+1} = \mathbf{s}_k - \alpha_k \mathbb{A}^\top \mathbf{q}_k$ 
10:   $\tilde{\mathbf{r}}_{k+1} = \mathbb{P}^{-1}\mathbf{r}_{k+1}$ ,            $\tilde{\mathbf{s}}_{k+1} = \mathbb{P}^{-\top}\mathbf{s}_{k+1}$ 
11:   $\beta_{k+1} = \frac{\mathbf{s}_{k+1}^\top \tilde{\mathbf{r}}_{k+1}}{\mathbf{s}_k^\top \tilde{\mathbf{r}}_k}$ 
12:   $\mathbf{p}_{k+1} = \tilde{\mathbf{r}}_{k+1} + \beta_{k+1}\mathbf{p}_k$ ,      $\mathbf{q}_{k+1} = \tilde{\mathbf{s}}_{k+1} + \beta_{k+1}\mathbf{q}_k$ 
13: end for

```