Parallel and Distributed Programming – Assignment 1 Group 15

Claude Carlsson Edvin Germer Ture Hassler

April 5, 2023



1 Introduction

For this assignment, the goal is to parallelize a provided c program that does a stencil computation in serial. In this context, a stencil is a pattern that performs operations on an array of data, and it is composed of two components: the shape of the stencil and weights used in the computations.

In the performance and results section of the report, we are going to present two different scaling approaches: strong scaling and weak scaling. The difference between these two scaling concepts is that for strong scaling, we hold the problem size constant while increasing the number of processes. We measure the stencil time, and the goal for strong scaling is to achieve a speedup. For weak scaling, the problem size is increased proportionally with the number of processes. The stencil time is measured, and the goal for weak scaling is to maintain a constant time.

1.1 Communication type

The communication type used in this assignment is going to be the message passing interface, or MPI, for short. MPI is a library that provides functions that enable communication between processes.

For this assignment, the reasons for using MPI are twofold. First is to learn how to convert a serial implementation of a program to parallel using MPI. Second, is that MPI is designed to be used in a distributed environment, which enables scalability across a cluster of machines. Furthermore, another parallelization technique such as using OpenMP might offer an easier implementation, however, it is limited to hardware that has shared memory (that is, a single machine), and not distributed memory.

2 Parallelization

We used MPI for our parallelization. We read the data with process 0 and then evenly distributed chunks of the array to each thread using MPI_Scatter. We also sent the total number of elements to each process using MPI_Bcast.

For each iteration the stencil was applied, each process needs to send the four outmost elements (two on each side) to their respective neighbours. We did this with non blocking point to point communication, using MPI_Isend and MPI_Irecv. This allows each process to send their message and then continue applying the stencil on the middle elements that do not depend on the other processes, while waiting to receive the values from the other processes.

After each process has performed all computations that are independent of their neighbours, we then use MPI_Waitall to synchronize all processes, making sure that each process has received the neighbouring values so they can apply the stencil on the edge elements.

After all processes has finished their computations we then use **MPI_Gather** to send back the data to process 0, and then process 0 continues to write the output to a file.

3 Performance experiments and results

We performed several performance measurements, both for weak and strong scaling. We also included the system specifications for reproducibility of the results.

3.1 System specifications

System: UU Linux Host "Tussilago"

Processor: AMD Opteron(tm) Processor 6282 SE

Operating System: Ubuntu 22.04 x86 64

Compiler version: gcc (Ubuntu 11.3.0-1ubuntu1 22.04) 11.3.0

MPI version: mpirun (Open MPI) 4.1.2

3.2 Performance experiments

To get an accurate measurement of the parallel performance we measured the time after reading and distributing the data among all processes, but before writing the output to a file. We then recorded the time as the longest time out of all the processes individual time measurements.

We performed the strong scalability study by measuring the time for varying number of threads, while keeping the size of the array fixed. From this we then calculated the speedup. This was done for three different problem sizes using 1, 4 and 8 million elements in the arrays.

We performed a weak scalability study by keeping the work per process constant at 1 million elements, and increasing both the number of processes and the number of elements in the array, from 1 to 8 processes and from 1 to 8 million elements.

3.3 Results

The results of the time measurements of the scalability study can be seen here, including both plots and tables.

3.3.1 Strong scaling

Table 1: The experiment values from strong scaling, 1M corresponds to one million data points, and the results is in milliseconds

Input size Processes	1M	4M	8M	100M
1	7.7	30.7	61.1	757.9
2	4.1	16.2	32.3	400.5
4	4.0	19.3	33.5	917.9
8	1.1	33.0	96.0	1280.8

Speed up for strong scaling with 1 000 000 data points

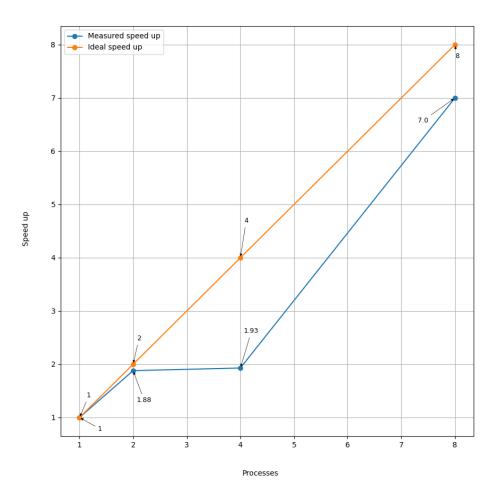


Figure 1: Speed up for strong scaling with one hundred million data points

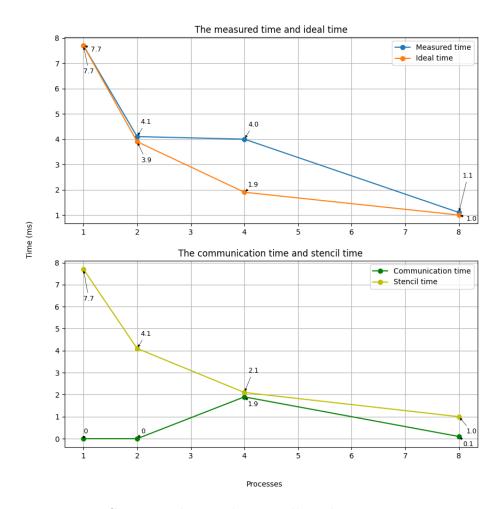


Figure 2: Strong scaling with one million data points as input

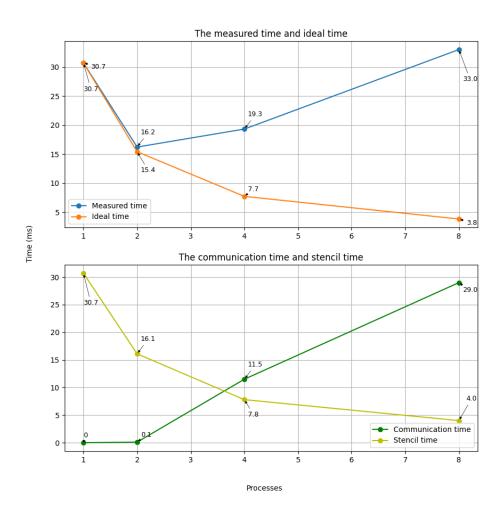


Figure 3: Strong scaling with four million data points as input

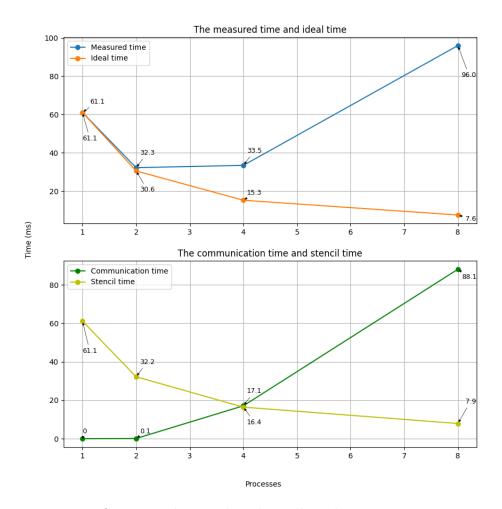


Figure 4: Strong scaling with eight million data points as input

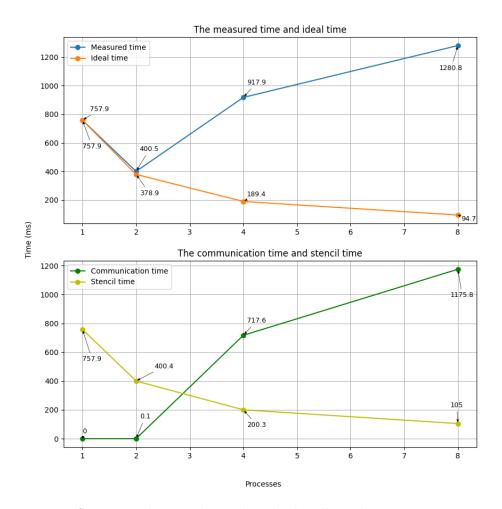


Figure 5: Strong scaling with one hundred million data points as input

3.3.2 Weak scaling

Table 2: The experiment values from strong weak scaling, 1M corresponds to one million data points, and, the result is in milliseconds

Processes & Input size	Time [ms]
1 & 1M	7.7
2 & 2M	8.1
4 & 4M	18.1
8 & 8M	93.4

Speed up for weak scaling

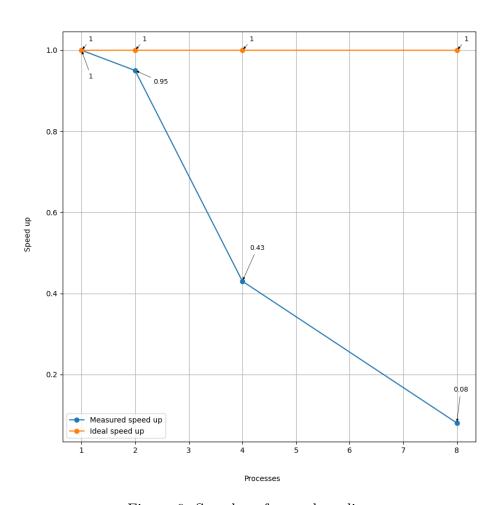


Figure 6: Speed up for weak scaling

Weak scaling

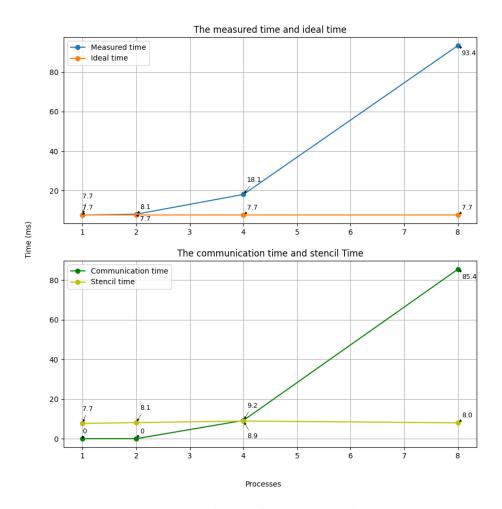


Figure 7: The weak scaling results

3.4 Verification

There are a few files associated with this project that can be used to verify our solution. For the input file "input96.txt" there are two output files "output96_1_ref.txt" and "output96_4_ref.txt" the can be compared to the output of our program when the stencil is applied once, or four times. Below, in figures 8 and 9 the program output is plotted together with the input data. The resulting plots were identical for one or many MPI processes.

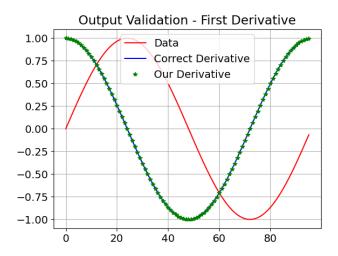


Figure 8: The first derivative of input96.txt

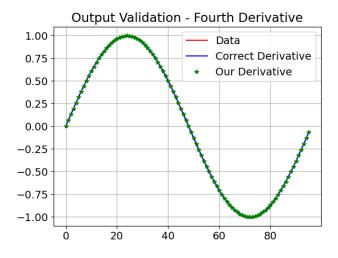


Figure 9: The fourth derivative of input96.txt

4 Discussion and analysis

In our experiments, we have near ideal speed up when going form one to two processes, regardless of file size. But when going beyond two processes, for large inputs, our solution is inefficient. The reason for this is the huge increase in communication time that arise when going beyond two parallel processes.

4.1 Communication time

When performing the stencil application with multiple MPI processes each process send four elements in total to the neighbouring processes. The first two elements are sent to one neighbour and the last two elements are sent to the opposite neighbour. So, with each process there are two send calls, and two receive calls and one MPI_Waitall call, which makes five MPI calls in total, for each stencil application. These five calls are performed in parallel with non-blocking communication so in the ideal case the total time should not increase with an increased number of processes. Even in the worst ideal case, where all communications are made serially, we would not expect the communication time to be worse than linearly increasing as T(t) = nt + C, where T is the total communication time, t is the communication time for one process, t is the amount of processes and t is some constant. However, this is not what we observe in our experiments, accurately described in the bottom graphs in figures 2, 3, 4 and 5.

In our experiments we measure close to zero communication time with two processes but when going beyond that, the communication time increases rapidly. The slopes between two-four and four-eight processes are different from each other but the different is not as dramatic as both of them compare to the slope of going form one to two process where we see a huge increase in time.

For smaller input arrays, such as 1.000.000 datapoints as we seen in figure 2, we observe very strange behavior of the communication time between the processes. We find close to zero communication time for one, two and eight processes but a huge increase (roughly 19 times) for four processes. This result was consistent over many runs and deviates greatly from the theory. We could also see that some processes completed significantly faster, and therefore a significant portion of their time waiting for the other threads to catch up. This was also unexpected since we (aside from process 0 when distributing the data) have a completely even load balance between the different processes.

We have thus determined the communication time to be non-linear, far from the ideal case and dependent on the size of the input array.

4.2 Stencil time

The stencil application section of the program is a perfectly parallelizable task (after receiving the neighbouring elements) and our experiments prove that. In the bottom graphs in figures 2, 3, 4 and 5 we can see that for strong scaling, the total stencil application time decreases close to $\frac{1}{n}$ where n is the number of parallel processes.

4.3 Other solutions

We have experimented with both blocking and non-blocking communication but our results where consistent with both of these methods and we saw no performance increase.

We have also experimented with first initiating the communication then starting with the middle elements in the array, that require no communication between other processes, and then to wait for any of the neighbouring elements to arrive and apply the stencil with those elements. This approach held the most promise of success but when performing the experiments we found no significant increase in the performance and the code became very unreadable.

We settled on an intermediate solution where we start by initiating communication, then applying the stencil on the middle elements and then waiting for ALL neighbouring elements to arrive before applying the stencil on the first and last elements. With this solution the code remains readable, and we find close to equivalent performance as the approach described in the paragraph above.