ECE459: Programming for Performance	Winter 2021
Lecture 27 — Memory Profiling, Cachegrind	
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Memory Profiling Return to Asgard

Thus far we have focused on CPU profiling. Other kinds of profiling got some mention, but they're not the only kind of profiling we can do. Memory profiling is also a thing, and specifically we're going to focus on heap profiling. We kind of touched on the subject a little bit earlier when we looked at finding memory leaks. The ideas are the same: we don't want to leak memory, but remember that last category (other than suppressed), "Still Reachable", things that remained allocated and we still had pointers to them, but were not properly deallocated? Right, we care about them too, and for that we want to do heap profiling.

If we don't look after those things, we're just using more and more memory over time. That likely means more paging and the potential for running out of heap space altogether. Again, the memory isn't really lost, because we could free it.

Well, let's start with where we left off. Returning to the realm of Asgard, we're going to call again on our old friend Valgrind. Except this time we're going to use a fourth tool in it: Massif. This is, obviously, a joke on "massive", combined with the name Sif, a Norse goddess associated with the earth (and in the Marvel movies, Shieldmaiden to Thor). While we're on the subject, Sif has an axe (shield?) to grind with Loki, because at some point he cut off her golden hair (and in the Marvel films, it grew back in dark). That Loki—what a trickster! Right, we're digressing...what do you mean the course isn't ECE 459: Norse Mythology?!

So what does Massif do? It will tell you about how much heap memory your program is using, and also how the situation got to be that way. So let's start with the example program from the documentation [Dev16]:

```
#include <stdlib.h>
void q ( void ) {
    malloc( 4000 );
void f ( void ) {
    malloc( 2000 ):
    a():
int main ( void ) {
    int i:
   int* a[10]:
    for ( i = 0; i < 10; i++ ) {
        a[i] = malloc(1000);
    f();
   g();
    for ( i = 0; i < 10; i++ ) {
        free( a[i] );
    return 0:
}
```

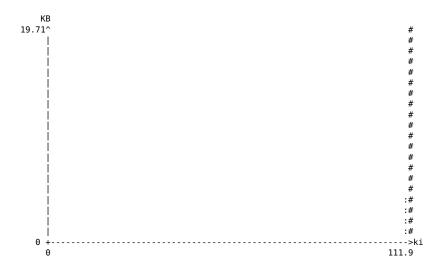
After we compile (remember the -q option for debug symbols), run the command:

```
jz@Loki:~/ece459$ valgrind --tool=massif ./massif
==25187== Massif, a heap profiler
==25187== Copyright (C) 2003-2013, and GNU GPL'd, by Nicholas Nethercote
==25187== Using Valgrind-3.10.1 and LibVEX; rerun with -h for copyright info
```

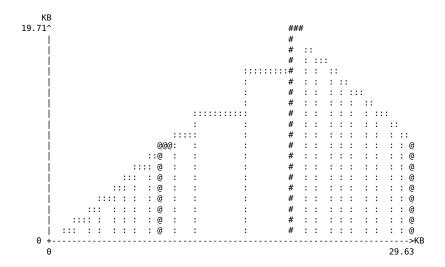
```
==25187== Command: ./massif
==25187==
==25187==
```

Doesn't that look useful?! What happened? Your program executed slowly, as is always the case with any of the Valgrind toolset, but you don't get summary data on the console like we did with Valgrind or helgrind or cachegrind. Weird. What we got instead was the file massif.out.25187 (matches the PID of whatever we ran). This file, which you can open up in your favourite text editor is not especially human readable, but it's not incomprehensible like the output from cachegrind ("Aha, a 1 in column 4 of line 2857. That's what's killing our performance!"). There is an associated tool for summarizing and interpreting this data in a much nicer way: ms_print, which has nothing whatsoever to do with Microsoft. Promise.

If we look at the output there (hint: pipe the output to less or something, otherwise you get a huge amount of data thrown at the console), it looks much more user friendly.



Now wait a minute. This bar graph might be user friendly but it's not exactly what I'd call... useful, is it? For a long time, nothing happens, then... kaboom! According to the docs, what actually happened here is, we gave in a trivial program where most of the CPU time was spent doing the setup and loading and everything, and the trivial program ran for only a short period of time, right at the end. So for a relatively short program we should tell Massif to care more about the bytes than the CPU cycles, with the --time-unit=B option. Let's try that.



Neat. Now we're getting somewhere. We can see that 25 snapshots were taken. It will take snapshots whenever there are appropriate allocation and deallocation statements, up to a configurable maximum, and for a long running program, toss some old data if necessary. Let's look in the documentation to see what the symbols mean (they're not just to look pretty). So, from the docs [Dev16]:

- Most snapshots are normal (they have just basic information) They use the ':' characters.
- Detailed snapshots are shown with '@' characters. By default, every 10th snapshot is detailed.
- There is at most one peak snapshot. The peak snapshot is a detailed snapshot, and records the point where memory consumption was greatest. The peak snapshot is represented in the graph by a bar consisting of '#' characters.

As a caveat, the peak can be a bit inaccurate. Peaks are only recorded when a deallocation happens. This just avoids wasting time recording a peak and then overwriting it; if you are allocating a bunch of blocks in succession (e.g., in assignment 1, a bunch of structs that have a buffer) then you would constantly be overwriting the peak over and over again. Also, there's some loss of accuracy to speed things up. Well, okay.

So let's look at the snapshots. We'll start with the normal ones. There are 9 of those, numbers 0 through 8:

n	time(B)	total(B)	useful-heap(B)	extra-heap(B)	stacks(B)
0	0	0	0	0	0
1	1,016	1,016	1,000	16	Θ
2	2,032	2,032	2,000	32	Θ
3	3,048	3,048	3,000	48	Θ
4	4,064	4,064	4,000	64	0
5	5,080	5,080	5,000	80	Θ
6	6,096	6,096	6,000	96	Θ
7	7,112	7,112	7,000	112	Θ
8	8,128	8,128	8,000	128	0

The columns are pretty much self explanatory, with a couple exceptions. The time(B) column corresponds to time measured in allocations thanks to our choice of the time unit at the command line. The extra-heap(B) represents internal fragmentation¹ in the blocks we received. The stacks column shows as zero because by default, Massif doesn't look at the stack. It's a heap profiler, remember?

Number 9 is a "detailed" snapshot, so I've separated it out, and reproduced the headers there to make this a little easier to remember what they are.

n	time(B)	total(B)	useful-heap(B)	extra-heap(B)	stacks(B)		
9	9,144	9,144	9,000	144	0		
98.43% (9,000B) (heap allocation functions) malloc/new/new[],alloc-fns, etc.							
->98.43% (9,000B) 0x4005BB: main (massif.c:17)							

So the additional information we got here is a reflection of where our heap allocations took place. Thus far, all the allocations took place on line 17 of the program, which was a[i] = malloc(1000); inside that for loop.

Then let's look at the peak snapshot (again, trimmed a bit to call out exactly what we need to see here):

 $^{^{1}}$ Remember from operating systems: if the user asked for some n bytes where n is not a nice multiple the returned block may be "rounded up". So a request for 1000 bytes is bumped up to 1016 bytes in this example. The extra space is "wasted" but it's nicer than having a whole bunch of little tiny useless fragments of the heap to be managed.

```
total(B) useful-heap(B) extra-heap(B)
                                                          stacks(B)
20,184
                       20,184
                                      20,000
                                                    184
                                                                 0
99.09% (20,000B) (heap allocation functions) malloc/new/new[], --alloc-fns, etc.
->49.54% (10,000B) 0x4005BB: main (massif.c:17)
->39.64% (8,000B) 0x400589: g (massif.c:4)
| ->19.82% (4,000B) 0x40059E: f (massif.c:9)
| | ->19.82% (4,000B) 0x4005D7: main (massif.c:20)
| ->19.82% (4,000B) 0x4005DC: main (massif.c:22)
->09.91% (2,000B) 0x400599: f (massif.c:8)
 ->09.91% (2,000B) 0x4005D7: main (massif.c:20)
```

Massif has found all the allocations in this program and distilled them down to a tree structure that traces the path through which all of these various memory allocations occurred. So not just where the malloc call happened, but also how we got there.

When program termination occurs we get a final output of what blocks remains allocated and where they come from. These point to memory leaks, incidentally, and valgrind would not be amused with us.

```
24 30,344 10,024 10,000 24 0
99.76% (10,000B) (heap allocation functions) malloc/new/new[], --alloc-fns, etc.
->79.81% (8,000B) 0x400589: g (massif.c:4)
| ->39.90% (4,000B) 0x40059E: f (massif.c:9)
| | ->39.90% (4,000B) 0x4005D7: main (massif.c:20)
| |
| ->39.90% (4,000B) 0x4005DC: main (massif.c:22)
|
| ->19.95% (2,000B) 0x400599: f (massif.c:8)
| ->19.95% (2,000B) 0x4005D7: main (massif.c:20)
|
|->00.00% (0B) in 1+ places, all below ms_print's threshold (01.00%)
```

In fact, if I ask valgrind what it thinks of this program, it says:

```
jz@Loki:~/ece459$ valgrind ./massif
==25775== Memcheck, a memory error detector
==25775== Copyright (C) 2002-2013, and GNU GPL'd, by Julian Seward et al.
==25775== Using Valgrind-3.10.1 and LibVEX; rerun with -h for copyright info
==25775== Command: ./massif
==25775==
==25775==
==25775== HEAP SUMMARY:
==25775== in use at exit: 10,000 bytes in 3 blocks
==25775== total heap usage: 13 allocs, 10 frees, 20,000 bytes allocated
==25775==
==25775== LEAK SUMMARY:
==25775== definitely lost: 10,000 bytes in 3 blocks
           indirectly lost: 0 bytes in 0 blocks possibly lost: 0 bytes in 0 blocks
==25775==
==25775==
           still reachable: 0 bytes in 0 blocks
==25775==
                  suppressed: 0 bytes in 0 blocks
==25775==
==25775== Rerun with --leak-check=full to see details of leaked memory
==25775==
==25775== For counts of detected and suppressed errors, rerun with: -v
==25775== ERROR SUMMARY: 0 errors from 0 contexts (suppressed: 0 from 0)
```

So probably a good idea to run valgrind first and make it happy before we go into figuring out where heap blocks are going with Massif. Okay, what to do with the information from Massif, anyway? It should be pretty easy to act upon this information. Start with the peak snapshot (worst case scenario) and see where that takes you (if anywhere). You can probably identify some cases where memory is hanging around unnecessarily.

Things to watch out for:

- memory usage climbing over a long period of time, perhaps slowly, but never really decreasing—memory is filling up somehow with some junk?
- large spikes in the graph—why so much allocation and deallocation in a short period?

Other cool things we can do with Massif [Dev16]:

- Look into stack allocation (--stacks=yes) option. This slows stuff down a lot, and not really necessary since we want to look at heap.
- Look at the children of a process (anything split off with fork) if desired.
- Check low level stuff: if we're doing something other than malloc, calloc, new, etc. and doing low level stuff like mmap or brk that is usually missed, but we can do profiling at page level (--pages-as-heap=yes).

As is often the case, we have examined how the tool works on a trivial program. As a live demo, let's see what happens when we take the program complexity up a little bit by (1) looking at the search program we saw in the earlier talk about valgrind; and (2) looking at the original (unmodified) paster.c file from assignment 1 (and then perhaps fixing it and going on). Depending on time available, we may look at some more complex programs.

Cachegrind

Cachegrind is another tool in the package and this one is much more performance oriented than the other two tools. Yes, Valgrind's memcheck and Helgrind look for errors in your program that are likely to lead to slowdowns (memory leaks) or make it easier to parallelize (spawn threads) without introducing errors. Cachegrind, however, does a simulation of how your program interacts with cache and evaluates how your program does on branch prediction. As we discussed earlier, cache misses and branch mispredicts have a huge impact on performance.

Recall that a miss from the fastest cache results in a small penalty (perhaps, 10 cycles); a miss that requires going to memory requires about 200 cycles. A mispredicted branch costs somewhere between 10-30 cycles. All figures & estimates from the cachegrind manual [Dev15].

Cachegrind reports data about:

- The First Level Instruction Cache (I1) [L1 Instruction Cache]
- The First Level Data Cache (D1) [L1 Data Cache]
- The Last Level Cache (LL) [L3 Cache].

Unlike for normal Valgrind operation, you probably want to turn optimizations on (-02 or perhaps -03 in gcc). You still want debugging symbols, of course, but enabling optimizations will tell you more about what is going to happen in the released version of your program.

If I instruct cachegrind to run on the search example (same one from above), using the -branch-sim=yes option because by default it won't show it:

```
jz@Loki:~/ece254$ valgrind --tool=cachegrind --branch-sim=yes ./search
==16559== Cachegrind, a cache and branch-prediction profiler
==16559== Copyright (C) 2002-2013, and GNU GPL'd, by Nicholas Nethercote et al.
```

```
==16559== Using Valgrind-3.10.0.SVN and LibVEX; rerun with -h for copyright info
==16559== Command: ./search
==16559==
--16559-- warning: L3 cache found, using its data for the LL simulation.
Found at 11 by thread 1
Found at 22 by thread 3
==16559==
==16559== I
              refs:
                         310,670
                           1,700
==16559== I1 misses:
==16559== LLi misses:
                           1.292
==16559== I1 miss rate:
                            0.54%
==16559== LLi miss rate:
                            0.41%
==16559==
                                  (77,789 rd
==16559== D
              refs:
                         114,078
                                               + 36,289 wr)
==16559== D1 misses:
                           4,398
                                  (3,360 rd
                                                 1.038 wr)
==16559== LLd misses:
                           3,252 ( 2,337 rd
                                                    915 wr)
==16559== D1 miss rate:
                             3.8% (
                                      4.3%
                                                    2.8%)
==16559== LLd miss rate:
                             2.8% (
                                                    2.5%)
                                      3.0%
==16559==
                           6,098
==16559== LL refs:
                                 ( 5,060 rd
                                               + 1,038 wr)
==16559== LL misses:
                           4,544 (3,629 rd
                                                    915 wr)
                             1.0% (
                                      0.9%
                                                    2.5%)
==16559== LL miss rate:
==16559==
==16559== Branches:
                          66,622
                                 (65,097 cond +
                                                  1,525 ind)
==16559== Mispredicts:
                           7,202 (6,699 cond +
                                                    503 ind)
==16559== Mispred rate:
                            10.8% ( 10.2%
                                                   32.9%
```

So we see a breakdown of the instruction accesses, data accesses, and how well the last level of cache (L3 here) does.

Why did I say enable optimization? Well, here's the output of the search program if I compile with the -02 option:

```
jz@Loki:~/ece254$ valgrind --tool=cachegrind --branch-sim=yes ./search
==16618== Cachegrind, a cache and branch-prediction profiler
==16618== Copyright (C) 2002-2013, and GNU GPL'd, by Nicholas Nethercote et al.
==16618== Using Valgrind-3.10.0.SVN and LibVEX; rerun with -h for copyright info
==16618== Command: ./search
==16618==
--16618-- warning: L3 cache found, using its data for the LL simulation.
Found at 11 by thread 1
Found at 22 by thread 3
==16618==
==16618== I
              refs:
                         306,169
==16618== I1 misses:
                           1,652
==16618== LLi misses:
                           1,286
==16618== I1 miss rate:
                            0.53%
==16618== LLi miss rate:
                            0.42%
==16618==
==16618== D
              refs:
                         112,015
                                  (76,522 rd
                                               + 35,493 wr)
==16618== D1 misses:
                           4,328
                                  ( 3,353 rd
                                                     975 wr)
==16618== LLd misses:
                           3,201
                                  ( 2,337 rd
                                                    864 wr)
==16618== D1 miss rate:
                             3.8% (
                                      4.3%
                                                    2.7%)
==16618== LLd miss rate:
                             2.8% (
                                      3.0%
                                                    2.4%)
==16618==
```

```
==16618== LL refs:
                            5,980 (5,005 rd
                                                      975 wr)
==16618== LL misses:
                            4,487 (3,623 rd
                                                      864 wr)
==16618== LL miss rate:
                              1.0% (
                                       0.9%
                                                      2.4% )
==16618==
==16618== Branches:
                           65,827
                                   (64,352 \text{ cond } +
                                                    1,475 ind)
==16618== Mispredicts:
                            7,109 (6,596 cond +
                                                      513 ind)
==16618== Mispred rate:
                             10.7% ( 10.2%
                                                     34.7%
```

Interesting results: our data and instruction miss rates went down marginally but the branch mispredict rates went up! Well, sort of—there were fewer branches and thus fewer we got wrong as well as fewer we got right. So the total cycles lost to mispredicts went down. Is this an overall win for the code? Yes.

In some cases it's not so clear cut, and we could do a small calculation. If we just take a look at the LL misses (4 544 vs 4 487) and assume they take 200 cycles, and the branch miss penalty is 200 cycles, it went from 908 800 wasted cycles to 897 400; a decrease of 11 400 cycles. Repeat for each of the measures and sum them up to determine if things got better overall and by how much. Also be sure that you're reasoning about a realistic workload.

Cachegrind also produces a more detailed output file, titled cachegrind.out.<pid> (the PID in the example is 16618). This file is not especially human-readable, but we can ask the associated tool cg_annotate to break it down for us, and if we have the source code available, so much the better, because it will give you line by line information. That's way too much to show even in the notes, so it's the sort of thing I can show in class (or you can create for yourself) but here's a small excerpt from the search.c example:

```
-- Auto-annotated source: /home/jz/ece254/search.c
Ir Ilmr ILmr Dr Dlmr DLmr Dw Dlmw DLmw Bc Bcm Bi Bim
127
           1 96
                3
                       0 4
                               Θ
                                   0 23 11 0 0
                                                        for ( int i = arg->startIndex; i < arg->endIndex; ++i ) {
147
           0 84
                  3
                       2 0
                               0
                                   0 21
                                          9
                                                           if ( array[i] == arg->searchValue ) {
                                                               *result = i;
 6
           0 4
                  0
                       0
                         2
                               0
                                   0 0
                                          0 0
                                                0
           0 0
                       0 0
 2
      0
                  0
                               0
                                   0 0
                                          0 0 0
                                                               break:
                                                       }
```

Cachegrind is very...verbose...and it can be very hard to come up with useful changes based on what you see...assuming your eyes don't glaze over when you see the numbers. Probably the biggest performance impact is last level cache misses (those appear as DLmr or DLmw). They have the highest penalty. You might also try to look at the Bcm and Bim (branch mispredictions) to see if you can give some better hints about what the likelihood of branch prediction is [Dev15]. Of course, to learn more about how Cachegrind actually does what it does and how it runs the simulation, the manual is worth reading. Not that anybody reads manuals anymore...Just give it a shot, when you get stuck, google the problem, click the first stack overflow link result...

References

- [Dev15] Valgrind Developers. Cachegrind: a cache and branch-prediction profiler, 2015. Online; accessed 25-November-2015. URL: http://valgrind.org/docs/manual/cg-manual.html.
- [Dev16] Valgrind Developers. Massif: a heap profiler, 2016. Online; accessed 23-January-2016. URL: http://valgrind.org/docs/manual/ms-manual.html.