

# implementation of Krylov subspace method with sparse matrix

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## sparse matrix format : 1/3

$n$  : # of rows

$nnz$  : # of nonzeros

$[A]_{ij}$  : nonzero entries at  $(i, j)$

- ▶ COO (Coordinate) format MUMPS

```
structure COOformat {  
    int n, nnz;  
    int irow[nnz];  
    int jcol[nnz];  
    double coef[nnz];  
};
```

- ▶ CSR (Compressed Sparse Row) /  
CRS (Compressed Row Storage) format Pardiso

```
structure CSRformat {  
    int n, nnz;  
    int ptrow[n+1];  
    int indcol[nnz];  
    double coef[nnz];  
}
```

$$[A]_{ij} = \text{coef}[k] \\ j = \text{indcol}[k], \text{ptrow}[i] \leq k < \text{ptrow}[i + 1]$$

## sparse matrix format, zero-based index : 2/3

an example,  $5 \times 5$  unsymmetric matrix,  $n = 5$ ,  $nnz = 15$ .

	0	1	2	3	4
1.1 1.2 1.4	0	0	1	2	
2.1 2.2 2.3 2.5	1	3	4	5	6
3.2 3.3	2		7	8	
4.1 0.0 4.5	3	9		10	11
5.2 5.4 5.5	4		12	13	14

$i$	0	1	2	3	4	5
ptrow[ $i$ ]	0	3	7	9	12	15
indcol[ $k$ ]	0	1	3	0	1	
coef[ $k$ ]	1.1	1.2	1.4	2.1	2.2	

- ▶ diagonal entry should exist even if the value is 0
- ▶ `indcol[]` should be in ascending order in each row

## sparse matrix format, zero-based index : 3/3

$5 \times 5$  symmetric matrix, upper triangular,  $n = 5$ ,  $nnz = 10$ .

	0	1	2	3	4
1.1	0	0	1	2	
1.2					
1.4					
2.2	1		3	4	5
2.3					
2.5					
3.3	2			6	
0.0	3			7	8
4.5					
5.5	4				9

$i$	0	1	2	3	4	5
ptrow[ $i$ ]	0	3	6	7	9	10
indcol[ $k$ ]	0	1	3	1	2	4
coef[ $k$ ]	1.1	1.2	1.4	2.2	2.3	2.5

- ▶ diagonal entry should exist even if the value is 0
- ▶ `indcol[]` should be in ascending order in each row
- ▶ upper triangular matrix is accepted by `Pardiso`

## SpMV : sparse matrix vector multiplication : 1/3

$A$  : CSRformat { ptrow, indcol, coef },  $\vec{x}, \vec{y} \in \mathbb{R}^N$   
to calculate  $\vec{y} = A \vec{x}$

$$\begin{aligned}[A\vec{x}]_i &= \sum_j [A]_{i,j} [\vec{x}]_j \\&= \sum_{j \in \{j ; [A]_{i,j} \neq 0\}} [A]_{i,j} [\vec{x}]_j \\&= \sum_{\text{ptrow}[i] \leq k < \text{ptrow}[i+1]} \text{coef}[k] \times [\vec{x}]_{\text{indcol}[k]}\end{aligned}$$

```
for (i = 0; i < n; i++) {
    y[i] = 0.0;
    for (k = ptrow[i]; k < ptrow[i + 1]; k++) {
        int j = indcol[k];
        y[i] += coef[k] * x[j];
    }
}
```

## SpMV : sparse matrix vector multiplication : 2/3

$A$  : CSRformat { ptrow, indcol, coef }, upper part is stored,  $\vec{x}, \vec{y} \in \mathbb{R}^N$   
to calculate  $\vec{y} = A \vec{x}$   
assumption

- ▶ coefficient of diagonal value of  $A$  is stored
- ▶ the first entry of the  $i$ -th row  $\text{indcol}[\text{ptrow}[i]] == i$

```
for (i = 0; i < n; i++) {
    y[i] = 0.0;
}
for (i = 0; i < n; i++) {
    for (k = ptrow[i] + 1; k < ptrow[i + 1]; k++) {
        int j = indcol[k];
        y[i] += coef[k] * x[j];
        y[j] += coef[k] * x[i];
    }
    int k = ptrow[i];
    int j = indcol[k] ( == i )
    y[i] += coef[k] * x[i];
}
```

## SpMV : sparse matrix vector multiplication : 3/3

for calculation of  $\vec{y} = \alpha A \vec{x} + \beta \vec{y}$  with general sparse matrix  $A$  stored in CSRformat

```
void SparseGEMV(struct CSRformat &A,
                  const double &alpha, std::vector<double> &x,
                  const double &beta, std::vector<double> &y)
{
    int nrow = A.n;
    for (int i = 0; i < nrow; i++) {
        y[i] *= beta;
    }
    double tmp;
    for (int i = 0; i < nrow; i++) {
        tmp = 0.0;
        for (k = ptrow[i]; k < ptrow[i + 1]; k++) {
            int j = indcol[k];
            tmp += coef[k] * x[j];
        }
    }
    y[i] += alpha * tmp;
}
```

## inner product

two vectors :  $\vec{x}, \vec{y} \in \mathbb{R}^N$

$$(\vec{x}, \vec{y}) = \sum_{1 \leq i \leq N} [\vec{x}]_i [\vec{y}]_i$$

```
double tmp = 0.0;
for (i = 0; i < n; i++) {
    tmp += x[i] * y[i];
}
```

BLAS level 1 subroutine ddot

```
double cblas_ddot(const int n,
                   const double *x, const int incx,
                   const double *y, const int incy);

tmp = cblas_ddot(n, &x[0], 1, &y[0], 1);
```

C++ STL std::vector< > class

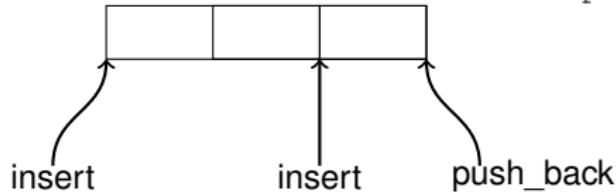
```
#include <vector>
std::vector<double> x(100); // allocation with size = 100
x.resize(200); // enlarging array with keeping 100 entries
x.clean(); // deallocation
&x[0]; // double pointer to the first entry
x.data(); //
```

## conversion of Sparse matrix format COO to CSR : 1/2

an example,  $5 \times 5$  unsymmetric matrix,  $n = 5$ ,  $nnz = 15$ .

					pcol				pcoef				
1.1	1.2		1.4		0	0	1	2	1.1	1.2	1.4		
2.1	2.2	2.3		2.5	1	0	1	2	4	2.1	2.2	2.3	2.5
	3.2	3.3			2	1	2			3.2	3.3		
4.1			0.0	4.5	3	0	3	4		4.1	0.0	4.5	
	5.2		5.4	5.5	4	1	3	4		5.2	5.4	5.5	

column index is stored in `list<int> pcol[i]` data i-th each row



- ▶ array of list for each row are stored in `vector<list<int> >`
- ▶ insertion using `std::list` iterator stores data in ascending order
- ▶ one dimensional array `int indcol[k]` is copied from `pcol`

list data structure will manage dynamic memory allocation but is not efficient for HPC application

## conversion of Sparse matrix format COO to CSR : 2/2

```
std::vector<std::list<int>> pcol(nrow);
std::vector<std::list<double>> pcoef(nrow);
for (int k = 0; k < nnz; k++) {
    int i = Acoo.irow[k], j = Acoo.jcol[k];
    double coef = Acco.coef[k];
    if (pcol[i].empty())
        pcol[i].push_back(j); pcoef[i].push_back(coef);
    else {
        if (pcol[i].back() < j)
            pcol[i].push_back(j); pcoef[i].push_back(coef);
        else {
            std::list<double>::iterator iv = pcoef[i].begin();
            std::list<int>::iterator it = pcol[i].begin();
            for (; it != pcol[i].end(); ++it, ++iv) {
                if ((*it) > j)
                    pcol[i].insert(it, j); pcoef[i].insert(iv, coef);
            }
        }
    }
}
Acsr.ptrow[0] = 0;
int k = 0;
for (int i = 0; i < n; i++) {
    ptrow[i + 1] = ptrow[i] + pcol[i].size();
    std::list<double>::iterator iv = pcoef[i].begin();
    std::list<int>::iterator it = pcol[i].begin();
    for (; it != pcol[i].end(); ++it, ++iv, k++) {
        Acsr.pcol[k] = (*it); Acsr.coef[k] = (*iv); } }
```

## conversion of Sparse matrix format CSR to COO

```
int nrow = Acsr.n;
for (int i = 0; i < n; i++) {
    for (int k = Acsr.ptrow[i]; k < Acsr.ptrow[i + 1]; k++) {
        Acoo.irow[k] = i;
        Acoo.jcol[k] = Acsr.indcol[k];
        Acoo.coef[k] = Acsr.coef[k];
    }
}
```

## exercise : implement GMRES/CG using IML++ : 1/2

GMRES is written by C++ template in IML++,

<https://math.nist.gov/iml++/gmres.h.txt>

first version without preconditioner const Preconditioner &M  
replacing Operator, Vector, and Matrix classes by simpler ones

- ▶ Real → double
- ▶ Vector → std::vector<double>
- ▶ Operator → structure CSRformat
- ▶ Matrix → std::vector<double> as one-dimensionalized array

```
std::vector<double> HH((max_iter + 1) * (max_iter + 1));  
#define H(i, j) HH((i) + (j) * (max_iter + 1))
```

matrix vector product to compute the residual as

```
r = b - A * x;
```

will be replaced by

```
std::vector<double> r(b);  
SparseGEMV(A, (-1.0), x, 1.0, r);
```

CG in IML++, <https://math.nist.gov/iml++/cg.h.txt>

- ▶ CSRformat for symmetric matrix with storing upper part
- ▶ `cblas_daxy` for  $\vec{x} += \alpha \times \vec{p}$

```
double cblas_daxpy(const int n, const double alpha,  
                    const double *x, const int incx,  
                    double *y, const int incy);
```

## exercise : implement GMRES/CG using IML++ : 2/2

```
template < class Operator, class Vector, class Preconditioner,
           class Matrix, class Real >
int
GMRES(const Operator &A, Vector &x, const Vector &b,
      const Preconditioner &M, Matrix &H, int &m, int &max_iter,
      Real &tol)
{
    Real resid;
    int i, j = 1, k;
    Vector s(m+1), cs(m+1), sn(m+1), w;

    Real normb = norm(M.solve(b));
    Vector r = M.solve(b - A * x);
```

→ without template

```
int
GMRES(csr_matrix &A, std::vector<double> &x, std::vector<double> &b,
       int &m, int &max_iter,
       double &tol)
{
    double resid;
    int nrow = x.size();
    int i, j = 1, k;
    std::vector<double> s(m+1, 0.0), cs(m+1), sn(m+1), w(nrow);
    std::vector<double> HH((max_iter + 1) * (max_iter + 1));

    double normb = norm(b);
    // r = Precond(A, b, x);
```