DELFT UNIVERSITY OF TECHNOLOGY

Introduction to High Performance Computing WI4049TU

Lab Report

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General Remarks

This final Lab report includes the answers for the exercises (base grad denoted in paranthesis):

- 0. Introductory exercise (0.5)
- 1. Poisson solver (1.75)
- 2. Finite elements simulation (1.0)
- 3. Eigenvalue solution by Power Method on GPU (1.75)

The optional **shining points** (e.g., performance analysis, optimization, discussion, and clarifying figures) which yield further points are usually marked by a small blue heading in the text or an additional note is added under a figure or table. For example:

This is a shining point.

0 Introductory exercise

In the introductory lab session, we are taking a look at some basic features of MPI. We start out very simple with a hello world program on two nodes.

Hello World

```
#include "mpi.h"
#include <stdio.h>

int np, rank;

int main(int argc, char **argv)

{
    MPI_Init(&argc, &argv);
    MPI_Comm_size(MPI_COMM_WORLD, &np);
    MPI_Comm_rank(MPI_COMM_WORLD, &rank);

printf("Node %d of %d says: Hello world!\n", rank, np);

MPI_Finalize();
return 0;

}
```

This program can be compiled with the following command:

```
mpicc -o helloworld1.out helloworld1.c
```

And run with:

```
srun -n 2 -c 4 --mem-per-cpu=1GB ./helloworld1.out
```

We get the following output:

```
Node 0 of 2 says: Hello world!
Node 1 of 2 says: Hello world!
```

From now on I'll skip the compilation and only mention on how many nodes the program is run and what the output is / interpretation of the output.

0.a) Ping Pong

I used the template to check how long MPI_Send and MPI_Recv take. The code can be found in the appendix for this section.

I've modified the printing a bit to make it easier to gather the information. Then I piped the program output into a textfile for further processing in python. I ran it first on one and then on two nodes as specified in the

assignment sheet. Opposed to the averaging over 5 send / receive pairs, I've done 1000 pairs. Furthmore I reran the whole programm 5 times to gather more data. All this data is shown in the following graph:



Figure 1: Ping Pong: Number of bytes sent vs. average time taken from 1000 pairs of send / receive. 5 runs shown for each size as scatter plot. Mean of these 5 runs shown as line. Blue small fit includes all data points up to 131072 bytes, blue large from there. Red small fit includes all data points up to 32768 bytes, red large from there.

As can be seen in the data and the fits, there are outliers especially for the larger data sizes. For our runs we get the following fits and Rš values:

Run Type	Data Size	Fit Equation	Rš Value
Single Node	Small (<=131072)	$5.95 \times 10^{-7} \cdot x + 7.97 \times 10^{-4}$	0.92
Single Node	Large ($>= 131072$)	$4.61 \times 10^{-7} \cdot x + 1.23 \times 10^{-2}$	0.89
Two Node	Small (<=32768)	$1.07 \times 10^{-6} \cdot x + 2.60 \times 10^{-3}$	0.97
Two Node	Large (>=32768)	$4.41 \times 10^{-7} \cdot x + 3.42 \times 10^{-3}$	0.97

Table 1: Fit Equations and Rš Values for Single Node and Two Node Runs

Note: Each run was performed 5 times (for 1 and 2 nodes) to get a fit on the data and calculate a Rš value. TODO: **Further analysis needed?**

Extra: Ping Pong with MPI_SendRecv

We do the same analysis for the changed program utilizing MPI_SendRecv. The code can be found in the appendix for this section.

We get the following graph from the measurements which were performed in the same way as for the previous program:



Figure 2: Ping Pong with MPI_SendRecv: Number of bytes sent vs. average time taken from 1000 pairs of send / receive. 5 runs shown for each size as scatter plot. Mean of these 5 runs shown as line. Blue small fit includes all data points up to 32768 bytes, blue large from there. Red small fit includes all data points up to 32768 bytes, red large from there.

We get the following fits and Rš values for the runs:

Run Type	Data Size	Fit Equation	Rš Value
Single Node	Small (<=32768)	$3.51 \times 10^{-7} \cdot x + 1.19 \times 10^{-3}$	0.98
Single Node	Large (>= 32768)	$1.69 \times 10^{-7} \cdot x + 4.00 \times 10^{-3}$	1.00
Two Node	Small (<=32768)	$6.65 \times 10^{-7} \cdot x + 9.47 \times 10^{-4}$	0.98
Two Node	Large (>= 32768)	$1.66 \times 10^{-7} \cdot x + 4.93 \times 10^{-3}$	1.00

Table 2: Fit Equations and Rš Values for Single Node and Two Node Runs

TODO: Further analysis needed?

0.b) MM-product

After an introduction of the matrix-matrix multiplication code in the next section, the measured speedups are discussed in the subsequent section.

Explanation of the code

For this excercise I've used the template provided in the assignment sheet as a base to develop my parallel implementation for a matrix-matrix multiplication. The code can be found in the appendix for this section.

The porgam can be run either in sequential (default) or parallel mode (parallel as a command line argument). For the sequential version, the code is practically unchanged and just refactored into a function for timing purposes. The parallel version is more complex and works as explained bellow:

First, rank 0 computes a sequential reference solution. Then rank 0 distributes the matrices in the following way in splitwork:

- Matrix A is split row-wise by dividing the number of rows by the number of nodes.
- The first worker (=rank 1) gets the most rows starting from row 0: total_rows (nr_workers 1) $\cdot floor(\frac{total_rows}{nr_workers})$.
- All other workers and the master (= rank 0) get the same number of rows: $floor(\frac{\text{total_rows}}{\text{nr workers}})$.
- The master copies the corresponding rows of matrix A and the whole transposed matrix B* into a buffer (for details on MM_input buffer see bellow) for each worker and sends them off using MPI_ISend.
- The workers receive the data using MPI_Recv and then compute their part of the matrix product and send only the rows of the result matrix back to the master using MPI_Send.
- In the meanwhile the master computes its part of the matrix product.
- Using MPI_Waitall the master waits for all data to be sent to the workers and only afterwards calls MPI_Recv to gather the results from the workers.
- Finally all results are gathered by the master in the result matrix.

Assume we have a 5x5 matrix A and 2 workers (rank 1 and rank 2) and master (rank 0). The partitioning is done row-wise as follows:

Partitioning Example

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \\ a_{41} & a_{42} & a_{43} & a_{44} & a_{45} \\ a_{51} & a_{52} & a_{53} & a_{54} & a_{55} \end{pmatrix} \rightarrow \begin{pmatrix} \text{Worker 1} \\ \text{Worker 1} \\ \text{Worker 1} \\ \text{Master} \\ \text{Master} \end{pmatrix}$$

- Rank 0 (Master): Rows 4 and 5 (last two rows)
- Rank 1 (Worker 1): Rows 1 to 3 (first three rows) Worker 1 always gets the most rows

This partitioning can be visually represented as:

Master (rank 0):
$$\begin{pmatrix} a_{41} & a_{42} & a_{43} & a_{44} & a_{45} \\ a_{51} & a_{52} & a_{53} & a_{54} & a_{55} \end{pmatrix}$$
Worker 1 (rank 1):
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \end{pmatrix}$$

Each worker computes its part of the matrix product, and the master gathers the results at the end and compiles them into the final matrix.

The MM_input buffer is used to store the rows of matrix A and the whole matrix B for each worker. It is implemented using a simple struct:

```
typedef struct MM_input {
    size_t rows;
double *a;
double *b;
} MM_input;
```

*[Optimization] Note on transposed matrix B: It is usually beneficial from a cache perspective to index arrays sequentially or in a row-major order. However, in the matrix-matrix multiplication, we access the elements of matrix B in a column-wise order. This leads to cache misses and is not optimal. To mitigate this, we can transpose matrix B and then access it in a row-wise order. This is done in the code by the master before sending the data to the workers.

Discussion of the speedups

The code was run on Delft's cluster with 1, 2, 4, 8, 16, 24, 32, 48, and 64 nodes. For the experiments the matrix size of A and B was set to 2000×2000 . This means that the program has to evaluate 2000 multiplications and 1999 additions for each element of the resulting matrix C. In total this results in $\approx 2000^3 = 8 \times 10^9$ operations. The command looked similar to the following for the different node counts:

srun -n 48 --mem-per-cpu=4GB --time=00:02:00 ./MM.out parallel

For this experiment, the execution time was measured and the speedup was calculated. The results are shown in Table 3 and Figure 3.

CPU Count	Execution Time / s	Approx. Speedup
1	47.11	1.0
2	10.26	4.6
4	10.30	4.6
8	5.20	9.1
16	2.97	15.9
24	2.54	18.5
32	2.29	20.6
48	2.98	15.8
64	1.72	27.4

Table 3: Execution Time vs CPU Count



Figure 3: Speedup vs CPU Count Black \times marks the average of the rerun for n=48.

Note: The speedup is calculated as $S = \frac{T_1}{T_p}$, where T_1 is the execution time on 1 node and T_p is the execution time on p nodes.

Discussion:

As one can cleary discern from the data in Table 3 and Figure 3, the speedup increases with the number of nodes (with the exception of n = 48). This is expected as the more nodes we have, the more work can be done in

parallel. However, the speedup is not linear. This is due to the overhead of communication between the nodes. The more nodes we have, the more communication is needed, and this overhead increases. This is especially visible in the data for n = 48. Here the speedup is lower than for n = 32. For this run the communication didn't went as smooth as for the other runs. This can potentially be attributed to the fact that one (or more) of the nodes or the network was under heavy load during this task.

[Further investigation] After observing this slower speed for the n=48, I reran the tests multiple times and got a runtime of around 1.9s which was to be expected initially. Therefore, this one run is an odd one out, most likely due to the reasons mentioned above! I've also added the averaged data of the reruns as a datapoint in Figure 3.

Another interesting fact can be seen when comparing the time taken for n = 1 and n = 2. They don't at all scale with the expected factor of 2. This is could be due to the fact, that the resource management system prefers runs with multiple nodes instead of a single node (= sequential).

Additional notes: The flag <code>-mem-per-cpu=<#>GB</code> was set depending on the number of nodes used. For 1-24 nodes 8GB was used, for 32-48 nodes 4GB, and for 64 nodes 3GB. This had to be done to comply with QOS policy on the cluster.

TODO: Data locality?

1 Poisson solver

In this section of the lab report, we will dicuss a prallel implementation of the Poisson solver. The Poisson solver is a numerical method used to solve the Poisson equation, which is a partial differential equation that is useful in many areas of physics.

Note: For local testing and development I'll run the code with mpirun instead of the srun command on the cluster.

1.1 Building a parallel Poisson solver

For the first part of the exercise we follow the steps lined out in the assignment sheet. I'll comment on the steps 1 through 10 and related questions bellow. The finished implementation can be found in the appendix for this section.

1. **Step:** After adding MPI_Init and MPI_Finalize, we can run the program with multiple processes. We can see that the program runs with 4 processes in Figure 4 via the quadrupeled output.

```
etschgi1@Deep-Thought:~/REPOS/HPC/01_lab1/src$ mpirun -np 4 ./mpi.out
Number of iterations : 2355
Number of iterations : 2355
Elapsed processortime: 0.133189 s
Number of iterations : 2355
Elapsed processortime: 0.134150 s
Elapsed processortime: 0.134474 s
Elapsed processortime: 0.135356 s
```

Figure 4: MPI Poisson after Step 1 - Running with 4 processes

2. **Step:** To see which process is doing what, I included the rank of the process for the print statements as shown in Figure 5.

Figure 5: MPI_Poisson after Step 2 - Running with 4 processes

3. **Step:** Next we define wtime as a global double and replace the four utility timing functions with the ones given on Brightspace. A quick verification as shown in Figure 6 shows that the program still runs as expected.

Figure 6: MPI Poisson after Step 3 - Running with 4 processes

- 4. **Step:** Next we check if two processes indeed give the same output. Both need 2355 iterations to converge and the diff command returned no output, which means that the files content is identical.
- 5. **Step:** Now only the process with rank 0 will read data from files and subsequently broadcast it to the others. Testing this again with 2 processes, we see an empty diff of the output files and the same number of iterations needed to converge.

6. **Step:** We create a cartesian grid of processes using MPI_Cart_create and use MPI_Cart_shift to find the neighbors of each process. We can see that the neighbors are correctly identified in Figure 7.

```
(0) (x,y)=(0,0)

(0) top 1, right -2, bottom -2, left 2

(1) (x,y)=(0,1)

(1) top -2, right -2, bottom 0, left 3

(2) (x,y)=(1,0)

(2) top 3, right 0, bottom -2, left -2

(3) (x,y)=(1,1)

(3) top -2, right 1, bottom 2, left -2
```

Figure 7: MPI_Poisson after Step 6 - Running with 4 processes on a 2x2 grid

When there is no neighbor in a certain direction, -2 (or MPI_PROC_NULL) is returned.

7. **Step:** We overhaul the setup to get a proper local grid for each process. Furthermore, we only save the relevant source fields in the local grid for each process.

With for instance 3 processes you should see that 1 or 2 processes do not do any iteration. Do you understand why?

If we have a look at the input file we see that there are only 3 source fields in the grid. This means that the process that does not have a source field in its local grid will not do any iterations (or only 1). Therefore, if we have 3 processes and the distribution of source fields as given in the input file only 1 process will do iterations if processes are ordered in x-direction and 2 if ordered in y-direction. From this we can conclude that indeed all processes have different local grids and perform different calculations.

```
• etschgi1@Deep-Thought:~/REPOS/HPC/02_lab1/src$ mpirun -np 3 ./mpi.out 3 1
(0) (x,y)=(0,0)
(0) top -2, right -2, bottom -2, left 1
(1) (x,y)=(1,0)
(2) (x,y)=(2,0)
(3) Number of iterations: 1
(2) Elapsed Wtime 0.000618 s (95.3% CPU)
(3) Elapsed Wtime 0.000477 s (95.2% CPU)
(4) Number of iterations: 695
(5) Elapsed Wtime 0.017636 s (95.3% CPU)
(6) Elapsed Wtime 0.017636 s (95.3% CPU)
(7) Elapsed Wtime 0.017636 s (95.3% CPU)
(8) Elapsed Wtime 0.017636 s (95.3% CPU)
(9) Elapsed Wtime 0.017636 s (95.3% CPU)
(10) Elapsed Wtime 0.017636 s (95.3% CPU)
(21) Elapsed Wtime 0.017636 s (95.3% CPU)
(22) Elapsed Wtime 0.017636 s (95.3% CPU)
(23) Elapsed Wtime 0.017636 s (95.3% CPU)
(24) Elapsed Wtime 0.017636 s (95.3% CPU)
(25) Elapsed Wtime 0.017636 s (95.3% CPU)
```

Figure 8: MPI_Poisson after Step 7 - Running with 3 processes on a 3x1 (left) vs. 1x3 (right) grid For the 3x1 grid, only rank 1 does iterations (> 1), for the 1x3 grid, ranks 0 and 2 do iterations (> 1).

- 8. **Step:** After defining and committing two special datatypes for vertical and horizontal communication, we setup the communication logic to exchange the boundary values between the processes. We call our Exchange_Borders function after each iteration (for both red / black grid points). Now we face the problem in which some processes may stop instantly (no source in their local grid). They will not supply any data to their neighbors, which will cause the program to hang. We shall fix this in the next step.
- 9. **Step:** Finally we need to implement the logic to check for convergence (in a global sense). We do this by using a MPI_Allreduce call with the MPI_MAX operation. This way we aggregate all deltas and choose the biggest one for the global delta which we use in the while-loop-condition to check for convergence. We can see that the program now runs as expected in Figure 9.

```
(0) (x,y)=(0,0)

(0) top -1, right 2, bottom 1, left -1

(1) (x,y)=(0,1)

(1) top 0, right 3, bottom -1, left -1

(2) (x,y)=(1,0)

(2) top -1, right -1, bottom 3, left 0

(3) (x,y)=(1,1)

(3) top 2, right -1, bottom -1, left 1

(0) Number of iterations : 2355

(1) Number of iterations : 2355

(2) Number of iterations : 2355

(3) Number of iterations : 2355

(3) Number of iterations : 2355

(1) Elapsed Wtime 0.287549 s (99.9% CPU)

(2) Elapsed Wtime 0.287537 s (100.0% CPU)

(3) Elapsed Wtime 0.287537 s (100.0% CPU)

(6) Elapsed Wtime 0.295957 s (99.9% CPU)
```

Figure 9: MPI_Poisson after Step 9 - Running with 4 processes on a 2x2 grid

Note that this run in Figure 9 was done with another pc and another MPI implementation. Therefore, we see -1 for cells without a neighbor! However, other than that cosmetic difference it has no impact on the programm.

10. **Step:** Now we only have to fix two remaining things. First we have to make sure that each process uses the right global coordinates for the output file in the end. Therefore, we change the function a bit to include the specific x-/y-offset for each processor. The second thing is the potential problem, that different processors might start with different (red/black) parities. In order to accomplish a global parity we simply have to change the calculation in the if in Do_Step from

```
if ((x + y) % 2 == parity && source[x][y] != 1)

to

if ((x + offset[X_DIR] + y + offset[Y_DIR]) % 2 == parity && source[x][y] != 1)
```

this guarantees that during a given iteration all processors are using the same parity.

This just leaves one question open: Are the results acutally the same?

Checking the output files of the MPI-implementation with the sequential reference indeed shows identical numerical values for the calculated points. Furthermore, the needed iteration count is also identical which isn't a big surprise, given that the two programms perform the exact same calculation steps.

1.2 Exercises, modifications, and performance aspects

For this subsection we'll define the following shorthand notation:

n:	the number of iterations		
<i>g</i> :	gridsize		
t:	time needed in seconds		
pt:	processor topology in form pxy , where:		
p:	number of processors used		
x:	number of processors in x-direction		
y:	number of processors in y-direction		

Table 4: Notation for this section

pt = 414 means 4 processors in a 1×4 topology.

Note on different Versions:

For the following exercises the implementation will be slightly adapted to measure different performance aspects. To facilitate this, we will use defines to switch between different versions of the code at compile time. The final version of the poissonsolver can be found in the appendix for this section.

1.2.1 Over-relaxation (SOR)

We start of by rewriting the Do_Step routine to facilitate SOR updates. Furthermore, we need h^2 , the grid spacing (which is 1 in our case) and the relaxation parameter ω to calculate the updated values. A quick test shows a speedup of roughly a factor of 10. More systematic tests will be done in the next section.

1.2.2 Optimal ω for 4 proc. on a 4x1 grid

With the power of a little python scripting we can easily test different values for ω and plot the results as seen in Figure 10.

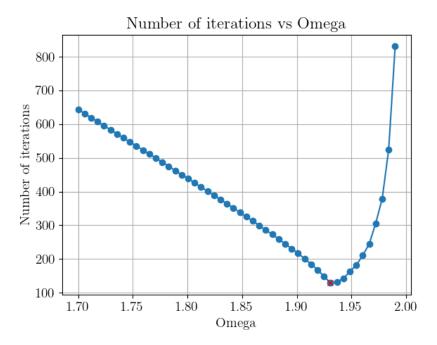


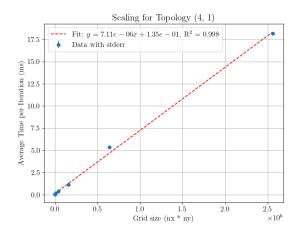
Figure 10: Optimal ω for 4 processors on a 4x1 grid

We find that the optimal ω is at about 1.93 for this setup with only 129 iterations. This constitutes a speedup of about 1825% compared to the sequential implementation.

N.B.: If not stated otherwise, we will use $\omega = 1.93$ for the following exercises.

1.2.3 Scaling behavior with larger grids

This investigation is carried out twice: Once with a 4×1 topology (as in the previous section) and once with a 2×2 topology. We use grid sizes of 10×10 , 25×25 , 50×50 , 100×100 , 200×200 , 400×400 , 800×800 and 1600×1600 and set $\omega = 1.95$ for all runs. The results are shown in Figure 11.



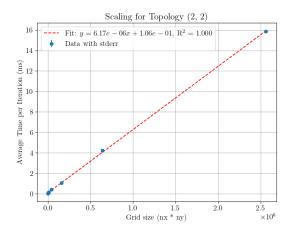


Figure 11: Scaling behavior of the Poisson solver with different grid sizes and processor topologies

As seen by the high R^2 values in the plots, the scaling behavior is very close to linear. We obtain the following scaling factors for the different grid sizes and topologies from the linear fits:

Topology	α	β
4×1	$1.35 + 10^{-1}$	$7.11 + 10^{-6}$
2×2	$1.06 + 10^{-1}$	$6.17 + 10^{-6}$

Table 5: Scaling factors for different processor topologies for the Poisson solver Using: $t(n) = \alpha + \beta \cdot n$ as a model

What can you conclude from the scaling behavior?

We see that the scaling behavior is very close to linear for both topologies. This means that the parallel implementation scales as expected with the number of grid points.

If we compare the scaling factors (β) for the two topologies we see that the 2×2 topology scales slightly better than the 4×1 topology. This is not surprising, as the 2×2 topology has a more balanced communication workload balance. In the 2×2 topology every processor has two neighbors, while in the 4×1 topology the processors at the ends only have one neighbor. This is a general trend: A topology which divides the grid into square / square-like parts will scale better than a topology which divides the grid into long and thin parts.

In essence: We want to keep the communication between processors as balanced as possible to achieve the best scaling behavior.

1.2.4 Scaling behavior [Theory - no measurements]

If I could choose between a 16×1 , 8×2 , 4×4 , 2×8 , 1×16 topology, I would choose the 4×4 topology. This is because the 4×4 topology has the most balanced communication workload balance, as detailed in the **Shining** in subsubsection 1.2.3.

1.2.5 Iterations needed for convergence scaling

We investigate the number of iterations needed for convergence using the 4×1 topology square grids with sidelength: 10, 25, 50, 100, 200, 400, 800, 1600. The results for different ω are shown in Figure 12.

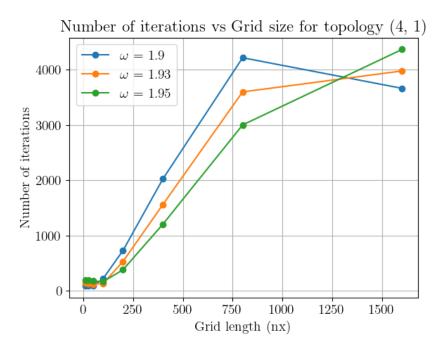


Figure 12: Iterations needed for convergence with different grid side lengths

We can clearly see that the number of iterations till convergence increases with the problem size. At first, I expected linear growth proportional to the number of gridpoints. However, it turns out that the number of iterations actually grow slower and in a square root like fashion. This can be seen by the linear behavior in the plot of grid-side length against iterations.

Why is the number of iterations needed for convergence $\propto \sqrt{g}$?

Our poisson problem is a discretized system in 2D space. The condition number of the matrix we have to solve is proportional to the number of gridpoints in our system. SOR uses the spectral properties of the matrix to solve in a way such that the dominant error mode takes time proportional to the diameter of the domain to converge. This means it is proportional to $\sqrt{g} = \sqrt{n_x \cdot n_y}$.

Why does omega with the best performance change with the grid size?

As can be seen in Figure 12 $\omega=1.9$ beats the other two values for very small and the largest gridsize. For different gridsizes we get differently sized matrices we have to solve. SOR overrelaxes high-frequency errors and underrelaxes low-frequency errors (the later for stability). The optimal ω is therefore dependent on the gridsize and the error modes present in the system. In our current example, it might be that $\omega=1.9$ is a good compromise for the grid sizes we are looking at.

1.2.6 Error as a function of the iteration number

With the same 4×1 topology and grid sizes of 800×800 the error for 15000 iterations is tracked using $\omega = 1.93$. The results are shown in Figure 13.

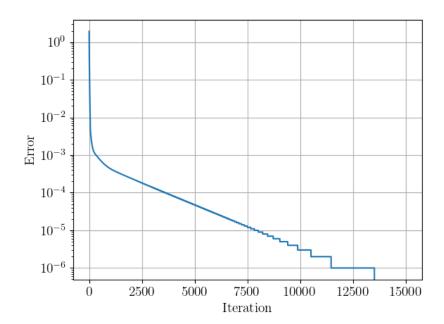


Figure 13: Error as a function of the iteration number

At first the error decreases rapidly in the first few iterations to about 10^{-3} (logarithmic scale!). After that the error decreases more slowly until it is below floating point precision. **Note:** All calculations are done using double precision floating point numbers and only the error recording was done using single precision which leaves the step-like artifacts in the plot.

1.2.7 Optional - Gain performance by reducing MPI_Allreduce calls

The last subsection showed us that the error reduces monotonically. We might be able to save some time by leaving out some checks and maybe check the global error every 10th or 100th iteration only.

First, we should benchmark if it is at all wise to optimize here, by measuring how long the MPI_Allreduce call takes. We can do this by measuring the time needed for the MPI_Allreduce call in the Do_Step function and summing up to get the total time spent in MPI_Allreduce calls.

We again solve with a 4×1 topology, $\omega = 1.93$ and a 800×800 grid: It takes roughly 20 seconds of which the processors spend around 1 - 2 seconds in the MPI_Allreduce call. This is a significant amount of time $((7.0 \pm 0.4)\%)$. This means we would save some time by reducing the number of MPI_Allreduce calls and calculating 9 (0.25%) of total) more iterations wouldn't hurt us too bad because it takes 3601 to converge!

We run the program three times with MPI_Allreduce calls every 1, 10 and 100 iterations and get the speedups in MPI_Allreduce calls as shown in Table 6.

Iterations	MPI_Allreduce - speedup (factor)	calculated overall speedup (%)	
1	1.00	0	
10	6.0 ± 2.0	5.9 ± 0.5	
100	62 ± 6	6.9 ± 0.4	

Table 6: Speedup in MPI_Allreduce calls for different iteration counts and calculated overall speedup (%)

As can be clearly seen from the table we can gain around 6 % using MPI_Allreduce calls every 10 iterations and around 7 % using MPI_Allreduce calls every 100 iterations. This is a significant speedup for a very small change in the code.

Note: The speedup is calculated to account for fluctuations in the runtime of the program, due to other processes running on the same machine / cluster.

1.2.8 Reduce border communication

Another way to reduce communication overhead is to reduce the number of border exchanges. To investigate if this yields a speedup we run the program on a 4×1 topology, $\omega = 1.93$ and different grid sizes and track the iterations and time as seen in Figure 14.

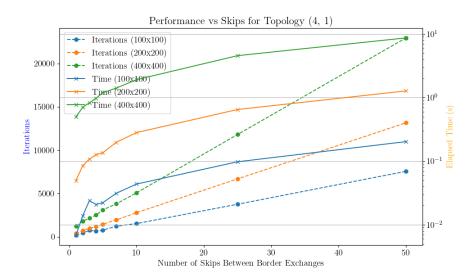


Figure 14: Speedup by reducing border exchanges

Running the with different numbers of skipped border exchanges naturally slows down convergence, meaning we need more iterations to reach the same error. For all tested grid sizes the initial SOR version without skipping border exchanges has the fewest iterations needed to convergence and also the fastes runtime.

What can you conclude from the results?

We can conclude that reducing the number of border exchanges does not yield a speedup. The reason for this is that we have to calculate more iterations to converge to the solution which outweighs the gains from reduced communication overhead. Interestingly, for the 100×100 grid there exists a local minimum in time at 4 skipped border exchanges compared to 3 skipped. This is likely due to our source field distribution and thus specific to our problem.

1.2.9 Optimize Do_Step loop

In Do_Step we iterate over the whole grid but only update one of the two parities at a time. This means we can split the loop into two loops, one for each parity. We start out with something like this:

```
for (x = 1; x < dim[X_DIR] - 1; x++){
    for (y = 1; y < dim[Y_DIR] - 1; y++){
        if ((x + offset[X_DIR] + y + offset[Y_DIR]) % 2 == parity && source[x][y] != 1){
        ...</pre>
```

and we change it to:

```
int start_y;
for (x = 1; x < dim[X_DIR] - 1; x++){
    start_y = ((1 + x + offset[X_DIR] + offset[Y_DIR]) % 2 == parity) ? 1 : 2;
    for (y = start_y; y < dim[Y_DIR] - 1; y += 2){
        if (source[x][y] != 1){
            ...</pre>
```

The basic idea is to avoid y-coordinates which are not in the parity we are currently updating. We measure 10 runs for a 800×800 grid and a 4×1 topology with $\omega = 1.93$ and get the following times:

```
t_{\text{no improvements}} = (5.59 \pm 0.05) \,\text{s} and t_{\text{loop improvements}} = (4.64 \pm 0.07) \,\text{s}
```

So we get a minimal speedup of about 17% by optimizing the loop which is a enormous speedup for such a small change.

Why does this make such a difference

The reason for this is that we avoid unnecessary looping and if statements. This means that we have less overhead in the loop and can therefore calculate faster by skipping the unnecessary loop entries.

1.2.10 Optional - Time spent within Exchange_Borders

We can measure the time spent in Exchange_Borders by adding a timer to the function. We run the program with $\omega = 1.93$ and different topologies and grid sizes and get the results shown in Figure 15.

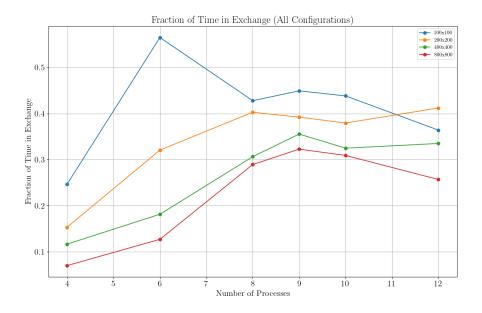


Figure 15: Fraction of total time spent in Exchange_Borders

As we can clearly see, the time spent in Exchange_Borders is initially smaller and grows with processor count. The curves are generally shifted downward for larger gridssizes.

Interpretation:

One would expect larger grid sizes to be more computationally expensive and we have already established that iterations take longer the bigger the grid. Communication obviously also takes longer for a larger grid because we have more data to sent. However, the circumference of a square grows linearly, while the area grows quadratically. The quadratic growth of the area is the reason for the downward shift in the curves for larger grid sizes because the communication overhead grows slower than the computational overhead for larger grids.

When is the time spent in Exchange_Borders significant / comparable to computation?

As can be seen in Figure 15 the time spent in Exchange_Borders is significant for all grid sizes from the start (between 5 and 25%). Thereafter it grows to around 30% to 45% for 9 processors. This means that the time spent in Exchange_Borders is significant for all grid sizes and processor counts, but especially as the processor count grows.

1.2.11 Latency and bandwith in Exchange_Borders

We use the configurations from subsubsection 1.2.3: 4×1 , 2×2 and 3×3 topologies with grid sizes of 200×200 , 400×400 and 800×800 and $\omega = 1.95$. We obtain the results in Table 7.

Table 7: Metrics for Exchange_Borders latency and bandwith

Topology	Grid Size	Latency (ms)	Latency (%)	Bandwidth $(B s^{-1})$	Total Data (B)
4x1	200x200	370 ± 44	6.6	447314206	135548032
4x1	400x400	$328\pm\ \ 39$	6.0	511 267 773	135548032
4x1	800x800	401 ± 38	7.1	390446955	135548032
2x2	200x200	346 ± 83	6.2	426689734	108546432
2x2	400x400	281 ± 74	5.2	576865986	108546432
2x2	800x800	288 ± 71	5.3	558562482	108546432
3x3	200x200	1054 ± 204	25.2	78 386 044	72364288
3x3	400x400	1147 ± 229	27.6	81 623 075	72364288
3x3	800x800	1082 ± 148	26.7	60 819 112	72499296

Interpretation:

As can be seen the latency is more or less comparable between the 4×1 and 2×2 topologies. The 3×3 topology has a significantly higher latency. This is due to the fact that the 3×3 topology has more neighbors to communicate with.

1.2.12 Exchange Border potential improvements

Indeed we communicate twice as much as we need after each Do_Step call. We shall analyze this considering the following points:

- address of the first point to exchange: This depends on the parity of the processor. It is simply the first or second point in the grid depending on the parity (leaving out the edge case where the processor only has one data-point).
- the number of points to exchange: This normally is the number of points in the grid minus ghost cells divided by two. However, this may change if there is an odd number of points in the grid (then we have to exchange one more or one less point).
- the number of points in between grid points that have to be exchanged: The stride of the data will change. Currently we exchange every point for one direction and every dim[Y_DIR]-th point for the other direction. This can be optimized to exchange every second point in one direction and every second dim[Y_DIR]-th point in the other direction.

Is it worth it?

For smaller gridsizes it is not worth it to optimize the border exchange. The time spent in the border exchange might be significant in relative terms, but the absolute time is still small. For larger gridsizes it might be worth it to optimize the border exchange. As we have seen, this becomes more significant as the processor count and gridsizes grow. For our current problem and similarly sized problems the effort put into optimizing the border exchange is certainly not worth it.

2 Finite elements simulation

We will now shift our focus to a more general grid which is based on triangulation. In this section we will compare our parallel implementation from the previous section and disect the differences and similarities. For that reason we shall use the same sources in our grid as given in sources.dat.

Note that the sections for the exercises will be labeled as (2.1, 2.2, 2.3, ...) corresponding to exercises (4.1, 4.2, 4.3, ...) in the lab manual.

2.1 Code understanding & Exchange_Borders

The first step is to read through the code in **MPI_Fempois.c** and to understand it. Furthermore, we have to implement the **Exchange_Borders** function for which only a skeleton is given. The function should exchange the border values of the local grid with the neighboring processes.

The implementation of this function is quite straight forward. We only have to loop over all the neighbors of a process and send out the border values and receive the border values from the neighbors. The function is implemented as follows:

2.2 Time benchmarking

Next we turn our attention to timing of different sections in the code. We have to measure:

- Time spent in computation
- Time spent exchanging information with neighbors
- Time spent doing global communication
- Idle time

We setup the following variable to measure / deduce the time spent in the different sections:

```
double total_time = 0.0;
double exchange_time_neighbors = 0.0;
double exchange_time_global = 0.0;
double compute_time = 0.0;
```

We will measure the time spent in computations by timing the solve function and subtracting the time spent in the MPI_Allreduce calls. The time spent in the MPI_Allreduce calls is the time spent in global communication. The time spent in exchanging information with neighbors is the time spent in the Exchange_Borders function. Finally, the idle time can be determined by summing up the differences between the cores total time and the slowest cores total time. Note that this way of calculating the idle time is an approximation since it is assumed that the slowest core doesn't have any idle time. However, I've discussed this with the TA and he said that this is a valid way of calculating the idle time for this exercise.

The commandline output for runs is as shown in Figure 16 for an example run:

```
(2) - Exchange time (neighbors): 0.007183
(2) - Exchange time (global): 0.002372
(2) - Sum of times (compute + exchange (global & local)): 0.062241
(2) - Total time: 0.069645
(0) - Total time: 0.069800
(1) - Compute time: 0.051027
(1) - Exchange time (neighbors): 0.005517
(1) - Exchange time (global): 0.004273
(1) - Sum of times (compute + exchange (global & local)): 0.060816
(1) - Total time: 0.069416
(3) - Compute time: 0.051977
(3) - Exchange time (neighbors): 0.006824
(3) - Exchange time (global): 0.003341
(3) - Sum of times (compute + exchange (global & local)): 0.062142
(3) - Total time: 0.069399
```

Figure 16: Timing for the different sections.
(x) ... denotes the process rank
Exchange time (neighbors) ...

3 Eigenvalue solution by Power Method on GPU

Appendix - Introductory exercise

The following code was used for the ping pong task:

```
#include <stdio.h>
#include <stdlib.h>
3 #include <mpi.h>
5 // Maximum array size 2^20= 1048576 elements
6 #define MAX_EXPONENT 20
7 #define MAX_ARRAY_SIZE (1<<MAX_EXPONENT)</pre>
8 #define SAMPLE_COUNT 1000
int main(int argc, char **argv)
11
      // Variables for the process rank and number of processes
12
13
      int myRank, numProcs, i;
      MPI_Status status;
14
15
      // Initialize MPI, find out MPI communicator size and process rank
16
      MPI Init(&argc, &argv);
17
      MPI_Comm_size(MPI_COMM_WORLD, &numProcs);
18
19
      MPI_Comm_rank(MPI_COMM_WORLD, &myRank);
20
21
      int *myArray = (int *)malloc(sizeof(int)*MAX_ARRAY_SIZE);
22
      if (myArray == NULL)
24
          printf("Not enough memory\n");
25
26
          exit(1);
27
      // Initialize myArray
28
      for (i=0; i<MAX_ARRAY_SIZE; i++)</pre>
29
          myArray[i]=1;
30
31
      int number_of_elements_to_send;
      int number_of_elements_received;
33
34
35
      // PART C
      if (numProcs < 2)
36
37
          printf("Error: Run the program with at least 2 MPI tasks!\n");
38
          MPI_Abort(MPI_COMM_WORLD, 1);
39
40
      double startTime, endTime;
41
42
43
      // TODO: Use a loop to vary the message size
      for (size_t j = 0; j <= MAX_EXPONENT; j++)</pre>
44
45
          number_of_elements_to_send = 1<<j;</pre>
46
          if (myRank == 0)
47
          {
              myArray[0]=myArray[1]+1; // activate in cache (avoids possible delay when sending
49
      the 1st element)
              startTime = MPI_Wtime();
              for (i=0; i<SAMPLE_COUNT; i++)</pre>
51
                  MPI_Send(myArray, number_of_elements_to_send, MPI_INT, 1, 0,
53
                       MPI_COMM_WORLD);
54
                  MPI_Probe(MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &status);
55
                  MPI_Get_count(&status, MPI_INT, &number_of_elements_received);
56
57
58
                  MPI_Recv(myArray, number_of_elements_received, MPI_INT, 1, 0,
                      MPI_COMM_WORLD, MPI_STATUS_IGNORE);
59
              } // end of for-loop
60
61
              endTime = MPI_Wtime();
62
              number_of_elements_received,(endTime - startTime)/(2*SAMPLE_COUNT));
64
          }
65
          else if (myRank == 1)
66
67
               // Probe message in order to obtain the amount of data
              MPI_Probe(MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &status);
68
```

```
MPI_Get_count(&status, MPI_INT, &number_of_elements_received);
69
70
                   for (i=0; i<SAMPLE_COUNT; i++)</pre>
71
                   {
                         MPI_Recv(myArray, number_of_elements_received, MPI_INT, 0, 0,
73
                         MPI_COMM_WORLD, MPI_STATUS_IGNORE);
74
                         {\tt MPI\_Send} \, ({\tt myArray} \, , \, \, {\tt number\_of\_elements\_to\_send} \, , \, \, {\tt MPI\_INT} \, , \, \, {\tt 0} \, , \, \, {\tt 0} \, , \, \,
75
76
                        MPI_COMM_WORLD);
77
                   } // end of for-loop
              }
78
79
80
        // Finalize MPI
81
        MPI_Finalize();
83
84
        return 0;
```

For the bonus task, the following code was used:

```
#include <stdio.h>
#include <stdlib.h>
3 #include <mpi.h>
5 // Maximum array size 2^20= 1048576 elements
6 #define MAX_EXPONENT 20
7 #define MAX_ARRAY_SIZE (1<<MAX_EXPONENT)</pre>
8 #define SAMPLE_COUNT 1000
int main(int argc, char **argv)
11 {
12
       // Variables for the process rank and number of processes
       int myRank, numProcs, i;
       MPI_Status status;
14
15
       // Initialize MPI, find out MPI communicator size and process rank
16
       MPI_Init(&argc, &argv);
17
       MPI_Comm_size(MPI_COMM_WORLD, &numProcs);
18
       MPI_Comm_rank(MPI_COMM_WORLD, &myRank);
19
20
21
22
       int *myArray = (int *)malloc(sizeof(int)*MAX_ARRAY_SIZE);
      if (myArray == NULL)
23
24
       {
25
           printf("Not enough memory\n");
           exit(1);
26
27
       // Initialize myArray
28
       for (i=0; i<MAX_ARRAY_SIZE; i++)</pre>
29
30
           myArray[i]=1;
31
       int number_of_elements_to_send;
32
       int number_of_elements_received;
33
34
       // PART C
35
       if (numProcs < 2)</pre>
36
       {
37
38
           printf("Error: Run the program with at least 2 MPI tasks!\n");
           MPI_Abort(MPI_COMM_WORLD, 1);
39
40
41
       double startTime, endTime;
42
       // TODO: Use a loop to vary the message size \,
43
       for (size_t j = 0; j <= MAX_EXPONENT; j++)</pre>
44
45
46
           number_of_elements_to_send = 1<<j;</pre>
47
           if (myRank == 0)
           {
48
               myArray[0]=myArray[1]+1; // activate in cache (avoids possible delay when sending
49
       the 1st element)
               startTime = MPI Wtime();
50
51
               for (i=0; i<SAMPLE_COUNT; i++)</pre>
               {
52
53
                    MPI_Sendrecv(myArray, number_of_elements_to_send, MPI_INT, 1,0,myArray,
```

```
number_of_elements_to_send, MPI_INT, 1, 0, MPI_COMM_WORLD, &status);
54
55
                endTime = MPI_Wtime();
56
                \label{lem:printf("Rank %2.1i: Received %i elements: Ping Pong took %f seconds \\ \normalfont{n", myRank,}
57
       number_of_elements_to_send,(endTime - startTime)/(2*SAMPLE_COUNT));
           }
58
59
           else if (myRank == 1)
60
            {
                for (i=0; i<SAMPLE_COUNT; i++)</pre>
61
62
                    MPI_Sendrecv(myArray, number_of_elements_to_send, MPI_INT, 0,0,myArray,
63
       number_of_elements_to_send , MPI_INT , 0 , 0 , MPI_COMM_WORLD , &status);
           }
65
66
67
       // Finalize MPI
68
       MPI_Finalize();
69
70
       return 0:
71
```

The matrix multiplication used the following code:

```
* FILE: mm.c
   * DESCRIPTION:
3
       This program calculates the product of matrix a[nra][nca] and b[nca][ncb],
       the result is stored in matrix c[nra][ncb].
5
      The max dimension of the matrix is constraint with static array
   *declaration, for a larger matrix you may consider dynamic allocation of the
   *arrays, but it makes a parallel code much more complicated (think of
   *communication), so this is only optional.
10
11
13 #include <math.h>
14 #include <mpi.h>
#include <stdbool.h>
16 #include <stdio.h>
#include <stdlib.h>
18 #include <string.h>
19
20 #define NRA 2000 /* number of rows in matrix A */
#define NCA 2000 /* number of columns in matrix A */
_{\rm 22} #define NCB 2000 /* number of columns in matrix B */
23 // #define N 1000
#define EPS 1e-9
#define SIZE_OF_B NCA*NCB*sizeof(double)
26
27 bool eps_equal(double a, double b) { return fabs(a - b) < EPS; }</pre>
28
void print_flattened_matrix(double *matrix, size_t rows, size_t cols, int rank) {
      printf("[%d]\n", rank);
30
      for (size_t i = 0; i < rows; i++) {</pre>
31
          for (size_t j = 0; j < cols; j++) {
32
              printf("%10.2f ", matrix[i * cols + j]); // Accessing element in the 1D array
33
34
          printf("\n"); // Newline after each row
35
36
37 }
38
  int checkResult(double *truth, double *test, size_t Nr_col, size_t Nr_rows) {
39
      for (size_t i = 0; i < Nr_rows; ++i) {</pre>
40
41
          for (size_t j = 0; j < Nr_col; ++j) {</pre>
              size_t index = i * Nr_col + j;
42
              if (!eps_equal(truth[index], test[index])) {
43
                  return 1;
44
              }
45
          }
46
      return 0;
48
49 }
```

```
51 typedef struct {
       size_t rows;
       double *a;
       double *b:
54
55 } MM_input;
56
57 char* getbuffer(MM_input *in, size_t size_of_buffer){
58
       char* buffer = (char*)malloc(size_of_buffer * sizeof(char));
       if (buffer == 0)
59
60
            printf("Buffer couldn't be allocated.");
61
            return NULL;
62
63
       size_t offset = 0;
64
       memcpy(buffer + offset, &in->rows, sizeof(size_t));
65
66
       offset += sizeof(size_t);
       size_t matrix_size = in->rows * NCA * sizeof(double);
67
68
       memcpy(buffer + offset, in->a, matrix_size);
69
       offset += matrix_size;
       memcpy(buffer + offset, in->b, NCA*NCB*sizeof(double));
70
71
       return buffer;
72 }
73
   MM_input* readbuffer(char* buffer, size_t size_of_buffer){
74
       MM_input *mm = (MM_input*)malloc(sizeof(MM_input));
75
76
77
       mm->rows = ((size_t*)buffer)[0];
       size_t offset = sizeof(size_t);
78
       size_t matrix_size = mm->rows * NCA;
79
       mm->a = (double*)malloc(sizeof(double)*matrix_size);
80
       mm->b = (double*)malloc(sizeof(double)*matrix_size);
81
       memcpy(mm->a, &(buffer[offset]), matrix_size);
82
       offset += matrix_size;
83
84
       memcpy(mm->b, &(buffer[offset]), NCA*NCB*sizeof(double));
       free(buffer);
85
86
       return mm;
87 }
88
89
90
   void setupMatrices(double (*a)[NCA], double (*b)[NCB], double (*c)[NCB]){
       for (size_t i = 0; i < NRA; i++) {</pre>
91
92
            for (size_t j = 0; j < NCA; j++) {</pre>
93
                a[i][j] = i + j;
94
96
       for (size_t i = 0; i < NCA; i++) {</pre>
97
            for (size_t j = 0; j < NCB; j++) {</pre>
                b[i][j] = i * j;
99
            }
100
101
102
       for (size_t i = 0; i < NRA; i++) {</pre>
            for (size_t j = 0; j < NCB; j++) {
   c[i][j] = 0;</pre>
104
105
106
108 }
109
double multsum(double* a,double* b_transposed, size_t size){
       double acc = 0;
       for (size_t i = 0; i < size; i++)</pre>
            acc += a[i]*b_transposed[i];
114
116
       return acc;
117 }
118
double productSequential(double *res) {
       // dynamically allocate to not run into stack overflow - usually stacks are
120
        // 8192 bytes big \rightarrow 1024 doubles but we have 1 Mio. per matrix
       double(*a)[NCA] = malloc(sizeof(double) * NRA * NCA);
```

```
double(*b)[NCB] = malloc(sizeof(double) * NCA * NCB);
123
       double(*c)[NCB] = malloc(sizeof(double) * NRA * NCB);
124
125
       /*** Initialize matrices ***/
126
       setupMatrices(a,b,c);
128
       /* Parallelize the computation of the following matrix-matrix
129
130
      multiplication. How to partition and distribute the initial matrices, the
      work, and collecting final results.
132
       // multiply
133
       double start = MPI_Wtime();
for (size_t i = 0; i < NRA; i++) {</pre>
134
135
            for (size_t j = 0; j < NCB; j++) {</pre>
136
                for (size_t k = 0; k < NCA; k++) {</pre>
137
                    res[i * NCB + j] += a[i][k] * b[k][j];
138
139
           }
140
141
       /* perform time measurement. Always check the correctness of the parallel
142
          results by printing a few values of c[i][j] and compare with the
143
           sequential output.
145
       double time = MPI_Wtime()-start;
146
147
       free(a);
       free(b):
148
       free(c);
149
       return time;
150
151 }
   double splitwork(double* res, size_t num_workers){
153
154
       if (num_workers == 0) // sadly noone will help me :((
            printf("Run sequential!\n");
156
            return productSequential(res);
158
159
       double(*a)[NCA] = malloc(sizeof(double) * NRA * NCA);
       double(*b)[NCB] = malloc(sizeof(double) * NCA * NCB);
161
       double(*c)[NCB] = malloc(sizeof(double) * NRA * NCB);
162
       // Transpose matrix b to make accessing columns easier - in row major way - better cache
       performance
       setupMatrices(a,b,c);
164
       double start_time = MPI_Wtime();
166
       double (*b_transposed)[NCA] = malloc(sizeof(double) * NCA * NCB);
       for (size_t i = 0; i < NCA; i++) {</pre>
168
            for (size_t j = 0; j < NCB; j++) {</pre>
169
                b_transposed[j][i] = b[i][j];
170
172
       /*** Initialize matrices ***/
174
       // given number of workers I'll split
       size_t rows_per_worker = NRA / (num_workers+1); //takes corresponding columns from other
176
       matrix
       printf("rows per worker: %zu\n", rows_per_worker);
       size_t row_end_first = NRA - rows_per_worker*num_workers;
178
       printf("first gets most: %zu\n", row_end_first);
179
180
       //setup requests
181
       MPI_Request requests[num_workers];
       MM_input *data_first = (MM_input*)malloc(sizeof(MM_input));
183
184
       data_first->rows = row_end_first;
       data_first->a = (double*)a; //they both start of with no offset!
       data_first->b = (double*)b_transposed;
186
187
       size_t total_size = sizeof(size_t) + (data_first->rows * NCA)*sizeof(double)+SIZE_OF_B;
       char* buffer = getbuffer(data_first, total_size);
                                                             //first one
188
189
        // Tag is just nr-cpu -1
       MPI_Isend(buffer, total_size, MPI_CHAR, 1, 0, MPI_COMM_WORLD, &requests[0]);
191
192
       free(data_first);
       total_size = sizeof(size_t) + (rows_per_worker * NCA)*sizeof(double) + SIZE_OF_B; //size
```

```
is the same for all other - just compute once!
194
       size t i;
       for (i = 0; i < (num_workers-1); ++i)</pre>
195
196
            MM input *data = (MM input*)malloc(sizeof(MM input)):
197
198
            data->rows = rows_per_worker;
            data->a = (double*)(a + (row_end_first + rows_per_worker*i));
199
            data->b = (double*)(b_transposed); // send everyting - all needed
200
            buffer = getbuffer(data, total_size);
201
            printf("nr_worker - %zu\n", i);
202
            MPI_Isend(buffer, total_size, MPI_CHAR, i+2, i+1, MPI_COMM_WORLD, &requests[i+1]);
203
204
            free(data);
205
       double* my_a = (double*)(a + (row_end_first + rows_per_worker*i));
206
207
208
       //I multiply the rest
       size_t offset = 0;
209
       for (size_t row = (NRA-rows_per_worker); row < NRA; row++)</pre>
210
211
            for (size_t col = 0; col < NCB; col++)</pre>
212
213
214
                res[row * NCB + col] = multsum(my_a+offset, (((double*)b_transposed)+col*NCA), NCA
       );
215
           }
216
            offset += NCA;
217
       printf("My c: \n");
218
        //wait for rest
219
       MPI_Status stats[num_workers];
220
       if(MPI_Waitall(num_workers, requests, stats) == MPI_ERR_IN_STATUS){
221
            printf("Communication failed!!! - abort\n");
222
223
       printf(">>>Everything sent and recieved\n");
224
225
226
       // reviece rest
       size_t buf_size = sizeof(double)*row_end_first*NCB;
227
228
       double* revbuf;
       offset = 0;
       for (size_t worker = 0; worker < num_workers; worker++)</pre>
230
231
232
            revbuf = (double*)malloc(buf_size); //first gets largest buffer
            MPI_Recv(revbuf, buf_size/sizeof(double), MPI_DOUBLE, worker+1, worker, MPI_COMM_WORLD
233
        ,&stats[worker]);
            memcpy(&res[offset/sizeof(double)], revbuf, buf_size);
234
235
            free(revbuf):
            offset += buf_size;
236
            buf_size = sizeof(double)*rows_per_worker*NCB;
237
238
       double time = MPI_Wtime()-start_time;
239
       //free all pointers!
240
241
       free(a):
       free(b);
242
       free(b_transposed);
243
       free(c);
       return time;
245
246 }
247
248
249
double work(int rank, size_t num_workers){
       size_t rows_per_worker = NRA / (num_workers+1);
251
252
       char* buffer;
       MPI_Status status;
253
       if (rank == 1) // first always get's most work
254
255
            rows_per_worker = NRA - rows_per_worker*num_workers;
256
257
258
       size_t size_of_meta = sizeof(size_t);
       size_t size_of_a = sizeof(double)*rows_per_worker*NCA;
259
       size_t buffersize = size_of_meta+size_of_a + SIZE_OF_B;
260
       buffer = (char*)malloc(buffersize);
261
262
       MPI_Recv(buffer, buffersize, MPI_CHAR, 0, rank-1, MPI_COMM_WORLD, &status);
```

```
264
       double start = MPI_Wtime();
265
       int count;
       MPI_Get_count(&status, MPI_CHAR, &count);
       printf("I'm rank %d and I got %d bytes (%ld doubles) of data from %d with tag %d.\n", rank
267
        , count, (count-sizeof(size_t))/sizeof(double), status.MPI_SOURCE, status.MPI_TAG);
268
       MM_input *mm = (MM_input*)malloc(sizeof(MM_input));
269
270
       mm->a = (double*)&buffer[size_of_meta];
271
       mm->b = (double*)&buffer[size_of_meta+size_of_a];
272
       double *res =(double*)malloc(sizeof(double)*rows_per_worker*NCB);
273
274
       size t offset = 0:
275
       for (size_t row = 0; row < rows_per_worker; row++)</pre>
276
277
            for (size_t col = 0; col < NCB; col++)</pre>
278
279
            {
                res[row * NCB + col] = multsum(mm->a+offset, (((double*)mm->b)+col*NCA), NCA);
280
281
           offset += NCA;
282
283
284
       MPI_Send(res, rows_per_worker*NCB, MPI_DOUBLE, 0,rank-1, MPI_COMM_WORLD);
       printf("[%d] sent res home\n",rank);
285
286
       free (res);
287
       return MPI_Wtime() - start;
288 }
289
   int main(int argc, char *argv[]) {
290
       int tid, nthreads;
291
       /* for simplicity, set NRA=NCA=NCB=N */
292
       // Initialize MPI, find out MPI communicator size and process rank
293
294
       int myRank, numProcs;
       MPI_Status status;
295
       MPI_Init(&argc, &argv);
296
       MPI_Comm_size(MPI_COMM_WORLD, &numProcs);
297
       MPI_Comm_rank(MPI_COMM_WORLD, &myRank);
298
       int num_Workers = numProcs-1;
299
        if (argc > 1 && strcmp(argv[1], "parallel") == 0) {
           // Variables for the process rank and number of processes
301
           if (myRank == 0) {
302
303
                printf("Run parallel!\n");
                double *truth = malloc(sizeof(double) * NRA * NCB);
304
                double time = productSequential(truth);
305
                printf("Computed reference results in %.6f s\n", time);
306
                printf("Hello from master! - I have %d workers!\n", num_Workers);
307
                // send out work
308
                double *res = malloc(sizeof(double)*NRA*NCB);
309
310
                time = splitwork(res, num_Workers);
                if (checkResult(res, truth, NCB, NRA)) {
311
                    printf("Matrices do not match!!!\n");
312
                    return 1;
313
314
                printf("Matrices match (parallel [eps \%.10f])! - took: \%.6f s\n", EPS, time);
315
                free(truth);
                free(res);
317
318
           } else {
319
                double time = work(myRank, num_Workers);
                printf("Worker bee %d took %.6f s (after recv) for my work\n", myRank, time);
320
321
322
       } else // run sequantial
323
           printf("Run sequantial!\n");
325
            double *res = malloc(sizeof(double) * NRA * NCB);
326
            double time = productSequential(res);
327
           if (checkResult(res, res, NCB, NRA)) {
328
                printf("Matrices do not match!!!\n");
329
                return 1;
330
331
            printf("Matrices match (sequantial-trivial)! - took: %.6f s\n", time);
332
           free(res);
333
       }
334
```

```
336     MPI_Finalize();
337     return 0;
338 }
```

Appendix - Poisson solver

The parallel Poisson solver used the following code:

```
1 /*
   * MPI_Poisson.c
  * 2D Poison equation solver (parallel version)
6 #include <stdio.h>
7 #include <stdlib.h>
8 #include <math.h>
9 #include <time.h>
10 #include <mpi.h>
#include <assert.h>
#define DEBUG 0
14
#define max(a,b) ((a)>(b)?a:b)
17
18 // defines for Exercises!
19
20 #define SOR 1
21 #define MONITOR_ERROR 1
22 #define FAST_DO_STEP_LOOP
23 // #define MONITOR_ALLREDUCE 1
24 // #define ALLREDUCE_COUNT 100
#define MONITOR_EXCHANGE_BORDERS
26 #define SKIP_EXCHANGE
28 #define DEFINES_ON (SOR || MONITOR_ERROR || 0)
29 //defines end
30
31 enum
32 {
      X_DIR, Y_DIR
33
34 };
36 // only needed for certain configs!
37 #ifdef SOR
38 double sor_omega = 1.9;
39 #endif
40 #ifdef MONITOR_ERROR
41 double *errors=NULL:
42 #endif
43 #ifdef MONITOR_ALLREDUCE
44 double all_reduce_time = 0;
45 #endif
46 #ifdef MONITOR_EXCHANGE_BORDERS
                                      // Total time spent in exchanges
47 double total_exchange_time = 0.0;
                                      // Total latency
double total_latency = 0.0;
49 double total_data_transferred = 0.0; // Total data transferred
                                      // Number of exchanges
50 int num_exchanges = 0;
51 #endif
52 #ifdef SKIP_EXCHANGE
size_t skip_exchange;
54 #endif
/* global variables */
57 int gridsize[2];
58 double precision_goal;
                            /* precision_goal of solution */
59 int max_iter; /* maximum number of iterations alowed */
int P; //total number of processes
int P_grid[2]; // process grid dimensions
62 MPI_Comm grid_comm; //grid communicator
63 MPI_Status status;
64 double hx, hy;
```

```
66 /* process specific globals*/
67 int proc_rank;
68 double wtime;
69 int proc_coord[2]; // coords of current process in processgrid
70 int proc_top, proc_right, proc_bottom, proc_left; // ranks of neighboring procs
71 // step 7
72 int offset[2] = {0,0};
73 // step 8
74 MPI_Datatype border_type[2];
/* benchmark related variables */
77 clock_t ticks; /* number of systemticks */
78 int timer_on = 0; /* is timer running? */
79
80 /* local grid related variables */
81 double **phi; /* grid */
                     /* TRUE if subgrid element is a source */
82 int **source;
83 int dim[2];
                   /* grid dimensions */
85 void Setup_Grid();
86 double Do_Step(int parity);
87 void Solve();
88 void Write_Grid();
89 void Clean_Up();
90 void Debug(char *mesg, int terminate);
91 void start_timer();
92 void resume_timer();
93 void stop_timer();
94 void print_timer();
95
96 void start_timer()
97 {
       if (!timer_on){
98
99
           MPI_Barrier(grid_comm);
           ticks = clock();
100
           wtime = MPI_Wtime();
101
           timer_on = 1;
102
103
104 }
105
void resume_timer()
107 {
108
       if (!timer_on){
           ticks = clock() - ticks;
109
           wtime = MPI_Wtime() - wtime;
110
           timer_on = 1;
112
113 }
114
void stop_timer()
116 {
       if (timer_on){
117
           ticks = clock() - ticks;
118
           wtime = MPI_Wtime() - wtime;
119
120
           timer_on = 0;
121
122 }
123
124 void print_timer()
125 {
126
       if (timer_on){
           stop_timer();
127
           printf("(%i) Elapsed Wtime %14.6f s (%5.1f%% CPU)\n", proc_rank, wtime, 100.0 * ticks
128
       * (1.0 / CLOCKS_PER_SEC) / wtime);
          resume_timer();
129
       }
130
      else{
131
           printf("(%i) Elapsed Wtime %14.6f s (%5.1f%% CPU)\n", proc_rank, wtime, 100.0 * ticks
132
       * (1.0 / CLOCKS_PER_SEC) / wtime);
133
134 }
135
```

```
void Debug(char *mesg, int terminate)
137 {
        if (DEBUG || terminate){
138
            printf("%s\n", mesg);
139
140
        if (terminate){
141
            exit(1);
142
143
144 }
145
void Setup_Proc_Grid(int argc, char **argv){
        int wrap_around[2];
147
        int reorder;
148
149
        Debug("My_MPI_Init",0);
150
152
        // num of processes
        MPI_Comm_size(MPI_COMM_WORLD, &P);
154
        //calculate the number of processes per column and per row for the grid
156
        if (argc > 2) {
            P_grid[X_DIR] = atoi(argv[1]);
            P_grid[Y_DIR] = atoi(argv[2]);
158
159
            if(P_grid[X_DIR] * P_grid[Y_DIR] != P){
                 Debug("ERROR Proces grid dimensions do not match with P ", 1);
160
161
            #ifdef SOR
            if (argc > 3)
164
                 // get sor from args
                 sor_omega = atof(argv[3]);
166
167
                 printf("Set sor_omega over argv to %1.4f\n", sor_omega);
168
            #endif
169
            #ifdef SKIP_EXCHANGE
            if (argc > 4)
171
172
                 skip_exchange = atoi(argv[4]);
                 printf("Set skip_exchange over argv to %zu\n", skip_exchange);
174
            }
175
176
            else{
                 skip_exchange = 1;
177
                 printf("Set skip_exchange to default value 1\n");
178
179
            #endif
180
        else{
182
            Debug("ERROR Wrong parameter input",1);
183
184
185
        // Create process topology (2D grid)
186
        wrap_around[X_DIR] = 0;
187
        wrap_around[Y_DIR] = 0;
188
        reorder = 1; //reorder process ranks
190
191
        // create grid_comm
        int ret = MPI_Cart_create(MPI_COMM_WORLD, 2, P_grid, wrap_around, reorder, &grid_comm);
        if (ret != MPI SUCCESS){
193
            Debug("ERROR: MPI_Cart_create failed",1);
194
195
        //get new rank and cartesian coords of this proc
196
        MPI_Comm_rank(grid_comm, &proc_rank);
        MPI_Cart_coords(grid_comm, proc_rank, 2, proc_coord);
printf("(%i) (x,y)=(%i,%i)\n", proc_rank, proc_coord[X_DIR], proc_coord[Y_DIR]);
199
200
        //calc neighbours
        // MPI_Cart_shift(grid_comm, Y_DIR, 1, &proc_bottom, &proc_top);
MPI_Cart_shift(grid_comm, Y_DIR, 1, &proc_top, &proc_bottom);
201
202
        MPI_Cart_shift(grid_comm, X_DIR, 1, &proc_left, &proc_right);
203
         printf("(\%i) top \%i, right \%i, bottom \%i, left \%i \n", proc_rank, proc_top, \\
204
        proc_right, proc_bottom, proc_left);
205 }
206
207 void Setup_Grid()
```

```
208 {
209
       int x, y, s;
        double source_x, source_y, source_val;
210
       FILE *f;
211
212
        Debug("Setup_Subgrid", 0);
213
214
215
       if(proc_rank == 0){
216
            f = fopen("input.dat", "r");
            if (f == NULL){
217
                Debug("Error opening input.dat", 1);
218
219
            fscanf(f, "nx: %i\n", &gridsize[X_DIR]);
220
            fscanf(f, "ny: %i\n", &gridsize[Y_DIR]);
221
            fscanf(f, "precision goal: %lf\n", &precision_goal);
fscanf(f, "max iterations: %i\n", &max_iter);
222
223
224
       MPI_Bcast(&gridsize, 2, MPI_INT, 0, grid_comm);
225
       MPI_Bcast(&precision_goal, 1, MPI_DOUBLE, 0, grid_comm);
226
       MPI_Bcast(&max_iter, 1, MPI_INT, 0, grid_comm);
227
       hx = 1 / (double)gridsize[X_DIR];
228
229
       hy = 1 / (double)gridsize[Y_DIR];
230
231
        /* Calculate dimensions of local subgrid */ //! We do that later now!
        // dim[X_DIR] = gridsize[X_DIR] + 2;
232
       // dim[Y_DIR] = gridsize[Y_DIR] + 2;
233
234
        //! Step 7
235
       int upper_offset[2] = {0,0};
236
        // Calculate top left corner cordinates of local grid
237
       offset[X_DIR] = gridsize[X_DIR] * proc_coord[X_DIR] / P_grid[X_DIR];
238
        offset[Y_DIR] = gridsize[Y_DIR] * proc_coord[Y_DIR] / P_grid[Y_DIR];
       upper_offset[X_DIR] = gridsize[X_DIR] * (proc_coord[X_DIR] + 1) / P_grid[X_DIR];
240
       upper_offset[Y_DIR] = gridsize[Y_DIR] * (proc_coord[Y_DIR] + 1) / P_grid[Y_DIR];
241
242
        // dimensions of local grid
243
       dim[X_DIR] = upper_offset[X_DIR] - offset[X_DIR];
244
        dim[Y_DIR] = upper_offset[Y_DIR] - offset[Y_DIR];
        // Add space for rows/columns of neighboring grid
246
247
        dim[X_DIR] += 2;
       dim[Y_DIR] += 2;
248
        //! Step 7 end
249
250
        /* allocate memory */
251
       if ((phi = malloc(dim[X_DIR] * sizeof(*phi))) == NULL){
252
            Debug("Setup_Subgrid : malloc(phi) failed", 1);
253
254
       if ((source = malloc(dim[X_DIR] * sizeof(*source))) == NULL){
255
            Debug("Setup_Subgrid : malloc(source) failed", 1);
256
257
        if ((phi[0] = malloc(dim[Y_DIR] * dim[X_DIR] * sizeof(**phi))) == NULL){
258
            Debug("Setup_Subgrid : malloc(*phi) failed", 1);
259
       }
260
       if ((source[0] = malloc(dim[Y_DIR] * dim[X_DIR] * sizeof(**source))) == NULL){
261
            Debug("Setup_Subgrid : malloc(*source) failed", 1);
262
263
       7
        for (x = 1; x < dim[X_DIR]; x++)</pre>
264
265
            phi[x] = phi[0] + x * dim[Y_DIR];
266
            source[x] = source[0] + x * dim[Y_DIR];
267
268
        /* set all values to '0' */
270
       for (x = 0; x < dim[X_DIR]; x++){</pre>
271
            for (y = 0; y < dim[Y_DIR]; y++)</pre>
272
            {
273
274
                phi[x][y] = 0.0;
275
                source[x][y] = 0;
            }
276
277
        /* put sources in field */
278
279
       do-f
           if (proc_rank==0)
```

```
281
             {
                 s = fscanf(f, "source: %lf %lf %lf n", &source_x, &source_y, &source_val);
282
             MPI_Bcast(&s, 1, MPI_INT, 0, grid_comm);
284
285
             if (s==3){
                 MPI_Bcast(&source_x, 1, MPI_DOUBLE, 0, grid_comm);
286
                 MPI_Bcast(&source_y, 1, MPI_DOUBLE, 0, grid_comm);
287
                 MPI_Bcast(&source_val, 1, MPI_DOUBLE, 0, grid_comm);
288
                 x = source_x * gridsize[X_DIR];
289
                 y = source_y * gridsize[Y_DIR];
290
                 x = x + 1 - offset[X_DIR]; // Step 7 --> local grid transform
291
                 y = y + 1 - offset[Y_DIR]; // Step 7 --> local grid transform
if(x > 0 && x < dim[X_DIR] -1 && y > 0 && y < dim[Y_DIR]-1){ // check if in local</pre>
292
293
        grid
                      phi[x][y] = source_val;
294
295
                      source[x][y] = 1;
                 }
296
             }
297
298
        while (s==3);
299
300
301
        if (proc_rank == 0) {
            fclose(f);
302
303
304 }
305
   void Setup_MPI_Datatypes()
306
307
   {
        Debug("Setup_MPI_Datatypes",0);
308
309
        // vertical data exchange (Y_Dir)
310
         \texttt{MPI\_Type\_vector}(\texttt{dim}\,[\texttt{X\_DIR}] \ - \ 2, \ 1, \ \texttt{dim}\,[\texttt{Y\_DIR}], \ \texttt{MPI\_DOUBLE}, \ \&\texttt{border\_type}\,[\texttt{Y\_DIR}]); 
311
        // horizontal data exchange (X_Dir)
312
        MPI_Type_vector(dim[Y_DIR] - 2, 1, 1, MPI_DOUBLE, &border_type[X_DIR]);
313
314
        MPI_Type_commit(&border_type[Y_DIR]);
315
316
        MPI_Type_commit(&border_type[X_DIR]);
317 }
318
319 int Exchange_Borders()
320
        #ifdef MONITOR_EXCHANGE_BORDERS
321
        double start_time, latency_start, latency;
322
        double data_size_top, data_size_left;
323
        double exchange_time;
324
        // Measure latency with a small dummy message
326
        latency_start = MPI_Wtime();
327
        double dummy;
328
        MPI_Sendrecv(&dummy, 1, MPI_DOUBLE, proc_top, 0, &dummy, 1, MPI_DOUBLE, proc_bottom, 0,
329
        grid_comm, &status)
        latency = MPI_Wtime() - latency_start;
330
        total_latency += latency;
331
        // Calculate data sizes
333
        data_size_top = dim[X_DIR] * sizeof(double); // Top and bottom rows
data_size_left = dim[Y_DIR] * sizeof(double); // Left and right columns
334
335
        double data_transferred = 2 * (data_size_top + data_size_left); // Total data for this
336
        exchange
337
        total_data_transferred += data_transferred;
        #endif
338
339
        Debug("Exchange_Borders",0);
340
        #ifdef MONITOR_EXCHANGE_BORDERS
341
        start_time = MPI_Wtime();
342
        #endif
343
        // top direction
344
        MPI_Sendrecv(&phi[1][1], 1, border_type[Y_DIR], proc_top, 0, &phi[1][dim[Y_DIR] - 1], 1,
345
        border_type[Y_DIR], proc_bottom, 0, grid_comm, &status);
        // bottom direction
        MPI_Sendrecv(&phi[1][dim[Y_DIR] - 2], 1, border_type[Y_DIR], proc_bottom, 0, &phi[1][0],
347
        1, border_type[Y_DIR], proc_top, 0, grid_comm, &status);
        // left direction
```

```
MPI_Sendrecv(&phi[1][1], 1, border_type[X_DIR], proc_left, 0, &phi[dim[X_DIR]-1][1], 1,
349
       border_type[X_DIR], proc_right, 0, grid_comm, &status);
        // right direction
        MPI_Sendrecv(&phi[dim[X_DIR]-2][1], 1, border_type[X_DIR], proc_right, 0, &phi[0][1], 1,
351
        border_type[X_DIR], proc_left, 0, grid_comm, &status);
352
       #ifdef MONITOR_EXCHANGE_BORDERS
353
354
        exchange_time = MPI_Wtime() - start_time;
355
        total_exchange_time += exchange_time;
356
       num_exchanges++;
       #endif
357
       return 1;
358
359 }
360
   double Do_Step(int parity)
361
362
363
        int x, y;
        double old_phi, c_ij;
364
        double max_err = 0.0;
365
366
       #ifdef FAST_DO_STEP_LOOP
367
368
        int start_y;
        for (x = 1; x < dim[X_DIR] - 1; x++){</pre>
369
            start_y = ((1 + x + offset[X_DIR] + offset[Y_DIR]) % 2 == parity) ? 1 : 2;
370
            for (y = start_y; y < dim[Y_DIR] - 1; y += 2){</pre>
371
                if (source[x][y] != 1){
372
                     old_phi = phi[x][y];
373
                     #ifndef SOR
374
                     phi[x][y] = (phi[x + 1][y] + phi[x - 1][y] + phi[x][y + 1] + phi[x][y - 1]) *
375
       0.25;
376
                     #endif
                    #ifdef SOR
377
                     c_{ij} = (phi[x + 1][y] + phi[x - 1][y] + phi[x][y + 1] + phi[x][y - 1] + hx*hy*
378
       source[x][y]) * 0.25 - phi[x][y];
                     phi[x][y] += sor_omega*c_ij;
                     #endif
380
                     if (max_err < fabs(old_phi - phi[x][y])){</pre>
381
                         max_err = fabs(old_phi - phi[x][y]);
382
                    }
383
384
                }
385
            }
       }
386
387
        return max_err;
        #endif
388
389
       #ifndef FAST_DO_STEP_LOOP
390
        /* calculate interior of grid */
391
        for (x = 1; x < dim[X_DIR] - 1; x++){
392
            for (y = 1; y < dim[Y_DIR] - 1; y++){</pre>
393
                if ((x + offset[X_DIR] + y + offset[Y_DIR]) % 2 == parity && source[x][y] != 1){
394
395
                     old_phi = phi[x][y];
                     #ifndef SOR
396
                    phi[x][y] = (phi[x + 1][y] + phi[x - 1][y] + phi[x][y + 1] + phi[x][y - 1]) *
397
       0.25;
                     #endif
398
399
                    #ifdef SOR
                     c_{ij} = (phi[x + 1][y] + phi[x - 1][y] + phi[x][y + 1] + phi[x][y - 1] + hx*hy*
400
       source[x][y]) * 0.25 - phi[x][y];
                     phi[x][y] += sor_omega*c_ij;
401
402
                     if (max_err < fabs(old_phi - phi[x][y])){</pre>
403
                         max_err = fabs(old_phi - phi[x][y]);
405
                }
406
            }
407
408
409
     return max_err;
     #endif
410
411 }
413 void Solve()
414 {
       int count = 0;
```

```
double delta;
416
       double global_delta;
417
        double delta1, delta2;
418
419
       Debug("Solve", 0);
420
421
        /* give global_delta a higher value then precision_goal */
422
423
       global_delta = 2 * precision_goal;
424
425
       while (global_delta > precision_goal && count < max_iter)</pre>
426
            Debug("Do_Step 0", 0);
427
            delta1 = Do_Step(0);
428
            #ifdef SKIP_EXCHANGE
429
            if (count % skip_exchange == 0 && Exchange_Borders()) // use short circuit evaluation
430
431
            #endif
432
            #ifndef SKIP_EXCHANGE
            Exchange_Borders();
433
            #endif
434
            Debug("Do_Step 1", 0);
435
            delta2 = Do_Step(1);
436
437
            #ifdef SKIP_EXCHANGE
            if (count % skip_exchange == 0 && Exchange_Borders())
438
439
            #endif
            #ifndef SKIP_EXCHANGE
440
441
            Exchange_Borders();
            #endif
442
            delta = max(delta1, delta2);
443
            #ifdef MONITOR_ALLREDUCE
444
            double time_ = MPI_Wtime();
            #endif
446
            #ifdef ALLREDUCE_COUNT
447
            if (count % ALLREDUCE_COUNT == 0) {
448
                MPI_Allreduce(&delta, &global_delta, 1, MPI_DOUBLE, MPI_MAX, grid_comm);
449
450
            #endif
451
            #ifndef ALLREDUCE_COUNT
452
            MPI_Allreduce(&delta, &global_delta, 1, MPI_DOUBLE, MPI_MAX, grid_comm);
            #endif
454
            #ifdef MONITOR_ALLREDUCE
455
456
            all_reduce_time += MPI_Wtime() - time_;
            #endif
457
            #ifdef MONITOR_ERROR
458
            if (proc_rank == 0)
459
            {
460
                errors[count] = global_delta;
461
462
            #endif
463
464
            count++;
465
466
       printf("(%i) Number of iterations : %i\n", proc_rank, count);
467
       #ifdef MONITOR_ALLREDUCE
468
       printf("(%i) Allreduce time: %14.6f\n", proc_rank, all_reduce_time);
       #endif
470
471
       #ifdef MONITOR_EXCHANGE_BORDERS
       printf("(%i) Exchange time: %14.6f\n", proc_rank, total_exchange_time);
472
       #endif
473
474 }
475
476 double* get_Global_Grid()
477 {
       Debug("get_Global_Grid", 0);
478
        //!! DEBUG only
479
        for (size_t i = 0; i < dim[X_DIR]; i++)</pre>
480
481
            for (size_t j = 0; j < dim[Y_DIR]; j++)</pre>
482
            {
483
                phi[i][j] = proc_rank;
484
            }
485
486
       }
487
```

```
// only process 0 needs to store all data!
489
490
       double* global_phi = NULL;
        if (proc_rank == 0) {
491
            global_phi = malloc(gridsize[X_DIR] * gridsize[Y_DIR] * sizeof(double));
492
            if (global_phi == NULL) {
493
                Debug("get_Global_Grid : malloc(global_phi) failed", 1);
494
495
496
497
498
       // copy own part into buffer - flatten!
       size_t buf_size = (dim[X_DIR] - 2) * (dim[Y_DIR] - 2) * sizeof(double);
499
       double* local_phi = malloc(buf_size);
500
       int idx = 0:
501
       for (int x = 1; x < dim[X_DIR] - 1; x++) {</pre>
502
            for (int y = 1; y < dim[Y_DIR] - 1; y++) {</pre>
503
504
                local_phi[idx++] = phi[x][y];
505
506
       printf("I'm proc %d and i have a buffer of size %zu\n", proc_rank, buf_size);
507
509
510
       // only proc 0 needs sendcounts and displacements for the gatherv operation
       int* sendcounts = NULL;
511
512
       int* displs = NULL;
513
       if (proc_rank == 0)
            sendcounts = malloc(P * sizeof(int));
514
            displs = malloc(P * sizeof(int));
515
516
            // size and offset of different subgrids
517
            //! Note that this only works if every process has the same subgrid
518
            if (gridsize[X_DIR] % P_grid[X_DIR] != 0 || gridsize[Y_DIR] % P_grid[Y_DIR] != 0)
519
520
521
                Debug("!!!A grid dimension is not a multiple of the P_grid in this direction!", 1)
           }
            int subgrid_width = gridsize[X_DIR] / P_grid[X_DIR];
524
            int subgrid_height = gridsize[Y_DIR] / P_grid[Y_DIR];
            for (int px = 0; px < P_grid[X_DIR]; px++) {</pre>
526
                for (int py = 0; py < P_grid[Y_DIR]; py++) {</pre>
527
                    int rank = px * P_grid[Y_DIR] + py;
528
                    sendcounts[rank] = subgrid_width * subgrid_height;
529
                    displs[rank] = (px * subgrid_width * gridsize[Y_DIR]) + (py * subgrid_height);
530
                }
531
            }
532
533
       Debug("get_Global_Grid : MPI_Gatherv", 0);
       //! TODO this Gatherv does something wrong - all local grids are alright!!!
535
       MPI_Gatherv(local_phi, (dim[X_DIR] - 2) * (dim[Y_DIR] - 2), MPI_DOUBLE, global_phi,
536
       sendcounts, displs, MPI_DOUBLE, 0, MPI_COMM_WORLD);
       free(local_phi);
538
       if (proc_rank == 0) {
539
            free(sendcounts);
            free(displs);
541
542
544
       return global_phi;
545 }
546
   void Write_Grid_global(){
547
       int x, y;
       FILE *f;
549
       char filename[40]; //seems danagerous to use a static buffer but let's go with the steps
550
       sprintf(filename, "output_MPI_global_%i.dat", proc_rank);
551
       if ((f = fopen(filename, "w")) == NULL){
552
            Debug("Write_Grid : fopen failed", 1);
554
555
       Debug("Write_Grid", 0);
556
557
       for (x = 1; x < dim[X_DIR]-1; x++){</pre>
558
            for (y = 1; y < dim[Y_DIR]-1; y++){</pre>
```

```
int x_glob = x + offset[X_DIR];
560
                 int y_glob = y + offset[Y_DIR];
fprintf(f, "%i %i %f\n", x_glob, y_glob, phi[x][y]);
561
562
563
564
        fclose(f);
565
566 }
567
568 void Write_Grid()
569 €
570
        double* global_phi = get_Global_Grid();
        if (proc_rank != 0) {
571
            assert (global_phi == NULL);
572
573
574
575
        int x, y;
576
        FILE *f;
        char filename[40]; //seems danagerous to use a static buffer but let's go with the steps
sprintf(filename, "output_MPI%i.dat", proc_rank);
577
578
        if ((f = fopen(filename, "w")) == NULL){
579
            Debug("Write_Grid : fopen failed", 1);
580
581
582
583
        Debug("Write_Grid", 0);
584
        for (x = 0; x < gridsize[X_DIR]; x++){</pre>
585
             for (y = 0; y < gridsize[Y_DIR]; y++){</pre>
586
                 fprintf(f, "\%i \%i \%f\n", x+1, y+1, global_phi[x*gridsize[Y_DIR] + y]);
587
588
589
        fclose(f);
590
591
        free(global_phi);
592 }
593
594
   void Clean_Up()
   }
595
        Debug("Clean_Up", 0);
596
597
        free(phi[0]);
598
599
        free(phi);
600
        free(source[0]);
        free (source);
601
602
        #ifdef MONITOR_ERROR
        free(errors);
603
        #endif
604
605 }
void setup_error_monitor(){
607
       if (proc_rank != 0)
608
609
             return:
610
611
        errors = malloc(sizeof(double)*max_iter);
612
613 }
614 void write errors(){
615
        if(proc_rank != 0){
616
617
        FILE *f;
618
        char filename[40]; //seems danagerous to use a static buffer but let's go with the steps
619
        sprintf(filename, "errors_MPI.dat");
620
621
        if ((f = fopen(filename, "w")) == NULL){
            Debug("Write_Errors : fopen failed", 1);
622
623
624
        Debug("Write_Errors", 0);
625
626
627
        for (size_t i = 0; i < max_iter; ++i)</pre>
628
             fprintf(f, "%f\n", errors[i]);
630
        fclose(f):
631
632 }
```

```
633
634 void Print_Aggregated_Metrics()
635 {
        #ifdef MONITOR_EXCHANGE_BORDERS
636
        if (num_exchanges > 0) {
637
            double avg_exchange_time = total_exchange_time / num_exchanges;
638
            double avg_latency = total_latency / num_exchanges;
639
            double avg_bandwidth = total_data_transferred / total_exchange_time;
640
641
            printf("
--- Aggregated Metrics ---
);
642
            printf("Total Exchanges: %d\n", num_exchanges);
printf("Total Data Transferred: %.2f bytes\n", total_data_transferred);
644
            printf("Total Exchange Time: %.9f s\n", total_exchange_time);
645
            printf("Average Exchange Time per Call: %.9f s\n", avg_exchange_time);
            printf("Average Latency per Call: %.9f s\n", avg_latency);
printf("Average Bandwidth: %.2f bytes/s\n", avg_bandwidth);
647
648
        } else {
649
            printf("No exchanges recorded.\n");
650
651
652
        #endif
653 }
654
int main(int argc, char **argv)
656 {
657
        MPI_Init(&argc, &argv);
        Setup_Proc_Grid(argc,argv); // was earlier MPI_Comm_rank(MPI_COMM_WORLD, &proc_rank);
658
659
        start_timer();
660
        Setup_Grid();
661
        Setup_MPI_Datatypes();
662
663
        #ifdef SOR
664
        if (proc_rank == 0)
665
666
667
             printf("SOR using omega: %.5f\n", sor_omega);
668
669
        #endif
        #ifdef MONITOR_ERROR
        setup_error_monitor();
671
        #endif
672
673
        Solve():
674
        #ifdef MONITOR_ERROR
675
        write_errors();
676
        #endif
677
        // Write_Grid();
678
        Write_Grid_global();
679
        Print_Aggregated_Metrics();
680
        print_timer();
681
682
683
        Clean_Up();
        MPI_Finalize();
684
        return 0;
685
686 }
```