ENCE464 Assignment 2: Computer Architecture

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The result of running the completed program over a range of cube sizes for 300 iterations using 20 threads can be found in Figure 1.

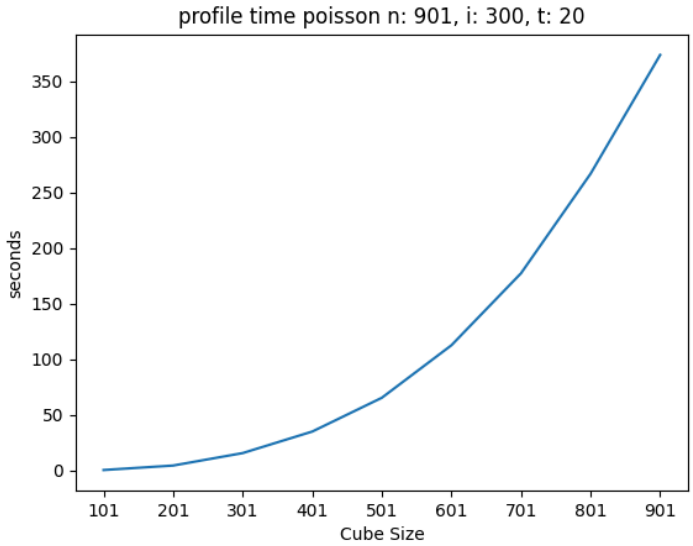

Figure 1: A complete run of the firmware across all cube sizes with 300 iterations and 20 threads.

Processor Architecture

Overview

The Central Processing Unit (CPU) described in this section is the AMD Ryzen 9 6900HX. Released in 2022, this CPU 8 identical cores with 2 threads per core for a total of 16 processing units. The CPU uses the x86-64 instruction set architecture. The CPU structure is shown in Figure 2.

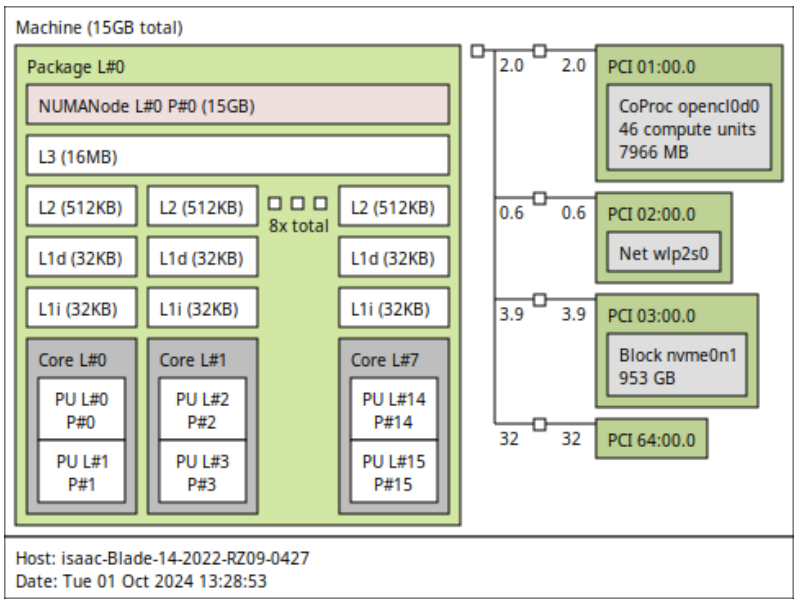

Figure 2: Central Processing Unit (CPU) architecture for the x86-64 AMD Ryzen 9 6900HX.

Cores

Memory

Instruction Set

Multithreading - easy - Daniel

row selection memcopy barrier

Cache - hard

Profiling - easy - Jack

Profiling was used through-out all stages of this projects development. This was done to identify which areas of the program where the slowest and how often these slow areas where called. From these results optimisations where made to the code to reduce execution time. When selecting areas of the code to optimise the sections called most often were prioritised as these give a larger performance benefit then optimising slower less frequent functions. To make profiling easier the various components of the code where compartmentalised into functions, while this does add some execution time (due to stack overheads) it allows the profiling tool gprof to provide more granular results.

Profiling was conducted on both optimised and non-optimised code to gain a wholistic understanding of the programs execution. A breakdown of the execution times and call counts for a non-optimised run of the program with a 201 node cube over 300 iterations using 20 threads can be seen in Table 1. The result of profiling using final optimisation parameters (-03) on the same cube size as before can be found in Table 2.

Table 1: GProf results for a non-optimised run of the program with 201 nodes 300 iterations and 30 threads.

Function	Call Count	Time per call (ms)
`poisson_iteration_inner_slice`	5965	1.25
`memcopy_3D`	5977	0.61
`apply_von_neuman_boundary_slice`	5956	0.05
Barrier waits cumulative	11945	0
Setup	0	0

Table 2: GProf results for a O3 optimised run of the program with 201 nodes 300 iterations and 30 threads.

Function	Call Count	Time per call (us)
`poisson_iteration_inner_slice`	5958	676.40
`memcopy_3D`	5971	410.32
`apply_von_neuman_boundary_slice`	5926	37.12
Barrier waits cumulative	N/A	N/A
Setup	0	0

The results found in Table 1 and 2 show that in both runs the largest time cost is the iteration over the inner slice of the cube. This is expected as it performs the majority of the floating point operations in the software. The next highest execution time is the application of the Von Neumann boundary.

In earlier iterations of the program the Von Neumann boundary was called at every inner loop of the main poisson iteration. Based on profiling the team was able to identify this as a bottle neck as it is unnecessary to call this for all if the inner nodes and move the updates to its own self contained iteration that only iterates over the outside nodes. Another example of profiling helping in optimisation of code is with the barrier waits that are used to synchronise the threads. Originally the team hypothesised that these waits would greatly increase the execution time as threads take different amounts of time to complete due to cpu allocation and the way the cube nodes are divided amongst them. By profiling the code with these barriers implemented it was discovered as can be seen in Table 1 that the barrier waits do not add any appreciable execution time and in the optimised version of the code seen in Table 2 are even expanded out of there respective functions and executed in the code itself with no function call overhead. Without profiling this would have been much harder to identify and solve.

- Python script
- gprof outputs and how they were used

Compiler Optimisation - easy

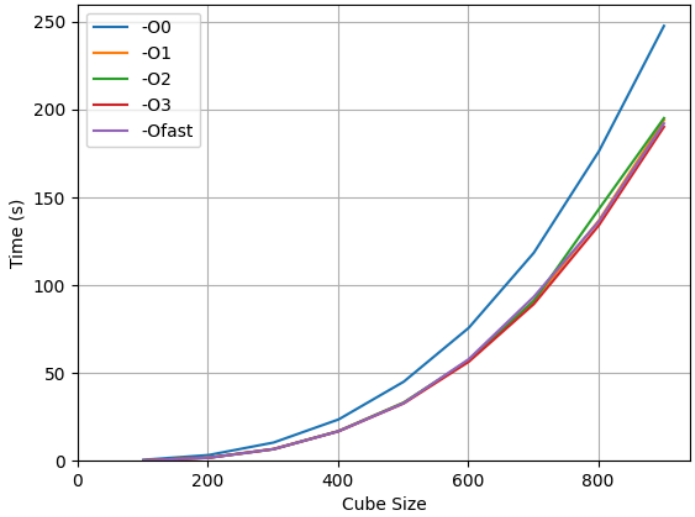

Figure 3: A comparison of the poisson solver software execution times with different compiler optimisations.

Individual Topic 1 Jack Duignan - Branch Prediction

Execution on a CPU is optimised using pipelineing which allows single clock cycle, single instruction excution. It does this by loading the next set of instructions while the previous instructions are being executed. When branch instructions occur a control hazard develops as the CPU is unsure which set of instructions to load. This can cause costly pipeline flushes if the CPU predicts wrong. To attempt to mitigate the number of times this pipeline flush occurs the CPU predicts which way execution will go before the conditional branch is executed. This is branch prediction and is implemented in various ways across CPU architectures.

In the poission iteration software there are several control hazards caused by conditional branching. One of the most significant of these is during the inner iteration of the poisson equation. In this operation the code has use several conditional instructions to check the location of the current node and apply the correct iteration. This results in our implementing of the iteration to have 12 conditional jump instructions per node update (if complied without optimisations). To improve the speed of our implementation it was decided to reduce the number of these jumps where possible to reduce the amount of branch prediction required by the CPU.

The reason our implementation required so many conditional instructions is that we use the same nested `for` loop to iterate over both the Von Neumann boundary (the dircilc boundary can be applied once during initialisation) and the inner nodes of the cube. It was hypothesised that moving these out of the same loop and reducing the Von Neumann iteration to only over the outer nodes would reduce the number of condition branch control hazards per iteration and thus overall. This was achieved by splitting each iteration into to components first the Von Neumann boundary is applied to only the nodes required then the nested loop only updated the inner nodes. This change completely removed the conditional instructions in the main iteration loop (which is called most often) removing the largest control hazard from the program.

The a comparison of the old program with conditional control hazards and without can be seen in Figure 4. This figure shows that the programs execution has been reduced by 10%. With the real benefits occur at the large cube sizes as the more iterations mean more conditional branch issues. This result is expected as by reducing the number of conditional branches the CPU can optimise use of pipelining as the number of possible pipeline flushes is reduced. This change also has the added benefit of reducing the number of instructions per iteration. This is due to the Von Neumann boundary application only iterating over nodes that will need to be applied to as opposed to all nodes in the cube.

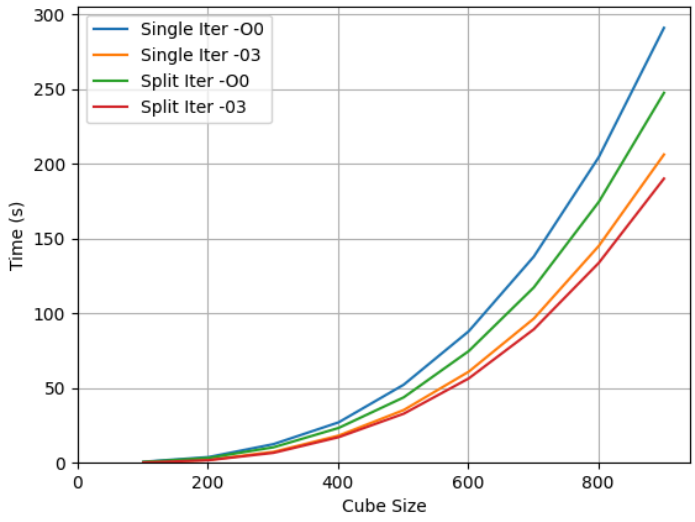

Figure 4: A comparison of the poisson solver software execution times with and without conditional branch reduction to reduce control hazards.

Individual Topic 2 Isaac Cone - GPU

Individual Topic 3 Daniel Hawes - SIMD

References