ME 759

High Performance Computing for Engineering Applications Assignment 7 Due Thursday 10/27/2022 at 0:00 PM

Due Thursday 10/27/2022 at 9:00 PM

Submit responses to all tasks which don't specify a file name to Canvas in a file called assignment 7. {txt, docx, pdf, rtf, odt} (choose one of the formats). Submit all plots (if any) on Canvas. Do not zip your Canvas submission.

All source files should be submitted in the HW07 subdirectory on the main branch of your homework git repo with no subdirectories. For this assignment, your HW07 folder should contain task1_cub.cu, task1_thrust.cu, task2.cu, count.cu and task3.cpp.

Important note: All commands or code must work on *Euler* with the nvidia/cuda/11.6.0.lua module and gcc/9.4.0 module loaded. Loading the modules is done via \$ module load nvidia/cuda gcc/9.4.0

This is because *Euler* may be currently experiencing an environment bug that requires you to use gcc/9.4.0 for compiling Thrust- or CUB-related code. gcc/9.4.0 may not be a requirement in other working environments.

We encourage you to test on *Euler* before you submit your homework.

Please submit clean code. Consider using a formatter like clang-format.

IMPORTANT: Before you begin, copy any provided files from Assignments/HW07 directory of the ME759 Resource Repo. Do not change any of the provided files since these files will be overwritten with clean, reference copies when grading.

- 1. (40 pts) In HW05, you have implemented a reduction using the first add during load approach. In this task, you will compare the performance of Thrust and CUB with the previous GPU implementation by performing a scaling analysis for the reduction problem.
 - a) Implement in a file called task1_thrust.cu the Thrust version of reduction. It's expected to do the following (some details about copying between host and device are not included here but should be implemented in your code when necessary):
 - Create and fill with random float numbers in the range [-1.0, 1.0] a thrust::host_vector of length n, where n is the first command line argument as below.
 - Use the built-in function in Thrust to copy the thrust::host_vector into a thrust::device_vector.
 - Call the thrust::reduce function to perform a reduction on the previously generated thrust::device_vector.
 - Print the result of reduction.
 - Print the time taken to run the thrust::reduce function in *milliseconds* using CUDA events.
 - Compile: nvcc task1_thrust.cu -Xcompiler -O3 -Xcompiler -Wall -Xptxas -O3 -std c++17 -o task1_thrust
 - Run by submitting a sbatch script (where n is a positive integer): ./task1_thrust n
 - Example expected output: 3141 0.012
 - b) Implement in a file called task1_cub. cu the CUB version of the reduction based on the code example from this link. Specifically, you should do the following.
 - Stick with the same device memory allocation pattern as the code example (DeviceAllocate() and cudaMemcpy()). Do not use unified memory.
 - Modify the example program so that the host array h_in has length n where n is the first command line argument as below, then fill in h_in with random float numbers in the range [-1.0, 1.0].

- Call the DeviceReduce::Sum function that outputs the reduction result to the output array.
- Print the reduction result
- Print the time taken to run the <code>DeviceReduce::Sum</code> function (the actual one, not the one that's used to find the size of temporary storage needed) in <code>milliseconds</code> using CUDA events. It's recommended to remove the debug function wrapper when gauging the performance of the reduction.
- Compile: nvcc task1_cub.cu -Xcompiler -O3 -Xcompiler -Wall -Xptxas -O3 -std c++17 -o task1_cub
- Run by submitting a sbatch script (where n is a positive integer): ./task1_cub n
- Example expected output: 3141 0.012
- c) On an $Euler\ compute\ node$:
 - Run task1_thrust for value $n=2^{10},2^{11},\cdots,2^{30}$ and generate a pattern of time vs. n in $\log-\log$ scale.
 - Run task1_cub for value $n=2^{10},2^{11},\cdots,2^{30}$ and generate a pattern of time vs. n in $\log-\log$ scale.
 - Overlay the above two patterns on top of the plot you generated for HW05 task1 in a file called task1.pdf. If you dropped HW05, then using information derived from this post or other Piazza posts is fine.
- d) Comment on the performance of the three implementations you came up with above.

2. (35 pts)

- (a) Implement in a file called count.cu the function count as declared and described in count.cuh. Your count function should be able to take a thrust::device_vector, for instance, named d_in (filled by integers), and fill the output values array with the unique integers that appear in d_in in ascending order, as well as the output counts array with the corresponding occurrences of these integers. A brief example is shown below:
 - Example input: d_in = [3, 5, 1, 2, 3, 1]
 Expected output: values = [1, 2, 3, 5]
 Expected output: counts = [2, 1, 2, 1]

Hints:

- Since the length of values and counts may not be equal to the length of d_in, you may want to use thrust::inner_product to find the number of "jumps" (when a[i-1] != a[i]) as you step through the sorted array (the input array is not sorted, so you would have to do a sort using Thrust built-in function). You can refer to Lecture 18 and 19 for thrust::sort examples. There are other valid options as well, for instance, thrust::unique.
- thrust::reduce_by_key could be helpful.
- (b) Write a test program task2.cu which does the following:
 - Create and fill with random int numbers in the range [0, 500] a thrust::host_vector of length n where n is the first command line argument as below.
 - Use the built-in function in Thrust to copy the thrust::host_vector into a thrust::device_vector as the input of your count function.
 - Allocate two other thrust::device_vectors, values and counts, then call your count function to fill these two arrays with the results of this counting operation.
 - Print the last element of values array.
 - Print the last element of counts array.
 - Print the time taken to run the count function in milliseconds using CUDA events.
 - Compile: nvcc task2.cu count.cu -Xcompiler -03 -Xcompiler -Wall -Xptxas -03 -std c++17 -o task2
 - Run by submitting a sbatch script (where n is a positive integer): ./task2 n
 - Example expected output: 370
 - 23 0.13
- (c) On an Euler *compute node*, run task2 for value $\mathbf{n} = 2^5, 2^6, \dots, 2^{24}$ and generate a plot of time vs. \mathbf{n} in log log scale in a file called task2.pdf.

- 3. (25 pts) Write a C++ program in a file called task3.cpp which does the following:
 - Launches four OpenMP threads.
 - Prints out the number of threads launched, with the format Number of threads: x (followed by a newline), where x is the total number of threads. This should be printed only once.
 - Lets each thread introduce itself, with the print format I am thread No. i (followed by a newline), where i is the thread number. Each thread should do that only once.
 - Computes and prints out the factorial of integers from 1 to 8, a!=b (followed by a newline), where a is one of the 8 integers, and b is the result of a!. This should be done in parallel with all 4 threads.

How to go about it, and what the expected output looks like:

```
• Compile: g++ task3.cpp -Wall -03 -std=c++17 -o task3 -fopenmp
```

- Run by submitting a sbatch script: ./task3
- Example expected output (as you can see, the order matters not):

```
Number of threads: 4
I am thread No. 0
I am thread No. 1
I am thread No. 1
I am thread No. 2
3!=6
5!=120
4!=24
6!=720
7!=5040
8!=40320
1!=1
2!=2
```

Important note: this problem is meant to get you started with OpenMP, it is not CUDA anymore, so the following changes need to be made to your slurm script:

- #SBATCH --gres=gpu:1 should be removed since GPU is not required in this assignment.
- #SBATCH --nodes=1 --cpus-per-task=4 (or -N 1 -c 4 for short) should be added, which requests one node with 4 cores. In this course, --cpus-per-task should generally be no more than 20.