Parallelism - Lab 1

1.- Node architecture and memory

Number of sockets per node: 1 Number of cores per socket: 6 Number of threads per core: 2

Maximum core frequency: 2395 MHz L1-I cache size (per-core): 32K L1-D cache size (per-core): 32K L2 cache size (per-core): 256K

Last-level cache size (per-socket): 12288K Main memory size (per socket): 12GB Main memory size (per node): 24GB

2.- Strong vs. weak scalability

In strong scalability the number of threads is changed with a fixed problem size. In this case parallelism is used to reduce the execution time of the program. Strong scalability means that increasing the number of threads or processors (for example) doubling the cores should, ideally, reduce by half the time elapsed in its execution.

In weak scalability the problem size is proportional to the number of threads. In this case parallelism is used to increase the problem size for which the program is executed. In other words, weak scalability means that the improvement is not in reducing time, but the ability to process more operations. So, if we double the processing power, the problem size for which the program is executed should be doubled (or close to it) as well.

3.- Analysis of task decompositions for 3DFFT

Version	t1 (ns)	t∞ (ns)	Parallelism
seq	639.780.001	639.707.001	1,00
v1	639.780.001	639.707.001	1,00
v2	639.780.001	361.439.001	1,77
v3	639.780.001	154.939.001	4,13
v4	639.780.001	64.614.001	9,90
v5	639.780.001	39.104.001	16,36

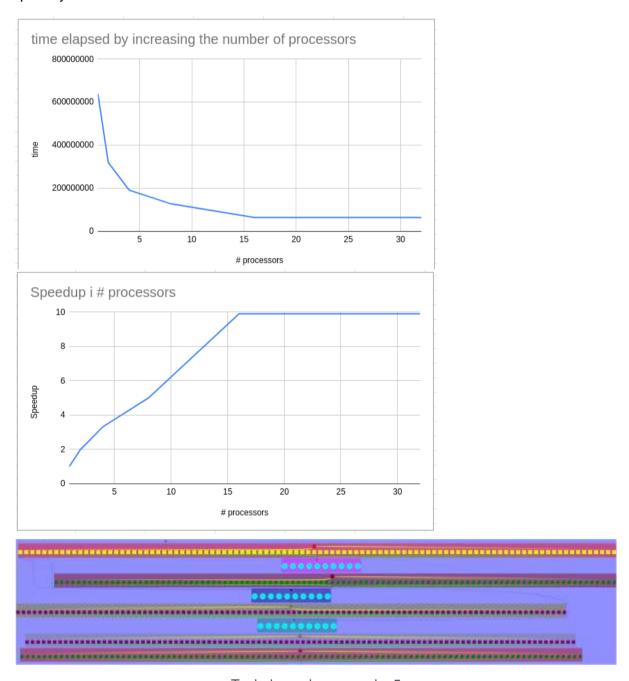
In v4 we can see that it would need 10 processors in order to reach $t\!\!\!/^{\,}_{\,}$

Instead of parallelising loop k, we have parallelising loop j. This way, there are more tasks to be parallelized. As a result, it would need 101 processors in order to reach t^{∞} .

The times obtained by running the simulations on v4 are the following:

1 CPU: 639.780.001 2 CPUs: 320.257.001 4 CPUs: 191.882.001 8 CPUs: 127.992.001 16 CPUs: 64.614.001 32 CPUs: 64.614.001

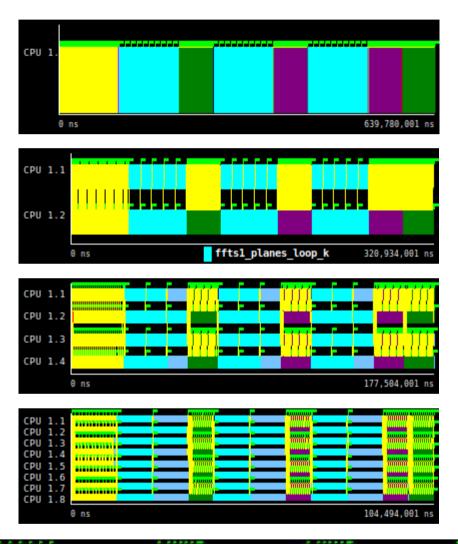
Following are two plots that describe through execution time and speedup how the improvement by increasing the number of processors diminishes as we increase the processor quantity.

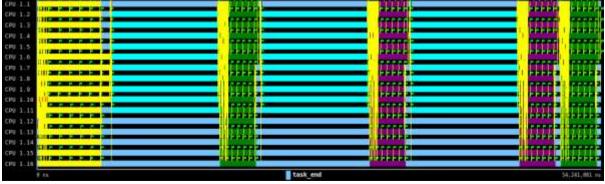


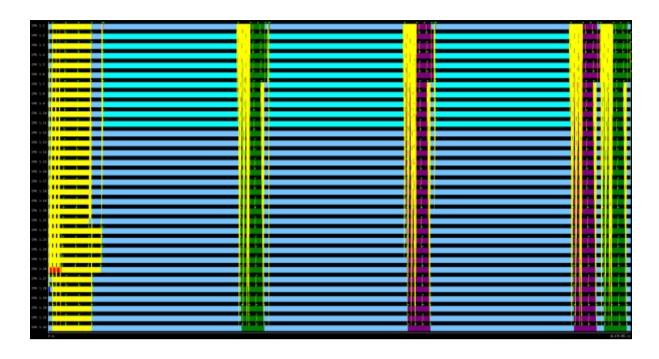
Task dependence graph v5

Following are the simulations given by Tareador regarding v5 and changing the number of processors. In the images of the 16 and 32 processor runs, the time elapsed isn't quite clear, here is every value recorded in nanoseconds:

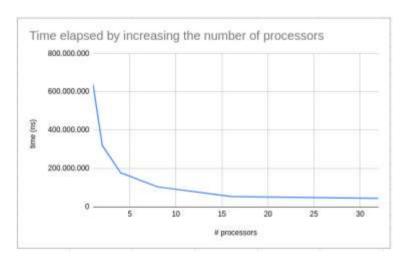
1 CPU: 639.780.001 2 CPUs: 320.934.001 4 CPUs: 177.504.001 8 CPUs: 104.494.001 16 CPUs: 54.241.001 32 CPUs: 44.476.001

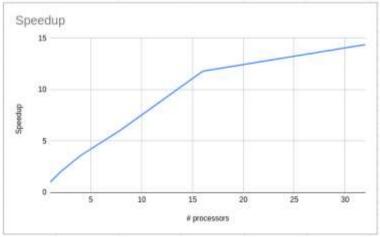






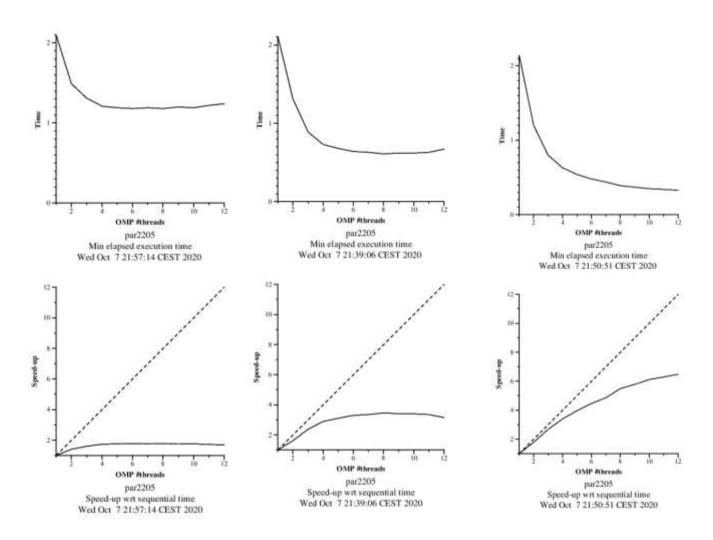
And the plots:





4.- Understanding the parallel execution of 3DFFT

The strong scalability plots that were obtained by executing the submit-strong-omp.sh script with the different modifications to 3dfft_omp (no changes, improved φ 3dfft_omp version and the version that reduces parallelisation overheads) are the following:



As shown in these plots, the code optimization applied resulted in great improvements regarding the scalability of the program. In the first version, parallelisation is barely taken advantage of, showing that, no matter the amount of cores a computer has, these are useless if we don't use them correctly.

When ϕ is improved, and therefore the percentage of parallelizable code in our program increases, a better result is obtained, but the speedup seems to reach its highest value from 4 to 12 threads, not seeing the strong scalability wanted.

In the third execution is where the slope is much better, showing that increasing the number of threads does make a difference and the program is strongly scalable.

Version	φ	S∞	T1 (ns)	T8 (ns)	S8
initial version	0,664	2,97	2.441.509.472	1.442.448.356	1,69
improved φ	0,905	10,57	2.392.321.907	902.009.720	2,65
reduced parallelisation overheads	0,906	10,63	2.293.938.687	597.999.871	3,84