Assignment 3: Task Farming

Practical information

Deadline: Sunday 27/2 20.00

Resources:

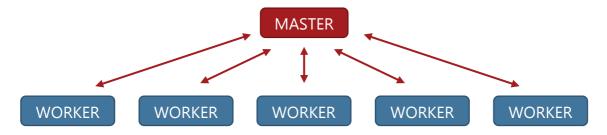
- ERDA for file storage
- Jupyter for the Terminal to access MODI
- Benchmarks through SLURM on MODI

Handin:

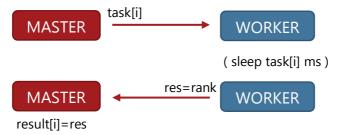
- Total assignment: a report of up to 3 pages in length (excluding the code)
- Part I: Your task farm code
- Use the template on Absalon to include your code in the report

Introduction

Task farming is used in many branches of science and computing to process large amounts of individual tasks in an effective and rapid manner. The idea is to have one controlling process (the "master", "conductor", "farmer", "controller", "provider") and many executing processes ("workers", "agents", "consumer"). In older literature you will also see the terminology master-slaves.



In this exercise you will implement a task farming algorithm in C++ using the message passing interface (MPI) to process data packages from high energy physics. The first step is to make a simplified version. In this version the master creates an array task[NTASKS] with random numbers. The master will distribute the tasks to the workers. Task no I is just an integer task[i]. The "work" that has to be performed is that the worker sleeps (e.g. do nothing) for task[i] milliseconds. The "result" of carrying out the task is that the worker returns an integer which contains the rank of the worker. The life cycle of carrying out a single task is sketched below

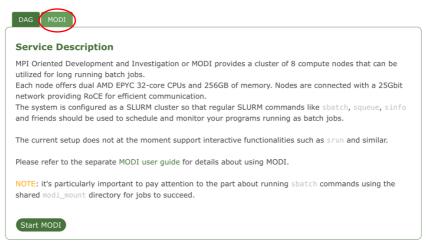


In the end the master loops over the result array and counts how many tasks and workunits (sum of tasks) were carried out by each worker.

MODI

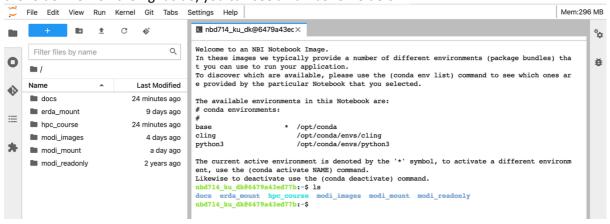
For this assignment we need g++ with the MPI library. This can be obtained by using the so called wrapper command mpic++ It is provided on ERDA using MODI (not DAG!).





You can read more about MODI in the user guide: https://erda.dk/public/MODI-user-guide.pdf

Spin up a Jupyter session on MODI selecting the "HPC notebook" notebook image. In the terminal (or the folder view on the right side) you can see a number of folders.



The different folders contain:

erda mount: your own files.

hpc course: course files.

modi images: images of virtual machines that can be used when submitting jobs.

modi_mount: this folder is the only one that can be seen from the cluster nodes. You need to copy any executables you use (this is called *staging*) to this filesystem before submitting a job.

modi_readonly: this folder can be ignored.

PREPARATIONS

Start by copying the exercise to your storage area and enter in to the folder. You can write 'ls' to get a file listing of the folder.

```
cd erda_mount/HPPC
cp -a ~/hpc_course/module3 .
cd module3
ls
```

To be able to edit the files for the exercise navigate to the same folder in the file view. Here you can open six files:

- Makefile
- task_farm_skeleton.cpp
- task_farm.cpp
- task_farm_HEP.cpp
- task_farm_HEP_seq.cpp
- job.sh

Before you can run the code, you need to compile it. This can be done with make. You should see something like this in the terminal:

```
$ make
mpic++ task_farm.cpp -03 -Wall -Wno-unused-const-variable -std=c++14 -march=native -o task_farm
mpic++ task_farm_HEP.cpp -03 -Wall -Wno-unused-const-variable -std=c++14-march=native -o task_farm_HEP
mpic++ task_farm_HEP_seq.cpp -03 -Wall -Wno-unused-const-variable -std=c++14 -march=native -o task_farm_HEP_seq
mpic++ task_farm.cpp -03 -Wall -march=native -g -std=c++14 -o task_farm
```

For the first task, one binary is produced: task_farm. The task_farm_skeleton.cpp code is there to give you a backup. To run the code on the login machine of the MODI cluster you need to use the mpiexec command. F.x. to run it with 8 processes you do:

```
$ mpiexec -np 8 ./task_farm
```

Code structure:

The code is relatively simple. It contains a main program with the commands to start up MPI, get the rank of the process and the total number of processes. It then calls either master or worker, depending on the rank number.

The first task of the assignment is to use MPI message passing commands to implement communication between master and workers. You can use MPI_Send and MPI_Recv. For non-blocking you can use MPI_Isend and MPI_Irecv. You can read more about how they work in chapter 8.2.

Task 1: Write a functioning master-worker program (points 4)

Use the template and our discussion Monday (see updated slides that includes the solution and the excellent sketches on slack) as a starting point for implementing a working master-worker program. In this version the master creates an array task[NTASKS] with random numbers. The master will distribute the tasks to the workers. It is just an integer task[i]. The "work" that has to be performed is that the worker sleeps (e.g. do nothing) for task[i] milliseconds. The "result" of carrying out the task is that the worker returns an integer which contains the rank of the worker.

Besides your implementation submitted as the code and in pdf, you also submit a report through Absalon. In the report, you should explain how you have parallelised the program. Remember to discuss when and what data you exchange between the MPI-processes and how that impacts the performance. It is up to you if you use blocking or non-blocking messages, or maybe both versions.

Task 2: Master-worker program for HEP data processing (points 2)

Use the template program $task_farm_HEP.cpp$ and the results of task 1 to implement a master—worker program for analysing high energy physics events data. In this version all ranks (both master and worker processes) reads in the data set. The master then creates a large set of possible cuts, that can be used to determine if an event was a background event or a real signal. For each set of cut an accuracy has to be computed. The master has to distribute settings for the cuts to the workers. Each setting contains 8 double precision variables. The settings are stored in a std::array<double, 8> variable, but MPI functions need a pointer to the data. Therefore settings[k].data() (or equivalently &settings[k][0]), which is a pointer to the underlying data array, has to be passed to the MPI function that sends the data from Master. Also be aware that you are now sending double precision variables (of MPI datatype MPI_DOUBLE) instead of the integers (MPI_INT) in task 1.

The "work" that has to be performed is that the worker computes the accuracy, and that is returned as a result to the master. The results can be compared and validated to the output from the reference code produced by compiling and running task farm HEP seq.cpp.

Task 3: Strong scaling of HEP processing using SLURM (points 4)

When benchmarking the performance of your program, use the MODI servers also through Jupyter. However, in order to get exclusive access to a machine you need to submit your run through SLURM. An example of a SLURM script, job.sh, that runs the parallel program on 8 cores on two nodes (2x4 cores) is:

The "exclusive" flag means that there will only be one user on the nodes. "modi_short" indicates the queue. Each node has 32 cores, so this is highly wasteful way to run, but it is a good way to get reliable benchmarks. The singularity image is needed because each notebook image has a different set of software installed.

You should provide experimental results of running using 1, 2, 4, ..., 64 MPI-processes. You are also welcome to go higher. The maximu is 512. The number of settings that are explored depend on n_{cut}^{8} . At a high number of ranks you may need to increase n_{cut} from the default 3 to 4 or even 5 in order to obtain reasonable run time and good performance. How you choose to distribute the MPI-processes between the MODI compute nodes and their CPU-cores is up to you – remember to discuss briefly your decision in the report.

The results should be presented as an easy to read graph, which includes the absolute and relative performance of the parallel code measured as the wall clock time it takes to update a single task. Given that the workload is fixed, this is what we call strong scaling. If you multiply the number by the number of (worker) CPU cores used, you obtain the CPU time spend per task. In the ideal case, this number should be fixed (e.g. a flat curve).