#### Lab 5

### Objective:

• The purpose of this lab is to introduce us to the idea of *benchmarking* and to familiarize ourselves with some of the tools used to benchmark code.

# What is Code Benchmarking

Wikipedia says that "in <u>computing</u>, a **benchmark** is the act of running a computer program in order to assess its relative **performance**, normally by running a number of standard tests and trials against it."

In this lab, we shall benchmark our program to assess its time performance as it is executing a few tasks on an array of various sizes. More specifically, we shall look at how long our program takes to ...

- randomize an array of various sizes, and
- quicksort this array, using two different partitioning algorithms.

### Part 1: Benchmarking a Loop

- Login to a CSIL machine in the Linux environment, download Lab5-files.zip and unpack it into sfuhome/CMPT295. Change directory into Lab5 where you will find the usual makefile and a collection of source files.
- Start by opening main.c in your favourite editor. The program's main purpose is to initialize an array of N integers, in the range [0, (N-1)/100], in a randomly permuted order.
- When you build the executable and run it, it will display the time it took (in microseconds (μs)) for the main loop to complete. This is done via the getrusage() system call, which asks the Linux operating system how much processor time has elapsed since the beginning of the program. Both calls to getrusage() record timing information in their parameters start and end.
- Important step: To know exactly how this timing information is processed, have a look at the printf statement in main.c where you will find both parameters start and end being used. To understand how they are used in this statement, first notice that they are both of type struct rusage, then have a look at the content of this struct by either using the man page: man getrusage, or having a look at <a href="http://man7.org/linux/man-pages/man2/getrusage.2.html">http://man7.org/linux/man-pages/man2/getrusage.2.html</a>.

As you can see from these two sources, the struct rusage makes use of the struct timeval (and the field of this struct type is called ru\_utime) to keep track of timing information.

You now need to see what struct timeval contains in order to have the complete picture. To do so, you can search the Internet. You will discover that this structure has

two fields. Read all about them as this knowledge will help you later on in this lab when you are debugging the code.

- An alternative to using <code>getrusage()</code> is to use the Linux time command. To try it, run time ./x. It should report the amount of *user* time for the entire program in seconds, to the nearest millisecond (ms). There should be a fairly close agreement between this number and the number of microseconds that the program reports for the main loop. Why do you think they might differ?
- You are going to time the main loop using different values of N. Initially, N is 1000000 (10<sup>6</sup>). Run the program 5 times, and record the time in μs for each in the <u>Main Loop</u> table of the data sheet found at the end of this lab.
- Next, change the value of N to 2000000 (2×10<sup>6</sup>). Run the program 5 times, and record the times in row 2 of the same table.
- Repeat for N equal to 4000000, 8000000, 16000000 and 32000000. Again run the program 5 times and record your data.
- Does the data make sense? To answer this question, we need to have an idea of what kind of results we are expecting. We are expecting the running time to roughly double for each doubling of N. Why? Because this main loop has a time complexity of O(N). This seems to occur except for the last value of N or two. This is because there is a bug in main.c, more specifically in how the times are computed. Hint: have a look at the printf statement and the two fields of struct timeval. Fix the bug and adjust your data accordingly.
- Calculate the average of each set of 5 numbers and report each number in the column marked μ.

Congratulations! You have just benchmarked the main loop of main.c, where the elements of the array A are randomized. You should observe a roughly linear progression. It isn't exactly linear, however. Do you have any suspicions as to why?

# Part 2: Benchmarking a Function

Next, you will benchmark two different versions of quicksort, each quicksort uses a different partitioning algorithm.

- Adjust your main.c so that it will report the time, in µs, of each of the quicksort functions. In doing so, do not delete or comment out the main loop since you still need a randomized array to benchmark each quicksort function.
  - What you need to do is move both calls to <code>getrusage()</code> so they wrap around the appropriate call to quicksort function and uncomment that call. You also need to move the print statement you just fixed in the previous Part of this lab.

- Just like in Part 1, you will run each quicksort function 5 different times using each of the 6 different values of N. Record your times in the appropriate table of the data sheet. Average your values and place each average in the column marked μ.
- Plot all three data sets on the graph paper found at the end of this lab. Use N (the sizes of your array) as the x axis and time (your recorded times - in seconds) as the y axis (since time is a function of N). Connect each data set with a sequence of 5 line segments. Finally, label each of the three resulting "curves" using the name of its corresponding data set.
- You can now move on to Part 3 below.

# Part 3: Challenge

Knowing that the behaviour of the main loop is O(N), is the shape of the curve on your graph representing the main loop as expected? What are we expecting?

Answer: A straight line since the relationship between N and the execution time of the main loop is linear. If this curve is not linear, does it have a concave upward curve to it? If so, this is probably due to the non-locality of the memory accesses performed during the execution of this loop (i.e., the number of cache misses increases non-linearly as N grows larger). We shall discuss this in more details when we cover memory caches in our last unit (Chapter 6 in our textbook).

Knowing that the average case behaviour of quicksort is O(N log N), have a look at your graph. Can you detect a slight concave upward tendency to the two curves on your graph representing *Quicksort 1* and *Quicksort 2*? This would match the O(N log N) curve (see: http://cooervo.github.io/Algorithms-DataStructures-BigONotation/big-O-notation.html). It may be very slight.

Lastly, looking at the two curves on your graph representing *Quicksort 1* and *Quicksort 2*, which one executes more quickly? It is said that Tony Hoare, the creator of quicksort back in 1960, was trying to write a simple sorting routine that took advantage of locality of memory references. He coded the original version in assembly, which might not be a surprise considering how tight the corresponding C code is! Although Hoare's algorithm is faster, Lomuto's algorithm appears in more CS textbooks because it is the easier of the two to understand.

Thank you to Brad Bart for having inspired this lab.

# Data Sheet

Data set: Main Loop

N	t1	t2	t3	t4	t5	μ
1000000	46602	43852	44094	62540	56866.	50790.8
2000000	126932	119775	122006	105691	117205	118321.8
4000000	395724	290186	386540	305442	277474.	331073.2
8000000	975809	914988	916591	975470	888193.	934210.2
16000000	1880462	1929428	1909624	1887427	1959963.	1913380.8
32000000	4349843	3909217	4042521	3890583	3998611	4038155

Data set: Quicksort 1

N	t1	t2	t3	t4	t5	μ
1000000	92918.	93186	93151	92815	101048.	94623.6
2000000	197229.	197806	198732	196937	197516.	197644
4000000	392410.	379534	389724	384463	384450.	386116.2
8000000	890122.	872534	865842	867222	871692.	873482.4
16000000	1830598	1700061	1712363	1731914	1714260.	1737839.2
32000000	3846885.	4010364	3539736	3581973	3616059.	3719003.4

Data set: Quicksort 2

N	t1	t2	t3	t4	t5	μ
1000000	166244.	167662.	166536.	165783.	166354.	166515.8
2000000	410179.	365236.	332439.	324771.	326528.	351830.6
4000000	710793.	715548.	716189	727486	716514.	717306
8000000	1546890.	1473936.	1503855.	1412370.	1391561.	1465722.4
16000000	3009707.	2878166.	2837206.	2961029.	3035807.	2944383
32000000	6352353.	6249141.	6283108	6263691.	6296984.	6289055.4

