

High Performance Computing Project Report

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Abstract

In this report the work on my HPC-project will be explained and the results will be shown.

1 Assignment

Consider the transient convection-diffusion equation in 1D:

$$-D \frac{\partial^2 c}{\partial x^2} - v_x \frac{\partial c}{\partial x} = \frac{\partial c}{\partial t} \quad 0 \leq x \leq L, \quad t \geq 0$$

with boundary conditions:

$$\begin{cases} c(0, x) = \bar{c} \\ D \frac{\partial c}{\partial x} \Big|_{x=L} = \bar{q} \end{cases}$$

Discretization leads to the algebraic system:

$$(H + B)c + P \frac{dc}{dt} - q = 0,$$

where H, B, P are tridiagonal matrices.

In the 2D case the convection-diffusion equation becomes:

$$\frac{\partial}{\partial x} \left(D_x \frac{\partial c}{\partial x} \right) + \frac{\partial}{\partial y} \left(D_y \frac{\partial c}{\partial y} \right) - v_x \frac{\partial c}{\partial x} - v_y \frac{\partial c}{\partial y} = \frac{\partial c}{\partial t} + f$$

with boundary conditions:

$$\begin{cases} c(x, y, t) = \bar{c}, (t) & \forall (x, y) \in \partial\Omega_1 \\ D_x \frac{\partial u}{\partial x} n_x + D_y \frac{\partial u}{\partial y} n_y = \bar{q}, & \forall (x, y) \in \partial\Omega_2 \end{cases}$$

The local contributions of the element e on the entries h_{ij} of the stiffness matrix H and p_{ij} of the capacity matrix P read:

$$h_{ij}^{(e)} = \int_{\Delta_e} \left(D_x \frac{\partial \xi_i^{(e)}}{\partial x} \frac{\partial \xi_j^{(e)}}{\partial x} + D_y \frac{\partial \xi_i^{(e)}}{\partial y} \frac{\partial \xi_j^{(e)}}{\partial y} \right) dx dy = \frac{D_x b_i b_j + D_y c_i c_j}{4\Delta_e}$$
$$p_{ij}^{(e)} = \int_{\Delta_e} \xi_i^{(e)} \xi_j^{(e)} dx dy = \begin{cases} \Delta_e/6 & i = j \\ \Delta_e/12 & i \neq j \end{cases}$$

where b_i, b_j, c_i e c_j are computed using the triangle e coordinates. If i is a node belonging to the Neumann boundary $\partial\Omega_2$, the contribution to the i -th right-hand side component arises from (4):

$$q_i^{(e)} = \int_{\partial\Omega_2^{(e)}} \bar{q} \xi_i^{(e)} dS = \bar{q} \frac{l^{(e)}}{2}$$

with $l^{(e)}$ the length of the side a triangle belonging to $\partial\Omega_2$ and the node i as vertex.

The b_{ij} entry of the convective matrix B is given by:

$$b_{ij} = \int_{\Omega} \left(v_x \frac{\partial \xi_j}{\partial x} + v_y \frac{\partial \xi_j}{\partial y} \right) \xi_i dx dy$$

To solve this problem two main tasks need to be handled:

1. The assembly of the linear system given a predefined mesh structure.
2. The efficient solution of the large linear system obtained

The primary objective of this report is to highlight the implementation of parallel programming paradigms as the preferred solution for these tasks.

2 Implementation Framework

2.1 Used OpenMP Paradigms

OpenMP is a programming model that helps in parallelizing tasks on shared memory architectures by dividing the work into smaller tasks that can be executed concurrently by multiple threads.

The number of threads can be set in the terminal with command:

```
export OMP_NUM_THREADS=n
```

2.1.1 for

The `for` paradigm is used to parallelize loops. It divides the iteration between the available threads.

```
#pragma omp parallel for
for (int i = 0; i < n; i++) {
    // do something
    // ...
}
```

2.1.2 schedule

The `schedule` command is used with `for` loops to decide how threads are scheduled. In this project, we use the `dynamic` scheduling approach as shown in the following example. Using dynamic scheduling helps reduce the amount of time threads spend waiting. The formula $n/1000+1$ determines the size of each chunk. After a thread finishes processing one chunk, it receives a smaller chunk to prevent a situation where all threads have to wait for one thread to finish.

```
#pragma omp parallel for schedule(dynamic, n/1000+1)
for (int i = 0; i < n; i++) {
    // do something
    // ...
}
```

2.1.3 reduction

`reduction` is used for reducing a simple operation, such as summation or multiplication, on a variable over multiple threads. Pairwise operations will be done such that the number of total operations is reduced by half each step.

```
int sum = 0;
#pragma omp parallel for reduction(+:sum)
for (int i = 0; i < n; i++) {
    sum += array[i];
}
```

2.1.4 atomic

`atomic` ensures that a special location in the memory is only accessed by one thread at a time to prevent data race. This leads to potential performance reduction as the threads may need to wait to access the desired memory location.

```
#pragma omp parallel for
for (int i = 0; i < n; i++) {
    #pragma omp atomic
    counter++;
}
```

2.1.5 critical

The clause `critical` makes sure that only one thread enters a special section in the code. This also reduces performance because threads may need to wait for executing the `critical` code.

```
#pragma omp parallel for
for (int i = 0; i < 10; i++) {
    // do something
    // ..
    #pragma omp critical
    {
        // do something only one thread is allowed once at a time
    }
}
```

2.2 Data Structures and Basic Functions

2.2.1 The Compressed Sparse Row Format (CSR)

The CSR format is widely used representation to store large sparse matrices in an efficient way and minimize the needed memory.

In the project, we only deal with square matrices. That's why, for the explanations, we only consider square matrices and look at an $n \times n$ matrix with n_{term} entries.

The standard approach in storing a sparse matrix would be storing three lists containing the column and row index and the value of the non-zero elements. This leads to a memory occupation of `n_term(2 * sizeof(int) + sizeof(double))`

As shown in Table 1 the CSR format also has three lists but needs `n_term - n` fewer elements. The larger the matrix and the more non-zero elements the matrix get this format gets more efficient.

Name	Data Type	Length	Explanation
values	double*	n_term	Array storing the non-zero values of the matrix
column_indices	int*	n_term	Array storing the column indices of the corresponding non-zero values
row_pointers	int*	nrow + 1	Array storing the indices where each row starts in the values and column_indices arrays last index for these arrays

Table 1: Arrays for CSR Format

2.2.2 The Sparse Class

The CSR format was utilized for creating an own data structure to store and manage large sparse matrices in the code all attributes and methods of this class are shown in Table 2. In the following special functions and their parallelization will be explained.

Matrix-vector Product postMV computes the matrix-vector product $Av = y$ in a parallel and efficient way.

Explanation:

- outer loop iterates over each row as each row gets multiplied with the vector
- inner loop iterates from the first element in the row `iat[i]` to the last element in the row `iat[i]-1`
- This is embarrassingly parallel as the rows have no relation as well as the saved component for each `y`

Code:

```
void post_MV(double* v, double* y ){
    #pragma omp parallel for \
        schedule(dynamic,nrow/1000+1)
    for(int i=0; i<nrow; i++){
        double sum = 0.0;
        // #pragma omp simd reduction(+:sum)
        for(int j=iat[i]; j<iat[i+1]; j++){
            sum += coef[j]*v[ja[j]];
        }
        y[i] = sum;
    }
}
```

Class Name	sparse	
Private Members		
nrow	int	Number of rows
ncol	int	Number of columns
n_term	int	Number of non-zero matrix elements
coef	double*	Pointer to non-zero matrix elements
iat	int*	Pointer to the first non-zero column index
ja	int*	Pointer to the column indices
Public Members		
sparse()	sparse	Default constructor
sparse(int nrow, int ncol, int n_term)	sparse	Parameterized constructor
~sparse()	void	Destructor
Get and Set Methods		
full2sparse(double** A, int n_row)	void	Convert full matrix to sparse
addition_update(sparse* B_ptr)	void	Addition and update operation
scalarMult(double alpha)	void	Scalar multiplication
post_MV(double* v, double* y)	void	Matrix-vector multiplication
pre_MV(double* v, double* y)	void	Vector-matrix multiplication
matrixProduct(sparse* A_ptr, sparse* result)	void	Matrix product
diag(double* v)	void	diagonal submatrix D
getJacobi(double* v)	void	Inverse of the Diagonal matrix D^{-1}
diag_x_sparse(double* diag)	void	Diagonal matrix times sparse
left_Jacobi()	double*	Left Jacobi preconditioning $D^{-1}A$
printMat()	void	Print matrix
printComponents()	void	Print components

Table 2: UML Diagram of the **sparse** Class

Jacobi Preconditioning Computes the Matrix multiplied by the Jacobi preconditioner from the left side. The left side multiplication utilizes the CSR format because it is easier to iterate over each row. This is done by four methods.

Explanation:

- the diagonal matrix is represented as an 1-D array, of which all values were set to zero before function call
 - loops are the same as in matrix-vector product
 - stop iterating in the row when diagonal index is met
 - also embarrassingly parallel as the rows have no relation as well as the saved component for each v
-
- compute inverse of diagonal matrix D
 - prevent division by zero and set diagonal element to one s.t. no change in that row will take place
 - count the zeros as the effectiveness of Jacobi reduces with increasing number of zeroes
-
- loops are the same as in Matrix-vector product
 - simplified matrix product because of diagonal structure
 - every element in row i is multiplied by $v[i]$
-
- allocate array with guaranteed zeroes in every entry
 - compute the preconditioned matrix

Code:

```
void diag(double* v) {
    int start, end;
    #pragma omp parallel for \
        schedule(dynamic, ncol/1000 + 1)
    for(int i = 0; i < ncol; i++) {
        start = iat[i];
        end = iat[i + 1];
        for(int j = start; j < end; j++) {
            if (ja[j] == i) {
                v[i] = coef[j];
                break;
            }
        }
    }
}

void getJacobi(double* v) {
    this->diag(v);
    int zeros = 0;
    #pragma omp parallel for \
        schedule(dynamic, ncol/1000 + 1)
    for(int i = 0; i < ncol; i++) {
        if (v[i] == 0) {
            v[i] = 1;
            #pragma omp atomic
            zeros++;
        }
        v[i] = 1 / v[i];
    }
    if (zeros > 0) {
        cout << "WARNING: Number of zeros\\
in diagonal: " << zeros << endl;
    }
}

void diag_x_sparse(double* diag) {
    int start, end;
    #pragma omp parallel for \
        schedule(dynamic, ncol/1000 + 1)
    for(int i = 0; i < ncol; i++) {
        start = iat[i];
        end = iat[i + 1];
        for(int j = start; j < end; j++) {
            coef[j] *= diag[i];
        }
    }
}

double* left_Jacobi() {
    double* v = (double*)calloc(ncol, sizeof(double));
    this->getJacobi(v);
    this->diag_x_sparse(v);
    return v;
}
```

2.2.3 Basic Linear Algebra Operations

In the following basic functions for standard operations are described. Some others are equivalent and can be found in the complete code.

Scalar Product calculate the standard scalar product of two vectors.

Explanation:

- perfect use for the reduction clause
- reduction does not need an atomic statement
- embarrassingly parallel

Code:

```
double scalar_prod(double *v1, double *v2, int nr){
    double sum = 0.0;
    #pragma omp parallel for reduction(+:sum)
    for(int i = 0; i < nr; i++){
        sum += v1[i] * v2[i];
    }

    return sum;
}
```

Vector Update Add a vector onto another and scale it with a factor.

Explanation:

- standard parallel for loop
- embarrassingly parallel

Code:

```
void vector_update(double *v1, double *p,\\
                  double alpha, int nr) {
    #pragma omp parallel for \\
        schedule(dynamic, nr/1000+1)
    for (int i = 0; i < nr; ++i) {
        v1[i] += alpha * p[i];
    }
}
```

Upper Triangular System Solver Solving an upper triangular system using backward substitution is straightforward. However, when working with large systems, efficient parallelization row-by-row is difficult due to the dependency on solving the rows below first. The basic operation is

$$x[\text{row}] = (b[\text{row}] - \sum_{\text{col}=\text{row}+1}^n A[\text{row}][\text{col}] \cdot x[\text{col}]) / A[\text{row}][\text{row}]$$

To parallelize this process, the reduction operator can be used as the summation grows with each iteration of row. However, it needs to be evaluated if the size of the system being solved is large enough to actually improve the performance using parallelization.

Code:

```
void triangularSolver(double** A, double* b,\\
                    double* x, int n) {
    for (int row = n - 1; row >= 0; row--) {
        double sum = 0.0;
        #pragma omp parallel for reduction(+:sum)
        for (int col = row + 1; col < n; col++) {
            sum += A[row][col] * x[col];
        }
        x[row] = (b[row] - sum) / A[row][row];
    }
}
```

Maximum Finding the maximum value in an array utilizing parallelization can be achieved by using the `critical` directive to prevent data races. This is necessary because other threads may have modified the `maxVal` variable in the meantime. The code snippet provided demonstrates the usage of `critical` to ensure that the comparison `v[i] > maxVal` is still valid before updating `maxVal` if needed.

```
double max(double* v, int n) {
    double maxVal = v[0];
    #pragma omp parallel for \
        schedule(dynamic, n/1000+1)
    for (int i = 1; i < n; i++) {
        if (v[i] > maxVal) {
            #pragma omp critical
            {
                if (v[i] > maxVal) {
                    maxVal = v[i];
                }
            }
        }
    }
    return maxVal;
}
```

3 Assembly

3.1 Mesh Data

For the solution we used a non overlapping triangulation of Ω with triangular finite elements. The input data is shown in Table 3.

File Name	File containing	Input Representation
mesh	number of elements ne the triangle topology	800
		1 23 22
		22 44 43
		43 65 64
		...
coord	number of nodes nn the node coordinates	441
		0.0 1.00
		0.0 0.95
		0.0 0.90
		...
bound	number of dirichlet nodes nb the set of Dirichlet nodes and their values	80
		1 1.0
		2 1.0
		3 1.0
		...

Table 3: Description of the input files describing the mesh.

3.2 Nonzero Structure of the System Matrix

Determining the nonzero structure of a matrix is essential before further computations can proceed, especially in the CSR format. Changing a zero to a nonzero value is not feasible without storing a new matrix. A matrix entry, denoted as $H_{i,j}$, is considered nonzero if and only if two finite elements share nodes i and j .

The mesh topology input allows for an easy derivation of this pattern. We construct the adjacency matrix A with **nn** rows and **ne** columns, where each element will have a 1 at every node that is part of it.

The final pattern of the system matrix is then received by the product: $H = A^T A$. This is shown in the code below.

```
void stiffness_struct(char* fname, int ne, int nn, double** topol) {
    double** Adj = mat_allocation(ne, nn);
    double* i_idx = (double*)malloc(static_cast<int>(nn * nn * 0.2));
```

```

double* j_idx = (double*)malloc(static_cast<int>(nn * nn * 0.2));

int idx;
#pragma omp parallel for schedule(dynamic, nn / 1000 + 1)
for (int i = 0; i < ne; i++) {
    for (int k = 0; k < 3; k++) {
        idx = static_cast<int>(floor(topol[i][k]) - 1);
        Adj[i][idx] = 1;
    }
}

double sum;
int nnz = 0;
std::string tempFile = "temp_file.txt";
std::ofstream file(tempFile, std::ofstream::trunc); // Open the temporary file in write mode with truncation
for (int i = 0; i < nn; i++) {
    for (int j = 0; j < nn; j++) {
        sum = 0;
        for (int k = 0; k < ne; k++) {
            sum += Adj[k][i] * Adj[k][j];
        }
        if (sum > 0) {
            nnz++;
            file << j + 1 << " " << i + 1 << " " << 0 << std::endl;
        }
    }
}
file.close();
}

```

This technique works perfectly fine for small meshes, but with growing meshsize the performance and memory consumption are the reason why this is not working.

Therefore another function with way less memory consumption and operations was implemented as shown below.

In this version the creation of the adjacency matrix is omitted and the indices are saved directly to be ordered afterwards and then added to the CSR matrix structure directly. This could be parallelized with only adjusting vector containers by using the `atomic` operator or creating subvectors and merging them after the loop, it would be solved.

This approach lead to unsolvable errors and was not continued because of time consumption.

```

void stiffness_struct_small_loops(sparse* H, int ne, int nn, double** topol) {
    // Create vectors to store the i_idx and j_idx pairs
    std::vector<int> i_idx;
    std::vector<int> j_idx;

    // Loop through each element
    for (int i = 0; i < ne; i++) {
        for (int k = 0; k < 3; k++) {
            int idx = static_cast<int>(floor(topol[i][k]) - 1);
            for (int l = 0; l < 3; l++) {
                int jdx = static_cast<int>(floor(topol[i][l]) - 1);
                i_idx.push_back(idx);
                j_idx.push_back(jdx);
            }
        }
    }

    // Combine i_idx and j_idx into pairs for sorting
    std::vector<std::pair<int, int>> pairs;
    for (int i = 0; i < i_idx.size(); i++) {
        pairs.push_back(std::make_pair(i_idx[i], j_idx[i]));
    }

    // Sort the pairs to bring duplicates together
    std::sort(pairs.begin(), pairs.end());
}

```



```

// Remove duplicates and copy the unique pairs back to i_idx and j_idx
int nnz = 0;
for (int i = 0; i < pairs.size(); i++) {
    if (i == 0 || pairs[i] != pairs[i - 1]) {
        i_idx[nnz] = pairs[i].first + 1;
        j_idx[nnz] = pairs[i].second;
        nnz++;
    }
}

// Allocate memory for the CSR arrays
int* ja_loc = new int[nnz];
int* iat_loc = new int[nn + 1];

// Copy the data from i_idx and j_idx to the CSR arrays
for (int i = 0; i < nnz; i++) {
    ja_loc[i] = j_idx[i];
}

// Compute the CSR arrays from i_idx and ja_loc
irow2iat(nn, nnz, &i_idx[0], iat_loc);

H->set_n_term(nnz);
H->set_ncol(nn);
H->copy_ja(ja_loc);
H->copy_iat(iat_loc);

delete[] ja_loc;
delete[] iat_loc;
}

```

3.3 Computation of the Right Hand Side

The right hand side is composed of the forcing function f and the boundary condition that is also given as input file. In this report the forcing function is zero always.

3.4 Boundary Condition Enforcement

The boundary condition is enforced with the penalty method with the value 10^{15}

```

void imposeBC(sparse* H, double** bound, double* q, int nb){
    double R = 1e15;
    int j, bound_idx, start, end;
    int* iat = H->get_iat();
    int* ja = H->get_ja();
    double* coef = H->get_coef();

    for(int i = 0; i < nb; ++i){
        bound_idx = static_cast<int>(bound[i][0])-1;
        start = iat[bound_idx];
        end = iat[bound_idx+1];

        for(j = start; j < end; j++){
            if (ja[j] == bound_idx){break;}
        }

        coef[j] = R;
        q[bound_idx] = R*bound[i][1];
    }
}

```

4 Generalized Minimal Residual Algorithm

GMRES is an iterative method to solve linear systems of equations. The solution is approximated by the Krylov subspace with minimal residual.

4.1 The Householder Version

The orthogonalization in the Arnoldi process is commonly performed using the Gram-Schmidt procedure, which can suffer from numerical instability. However, an alternative technique known as Householder projections provides improved numerical stability for the orthogonalization step. The Householder projection matrix is defined as $P = I - 2u^T u$, where u is a unit vector. This definition can be utilized in the GMRES algorithm as only the vector v needs to be stored to define a P_i as well as $P_i v = v - u^T u v$ where the matrix-vector product reduces to simple vector operations.

In Algorithm 1 GMRES with Householder projections is presented taken from [Saad, 2003], who developed this method in 1986.

Algorithm 1: GMRES Algorithm with Householder projections

Input : A, b, m, x_0
Output: x_m

```

1 Compute  $r_0 = b - Ax_0, z = r_0$ ;
2 for  $j = 1, \dots, m, m+1$  do
3   Compute the Householder unit vector  $w_j$  such that
4    $w_j(i) = 0, i = 1, \dots, j-1$  and  $(P_j z)(i) = 0, i = j+1, \dots, n$ 
5   where  $P_j = I - 2w_j w_j^T$ ;
6    $h_{j-1} = P_j z$ ;
7   if  $j = 1$  then
8     Let  $\beta = e_1^T h_0$ ;
9   end
10   $v = P_1 P_2 \dots P_j w_j$ ;
11  if  $j \leq m$  then
12    Compute  $z = P_j P_{j-1} \dots P_1 A v$ ;
13  end
14 end
15 Define  $\bar{H}_m$  as the  $(m+1) \times m$  upper part of the matrix  $[h_1, \dots, h_m]$ ;
16 Compute  $y_m = \text{Argmin}_y \|\beta e_1 - \bar{H}_m y\|_2$ ;
17 Let  $y_m = (\eta_1, \eta_2, \dots, \eta_m)^T$ ;
18  $z = 0$ ;
19 for  $j = m, m-1, \dots, 1$  do
20    $z = P_j(\eta_j e_j + z)$ ;
21 end
22 Compute  $x_m = x_0 + z$ ;
```

Givens Rotation Saad also suggests in his book to use Givens rotations to transform the linear system, that needs to be solved in line 16, from Householder structure to an upper triangular matrix making it easy to solve.

The rotation matrix Ω_i is defined as

$$\Omega_i = \begin{pmatrix} 1 & & & & & & \\ & \ddots & & & & & \\ & & 1 & & & & \\ & & & c_i & s_i & & \\ & & & -s_i & c_i & & \\ & & & & & 1 & \\ & & & & & & \ddots & \\ & & & & & & & 1 \end{pmatrix} \quad \text{with } c_i^2 + s_i^2 = 1 \text{ and } c_i \text{ and } s_i \text{ of the } i^{\text{th}} \text{ rotation}$$

$$s_i = \frac{h_{i+1,i}}{\sqrt{(h_{ii}^{(i-1)})^2 + h_{i+1,i}^2}}, \quad c_i = \frac{h_{ii}^{(i-1)}}{\sqrt{(h_{ii}^{(i-1)})^2 + h_{i+1,i}^2}}.$$

Applying the rotation matrices on \bar{H}_4 the transformed system is shown below.

$$\bar{H}_4 = \begin{pmatrix} h_{11} & h_{12} & h_{13} & h_{14} \\ h_{21} & h_{22} & h_{23} & h_{24} \\ & h_{32} & h_{33} & h_{34} \\ & & h_{43} & h_{44} \\ & & & h_{54} \end{pmatrix}, \quad \bar{g}_0 = \begin{pmatrix} \beta \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}.$$

$$\bar{H}_4^{(4)} = \begin{pmatrix} h_{11}^{(4)} & h_{12}^{(4)} & h_{13}^{(4)} & h_{14}^{(4)} \\ & h_{22}^{(4)} & h_{23}^{(4)} & h_{24}^{(4)} \\ & & h_{33}^{(4)} & h_{34}^{(4)} \\ & & & h_{44}^{(4)} \\ & & & & 0 \end{pmatrix}, \quad \bar{g}_4 = \begin{pmatrix} \gamma_1 \\ \gamma_2 \\ \gamma_3 \\ \gamma_4 \\ \gamma_5 \end{pmatrix}$$

Restarting Another suggestion from Saad is to apply restarting after some iterations to the GMRES algorithm. This is beneficial since the start vector improves by every restart and so does the generated Krylov subspace. Another improvement is that the triangular system to solve is relatively small as well as the matrices that store projections or the transformed Hessenberg matrix. For the standard GMRES the convergence is guaranteed after n iterations. This property gets lost with the restarted version, but since n is large this property is impractical.

The algorithm 2 is also taken from [Saad, 2003].

Algorithm 2: Restarted GMRES

Input : A, b, m, x_0

Output: x_m

- 1 Compute $r_0 = b - Ax_0$, $\beta = \|r_0\|_2$, and $v_1 = r_0/\beta$;
 - 2 Generate the Arnoldi basis and the matrix \bar{H}_m using the Arnoldi algorithm starting with v_1 ;
 - 3 Compute y_m which minimizes $\|\beta e_1 - \bar{H}_m y\|_2$;
 - 4 If satisfied, then Stop. Otherwise, set $x_0 := x_m$ and go to Step 1;
-

4.2 Implementation

In the following, parts of the implemented GMRES will be written down in pseudo-code and details of the implementation will be explained. When an operation is parallelized, it is mentioned once. Subsequent occurrences of the same operation are parallelized as well, without explicitly mentioning.

The restarted part will be omitted in this explanation since it is straightforward.

The full GMRES implementation can be viewed in the appendix Code A.2.

The input values can be viewed in table 4. The function does not have a return value the solution vector is stored in start vector from the input.

Name	Data Type	Explanation
z_ptr	sparse*	Pointer to the sparse matrix of the linear system to be solved
b	double*	Pointer to the right-hand side array
x	double*	Pointer to the start and solution array
preconditioning	bool	Boolean indicating whether preconditioning is used
restarting	bool	Boolean indicating whether restarting is used
maxit	int	Maximum number of iterations

Table 4: Input variables for the GMRES function.

Arnoldi process In Algorithm 3 the Arnoldi process with Householder projections of my implementation is presented and parts of the parallelization are mentioned. The parallelization can just be applied on single operations as there is data dependence on the outer loops or even the application of the projection matrices.

Givens Rotation In Algorithm 4 the givens rotation to obtain the triangular system is explained. In this part no real parallelization can be done because the operations are mostly on scalars and not Vectors that can be divided in parallel calculations.

Evaluation of x_i This part solves the triangular system and computes the approximation x_i of the solution vector. This part will be entered right after the Givens rotation if the process is stagnating or the norm smaller than the relative tolerance

Algorithm 3: Householder part

```
Input :  $A, x, b, U \in \mathbb{R}^{n \times inner}$ 
Output:  $x_m$ 
1 Compute  $r = b - Ax, u = r, normr = ||r||, \beta = \text{sig}(normr) \cdot normr, u = r$ 
2  $u_0 = u_0 + \beta, u = \frac{u}{||u||}, U_{0,:} = u, w_0 = -\beta$ ;
3 for  $j = 1, \dots, outer$  do
4   for  $i = 1, \dots, inner$  do
5      $v = -2u_i \cdot u^{(i)}$ ; // scalar multiplication done in parallel
6      $v_i += 1$ ; // just update one value
7     // do:  $v = P_1 \dots P_i e_i$ , but why ei?
8     for  $k = i - 1 \dots 0$  do
9        $v = v - 2u^{(k)T} u^{(k)} \cdot v$ ; // scalar product and vector update is parallelized
10    end
11     $v = \frac{v}{||v||}$ ; // normalization is parallel
12    // do:  $v = P_i \dots P_0 Av$ 
13     $v = Av$ ; // sparse Matrix-vector product is parallel
14    for  $k = 0 \dots i$  do
15       $v = v - 2u^{(k)T} u^{(k)} \cdot v$ ;
16    end
17    // compute  $u^{i+1}$  for  $P_{j+1}$ 
18     $u^{i+1} = v$ ;
19     $u_{0:inner} = 0$ ; // zero_out is parallel
20     $\alpha = ||u||$ ; // norm is parallel
21    if  $\alpha > 0$  then
22       $\alpha = v_{i+1} \alpha$ ;
23       $u_{i+1}^{(i+1)} = u_{i+1}^{(i+1)} + \alpha$ ;
24      // normalize  $u^{(i+1)}$ ;
25      // do  $v = P_{i+1} v$ 
26       $v_{i+1:n} = 0$ ;
27       $v_{i+1} = -\alpha$ ;
28    end
29    // Continue with Givens rotation...
30    // ...
31  end
32 end
```

$normr < tol_b$ or after the iteration of the inner loop.

With that the main outline of the GMRES is concluded. As mentioned the parallelization could just be applied on the single operations.

5 Numerical Experiments

For the experiments, a MacBook Pro 2019 with an Intel(R) Core(TM) i5-8279U CPU running at 2.40GHz is used. The system has 4 physical CPU cores and 8 logical CPU cores due to hyper-threading. Additionally, the system has 8 GB of RAM.

5.1 Performance Metrics

The following metrics are used to evaluate the performance of the parallelization in the implemented program.

5.1.1 Speedup

The speedup measures the improvement of the performance compared to the used threads. It compares the runtime of the program without parallelization divided by the runtime with parallelization.

$$S_p := \frac{T_{\text{one thread}}}{T_{p \text{ threads}}}$$

An optimal speedup is equal to the number of threads used, i.e. if 4 threads are used and the speed up is 4, the program runs 4 times faster which means the utilization of parallelization is optimal.

Algorithm 4: Givens rotation part

```
for  $j = 1, \dots, \text{outer}$  do
  for  $i = 1, \dots, \text{inner}$  do
    // ...
    // Previous code...
    // Continue with Givens rotation...
1    $j_0 = v_i$ ; // Store  $v_i$  in  $j_0$ 
2    $j_1 = v_{i+1}$ ; // Store  $v_{i+1}$  in  $j_1$ 
3    $\rho = \sqrt{j_0^2 + j_1^2}$ ; // Compute the norm
4    $j_0 = \frac{j_0}{\rho}$ ; // Normalize  $j_0$ 
5    $j_1 = \frac{j_1}{\rho}$ ; // Normalize  $j_1$ 
6    $J_{i,0} = j_0$ ; // Store the Givens rotation values
7    $J_{i,1} = j_1$ ; // Store the Givens rotation values
8    $w_{i+1} = -J_{i,1} \cdot w_i$ ; // Update  $w_{i+1}$ 
9    $w_i = J_{i,0} \cdot w_i$ ; // Update  $w_i$ 
10   $v_i = \rho$ ; // Update  $v_i$ 
11   $v_{i+1} = 0$ ; // Set  $v_{i+1}$  to 0
    //  $v$  is a row vector in  $R$  the matrix of the triangular system
    //  $w$  is the RHS of this system
12   $R_{:,i} = v$ ;
13   $\text{normr} = |w_{i+1}|$ ;
    // Following Code ...
    // ...
  end
end
end
```

5.1.2 Efficiency

The efficiency describes the utilization of parallelism in the program, i.e. the mean time a processor is really working.

$$E_p := \frac{S_p}{p} = \frac{T_1}{pT_p}$$

Following the definition an efficiency of 1 is optimal and means that the program is using the parallel resources 100% all the time.

5.1.3 Scalability

Scalability describes how well the program handles increasing workloads, in our case increased Matrix size or number of non zero elements. $r = \frac{n_{\text{increased}}}{n_{\text{reference}}}$ is the factor by which the data size increased.

$$SB_r = r \cdot \frac{T_{\text{reference}}}{T_{\text{increased}}}$$

If SB_r is greater than 1 it means the Performance decreases with increasing problem size and vice versa.

5.2 Performance of the GMRES

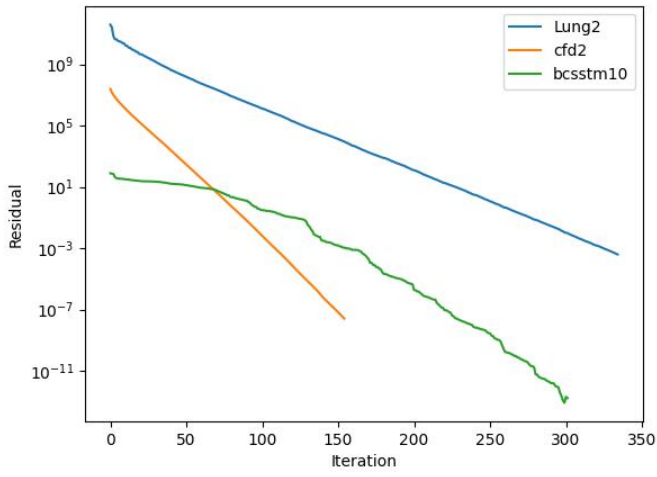
5.2.1 Convergence

For evaluating the quality of the implemented GMRES algorithm large matrices from [SuiteSparse Matrix Collection](#) were taken and their properties are shown in Table 5. For the right hand side of the linear systems a random vector with values between 0 and 1 with the suiting length is created.

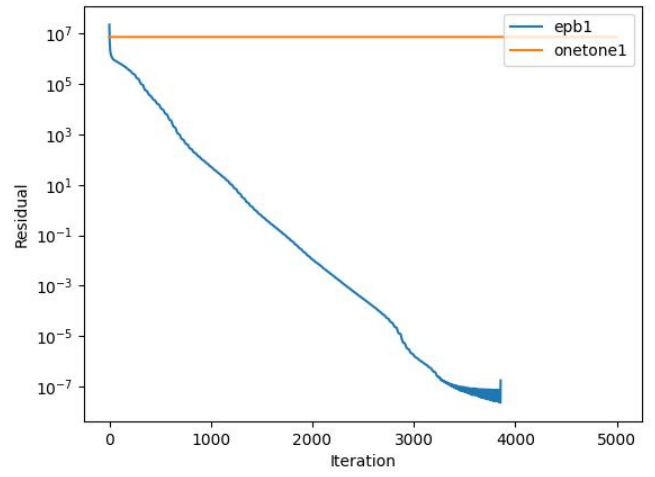
The convergence for the different test matrices can be seen in Figure 1. The positive symmetric matrices converge very fast and do not need many iterations with respect to their size as well as the relatively small *bcsstm10* matrix. Another interesting thing we can observe is that the convergence for the two SPD matrices follow a constant exponential rate and the convergence of *bcsstm10* is very irregular.

For *epb1* it takes many iterations to converge but it does. The oscilation at the end come from the described restarted version of GMRES. Comparing the sizes of *epb1* and *cf2* we see that the limiting factor in fast convergence is the non SPD property because the number of nonzeros is just $\frac{95053}{1605669} \approx 5.9\%$ compared to *cf2* that converges very fast.

GMRES is not capable of solving a linear system with *onetone1* as system matrix, an improved preconditioner would be needed to solve this system.



(a)



(b)

Figure 1: Convergence plots of GMRES for the test matrices. (a) shows fast convergences and (b) slow or no convergence of GMRES

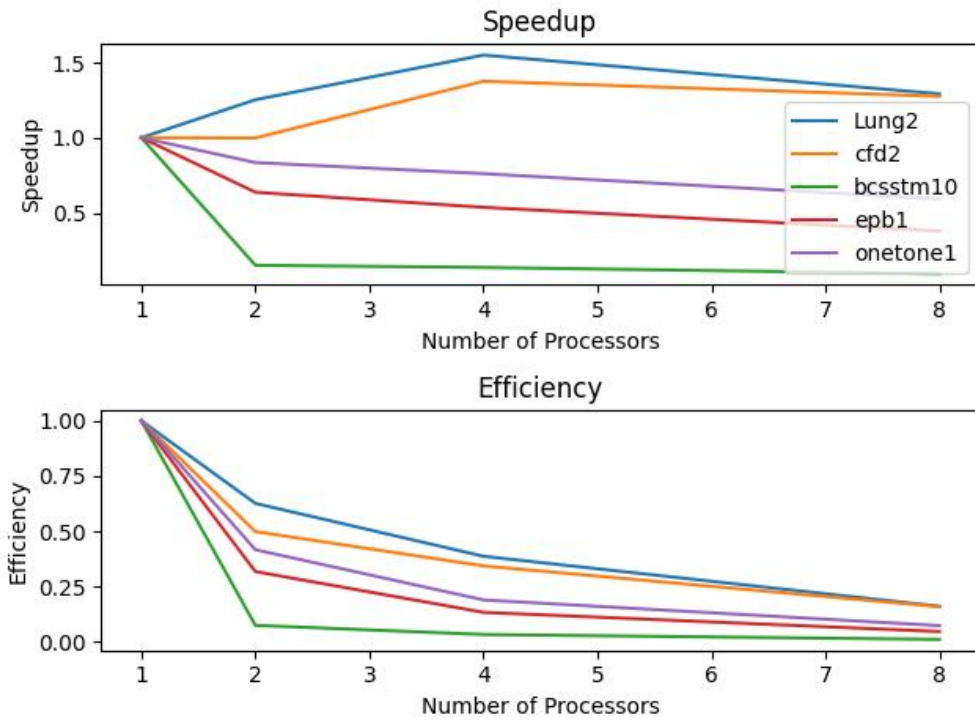


Figure 2: Speedup and Efficiency Metric for Test Matrices

Algorithm 5: Evaluating x_m

```
for  $j = 1, \dots, \text{outer}$  do
  for  $i = 1, \dots, \text{inner}$  do
    // ...
    // Previous code...
    // Continue with evaluation of  $x_m$ ...
1    $y = R^{-1}w$ ; // solve triangular system
    // Compute additive vector from  $x_m = x_0 + z$ , with  $z := P_j(\eta_j e_j + z)$ 
2    $z = -2u_i^{(i)} y_i u$ ;
3    $z_i = z_i + y_i$ ;
4   for  $k = i-1 \dots 0$  do
5      $z_k = z_k + y_k$ ;
6      $z = z - 2uu^T z$ ;
7   end
8    $x_i = x_0 + z$ ;
9    $r = b - Ax_i$ ;
    // Following Code ...
    // ...
  end
  // evaluation will be done here if not happened before
  // ...
end
```

Name	Dimension	Nonzeros	Condition Number
Symmetric Positive Definite (Spd) Matrices			
Lung2	109460	492564	-
cfd2	123440	1605669	-
Non-Symmetric Positive Definite Matrices			
bcstm10	1086	11589	1.258792×10^5
epb1	14734	95053	5.940657×10^3
onetone1	36057	335552	9.388846×10^6

Table 5: Test Matrices

5.2.2 Parallelization Performance

The runtimes for solving the linear system with the GMRES algorithm are shown in Table 6 and some performance metrics are displayed in 2.

For the SPD matrices *lung* and *cfd2* the runtime actually increases. This is the expected behaviour as the program has few steps where no parallelization can be applied and the datasize is very large (see Table 5). The speedup is still bad as the optimal speedup is equal to the number of used processors. Additionally, the runtime increases when 8 threads are used.

The non SPD matrices on the other side increase their runtime with increasing number of threads. This behaviour is inexplicable as the data size is also big and the parallelization overhead necessary to distribute the parallel processes on different threads should increase the runtime with such a large ratio. Additionally for comparable size the speedup is greater than 1 for the SPD matrices, but the influence on runtime should not depend on the property if SPD or not.

number of threads	1	2	4	8
Lung2	2.52803	2.01759	1.63059	1.95627
cfd2	1.50406	1.50657	1.09299	1.17903
bcsstm10	0.098761	0.65573	0.725824	1.04566
epb1	5.65513	8.87665	10.5441	14.9338
onetone1	15.0989	18.0937	19.8476	25.4774

Table 6: Execution times for different processes.

5.3 Solution of the Convection-Diffusion Equation

Consider the problem

$$\frac{\partial}{\partial x} \left(D_x \frac{\partial c}{\partial x} \right) + \frac{\partial}{\partial y} \left(D_y \frac{\partial c}{\partial y} \right) - v_x \frac{\partial c}{\partial x} - v_y \frac{\partial c}{\partial y} = \frac{\partial c}{\partial t} + f$$

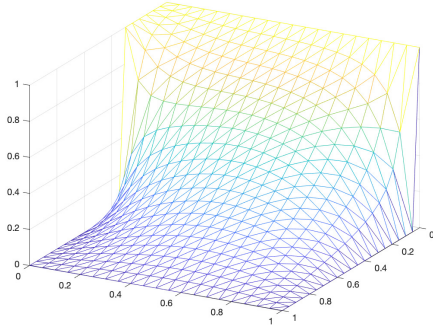
with the boundary conditions

$$\begin{aligned} c(x, t) &= [x \in \Omega_1] && \text{for } x \in \Omega_1, t > 0 \\ c(x, t) &= [x \in \Omega_2] && \text{for } x \in \Omega_2, t > 0 \end{aligned}$$

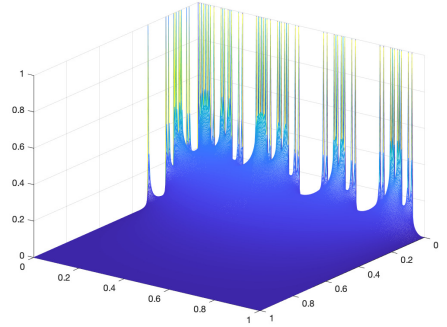
where the boundaries Ω_1 and Ω_2 are given by:

$$\begin{aligned} \Omega_1 &= \{(x, y) : [x = 0, 0 \leq y \leq 1] \cup [0 \leq x \leq 0.3, y = 0]\} \\ \Omega_2 &= \{(x, y) : [x = 1, 0 \leq y \leq 1] \cup [0 \leq x \leq 1, y = 1] \cup [0.3 \leq x \leq 1, y = 0]\} \end{aligned}$$

The correct solution of this problem on the two different meshes is shown in 3. The solutions for small and big mesh should look the same but there occurred an error in the mesh creation for the big mesh and the boundary conditions are not created correctly. Because of the small size of the problem, it is not necessary to look at the performance for parallelization. Unfortunately the evaluation for the big mesh could not be done because as mentioned the parallelization for the assembly which takes the most computational effort could not be resolved.



(a) Solution for the small mesh



(b) Solution for the big mesh

Figure 3: Solution of the Convection-Diffusion Problem.

References

[Saad, 2003] Saad, Y. (2003). *Iterative Methods for Sparse Linear Systems Second Edition*.

A C++ Files

A.1 main.cpp

```
1 #include <iostream>
2 #include <cmath>
3 #include <fstream>
4 #include <omp.h>
5 #include <cstring>
6
7 #include "sparse.h"
8 #include "read_mat.h"
9 #include "functions.h"
10 #include "GMRES.h"
11 #include "constants.h"
12 #include "fem_functions.h"
13
14
15 const int n_threads = 8;
16 using namespace std;
17
18 int main(int argc, char* argv[]) {
19     //allocate memory for the solution of the Matrix Vector prod
20     // char filename[] = "mesh/u_mattia.dat";
21
22     // sparse mat_mattia = sparse();
23     // double* b_mattia = getFEM_sparse(&mat_mattia);
24
25     // double *xstar_mattia = (double*)malloc(mat_mattia.get_nrow()*sizeof(double));
26
27     // // sparse* newmat = &mat_mattia;
28     // gmres(filename, &mat_mattia,b_mattia, xstar_mattia, true, true, 1000);
29
30     // save_solution(filename, xstar_mattia, mat_mattia.get_ncol());
31
32     // free(b_mattia);
33     // free(xstar_mattia);
34
35
36
37
38     // FEM assembly
39
40
41
42
43     // int num_threads = 4;
44     // omp_set_num_threads(num_threads);
45
46     // Enable dynamic threads to support task-based parallelism
47     // omp_set_dynamic(1);
48
49     // Get the actual number of threads being used by OpenMP
50     int actual_num_threads = omp_get_max_threads();
51     std::cout << "Actual number of threads: " << actual_num_threads << std::endl;
52
53
54     double start_time = omp_get_wtime();
55
56     double D[2] = {1,1};
57     double V[2] = {1,3};
58     int ne, nn, nb; // number of mesh (800), number of nodes(441), number of ? (80)
59
60     double **coord, **bound, *delta, *delta_node, *F;
61     double **topol, *temp, *q, *b;
62
63
64
```

```

65 sparse* FEM;
66 bool bigmesh = true;
67
68 char sol_path[100];
69
70
71 if (bigmesh){
72     string mesh_path = "bigmesh/mesh.dat";
73     string nodes_path = "bigmesh/dirtot.dat";
74     string coord_path = "bigmesh/xy.dat";
75
76     ifstream mesh(mesh_path);
77     ifstream nodes(nodes_path);
78     ifstream xy(coord_path);
79
80
81     xy >> nn;
82     nodes >> nb;
83     mesh >> ne;
84
85     bound = mat_allocation(nb,2);
86     coord = mat_allocation(nn,2);
87     topol = mat_allocation(ne,3);
88
89     for (int i = 0; i < nn; ++i){xy >> coord[i][0] >> coord[i][1];}
90     for (int i = 0; i < nb; ++i){nodes >> bound[i][0] >> bound[i][1];}
91     for (int i = 0; i < ne; ++i){mesh >> topol[i][0] >> topol[i][1] >> topol[i][2];}
92
93     xy.close();
94     nodes.close();
95     mesh.close();
96     char path[] = "bigmesh/u.dat";
97     strcpy(sol_path, path);
98
99
100 }else{
101     string mesh_path = "mesh/mesh.dat";
102     string nodes_path = "mesh/dirnod.dat";
103     string coord_path = "mesh/xy.dat";
104
105
106     ifstream mesh(mesh_path);
107     ifstream nodes(nodes_path);
108     ifstream xy(coord_path);
109
110     int not_used;
111
112     xy >> nn;
113     nodes >> nb;
114     mesh >> ne;
115
116
117     bound = mat_allocation(nb,2);
118     coord = mat_allocation(nn,2);
119     topol = mat_allocation(ne,3);
120
121     for (int i = 0; i < nn; ++i){xy >> coord[i][0] >> coord[i][1];}
122     for (int i = 0; i < nb; ++i){nodes >> bound[i][0] >> bound[i][1];}
123     for (int i = 0; i < ne; ++i){mesh >> topol[i][0] >> topol[i][1] >> topol[i][2]>>
        not_used;}
124
125
126     for (int i = 0; i < nn; ++i){xy >> coord[i][0] >> coord[i][1];}
127     for (int i = 0; i < nb; ++i){nodes >> bound[i][0] >> bound[i][1];}
128     for (int i = 0; i < ne; ++i){mesh >> topol[i][0] >> topol[i][1] >> topol[i][2];}
129
130     xy.close();
131     nodes.close();
132     mesh.close();

```

```

133     char path[] = "mesh/u.dat";
134     strcpy(sol_path, path);
135 }
136
137
138 //read data
139
140
141
142 // printMat(coord,nb,2);
143
144
145
146 double h = sqrt(2)*(coord[0][1]-coord[1][1]); // why sqrt2
147 double tau = 0.1; // whats tau?
148
149 delta = (double*)malloc(ne*sizeof(double));
150
151 delta_node = (double*)calloc(nn,sizeof(double));
152
153 F = (double*)malloc(nn*sizeof(double));
154 q = (double*)calloc(nn,sizeof(double));
155 b = (double*)malloc(nn*sizeof(double));
156
157
158 FEM = create_FEM_mat(coord, topol, ne, nn, D, h, tau, V, delta);
159
160 int idx;
161 #pragma omp for schedule(dynamic,nn/1000+1)
162 for(int i = 0; i<ne; i++){
163     for(int k=0; k<3; k++){
164         idx = static_cast<int>(floor(topol[i][k])-1);
165         delta_node[idx] += delta[k]/3;
166     }
167 }
168
169 // RHS
170 for(int i = 0; i<nn; i++){
171     F[i] = force(coord[i][0], coord[i][1]*delta_node[i]);
172 }
173
174 // BCs
175 imposeBC(&FEM[0], bound, q, nb);
176 FEM[1].addition_update(&FEM[2]);
177 FEM[0].addition_update(&FEM[1]);
178 // printVec(FEM[0].get_coef(),FEM[0].get_n_term());
179 double* x = (double*)calloc(FEM[0].get_ncol(),sizeof(double));
180
181
182 vector_update(q,F,1,nn);
183 // printVec(q,nn);
184
185 // FEM Solver
186 cout<< norm(q,nn)<<endl;
187 sparse* A = new sparse;
188 // b = getAssembledSystem(A);
189
190 // double* x = (double*)calloc(A->get_ncol(),sizeof(double));
191 // FEM[0].printComponents();
192
193
194 gmres(sol_path,&FEM[0],q, x, true, true, 1000);
195 double end_time = omp_get_wtime();
196 double runtime = (end_time - start_time);
197
198 save_solution(sol_path,x,FEM[0].get_ncol());
199
200
201 free(b);

```

```

202     free(x);
203     delete A;
204
205     return 0;
206
207 }

```

A.2 GMRES.h

```

1 void gmres(char* matname, sparse* A_ptr, double* b, double* x, bool preconditioning, bool
  restarting, int maxit){
2     /*
3     GMRES Algorithm with
4         - Householder projections
5         - Givens rotation
6
7     Input:
8         - A_ptr: pointer to sparse Matrix of the linear system to be solved
9         - b: ptr to rhs
10        - x: ptr to solution vector (and start vector?)
11
12
13
14
15
16    Flag is set to see the status of the Algorithm
17    flag = 0: succesful convergence
18        - inner loop: rormr_act < tol b
19        - outer loop: rormr_act < tol b after the last evaluation
20    flag = 1: maximum iterations reached without convergence
21        - initialization if it stays unchanged: max iterations reached
22    flag = 2: stagnation
23        - no significant change in norm reduction
24
25    */
26    sparse& A = *A_ptr;
27
28    int nc = A.get_ncol();
29    int nr = A.get_nrow();
30
31    if(nc!=nr){
32        cout<<"ERROR need square Matrix n != n"<<endl;
33        return;
34    }
35
36    const int n = nc;
37
38    double make_relres_smaller=1;
39    if (preconditioning){
40        double* M_inv = A.left_Jacobi();
41        elementwise_prod(b,M_inv,n);
42    }else{
43        make_relres_smaller = 1.0e-15;
44    }
45
46    int restart, outer;
47    if(restarting){
48        restart = 20;
49        outer = maxit/restart;
50        maxit = restart; // 10 is small
51    }
52    int inner = maxit;
53
54
55
56
57    double tol = 1e-15;

```

```

58     double n2b = norm(b,n);
59
60
61     int flag = 1;
62
63     int imin = 0;
64     int jmin = 0;
65     double tol_b = tol*n2b*make_relres_smaller;
66
67     int stag = 0;
68     int moresteps = 0;
69     int maxmsteps = 3;
70     int maxstagsteps = 3;
71
72
73     double* r = (double*)malloc((n)*sizeof(double));
74     double* Ax = (double*)malloc((n)*sizeof(double));
75     double* xm = (double*)malloc((n)*sizeof(double));
76     double* Axm = (double*)malloc((n)*sizeof(double));
77
78     A.post_MV(x,Ax);
79     vector_add(1.,b,-1.,Ax,r,n);
80     double normr = norm(r, n);
81
82     double* resvec = (double*)malloc((inner*outer+1)*sizeof(double));
83     resvec[0] = normr;
84     double normrmin = normr;
85
86     // preallocate
87
88     // J is of size (inner x 2)
89     // Values used in one iteration are in the same array
90     // working direction J[row][i] ist ja doch die falsche working direction oder
91     double** J = (double**)malloc(inner * sizeof(double*)); // Matrix for givens rotation J(2,
92     // inner)
93     for (int i = 0; i < inner; ++i) {
94         J[i] = (double*)malloc(2 * sizeof(double));
95     }
96
97     // U is of size (inner x n)
98     // Values used in one iteration are in the same array
99     // working direction U[row][i] = i;
100    double** U = (double**)malloc((inner+1) * sizeof(double*)); // Matrix for holding
101    // Householder reflectors u(w) U(n,inner)
102    for (int i = 0; i < (inner+1); ++i) {
103        U[i] = (double*)malloc(n * sizeof(double));
104    }
105
106
107    double* u = (double*)malloc(n * sizeof(double)); // row vector of u as helper;
108
109    double** R = (double**)malloc(inner * sizeof(double*)); // givens rotated Hessenbergmatrix
110    // R(inner,inner)
111    for (int i = 0; i < inner; ++i) {
112        R[i] = (double*)malloc(inner * sizeof(double));
113    }
114
115    double* w = (double*)malloc((n+1)*sizeof(double));
116
117    double* v = (double*)malloc((n+1)*sizeof(double));
118    double* vtemp = (double*)malloc((n+1)*sizeof(double));
119
120    double* jtemp = (double*)malloc(2*sizeof(double)); // there to save v(initer:initer+1)
121
122    double beta;
123    double alpha;

```

```

124 double relres; //relative residual
125
126 int initer; // so that they live outside the loop
127 int outiter;
128 int iterations = 0;
129 int idx;
130
131 double* y = (double*)malloc((inner+1)*sizeof(double));
132 double* additive = (double*)malloc(n*sizeof(double));
133
134
135 double start_time = omp_get_wtime();
136
137 normr = norm(r,n);
138 beta = scalarsgn(normr)*normr;
139 copyArray(u, r, n);
140 u[0] += beta;
141 normalize(u, n);
142
143 copyArr2Mat_col(U,u,0,n);
144 // Apply Householder projection to r.
145 // w = r - 2*u*u'*r;
146 w[0] = -beta;
147
148 for (outiter = 0; outiter < outer; outiter++){
149     for(initer=0; initer<inner; initer++){
150         iterations = (outiter)*inner+initer+1;
151         // copyArray(v, u, n); // wofuer das
152         vector_scalarMult(v, u, -2.0*u[initer], n);
153         v[initer] += 1;
154
155         // v = P1*P2*...Pjm1*(Pj*ej)
156         for(int k = initer - 1; k >= 0; k--){
157             // copyMat_col2Arr(U, u, k, n);
158             // vector_update(v, u, -2*scalar_prod(u,v,n), n);
159             vector_update(v, U[k], -2*scalar_prod(U[k],v,n), n);
160
161         }
162
163         normalize(v, n);
164         copyArray(vtemp, v, n);
165
166         A.post_MV(vtemp, v);
167
168         // Form Pj*Pj-1*...*P1*Av
169         for(int k=0; k<=initer; k++){
170             // copyMat_col2Arr(U, u, k, n);
171             // vector_update(v, u, -2*scalar_prod(u,v,n), n);
172             vector_update(v, U[k], -2*scalar_prod(U[k],v,n), n);
173
174         }
175
176         // determine Pj+1
177         if(initer!= n-1){ // if not last iteration // not necessary in restarted
178             version
179             // construct u for Pj+1
180             copyArray(u, v, n);
181             vector_zero_out(u, 0, initer, n);
182
183             alpha = norm(u,n);
184
185             if (alpha > eps){
186                 alpha *= scalarsgn(v[initer+1]);
187                 //u = v(initer+1:end) + sign(v(initer+1))*||v(initer+1:end)||*e_{initer
188                     +1)
189                 u[initer+1] += alpha;
190
191                 normalize(u, n);

```

```

191         copyArr2Mat_col(U, u, initer+1, n);
192
193         // Apply Pj+1 to v
194         vector_zero_out(v, initer+1, n-1, n);
195         v[initer+1] -= alpha;
196     }
197 }
198
199
200
201 for(int colJ = 0; colJ<initer; colJ++){
202     double tmpv = v[colJ];
203     v[colJ] = J[colJ][0]*v[colJ] + J[colJ][1]*v[colJ+1];
204     v[colJ+1] = -J[colJ][1]*tmpv + J[colJ][0]*v[colJ+1];
205 }
206
207
208 // compute Given's rotation Jm
209 if (initer!= n-1){
210     jtemp[0] = v[initer];
211     jtemp[1] = v[initer+1];
212
213     double rho = norm(jtemp,2);
214     normalize(jtemp,2);
215     copyArr2Mat_col(J,jtemp,initer,2);
216
217     w[initer+1] = -J[initer][1]*w[initer];
218     w[initer] *= J[initer][0];
219
220     v[initer] = rho;
221     v[initer+1] = 0;
222 }
223
224
225 copyArr2Mat_row(R,v,initer,inner);
226
227 normr = scalarsgn(w[initer+1])*w[initer+1]; // abs(w[initer+1])
228 resvec[iterations] = normr;
229
230
231 if (normr <= tolB || stag >= maxstagsteps || moresteps){
232     // normr smaller than relative tolerance
233     // stagnation larger than allowed
234     // if moresteps set to 1 to allow further improvement
235     // one of these is true: allow to compute the solution vector x
236
237     vector_zero_out(additive,n);
238     vector_zero_out(y,initer+1);
239
240     triangularSolver(R,w,y,initer+1);
241     copyMat_col2Arr(U, u, initer, n);
242     vector_scalarMult(additive,u,-2*y[initer]*U[initer][initer],n);
243     additive[initer] += y[initer];
244     for(int k=initer-1;k>=0;k--){
245         // copyMat_col2Arr(U, u, k, n);
246         // additive[k] += y[k];
247         // double tmpscalar = -2*scalar_prod(u,additive,n);
248         // vector_update(additive,u,tmpscalar,n);
249
250         additive[k] += y[k];
251         double tmpscalar = -2*scalar_prod(U[k],additive,n);
252         vector_update(additive,U[k],tmpscalar,n);
253     }
254     if(norm(additive,n)<eps*normr){ //check if the additive vector is big
255         enough to have an influence
256         stag += 1;
257     }else{
258         stag = 0;
259     }
260 }

```

```

259
260 // copyArray(xm,x,n);
261 // vector_update(xm,additive,1,n); // xm = x + additive
262 vector_add(1,x,1,additive,xm,n);
263
264
265
266
267 A.post_MV(xm,Axm);
268 vector_add(1.,b,-1.,Axm,r,n);
269 if (norm(r,n)<= tolb){
270     copyArray(x,xm,n);
271     flag = 0;
272     break;
273 }
274 normr = norm(r,n);
275
276 resvec[iterations] = normr;
277
278
279
280
281
282 if (normr <= tolb){
283     copyArray(x,xm,n);
284     flag = 0;
285     jmin = initer;
286     break;
287 }else{
288     if (stag >= maxstagsteps && moresteps ==0){
289         stag = 0;
290     }
291     moresteps += 1;
292     if (moresteps >= maxmsteps){
293         cout<<"stagnation"<<endl;
294         flag = 2;
295         jmin = initer;
296         break;
297     }
298 }
299 }//endif (normr <= tolb || stag >= maxstagsteps || moresteps)
300
301
302 if (stag>=maxstagsteps){
303     flag = 2;
304     break;
305 }
306 }// end for in
307
308 if (flag != 0){ // computes the x with the lowest norm
309     vector_zero_out(additive,n);
310
311     vector_zero_out(y,initer+1);
312     triangularSolver(R,w,y,initer);
313     // copyMat_col2Arr(U, u, initer, n);
314     // vector_scalarMult(additive,u,-2*y[initer]*U[initer][initer],n);
315
316     vector_scalarMult(additive,U[initer],-2*y[initer]*U[initer][initer],n);
317
318
319 additive[initer] += y[initer];
320 for(int k=initer-1;k>=0;k--){
321     // copyMat_col2Arr(U, u, k, n);
322     // additive[k] += y[k];
323     // double tmpscalar = -2*scalar_prod(u,additive,n);
324     // vector_update(additive,u,tmpscalar,n);
325
326     additive[k] += y[k];
327     double tmpscalar = -2*scalar_prod(U[k],additive,n);

```



```

328         vector_update(additive,U[k],tmpscalar,n);
329     }
330
331     vector_update(x,additive,1,n);
332
333     A.post_MV(x,Ax);
334     vector_add(1.0,b,-1.0,Ax,r,n);
335
336     normr = norm(r,n);
337
338     normr = norm(r,n);
339     beta = scalarsgn(normr)*normr;
340
341     vector_zero_out(u,inner);
342     mat_zero_out(U,n,inner+1);
343     copyArray(u, r, n);
344     u[0] += beta;
345     normalize(u, n);
346
347     copyArr2Mat_col(U,u,0,n);
348     // Apply Householder projection to r.
349     // w = r - 2*u*u'*r;
350     w[0] = -beta;
351 } // end if flag~=0
352
353
354 if (flag == 2) { //stagnation
355     break;
356 }
357
358 if (normr <= tolb) { // convergence
359     flag = 0;
360     break;
361 }
362
363
364 } //end for out
365
366
367
368
369 if (normr <= tolb){
370     flag = 0;
371 }
372
373
374 double end_time = omp_get_wtime();
375 double runtime = (end_time - start_time);
376
377 // writeCSV(matname,n, inner, iterations,b, x, w, resvec, U, R);
378 cout<<"GMRES finished, with residual norm: " << normr<< endl << "flag: " << flag<< "
379     iterations: " << iterations << endl;
380 // cout << "Runtime: " << runtime << " seconds" << endl;
381
382 free(r);
383 free(Ax);
384 free(xm);
385 free(Axm);
386
387 for (int i = 0; i < 2; ++i) {
388     free(J[i]);
389 }
390 free(J);
391
392 free(resvec);
393
394 for (int i = 0; i < inner+1; ++i) {
395     free(U[i]);
396 }

```

```

396     free(U);
397
398     free(u);
399
400     for (int i = 0; i < inner; ++i) {
401         free(R[i]);
402     }
403     free(R);
404
405     free(w);
406
407     free(v);
408
409     free(vtemp);
410
411     free(jtemp);
412
413     free(y);
414     free(additive);
415
416 }//this is end of gmres

```

A.3 functions.h

```

1  #ifndef FUNCTIONS_H
2  #define FUNCTIONS_H
3
4  #include <iostream>
5  #include <fstream>
6  #include <string>
7  #include <cstring>
8  #include <omp.h>
9
10 #include "constants.h"
11 #include "sparse.h"
12 #include "read_mat.h"
13
14 #include <fstream>
15 #include <string>
16
17 #include <cmath>
18
19
20 #include <random>
21
22
23 using namespace std;
24
25
26 double scalar_prod(double *v1, double *v2, int nr){
27     double sum = 0.0;
28
29     if(nr>par_threshold){
30         #pragma omp parallel for reduction(+:sum)
31         for(int i = 0; i < nr; i++){
32             sum += v1[i]*v2[i];
33         }
34         return sum;
35     }else{
36         for(int i = 0; i < nr; i++){
37             sum += v1[i]*v2[i];
38         }
39         return sum;
40     }
41
42 }
43

```

```

44 void elementwise_prod(double *v1, double *v2, int nr){
45     if(nr>par_threshold){
46         #pragma omp parallel for
47         for(int i = 0; i < nr; i++){
48             v1[i] *=v2[i];
49         }
50     }else{
51         for(int i = 0; i < nr; i++){
52             v1[i] *=v2[i];
53         }
54     }
55 }
56
57 double norm(double *v1, int nr){
58     // Euclidean norm
59     return sqrt(scalar_prod(v1,v1,nr));
60 }
61
62 void normalize(double* v1, int nr){
63     double v_norm = norm(v1,nr);
64     if(v_norm<eps){
65         cout<<"Warning: Norm very small: ";<<endl;
66     }else{
67         if(nr>par_threshold){
68             #pragma omp parallel for schedule(dynamic, nr/1000+1)
69             for(int i=0; i<nr; i++){
70                 v1[i] /= v_norm;
71             }
72         }else{
73             for(int i=0; i<nr; i++){
74                 v1[i] /= v_norm;
75             }
76         }
77     }
78 }
79
80
81
82 void vector_update(double *v1, double *p, double alpha, int nr){
83     if(nr>par_threshold){
84         #pragma omp parallel for schedule(dynamic, nr/1000+1)
85         for(int i = 0; i < nr; ++i){
86             v1[i] += alpha * p[i];
87         }
88     }else{
89         for(int i = 0; i < nr; ++i){
90             v1[i] += alpha * p[i];
91         }
92     }
93 }
94
95
96
97 void vector_scalarMult(double *v1, double *p, double alpha, int nr){
98     if(nr>par_threshold){
99         #pragma omp parallel for schedule(dynamic, nr/1000+1)
100         for(int i = 0; i < nr; ++i){
101             v1[i] = alpha * p[i];
102         }
103     }else{
104         for(int i = 0; i < nr; ++i){
105             v1[i] = alpha * p[i];
106         }
107     }
108 }
109
110 void vector_add(double alpha, double* v1, double beta, double* v2, double* result, int nr){
111     if(nr>par_threshold){
112         #pragma omp parallel for schedule(dynamic, nr/1000+1)

```

```

113     for(int i = 0; i < nr; ++i){
114         result[i] = alpha * v1[i] + beta * v2[i];
115     }
116 }else{
117     for(int i = 0; i < nr; ++i){
118         result[i] = alpha * v1[i] + beta * v2[i];
119     }
120 }
121 }
122
123 void vector_zero_out(double* v, int n){
124     if(n>par_threshhold){
125         #pragma omp parallel for schedule(dynamic, n/1000+1)
126         for(int i = 0; i <n; ++i){
127             v[i] = 0;
128         }
129     }else{
130         for(int i = 0; i <n; ++i){
131             v[i] = 0;
132         }
133     }
134 }
135
136 void vector_zero_out(double* v, int from, int to, int n){
137
138     if(n>par_threshhold){
139         #pragma omp parallel for schedule(dynamic, n/1000+1)
140         for(int i = from; i <= to; ++i){
141             v[i] = 0;
142         }
143     }else{
144         for(int i = from; i <= to; ++i){
145             v[i] = 0;
146         }
147     }
148 }
149
150
151 void mat_zero_out(double** A, int nr, int nc){
152     if(nc>par_threshhold){
153         #pragma omp parallel for schedule(dynamic, nr/1000+1)
154         for(int i = 0; i <nc; ++i){
155             for(int j = 0; j<nr; j++){
156                 A[i][j] = 0;
157             }
158         }
159     }else{
160         for(int i = 0; i <nc; ++i){
161             for(int j = 0; j<nr; j++){
162                 A[i][j] = 0;
163             }
164         }
165     }
166 }
167
168
169 void normalize_copy(double *normalized, double *original, int nr){
170     double v_norm = norm(original,nr);
171     if(v_norm<eps){
172         cout<<"Warning: Norm to small "<<v_norm<<endl;
173     }else{
174         if(nr>par_threshhold){
175             #pragma omp parallel for schedule(dynamic, nr/1000+1)
176             for(int i=0; i<nr; i++){
177                 normalized[i] = original[i]/v_norm;
178             }
179         }else{
180             for(int i=0; i<nr; i++){
181                 normalized[i] = original[i]/v_norm;

```

```

182     }
183 }
184 }
185 }
186
187
188 void copyArray(double* dest, const double* src, int size) {
189     if(size>par_threshold){
190         #pragma omp parallel for schedule(dynamic, size/1000+1)
191         for(int i=0; i<size; i++){
192             dest[i] = src[i];
193         }
194     }else{
195         for(int i=0; i<size; i++){
196             dest[i] = src[i];
197         }
198     }
199 }
200
201 void copyArr2Mat_col(double** matrix, const double* arr, int col, int size){
202     // #pragma omp parallel for schedule(dynamic, nr/1000+1)
203     for(int i=0; i<size; i++){
204         matrix[col][i] = arr[i];
205     }
206 }
207
208 void copyArr2Mat_col(double** matrix, const double* arr, int col, int size, int from, int to){
209     // #pragma omp parallel for schedule(dynamic, nr/1000+1)
210     for(int i=from; i<=to; i++){
211         matrix[col][i] = arr[i];
212     }
213 }
214
215 void copyArr2Mat_row(double** matrix, const double* arr, int row, int size){
216     // #pragma omp parallel for schedule(dynamic, nr/1000+1)
217     for(int i=0; i<size; i++){
218         matrix[i][row] = arr[i];
219     }
220 }
221
222 void copyArr2Mat_row(double** matrix, const double* arr, int row, int size, int from, int to){
223     // #pragma omp parallel for schedule(dynamic, nr/1000+1)
224     for(int i=from; i<=to; i++){
225         matrix[i][row] = arr[i];
226     }
227 }
228
229 void copyMat_col2Arr(double** matrix, double* arr, int col, int size){
230     // #pragma omp parallel for schedule(dynamic, nr/1000+1)
231     for(int i=0; i<size; i++){
232         arr[i] = matrix[col][i];
233     }
234 }
235
236
237 double scalarsgn(double x){
238     if (x >= 0){
239         return 1.0;
240     }
241     else{
242         return -1.0;
243     }
244 }
245
246
247
248
249
250

```

```

251 void printMat(double** matrix, int cols, int rows) {
252     for (int i = 0; i < cols; i++) {
253         for (int j = 0; j < rows; j++){
254             cout << matrix[i][j] << " ";
255         }
256         cout << endl;
257     }
258 }
259
260
261 void triangularSolver(double** A, double* b, double* x, int n){
262     for(int row = n-1; row>=0; row--){
263         double sum = 0.0;
264         //could be parralelized here or reduction operator
265         for(int col = row+1; col<n; col++){
266             sum+=A[row][col]*x[col];
267         }
268         x[row] = (b[row]-sum)/A[row][row];
269     }
270 }
271
272
273
274
275 void writeCSV(char* matname, int n, int inner, int iterations, double* b, double* xstar, double*
    w, double* resvec, double** U, double** R) {
276
277     char filename[100];
278     strcpy(filename, "data_output/");
279     strcat(filename, matname);
280     strcat(filename, "_data.csv");
281     ofstream file(filename, ofstream::trunc); // Open the file in write mode with truncation
282
283
284     // Write b (1D array)
285     file << "b";
286     for (int i = 0; i < n; i++) {
287         file << "," << b[i];
288     }
289     file << std::endl;
290
291     // Write xstar (1D array)
292     file << "xstar";
293     for (int i = 0; i < n; i++) {
294         file << "," << xstar[i];
295     }
296     file << std::endl;
297
298     // Write w (1D array)
299     file << "w";
300     for (int i = 0; i < n + 1; i++) {
301         file << "," << w[i];
302     }
303     file << std::endl;
304
305     // Write resvec (1D array)
306     file << "resvec";
307     for (int i = 0; i < iterations; i++) {
308         file << "," << resvec[i];
309     }
310     file << std::endl;
311
312     // Write U (2D array)
313     for (int i = 0; i < inner; i++) {
314         file << "U_" << i;
315         for (int j = 0; j < n; j++) {
316             file << "," << U[i][j];
317         }
318         file << std::endl;

```

```

319     }
320
321     // Write R (2D array)
322     for (int i = 0; i < inner; i++) {
323         file << "R_" << i;
324         for (int j = 0; j < inner; j++) {
325             file << "," << R[i][j];
326         }
327         file << std::endl;
328     }
329
330     file.close(); // Close the file
331 }
332
333
334
335 void createMat(sparse* mat, char* fname) {
336     std::ifstream file(fname);
337
338     if (file.is_open()) {
339         int nr, nc, nt;
340         file >> nr >> nc >> nt;
341
342         int* iat = (int*)malloc((nr + 1) * sizeof(int));
343         int* ja = (int*)malloc(nt * sizeof(int));
344         double* elem = (double*)malloc(nt * sizeof(double));
345
346         readCSRmat(fname, &nr, &nc, &nt, &iat, &ja, &elem, false);
347
348         mat->set_nrow(nr);
349         mat->set_ncol(nc);
350         mat->set_n_term(nt);
351         mat->set_iat(iat);
352         mat->set_ja(ja);
353         mat->set_coef(elem);
354
355         file.close();
356
357         // wo muss ich hier freeen? oder nicht weil in dekonstruktor gemacht wird?
358     } else {
359         cout << "FILE NOT OPENED " << fname << endl;
360     }
361 }
362
363 bool getRHS(char* fname, double* b){
364     std::ifstream file(fname);
365
366     if (file.is_open()) {
367         double temp_nr;
368         int nr;
369         file >> temp_nr;
370         nr = static_cast<int>(temp_nr);
371         for(int i=0;i<nr;i++){
372             file >> b[i];
373         }
374         cout<<"Vector file " << fname << endl;
375         cout<<"Vector length " << nr << endl;
376         return true;
377     } else {
378         cout << "FILE NOT OPENED " << fname << endl;
379         return false;
380     }
381 }
382
383 void make_b(char* fname, int nr){
384     std::ofstream file(fname, std::ofstream::trunc);
385
386     std::random_device rd;
387     std::mt19937 generator(rd());

```

```

388     std::uniform_real_distribution<double> distribution(0.0, 1.0);
389     file<<nr<<endl;
390     #pragma omp parallel for schedule(dynamic, nr/1000+1)
391     for(int i=0;i<nr;i++){
392         file<<distribution(generator)<<endl;
393     }
394
395     cout<<"random Vector of length created " << nr<<endl;
396     file.close();
397 }
398
399
400 void save_solution(char* fname, double* xstar, int n){
401     std::ofstream file(fname,std::ofstream::trunc);
402     file<<n<<endl;
403     for(int i=0;i<n;i++){
404         file<<xstar[i]<<endl;
405     }
406     cout<<"Vector saved in: " << fname << " length: " << n<<endl;
407     file.close();
408 }
409
410 double* getFEM_sparse(sparse* mat){
411     std::ifstream f_coef("FEM_output/coef.txt");
412     std::ifstream f_iat("FEM_output/iat.txt");
413     std::ifstream f_ja("FEM_output/Ja.txt");
414     std::ifstream f_b("FEM_output/b.txt");
415
416     int n_term, n_row_plus1, n_row;
417     f_coef >> n_term;
418     f_ja >> n_term;
419     f_iat >> n_row_plus1;
420     f_b >> n_row;
421
422     double* elem = (double*) malloc(n_term*sizeof(double));
423     int* ja = (int*) malloc(n_term*sizeof(int));
424     int* iat = (int*) malloc(n_row_plus1*sizeof(int));
425
426
427     double* b = (double*) malloc(n_row*sizeof(double));
428
429     for(int i = 0; i < n_term; i++){
430         f_coef >> elem[i];
431         f_ja >> ja[i];
432     }
433
434     for (int i = 0; i < n_row_plus1; i++){
435         f_iat >> iat[i];
436     }
437     for (int i = 0; i < n_row; i++){
438         f_b >> b[i];
439     }
440
441     mat->set_nrow(n_row);
442     mat->set_ncol(n_row);
443     mat->set_n_term(n_term);
444     mat->set_iat(iat);
445     mat->set_ja(ja);
446     mat->set_coef(elem);
447
448     printf("Matrix rows %d columns %d nterm %d\n",n_row,n_row,n_term);
449     f_coef.close();
450     f_iat.close();
451     f_ja.close();
452     f_b.close();
453     return b;
454 }
455
456 double** mat_allocation(int n_row , int n_col){

```



```

457     double **A = (double**) malloc(n_row*sizeof(double*));
458     for (int i = 0; i < n_row; ++i) {
459         A[i] = (double*)malloc(n_col * sizeof(double));
460     }
461     return A;
462 }
463
464 void mat_free(double** A, int n_col, int n_row){
465     for (int i = 0; i < n_row; ++i) {
466         free(A[i]);
467     }
468     free(A);
469 }
470
471
472 int getNZ(double** A, int n){
473     // number of nonzeros in mtx
474     // A is square
475     int nnz = 0;
476     #pragma omp parallel for schedule(dynamic,n/1000+1)
477     for(int i = 0; i < n; i++){
478         for(int j = 0; j < n; j++){
479             if(abs(A[i][j])>eps){
480                 #pragma omp atomic
481                 nnz++;
482             }
483         }
484     }
485     return nnz;
486 }
487
488 int getNZ_idx(double** A, double** idx, int n){
489     // number of nonzeros in mtx
490     // A is square
491     int nnz = 0;
492     for(int i = 0; i < n; i++){
493         for(int j = 0; j < n; j++){
494             if(abs(A[i][j])>eps){
495                 nnz++;
496                 idx[nnz][0] = i;
497                 idx[nnz][1] = j;
498             }
499         }
500     }
501     return nnz;
502 }
503
504 double max(double* v, int n){
505     double maxVal = v[0];
506     #pragma omp parallel for schedule(dynamic, n/1000+1)
507     for (int i = 1; i < n; i++) {
508         if (v[i] > maxVal) {
509             #pragma omp critical
510             {
511                 if (v[i] > maxVal) {
512                     maxVal = v[i];
513                 }
514             }
515         }
516     }
517     return maxVal;
518 }
519
520 #endif

```

A.4 fem_functions.h

```

1  #include <iostream>
2  #include <fstream>
3  #include <string>
4  #include <omp.h>
5
6  #include "constants.h"
7  #include "sparse.h"
8  #include "functions.h"
9
10 #include <fstream>
11 #include <string>
12 #include <cmath>
13
14 #include <algorithm>
15 #include <vector>
16
17 using namespace std;
18
19 void stiffness_struct(char* fname, int ne, int nn, double** topol) {
20     double** Adj = mat_allocation(ne, nn);
21     double* i_idx = (double*)malloc(static_cast<int>(nn*nn*0.2));
22     double* j_idx = (double*)malloc(static_cast<int>(nn*nn*0.2));
23
24     int idx;
25     #pragma omp parallel for schedule(dynamic, nn/1000+1)
26     for (int i = 0; i < ne; i++) {
27         for (int k = 0; k < 3; k++) {
28             idx = static_cast<int>(floor(topol[i][k]) - 1);
29             Adj[i][idx] = 1;
30         }
31     }
32
33     double sum;
34     int nnz = 0;
35     std::string tempFile = "temp_file.txt";
36     std::ofstream file(tempFile, std::ofstream::trunc); // Open the temporary file in write
37     mode with truncation
38     for (int i = 0; i < nn; i++) {
39         for (int j = 0; j < nn; j++) {
40             sum = 0;
41             for (int k = 0; k < ne; k++) {
42                 sum += Adj[k][i] * Adj[k][j];
43             }
44             if (sum > 0) {
45                 nnz++;
46                 file << j + 1 << " " << i + 1 << " " << 0 << std::endl;
47             }
48         }
49     }
50     file.close();
51
52     // Reopen the temporary file in read mode
53     std::ifstream inFile(tempFile);
54     if (!inFile) {
55         std::cerr << "Error opening the temporary file for reading." << std::endl;
56         return;
57     }
58
59     // Open the original file in write mode with truncation
60     std::ofstream outFile(fname, std::ofstream::trunc);
61     if (!outFile) {
62         std::cerr << "Error opening the file for writing." << std::endl;
63         inFile.close();
64         return;
65     }
66
67     // Write the new data at the beginning of the original file
68     outFile << nn << " " << nn << " " << nnz << std::endl;

```

```

69 // Copy the rest of the content from the temporary file to the original file
70 std::string line;
71 while (std::getline(inFile, line)) {
72     if (!line.empty()) {
73         outFile << line << std::endl;
74     }
75 }
76
77 // Close both files
78 inFile.close();
79 outFile.close();
80
81 // Delete the temporary file
82 std::remove(tempFile.c_str());
83 }
84
85
86 void stiffness_struct_para(char* fname, int ne, int nn, double** topol) {
87     double** Adj = mat_allocation(ne, nn);
88     double* i_idx = (double*)malloc(static_cast<int>(nn*nn*0.2));
89     double* j_idx = (double*)malloc(static_cast<int>(nn*nn*0.2));
90
91     int idx;
92     #pragma omp parallel for schedule(dynamic, nn/1000+1)
93     for (int i = 0; i < ne; i++) {
94         for (int k = 0; k < 3; k++) {
95             idx = static_cast<int>(floor(topol[i][k]) - 1);
96             Adj[i][idx] = 1;
97         }
98     }
99
100     double sum;
101     int nnz = 0;
102     std::string tempFile = "temp_file.txt";
103     std::ofstream file(tempFile, std::ofstream::trunc); // Open the temporary file in write
104     // mode with truncation
105
106     #pragma omp for schedule(dynamic, nn/1000+1)
107     for (int i = 0; i < nn; i++) {
108         for (int j = 0; j < nn; j++) {
109             sum = 0;
110             for (int k = 0; k < ne; k++) {
111                 if (Adj[k][i] * Adj[k][j] > 0) {
112                     nnz++;
113                     file << j + 1 << " " << i + 1 << " " << 0 << std::endl;
114                     break;
115                 }
116             }
117         }
118     }
119     file.close();
120
121     // Reopen the temporary file in read mode
122     std::ifstream inFile(tempFile);
123     if (!inFile) {
124         std::cerr << "Error opening the temporary file for reading." << std::endl;
125         return;
126     }
127
128     // Open the original file in write mode with truncation
129     std::ofstream outFile(fname, std::ofstream::trunc);
130     if (!outFile) {
131         std::cerr << "Error opening the file for writing." << std::endl;
132         inFile.close();
133         return;
134     }
135
136     // Write the new data at the beginning of the original file

```

```

137     outFile << nn << " " << nn << " " << nnz << std::endl;
138
139     // Copy the rest of the content from the temporary file to the original file
140     std::string line;
141     while (std::getline(inFile, line)) {
142         if (!line.empty()) {
143             outFile << line << std::endl;
144         }
145     }
146
147     // Close both files
148     inFile.close();
149     outFile.close();
150
151     // Delete the temporary file
152     std::remove(tempFile.c_str());
153 }
154
155
156 void stiffness_struct_perf(sparse* H, int ne, int nn, double** topol){
157     double** Adj = mat_allocation(ne+1, nn);
158
159     // Create vectors for i_idx and j_idx
160     std::vector<int> i_idx;
161     std::vector<int> j_idx;
162     int idx;
163
164     // Reserve an initial capacity for the vectors
165     double n_temp = static_cast<double>(nn);
166     int initialCapacity = static_cast<int>(n_temp * n_temp * 0.05 );
167     i_idx.reserve(initialCapacity);
168     j_idx.reserve(initialCapacity);
169
170     #pragma omp parallel for schedule(dynamic, nn/1000+1)
171     for (int i = 0; i < ne; i++) {
172         for (int k = 0; k < 3; k++) {
173             idx = static_cast<int>(floor(topol[i][k]) - 1);
174             Adj[i][idx] = 1;
175         }
176     }
177
178     double sum;
179     int nnz = 0;
180     #pragma omp parallel for schedule(dynamic, nn/1000+1)
181     for (int i = 0; i < nn; i++) {
182         for (int j = 0; j < nn; j++) {
183             sum = 0;
184             for (int k = 0; k < ne; k++) {
185                 if (Adj[k][i] * Adj[k][j] > 0) {
186                     j_idx[nnz] = (i + 1);
187                     i_idx[nnz] = (j);
188                     #pragma omp atomic
189                     nnz++;
190                     // Optionally, check the capacity and reserve more space if needed (e.g.,
191                     // double the current capacity)
192                     if (nnz >= i_idx.capacity()) {
193                         std::cout<<"capacity too small"<<nnz<<endl;
194                         int newCapacity = i_idx.capacity() * 1.5;
195                         i_idx.reserve(newCapacity);
196                         j_idx.reserve(newCapacity);
197                     }
198                     break;
199                 }
200             }
201         }
202     }
203     int* ja_loc = &i_idx[0];
204     int* irow = &j_idx[0];

```

```

205
206     int* iat_loc = (int*)malloc(nn*sizeof(int));
207     irow2iat(nn,nnz,irow,iat_loc);
208     H->set_n_term(nnz);
209     H->set_ncol(nn);
210     H->copy_ja(ja_loc);
211     H->copy_iat(iat_loc);
212
213     free(iat_loc);
214 }
215
216 void stiffness_struct_small_loops(sparse* H, int ne, int nn, double** topol) {
217     // Create vectors to store the i_idx and j_idx pairs
218     std::vector<int> i_idx;
219     std::vector<int> j_idx;
220
221     // Loop through each element
222     for (int i = 0; i < ne; i++) {
223         for (int k = 0; k < 3; k++) {
224             int idx = static_cast<int>(floor(topol[i][k]) - 1);
225             for (int l = 0; l < 3; l++) {
226                 int jdx = static_cast<int>(floor(topol[i][l]) - 1);
227                 i_idx.push_back(idx);
228                 j_idx.push_back(jdx);
229             }
230         }
231     }
232
233     // Combine i_idx and j_idx into pairs for sorting
234     std::vector<std::pair<int, int>> pairs;
235     for (int i = 0; i < i_idx.size(); i++) {
236         pairs.push_back(std::make_pair(i_idx[i], j_idx[i]));
237     }
238
239     // Sort the pairs to bring duplicates together
240     std::sort(pairs.begin(), pairs.end());
241
242     // Remove duplicates and copy the unique pairs back to i_idx and j_idx
243     int nnz = 0;
244     for (int i = 0; i < pairs.size(); i++) {
245         if (i == 0 || pairs[i] != pairs[i - 1]) {
246             i_idx[nnz] = pairs[i].first + 1;
247             j_idx[nnz] = pairs[i].second;
248             nnz++;
249         }
250     }
251
252     // Allocate memory for the CSR arrays
253     int* ja_loc = new int[nnz];
254     int* iat_loc = new int[nn + 1];
255
256     // Copy the data from i_idx and j_idx to the CSR arrays
257     for (int i = 0; i < nnz; i++) {
258         ja_loc[i] = j_idx[i];
259     }
260
261     // Compute the CSR arrays from i_idx and ja_loc
262     irow2iat(nn, nnz, &i_idx[0], iat_loc);
263
264     H->set_n_term(nnz);
265     H->set_ncol(nn);
266     H->copy_ja(ja_loc);
267     H->copy_iat(iat_loc);
268     // H->printComponents();
269     delete[] ja_loc;
270     delete[] iat_loc;
271 }
272
273 double get_loc(double** coord, double** topol, double *D, double*V, int el, double*** Loc){

```

```

274 double *t = topol[e1];
275 double x[3]; //i,j,m
276 double y[3]; //i,j,m
277 double a[3],b[3],c[3], delta, temp, mod_v;
278
279 for(int i = 0; i<3; i++){
280     x[i] = coord[static_cast<int>(t[i]-1)][0];
281     y[i] = coord[static_cast<int>(t[i]-1)][1];
282 }
283
284 a[0] = x[1]*y[2] - x[2]*y[1];
285 a[1] = x[2]*y[0] - x[0]*y[2];
286 a[2] = x[0]*y[1] - x[1]*y[0];
287
288 int idx1, idx2;
289 for(int i = 0; i<3;i++){
290     idx1 = (i+1)%3;
291     idx2 = (i+2)%3;
292     b[i] = y[idx1] - y[idx2];
293     c[i] = x[idx2] - x[idx1];
294 }
295
296 delta = (a[0]+a[1]+a[2])/2;
297 mod_v = norm(V,2);
298
299 if(mod_v>eps){// prevent zero division
300     double temp;
301     for(int i = 0; i<3; i++){
302         for(int j = 0; j<3; j++){
303             temp = (V[0]*b[j] + V[1]*c[j]);
304             Loc[0][i][j] = (D[0]*b[i]*b[j]+ D[1]*c[i]*c[j])/(4*delta);
305             Loc[1][i][j] = temp/6;
306             Loc[2][i][j] = (V[0]*b[i] + V[1]*c[i])*temp/(8*delta*mod_v*max(D,2));
307         }
308     }
309 }else{
310     std::cout<<" Norm v zero, just compute stiffness Matrix H \n";
311     for(int i = 0; i<3; i++){
312         for(int j = 0; j<3; j++){
313             Loc[0][i][j] = (D[0]*b[i]*b[j]+ D[1]*c[i]*c[j])/(4*delta);
314         }
315     }
316 }
317
318
319 return delta;
320 }
321
322 void loc2glob(sparse* A_ptr, double** Loc, double** topol, int e){
323     double* coef = A_ptr->get_coef();
324     int* ja = A_ptr->get_ja();
325     int* iat = A_ptr->get_iat();
326     int row, col;
327     double* top = topol[e];
328     for(int i = 0; i<3; i++){ // choose row
329         row = static_cast<int>(top[i]) - 1;
330         for(int j = 0; j<3; j++){// chose col
331             col = static_cast<int>(top[j]) - 1;
332             int k;
333             for(k = iat[row]; k<iat[row+1];k++){ //go over column indices in ja
334                 if(ja[k]==col){break;}
335             }
336             #pragma omp atomic
337             coef[k] += Loc[i][j];
338         }
339     }
340 }
341
342 sparse* create_FEM_mat(double** coord, double** topol, int ne, int nn, double* D,\

```

```

343         double h, double tau, double* V, double* delta){
344     sparse* mat_list;
345     mat_list = new sparse[3]; // H,B,S
346     double*** Loc = (double***) malloc(3*sizeof(double**));
347     // double ***Loc;
348     // Loc = new double**[3];
349     int buffer_length = static_cast<int>(nn * nn * 0.1);
350
351     int nterm, nr;
352
353     int* ja = (int*) calloc(buffer_length,sizeof(int));
354     int* iat = (int*) calloc(nn+1,sizeof(int));
355
356     //allocate space for the local H, B, S
357     for(int i = 0; i<3;i++){Loc[i] = mat_allocation(3,3);}
358
359     char fname[] = "fullmat.txt";
360     // stiffness_struct(fname, ne, nn, topol);
361     // createMat(&mat_list[0], fname);
362     stiffness_struct_small_loops(&mat_list[0], ne, nn, topol);
363
364     mat_list[2] = mat_list[1] = mat_list[0];
365
366     for(int i = 0; i<ne; i++){
367         delta[i] = get_loc(coord,topol,D,V,i,Loc);
368         for(int k = 0; k<3;k++){
369             loc2glob(&mat_list[k],Loc[k],topol,i);
370         }
371     }
372     // printVec(delta,ne);
373     mat_list[2].scalarMult(tau*h);
374
375     return mat_list;
376 }
377
378
379 double force(double x, double y){
380     double f = 0;
381     // do stuff
382     return f;
383 }
384
385
386 void imposeBC(sparse* H, double** bound, double* q, int nb){
387     double R = 1e15;
388     int j, bound_idx, start, end;
389     int* iat = H->get_iat();
390     int* ja = H->get_ja();
391     double* coef = H->get_coef();
392
393     for(int i = 0; i<nb; ++i){
394         bound_idx = static_cast<int>(bound[i][0])-1;
395         start = iat[bound_idx];
396         end = iat[bound_idx+1];
397
398         for(j = start; j<end; j++){
399             if (ja[j] == bound_idx){break;}
400         }
401
402         coef[j] = R;
403         q[bound_idx] = R*bound[i][1];
404     }
405 }
406
407
408
409 double* getAssembledSystem(sparse* A){
410     // FEM assembly
411     char mesh_path[] = "mesh/mesh.dat";

```

```

412 char node_path[] = "mesh/dirnod.dat";
413 char coord_path[] = "mesh/xy.dat";
414
415 double D[2] = {1,1};
416 double V[2] = {1,3};
417 int ne, nn, nb; // number of mesh (800), number of nodes(441), number of ? (80)
418
419 double **coord, **bound, *delta, *delta_node, *F;
420 double **topol, *temp, *q;
421
422 sparse* FEM;
423
424
425 ifstream mesh(mesh_path);
426 ifstream nodes(node_path);
427 ifstream xy(coord_path);
428
429 //read data
430 int not_used;
431
432 xy >> nn >> not_used;
433 nodes >> nb >> not_used;
434 mesh >> ne >> not_used >> not_used >> not_used;
435
436 bound = mat_allocation(nb,2);
437 coord = mat_allocation(nn,2);
438 topl = mat_allocation(ne,3);
439
440
441
442 for (int i = 0; i < nn; ++i){xy >> coord[i][0] >> coord[i][1];}
443 for (int i = 0; i < nb; ++i){nodes >> bound[i][0] >> bound[i][1];}
444 for (int i = 0; i < ne; ++i){mesh >> topl[i][0] >> topl[i][1] >> topl[i][2]>>not_used;}
445
446
447 // printMat(coord,nb,2);
448
449 xy.close();
450 nodes.close();
451 mesh.close();
452
453 double h = sqrt(2)*(coord[0][1]-coord[1][1]);// why sqrt2
454 double tau = 0.1; // whats tau?
455
456 delta = (double*)malloc(ne*sizeof(double));
457
458 delta_node = (double*)calloc(nn,sizeof(double));
459
460 F = (double*)malloc(nn*sizeof(double));
461 q = (double*)calloc(nn,sizeof(double));
462
463
464
465 FEM = create_FEM_mat(coord, topl, ne, nn, D, h, tau, V, delta);
466
467 int idx;
468 #pragma omp parallel for schedule(dynamic,nn/1000+1)
469 for(int i = 0; i<ne; i++){
470     for(int k=0; k<3; k++){
471         idx = static_cast<int>(floor(topl[i][k])-1);
472         delta_node[idx] += delta[k]/3;
473     }
474 }
475
476 // RHS
477 for(int i = 0; i<nn; i++){
478     F[i] = force(coord[i][0],coord[i][1]*delta_node[i]);
479 }
480

```



```

481 // BCs
482 imposeBC(&FEM[0], bound, q, nb);
483 FEM[1].addition_update(&FEM[2]);
484 FEM[0].addition_update(&FEM[1]);
485
486 *A = FEM[0];
487 vector_update(q,F,1,nn);
488
489 mat_free(bound,nb,2);
490 mat_free(coord,nn,2);
491 mat_free(topol,ne,3);
492 free(delta);
493 free(delta_node);
494 free(F);
495 delete[] FEM;
496
497
498 return q;
499 }

```

A.5 sparse.h

```

1  #ifndef SPARSE_H
2  #define SPARSE_H
3
4  #include <omp.h>
5  #include <iostream>
6  #include <cmath>
7
8  #include "constants.h"
9
10 class sparse{
11 private:
12     int nrow, ncol, n_term;
13     double* coef; // non zero matrixelements
14     int *iat, *ja; // first non zero column index iat, what is ja?
15
16
17 public:
18     // constructors
19     sparse();
20     sparse(int nrow, int ncol, int n_term);
21     ~sparse(); // destructor
22
23
24     // get
25     int get_nrow(){return nrow;}
26     int get_ncol(){return ncol;}
27     int get_n_term(){return n_term;}
28
29     double* get_coef(){return coef;}
30     int* get_iat(){return iat;}
31     int* get_ja(){return ja;}
32
33
34     // set
35     void set_nrow(int n){
36         nrow = n;
37         ncol = n;
38         free(iat);
39         iat = (int*) malloc(ncol*sizeof(int));
40     }
41     void set_ncol(int n){
42         nrow = n;
43         ncol = n;
44         free(iat);
45         iat = (int*) malloc(ncol*sizeof(int));

```

```

46     }
47     void set_n_term(int n){
48         n_term = n;
49         free(coef);
50         free(ja);
51         ja = (int*) malloc(n_term*sizeof(int));
52         coef = (double*) malloc(n_term*sizeof(double));
53     }
54
55     // void alloc_coef(int n){coef = (double*)malloc((n)*sizeof(double));}
56     // void alloc_iat(int n){iat = (int*)malloc((n+1)*sizeof(int));}
57     // void alloc_ja(int n){ja = (int*)malloc((n)*sizeof(int));}
58
59     void set_coef(double* v){coef = v;}
60     void set_iat(int* v){iat = v;}
61     void set_ja(int* v){ja = v;}
62
63     void copy_coef(double* v){
64         #pragma omp parallel for
65         for(int i = 0; i<n_term;i++){
66             coef[i]=v[i];
67         }
68     }
69
70     void copy_iat(int* v){
71         #pragma omp parallel for
72         for(int i = 0; i<nrow+1;i++){
73             iat[i]=v[i];
74         }
75     }
76
77     void copy_ja(int* v){
78         #pragma omp parallel for
79         for(int i = 0; i<n_term;i++){
80             ja[i]=v[i];
81         }
82     }
83
84
85     //operators
86     sparse& operator=(sparse&);
87     friend bool operator==(const sparse& lhs, const sparse& rhs);
88
89     //Data Operations
90     void full2sparse(double** A, int n_row);
91
92
93     // matrix funcitons
94     void addition_update(sparse* B_ptr);
95     void scalarMult(double alpha);
96     void post_MV(double* v, double* y); // A*b
97     void pre_MV(double* v, double* y); // b'*A
98     void matrixProduct(sparse* A_ptr, sparse* result);
99     void diag(double* v);
100    void getJacobi(double* v);
101    void diag_x_sparse(double* diag);
102    double* left_Jacobi();
103
104
105    // display mtx
106    void printMat();
107    void printComponents();
108 };
109
110
111 using namespace std;
112
113 sparse::sparse(){
114     // cout<<"empty constructor\n";

```

```

115     nrow = 0;
116     ncol = 0;
117     n_term = 0;
118
119     coef = nullptr;
120     ja = nullptr;
121     iat = nullptr;
122 };
123
124 sparse::sparse(int nr, int nc, int nt){
125     nrow = nr;
126     ncol = nc;
127     n_term = nt;
128
129     coef = (double*)malloc(nt*sizeof(double));
130     ja = (int*)malloc(nt*sizeof(int));
131     iat = (int*)malloc((nr+1)*sizeof(int)); //+1
132 };
133
134 sparse::~~sparse(){
135     free(coef);
136     free(ja);
137     free(iat);
138 }
139
140
141 void sparse::post_MV(double* v, double* y ){
142     // #pragma omp parallel for schedule(dynamic,nrow/1000+1)
143     for(int i=0; i<nrow; i++){
144         double sum = 0.0;
145         #pragma omp simd reduction(+:sum)
146         for(int j=iat[i]; j<iat[i+1]; j++){
147             sum += coef[j]*v[ja[j]];
148         }
149         y[i] = sum;
150     }
151 }
152
153 void sparse::pre_MV(double* v, double* y ){
154     #pragma omp parallel for schedule(dynamic,nrow/1000+1)
155     for(int j=0; j<ncol; j++){
156         double sum = 0.0;
157         // #pragma omp simd reduction(+:sum)
158         for(int i=0; i<nrow; i++){
159             for(int k=iat[i]; k<iat[i+1]; k++){
160                 if (ja[k] == j){
161                     sum += v[i]*coef[k];
162                 }
163             }
164         }
165         y[j] = sum;
166     }
167 }
168
169 sparse& sparse::operator=(sparse& A){
170     // std::cout<<"cpy"<<endl;
171     int n_row = A.get_nrow();
172     int n_term = A.get_n_term();
173
174     this->set_n_term(n_term);
175     this->set_ncol(n_row);
176     this->copy_coef(A.get_coef());
177     this->copy_iat(A.get_iat());
178     this->copy_ja(A.get_ja());
179
180     return *this;
181 }
182
183 bool operator==(const sparse& lhs, const sparse& rhs) {

```

```

184     if (lhs.nrow != rhs.nrow || lhs.ncol != rhs.ncol || lhs.n_term != rhs.n_term) {
185         cout<<" Shapes are not equal\n";
186         return false;
187     }
188 }
189
190 // Compare ja arrays
191 for (int i = 0; i < lhs.n_term; i++) {
192     if (lhs.ja[i] != rhs.ja[i]) {
193         cout<<" ja arrays are not equal\n";
194         return false; //
195     }
196 }
197 }
198
199 // Compare iat arrays
200 for (int i = 0; i <= lhs.nrow; i++) {
201     if (lhs.iat[i] != rhs.iat[i]) {
202         cout<<" iat arrays are not equal\n";
203         return false; //
204     }
205 }
206 }
207
208 return true; // Shapes, ja, and iat arrays are equal
209 }
210
211
212 void sparse::full2sparse(double** A, int n){
213     // A is square
214     // A is sparse --> suppose only max 10% of elements are nonzero
215     double sparse_ratio = 1;
216     int n_term_max = static_cast<int>(n * n * sparse_ratio);
217
218     double* coef_temp = (double*)malloc(n_term_max*sizeof(double));
219     int* Ja_temp = (int*)malloc(n_term_max*sizeof(int));
220     int* iat_temp = (int*)malloc((n+1)*sizeof(int));
221
222     this->set_ncol(n);
223     this->set_nrow(n);
224
225     int idx_nt = 0;
226     int idx_nr = 0;
227     bool new_row = true;
228
229     #pragma omp parallel for schedule(dynamic,n/1000+1)
230     for(int i = 0; i < n; i++){
231         for(int j = 0; j < n; j++){
232             if(abs(A[i][j])>eps){
233                 coef_temp[idx_nt] = A[i][j];
234                 Ja_temp[idx_nt] = j;
235                 if(new_row){
236                     iat_temp[idx_nr] = idx_nt;
237                     idx_nr++;
238                     new_row = false;
239                 }
240                 idx_nt++;
241             }
242         }
243         new_row = true;
244     }
245     iat_temp[idx_nr] = idx_nt;
246
247     this->set_n_term(idx_nt);
248     this->copy_coef(coef_temp);
249     this->copy_ja(Ja_temp);
250     this->copy_iat(iat_temp);
251
252     free(iat_temp);

```

```

253     free(coef_temp);
254     free(Ja_temp);
255 }
256
257 void sparse::addition_update(sparse* B_ptr){
258     if(*this==*B_ptr){
259         double* add = B_ptr->get_coef();
260         #pragma omp parallel for schedule(dynamic,n_term/1000 +1)
261         for(int i = 0; i<n_term; i++){
262             coef[i] += add[i];
263         }
264     }else{
265         cout<<"Sparsity structure is not the same \n";
266     }
267 }
268
269
270
271 void sparse::scalarMult(double alpha){
272     #pragma omp parallel for schedule(dynamic,n_term/1000 +1)
273     for(int i = 0; i<n_term; i++){
274         coef[i] *= alpha;
275     }
276 }
277 void sparse::matrixProduct(sparse* B_ptr, sparse* result){
278     // (nxm)*(mxk) = (nxk) but usually square anyways
279
280
281     sparse& B = *B_ptr;
282
283     if (nrow != B.ncol){
284         std::cout<<"Incompatible Size"<<endl;
285     }
286
287
288     for(int i = 0; i < nrow; i++){
289         double sum = 0.0;
290         for(int j = 0; j < B.ncol; j++){
291
292             sum += i+j;
293         }
294
295     }
296 }
297
298
299 }
300
301
302 void sparse::diag(double* v){
303     int start, end;
304     #pragma omp parallel for schedule(dynamic, ncol/1000+1)
305     for(int i = 0; i<ncol; i++){
306         start = iat[i];
307         end = iat[i+1];
308         for(int j = start; j<end; j++){
309             if(ja[j] == i){
310                 v[i] = coef[j];
311                 break;
312             }
313         }
314     }
315 }
316
317 void sparse::getJacobi(double* v){
318     this->diag(v);
319     int zeros = 0;
320     #pragma omp parallel for schedule(dynamic, ncol/1000 +1)
321     for(int i = 0; i<ncol; i++){

```

```

322     if (v[i] == 0){
323         v[i]=1;
324         #pragma omp atomic
325         zeros++;
326     }
327     v[i] = 1/v[i];
328 }
329 if(zeros>0){
330     cout<< "WARNING: Number of zeros in diagonal: "<<zeros<<endl;
331 }
332 }
333
334 void sparse::diag_x_sparse(double* diag){
335     int start, end;
336     #pragma omp parallel for schedule(dynamic, ncol/1000+1)
337     for(int i = 0; i<ncol; i++){
338         start = iat[i];
339         end = iat[i+1];
340         for(int j = start; j<end; j++){
341             coef[j] *= diag[i];
342         }
343     }
344 }
345 }
346
347
348 double* sparse::left_Jacobi(){
349     double* v = (double*) calloc(ncol,sizeof(double));
350     this->getJacobi(v);
351     this->diag_x_sparse(v);
352     return v;
353 }
354
355 #include <iomanip>
356
357 void sparse::printMat() {
358     // Create a 2D matrix to represent the sparse matrix
359     double** matrix = new double*[nrow];
360     for (int i = 0; i < nrow; i++) {
361         matrix[i] = new double[ncol];
362         // Initialize all elements to 0
363         for (int j = 0; j < ncol; j++) {
364             matrix[i][j] = 0.0;
365         }
366     }
367
368     // Fill the matrix with the non-zero elements from the sparse matrix
369     for (int row = 0; row < nrow; row++) {
370         int start = iat[row];           // Starting index for the current row
371         int end = iat[row + 1];         // Ending index for the current row
372
373         // Iterate over the non-zero elements in the current row
374         for (int index = start; index < end; index++) {
375             int column = ja[index];     // Column index of the non-zero element
376             double value = coef[index]; // Value of the non-zero element
377             matrix[row][column] = value; // Store the value in the matrix
378         }
379     }
380
381     // Print the matrix
382     for (int i = 0; i < nrow; i++) {
383         for (int j = 0; j < ncol; j++) {
384             std::cout << setw(8) << matrix[i][j] << " ";
385         }
386         std::cout << endl;
387     }
388
389     // Free the memory allocated for the matrix
390     for (int i = 0; i < nrow; i++) {

```

```

391         delete[] matrix[i];
392     }
393     delete[] matrix;
394 }
395
396 void sparse::printComponents(){
397     std::cout<<"Coef: ";
398     for (int i=0; i<n_term; i++){std::cout<<coef[i]<<" ";}
399     std::cout<<std::endl<<"ja: ";
400     for (int i=0; i<n_term; i++){std::cout<<ja[i]<<" ";}
401     std::cout<<std::endl<<"iat: ";
402     for (int i=0; i<nrow+1; i++){std::cout<<iat[i]<<" ";}
403     std::cout<<std::endl;
404 }
405
406 #endif

```
