Note:

All tests were run on 20 cores and 1 node. All C code is compiled with -Ofast flag. To run under basic naïve algorithm, run command:

make basic make run

To run under tiled version, run command:

make tiled make run

To collect tiled version data for each tiled size, run command:

python compile_N_run.py (generate raw.dat) python beautify.py (generate beauty.dat)

python final_step.py (generate perf.dat, l1_miss.dat, llc_miss.dat)

All generated data are stored in a folder called "data".

To run compile_N_run.py in compute node, you have to load icc compiler first.

Question 1:

Below is my exercise.c file. I used C preprocessor and the -D compiler flag to define values for N and b at compile-time, as well as to deter- mine at compile time whether the code runs the basic algorithm or the tiled algorithm. I also made a makefile for a easier compiling time. See Appendix A for my makefile.

```
#include <stdio.h>
//#include <sys/time.h>
#ifndef N
   #define N 17600
#endif
#ifndef b
   #define b 50
#endif
double A[N][N];
double B[N][N];
int main() {
   //struct timeval start, end;
    //gettimeofday(&start, NULL);
   int i, j;
#ifdef BASIC
    for (i=0; i<N; i++) {
       for (j=0; j<N; j++) {
    A[i][j] += B[j][i];
    //printf("Runing basic algorithm\n");
#endif
  #ifdef TILED
       }
    //printf("Runing tiled algorithm\n");
   //gettimeofday(&end, NULL);
   //printf("Seconds elapsed: %f\n", (end.tv_sec*1000000.0 + end.tv_usec - start.tv_sec*1000000.0 -
       start.tv_usec) / 1000000.0);
}
```

Figure 1: C source file for this assignment

I tried multiple times with basic algorithm, and found if I define N to be 17600, it runs about 2 seconds under 20-core 1-node environment.

Question 2:

Method:

I used python to collect the performance data. See Appendix B, C, D for my python code of automation data collecting. First, I used Compile_N_Run.py to

compile and run the C code for 10 trials for each tile size between 1 and 300. This python code will redirect the performance data to a file called raw.dat. Then I used beautify.py to make the data more readable. These data is stored in beauty.dat. At last step, I used final_step.py to calculate the average elapsed time for each tile size.

Result:

Here is the plot of average elapsed time vs. tile size:

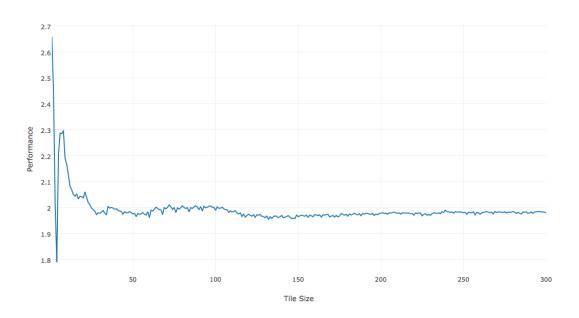


Figure 2: Elapsed Time VS. Tile Size

As clearly stated in the graph and data, for this program, tile size 4 achieves best elapsed time. With tile size 1, the elapsed time is the max. This is because with tile size 1, it is basically same with the basic algorithm. Moreover, the program has to compute extra variables, which increases the execution time. As tile size increasing, the elapsed time decreases generally. Before the tile size decreasing to 35, the elapsed time decreases greatly. With tile size between 35 and 100, it keeps around same elapsed time. With between 100 and 120, it decreases lightly. After the tile size bigger than 120, the elapsed time remains in a stable range.

With tile size of 4, the elapsed time reaches its minimum of 1.79 seconds. The speedup with regard to the basic version is 2/1.79 = 1.12.

Question 3:

Below is the graph of average L1 cache misses and LLC cache misses vs. tile size.

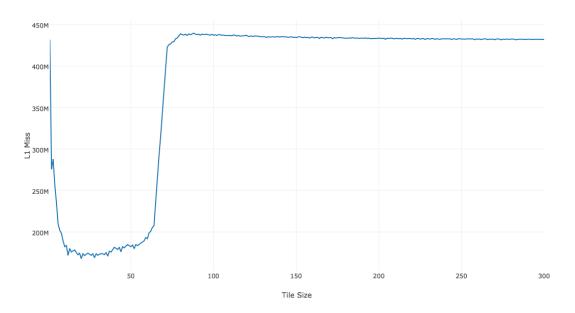


Figure 3: L1 Miss VS. Tile Size

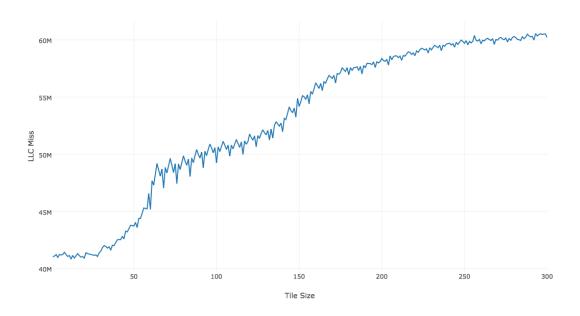


Figure 4: LLC Miss VS. Tile Size

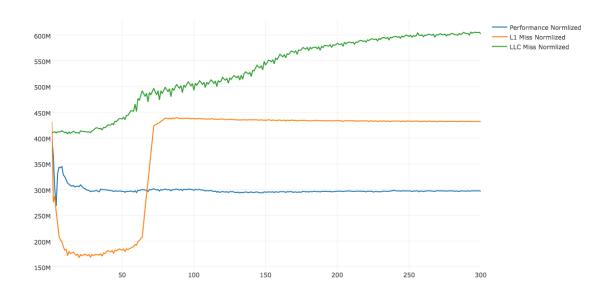


Figure 5: Elapsed Time, L1 Miss, LLC Miss VS. Tile size

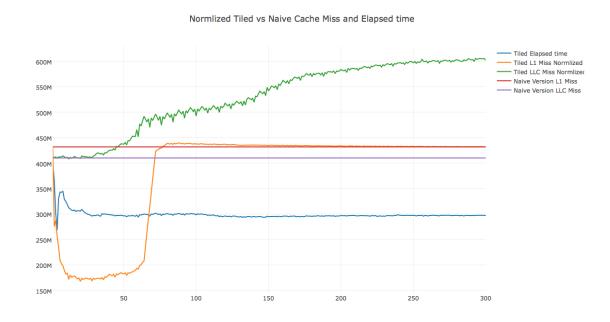


Figure 6: Tiled version VS. Naïve version

For the basic version the L1 cache miss is 431 million, LLC cache miss is 41 million, while with tiled tile size of 4 version, the L1 caches miss is 257 million, LLC cache miss is 41million. We can see that L1 dcache load misses is much smaller than the naïve version. While the LLC cache misses for both version starts

almost same, but the tiled version's LLC load misses increases as tile size grows. From the graph, we can see that, generally, the elapsed time decreases as L1 load misses decreases, which makes sense. However, there is some situation where this is not true. For example, our best tile size of 4, the L1 load miss is not in its lowest, which means the lowest L1 cache miss does not guarantee the lowest elapsed time. Based on this result, it is usually difficult to pick a decent tile size for best performance.

Appendix A

makfile:

```
CC=icc
CFLAGS=-mcmodel=medium -Ofast -c
PFLAGS=-D N=17600 -D b=4
all: exercise1
basic: CFLAGS += -D BASIC
basic: exercise1
tiled: CFLAGS += -D TILED
tiled: exercise1
exercise1: exercise1.0
       $(CC) exercise1.o -o exercise1
exercise1.o: exercise1.c
       $(CC) $(CFLAGS) $(PFLAGS) exercise1.c
run:
       ./exercise1
clean:
       rm *.o
real_clean:
       rm exercise1
```

Appendix B

compile_N_run.py:

```
from subprocess import call
def run_c_program():
    #call('rm output', shell=True)
for i in range(1,301):
        with open("./data/raw.dat", "a") as myfile:
             myfile.write("tile size=")
             myfile.write(str(i))
             myfile.write("\n")
         for j in range(10):
             compile_command = 'icc exercise1.c -o exercise1 -mcmodel=medium -Ofast -D 1
             compile_command += str(i)
             call(compile_command, shell=True)
             run_command =""
             run_command +="perf stat"
             run_command +=" -e L1-dcache-load-misses"
run_command +=" -e LLC-load-misses"
run_command +=" ./exercise1"
             run_command +=" 2>&1"
             run_command +=" | grep -E 'time|L1-dcache|LLC'"
             run_command +=" >> ./data/raw.dat"
             call(run_command, shell=True)
             #run = 'sbatch exercise1.slurm'
run_c_program()
```

Appendix C

beautify.py:

```
import re
This code beautify the raw data generate from 'perf stat | grep > file'
Such that it is easy to read and compute
def warehouse():
   out = open('./data/beauty.dat', 'w')
   with open('./data/raw.dat', 'r') as inputdata:
      line = ''
      item = ""
      count = 0
      for line in inputdata:
         if line =='':
            break
         if line.startswith('tile'):
            out.write(line)
            continue
         count += 1
         line = re.sub('[,@]', '', line)
         line = line.strip()
         line = line.split()
         for word in line:
            if is_number(word):
               item += str(word)
               if not count %3 == 0:
                  item += '\t'
         if count % 3 == 0:
            out.write(item)
            out.write("\n")
            item = ""
   out.close()
def is_number(s):
  try:
      float(s)
      return True
   except ValueError:
     try:
         int(s)
         return True
      except ValueError:
         return False
warehouse()
```

Appendix D

final_step.py:

```
def cal_ave():
    performance = open('./data/perf.dat', 'w')
    l1_miss = open('./data/l1_miss.dat', 'w')
    llc_miss = open('./data/llc_miss.dat', 'w')
    with open('./data/beauty.dat', 'r') as infile:
        line = ""
        count = 0
        perf = 0
        11 = 0
        llc = 0
        for line in infile:
            if line == '':
                break
            if line.startswith('tile'):
                continue
            count += 1
            line = line.strip().split()
            l1 += int(line[0])
            llc += int(line[1])
            perf += float(line[2])
            if count % 10 == 0:
                performance.write(str(perf/10))
                l1_miss.write(str(l1/10))
                llc_miss.write(str(llc/10))
                performance.write('\n')
                l1_miss.write('\n')
                llc_miss.write('\n')
                perf = 0
                llc = 0
                11 = 0
    performance.close()
    l1_miss.close()
    llc_miss.close()
cal_ave()
```