HIGH PERFORMANCE PROGRAMMING UPPSALA UNIVERSITY SPRING 2018

LAB 14: DEBUGGING AND PROFILING

Profiling is a process of looking for slow parts of the program and optimizing them. The aim of this lab is to give hands-on experience of different profiling and debugging tools.

This lab includes the following parts:

- (1) Using the GDB debugger
- (2) More about valgrind memcheck
- (3) valgrind cachegrind
- (4) valgrind callgrind
- (5) gprof

Note: valgrind is usually available or at least easy to install on Linux computers. It is available on the lab computers. If you're using your own Mac, it may happen that valgrind and/or gprof is not available. It should be possible to install them, but it may be easier for you to use the Linux lab computers for the tasks involving valgrind and gprof. Alternatively, you can use ssh to login to one of the university's Linux servers, see the list of available Linux hosts here: http://www.it.uu.se/datordrift/maskinpark/linux.

Start by unpacking the Lab14_PerfAnalysis.tar.gz file which contains the files needed for the tasks in this lab.

1. The GDB debugger

Task 1:

The code for this task is in the Task-1 directory.

In Lab 1 we saw how GDB can be used to detect where in your code a "segmentation fault" error happens. Now we will look at some other ways of using gdb.

We will use the program littlecode.c as our example here. Suppose that we are interested in the state of the program when a certain function is called. We can then use gdb to set a so-called *breakpoint* in that place in the code. Then gdb will run the code until it reaches the breakpoint, and then stop, giving us the opportunity of investigating e.g. what values different variables have at that point of our program's run.

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To test this, first compile the program using the -g compiler option to include the debugging information that gdb needs:

gcc -g littlecode.c -o littlecode

Run the program to see what it does; it computes a number x that is printed. Now run gdb giving out program as input argument, like this:

gdb ./littlecode

Now we get a gdb prompt "(gdb)" where we can give different commands. Suppose we are interested in the program's behavior when the hh() function is called. Then we can set a breakpoint there, like this:

(gdb) break hh

Then tell gdb to start our program by giving the command "run". Now gdb will run our program until it reaches the breakpoint, and then it will stop and show the "(gdb)" prompt again. So now we are at the breakpoint. Try the "bt" (backtrace) command to show the current call stack; the functions that have been called to reach this point. Compare to the code in littlecode.c. Does the call stack look as you expect?

When we have stopped at a breakpoint we can also check the current values of different variables using the "print" command. Try this, for example to check the values of the global1 and global2 variables. Does it work? We can use the command "c" (continue) to continue execution of the program. In this case, since the hh() function is called in a loop, the program will again stop at the same breakpoint in the next loop iteration. Try doing "c" several times and look at the output this gives. You can see that it stops at the breakpoint in the hh() function, with different values of the a input argument. Do the values of a shown match your expectations, considering how the code in littlecode.c is written?

Another useful thing we can do with gdb is to find out why a program seems to hang. This is a common type of bugs that can be difficult to track down – if your program just appears to freeze it is hard to know what went wrong. In this situation gdb can help.

To see how gdb can help us when our program seems to hang, start by introducing a bug in the littlecode.c code: change the loop condition in the loop in the gg() function from c<100 to c<200. Since c is of type char this means that the loop will never end. Now we will pretend as if we do not know that, we just have a program with a bug that causes it to seemingly freeze while it is running.

Run the program:

./littlecode

So now the program just seems to hang, it never finishes. We can see the process ID (PID) of the program if we run top in another terminal window. Do that, and note the PID of the littlecode process. Knowing the PID we can run gdb and attach it to the process that is already running. If the PID is 1234 then this is done as follows:

gdb -p 1234

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Try that. What happens?

If you get an error message saying "Could not attach to process", see the alternative approach under "Alternative" below.

When attaching gdb to a running process like this, gdb will stop that program and let us check what the program was actually doing. In this case we will directly see that the program is currently inside the gg() function. We can use the "bt" command to see which other functions were called to reach this point, and we can print values of variables that we are interested in. Hopefully this will help us understand what had happened, in this case that we have an infinite loop due to the c<200 code change.

Alternative: On some systems, depending on security settings in Linux, gdb may not be allowed to attach to another process in the way described above. In that case, if you have a program that appears to hang and you want to stop it and find out where and why it hangs, you will have to start the program using gdb, then when it hangs enter <ctrl>C (where <ctrl> means the ctrl key on your keyboard) to interrupt the program. Then gdb will detect the interrupt and stop the program, and you will then see the gdb command prompt and be able to check what is going on in the same way as if you had attached gdb to a running process.

There is much more you can do with gdb – look at man gdb and/or type "help" within gdb to get more information. For example, you can look up the gdb commands "next" and "step", and try using them.

2. Valgrind and gprof

Task 2:

The code for this task is in the Task-2 directory.

The program for this task is doing some operations with an array. The array is initialised with random integers and a result is printed out.

Here you will be using valgrind. (You have already tried valgrind earlier in the course; here we will look at the valgrind memcheck functionality again, in a bit more detail.)

In order to profile a program with valgrind you first need to generate debugging information when you compile it by specifying the -g option to the compiler.

This debugging information is stored in the object file. It contains the data type of each variable or function and the correspondence between source line numbers and addresses in the executable code.

Try to compile and run the ex1.c program.

The program may seem to run without problems but it contains some memory errors. Code with memory errors can be tricky to debug since it can appear to work for small tests but if applied to large problems or if routines are called many times, memory errors will likely cause the program to crash or to become extremely slow. Your tasks is to find memory errors using valgrind and fix them. Valgrind is installed on the Linux systems of the university.

Usage of valgrind: valgrind [options] prog-and-args

Valgrind can be used in different ways, with different so-called *tools*. In order to specify which tool to use you pass to valgrind the corresponding option. Memcheck is the default tool, therefore one does not need to specify it explicitly, so

```
valgrind ./ex1
is the same as
valgrind --tool=memcheck ./ex1
```

Look at the output. Every row has prefix like ==12345==, where the number is an identifier of the working process. The memory errors are described after the line

```
==12345== Command: ./ex1
```

Here you should see lines similar to:

```
==12345== Invalid write of size 4
==12345== at 0x400872: main (ex1.c:35)
==12345== Address 0x520307c is 0 bytes after a block of size 60 alloc'd
==12345== at 0x4C2DB8F: malloc (vg_replace_malloc.c:299)
==12345== by 0x40081B: main (ex1.c:29)
```

This output indicates that there is no storage beyond the end of the array of 60 bytes, which means that program is trying to reach a location outside the bounds of the allocated memory. The (ex1.c:35) tells you the location in the code: line 35 in ex1.c. Valgrind also shows where the allocated memory buffer came from: memory was allocated using malloc in the main function (in ex1.c line 29).

Look also under LEAK SUMMARY.

If there is something definitely lost, that means the program has memory leaks.

Rerun with --leak-check=full to see details of leaked memory. If you fix all memory leaks you should see the following message:

```
All heap blocks were freed -- no leaks are possible
```

Next, uncomment the line with the call to function66(N) and again use valgrind to check for memory errors. Now you will see a few more lines for each error. This is because now there are errors inside some function calls and valgrind is showing the call stack which lead to each error, e.g. main called function66 which in turn called function77 where the error occurred.

To find more information about options for valgrind tools type valgrind --help

Valgrind can detect when your program tries to access memory outside what was allocated, giving "invalid read" or "invalid write" messages. However, this does not mean that valgrind can detect all errors where a program is going outside of array bounds. To see this, use the example code ex2.c. Look at the code to see what it does: it works with a struct type A that includes an array arr followed by a long int x.

Compile the program and run it with valgrind to check for errors. Then introduce a bug by changing from i < 8 to i < 9 in the loop where a->arr is set. This means

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that the program is going beyond the array size. Compile and run it with valgrind. Is valgrind able to detect this error? What happens with the a->x output value? Can you understand why?

Then introduce the same bug for b, changing from i < 8 to i < 9 in the loop where b->arr is set. Is valgrind able to detect that error? Can you understand why?

Task 3:

The code for this task is in the Task-3 directory.

The program matmul.c multiplies two dense square matrices. Three versions of the loop ordering (ijk, kij, jik) for multiplication are implemented.

Here we will use two of valgrind's tools for profiling: callgrind and cachegrind.

Valgrind cachegrind simulates a CPU with a 2-level cache. For modern machines it simulates the first and the last level (LL) caches (if cachegrind can detect the cache configuration). Usually the LL cache is the L2 or L3 cache. Valgrind cachegrind gives detailed cache and branch profiling.

Compile matmul.c using the -g flag.

Run valgrind on the executable using the --tool=cachegrind and --branch-sim=yes options.

Set matrix size to 200.

You will see a table with data collected by valgrind during execution of the program. In the first lines cachegrind reports information about instruction caches. In the second part are presented counters for cache accesses for data (rd = reads, wr = writes). At the end one can see combined information for LL cache and branch prediction statistics.

Cachegrind also writes a file, cachegrind.out.<pid> (where <pid> is the ID of the process), in the current directory. This file contains all the data.

The program cg_annotate can produce a detailed presentation of this data. Run e.g. "cg_annotate cachegrind.out.1234" (replace 1234 with the process ID in your case). The program cg_merge can be used to combine data from different cachegrind files.

In the output from cg_annotate, in the first three rows, the cache configuration is presented. It looks something like this:

I1 cache: 32768 B, 64 B, 8-way associative
D1 cache: 32768 B, 64 B, 8-way associative
LL cache: 6291456 B, 64 B, 12-way associative

In the above example, "I1 cache" is the instruction cache and "D1 cache" is the L1 data cache. So the size of the L1 data cache is in the above example 32768 bytes, and the "64 B" means that the cache line size is 64 bytes. If you want, you can compare the cache sizes listed by cg_annotate to the cache sizes given by the lscpu command; you should normally see the same cache sizes there. In that

way you can also see if the "LL cache" corresponds to the L2 or L3 cache on the computer you are using.

After that comes different counters, both for the entire program and separately for each function. The meaning of all the counters can be found in table 2.1.

Ir	nr. of instructions executed	Dr	nr. of memory reads
I1mr	L1 instruction cache read misses	D1mr	L1 data cache read misses
ILmr	LL cache instruction read misses	DLmr	LL cache data read misses
Dw	nr. of memory writes	Bc	conditional branches executed
D1mw	L1 data cache write misses	Bcm	conditional branches mispredicted
DLmw	LL data cache write misses	Bi	indirect branches execute
		Bim	indirect branches mispredicted

Table 2.1. Cachegrind counters and their meaning.

Use cg_annotate to look at data cache read misses for the mul_xyz functions. According to this information, which version of the matrix multiplication algorithm is the most cache efficient regarding the L1 data cache (D1mr and D1mw numbers) for the matrix size yo uwere testing now?

Look at the source code in matmul.c to see how the three different matrix-matrix multiplication variants are implemented. From looking a the code, can you understand the D1mr and D1mw numbers given by cachegrind?

Note that for efficient cache usage it is in general best to have "stride 1" memory access in the inner loop, as in the mul_kij case in matmul.c (here, "stride 1" means moving by one step in memory, thereby directly using all the contents of each cache line). However, depending on the size of the cache and the matrix size, the cache usage can sometimes be OK even if we do not have "stride 1" memory access in the inner loop. Look at the mul_ijk case — there the inner loop variable is k so the memory access b[k][j] does not look good. However, if n is small, it may happen that the b[k][j] value is anyway in cache because it was loaded into the cache line earlier, for the previous j. Based on that, we would expect the mul_ijk code to have OK L1 cache usage for small matrices, but to become drastically worse when the matrix becomes too large so that the b[k][j] value no longer remains in cache. Based on the size of the L1 data cache and the size of each cache line, can you predict at what matrix size this should happen? Experiment with some different matrix sizes and try to see this happening. At approximately what matrix size does the mul_ijk code run into trouble with the L1 cache usage? How much worse does it become compared to the mul_kij variant?

The callgrind tool extends the functionality of Cachegrind by recording the function call history.

Run valgrind --tool=callgrind ./a.out

Use the program callgrind_annotate to post-process the file generated by callgrind. You can read about the differences between Cachegrind and Callgrind here:

http://valgrind.org/docs/manual/cl-manual.html

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Task 4:

The code for this task is in the Task-4 directory.

This task is about profiling using gprof.

Note about gprof and compiler optimization flags: Unfortunately, when some more advanced compiler optimization are turned on that often leads to less information from gprof. For this reason, first do this task using only the -01 compiler optimization flag, to see what information gprof gives then. Then you can try also with other optimization flags.

For a description of gprof and its options type

```
man gprof
```

In order to profile with gprof add the flag -pg when compiling:

```
gcc -01 -g -pg matmul.c
```

Run the program for matrix with size 900

```
./a.out
```

Then list all the files in the current directory using ls -1:

```
ls -1
```

Note the new file gmon.out where profiling data has been created. Note that the profiler shows results corresponding to the actual run of the program. If in this run some function was not called, then this function will not be included in the analysis. To include all the functions, even those that were never called, use the -z option.

Now run the profiler and examine the output:

```
gprof a.out gmon.out
```

The output information is divided into two parts. The *flat profile* shows how much time your program spent in each function, and how many times that function was called. The *call graph* shows relations between functions. Which functions called the current function, which other functions it called, and how many times. There is also an estimate of how much time was spent in the subroutines of each function.

Rerun the program and compare results of gprof for different matrix sizes.

- What are the most time consuming operations?
- Which loop ordering is more efficient?
- What is the most called function?
- Does the profiler output time include time while waiting for the user input?
- Which metric is used to order function in the call graph?
- What are the self and children columns representing?

Note that the <code>-pg</code> compiler option that is needed for <code>gprof</code> profiling means that extra instrumentation code is added in the program, and the performance may be affected by this; your program will probably be slower when compiled with the <code>-pg</code> option.