

CoE 163

Computing Architectures and Algorithms

03a: High-Level Optimization

MAXIMIZING ALGORITHMS

"The real problem is that programmers have spent far too much time worrying about efficiency in the wrong places and at the wrong times; premature optimization is the root of all evil (or at least most of it) in programming."

- Donald Knuth, "The Art of Computer Programming"



MAXIMIZING ALGORITHMS

It is much more important to create *correct* code than efficient code. Write correctly now, speed it up later.

Optimizations, however, are useful especially if a lot of the running time is spent on some piece of code.



AMDAHL'S LAW

- Expression for the maximum expected improvement of the whole system if a part of it is optimized
- Usually used in parallel programming, but we can still use it for our "non-parallel" programs



AMDAHL'S LAW

$$S = \frac{1}{(1 - f_E) + \frac{f_E}{f_I}}$$

If f_E (percent) of the code has been sped-up by f_l (times), then the whole program will have a maximum speedup of S (times).



AMDAHL'S LAW

Suppose that we have a raytracer program with the intersection algorithm (around 40% of the whole program) sped-up by 5 times.







$$S = \frac{1}{(1 - f_E) + \frac{f_E}{f_I}}$$

= $\frac{1}{(1 - 0.4) + \frac{0.4}{5}}$
 ≈ 1.47

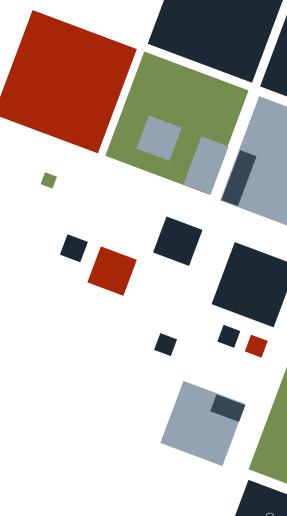
Was it worth it? Yes, if that code is commonly used.



GAINING SPEED-UPS

Knowing some basic code optimizations will come a long way in squeezing out less time from your code.

Code for correctness, but make obvious optimizations when opportunity comes.



C/C++ SPEED HACKS

- Use bit shift when 0 multiplying or dividing by two
- Simplify math expressions to 0 reduce the number of operations
- Take advantage of 0 short-circuit logic because conditionals are expensive



C/C++ SPEED HACKS

- Prefer pre-increment over post-increment
- Prefer iteration over recursion since function calls use the stack pointer
- Prefer pass-by-reference over pass-by-value
- ... a lot more!



PYTHON SPEED HACKS

- Some code parts may benefit from being coded into C and linked into Python
- Convert loops to list comprehensions or generators
- Take advantage of short-circuit logic
- ... and many more!



HOW WERE THEY DISCOVERED?

- Knowledge of assembly and the compiler
- Knowledge of computer architecture and microarchitectures
- Time and space profiling



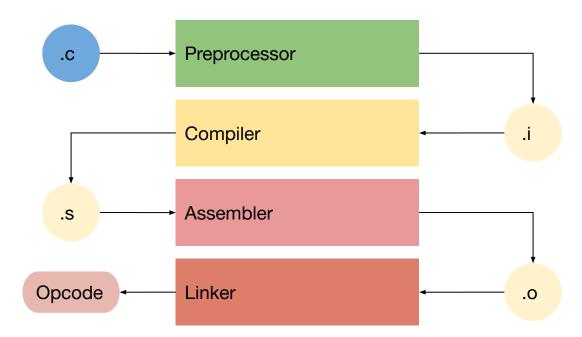
GCC

- The GNU Compiler
 Collection (GCC) is an
 optimizing compiler
- GCC has initially supported only C in 1987, but can now compile Go and D, among others
- It is an essential part of the GNU toolchain





GCC COMPILATION PIPELINE



GCC COMPILER

GCC compiles code in three-stages

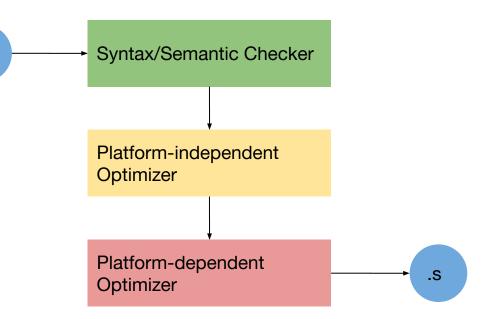
- Front syntax checking and parsing to an intermediate representation
- Middle platform-independent optimization
- Back platform-dependent optimization and conversion to assembly





GCC COMPILER

i.



GCC: ASSEMBLY

With the structure of GCC, it is possible to generate the intermediate preprocessor and assembly codes.

We can investigate how our code works at the low-level by reading the resulting assembly code.





GCC: ASSEMBLY

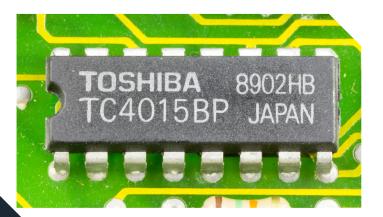
int main() {
 int x[] = {1, 6, 3};
 return x[0] + x[1] + x[2];

gcc -S arrays.c

push	rbp	Intel ASM
mov	rbp, rsp	
mov	DWORD PTR [rbp-12], 1	
mov	DWORD PTR [rbp-8], 6	
mov	DWORD PTR [rbp-4], 3	
mov	edx, DWORD PTR [rbp-12]	
mov	eax, DWORD PTR [rbp-8]	
add	edx, eax	
mov	eax, DWORD PTR [rbp-4]	
add	eax, edx	
pop	rbp	
ret		

CONSIDER...

Let's verify whether bit shifting is faster than division with a power of 2.







INTEGER DIVISION

int x = 16; int y = x / 4;

gcc -S div2.c

mov DWORD PTR [rbp-4], 16
mov eax, DWORD PTR [rbp-4]
lea edx, [rax+3]
test eax, eax
cmovs eax, edx
sar eax, 2
mov DWORD PTR [rbp-8], eax

Intel ASM

Extra test for integer division



BIT SHIFTING BY TWO

int x = 16;int y = x >> 2;gcc -S div2.c Intel ASM mov DWORD PTR [rbp-4], 16 mov eax, DWORD PTR [rbp-4] No extra test! sar eax, 2 mov DWORD PTR [rbp-8], eax

DIVISION BY POWER OF TWO

- When using integer division, the code still has to check whether the number is less than 0
- Load instructions are slow, but bitwise operations are fast
- Bit shifting is marginally faster than division by a power of two

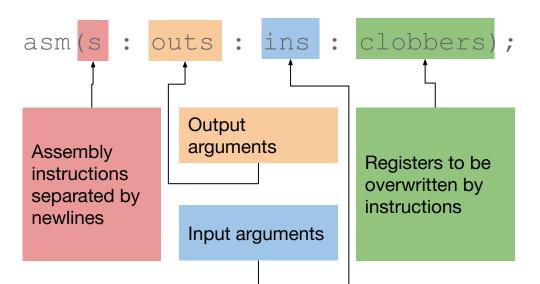


- GCC supports writing and compiling of assembly code within C/C++
- This is useful for systems development where some sections would run faster in assembly





INLINE ASSEMBLY IN C/C++





int a = 3, b = 3, c;

asm(

);

"mov %1, %%eax\n"
"mov %2, %%ebx\n"
"add %%eax, %%ebx\n"
"mov %%ebx, %0\n"

- : "=r" (c)
- : "r" (a), "r" (b)
- : "%eax", "%ebx"

Add a and b AT&T ASM

Although the feature is powerful, it is relatively easier to write a whole function in assembly.

It is also possible to write the assembly code separately as an . s file.





long add_me(long in, long in2); /* Prototype */

asm(/* Assembly function body */

"add_me:\n"

- " mov %rdi, %rax\n"
- " add %rsi, %rax\n"
- " ret\n"

);

int main(void) {
 return add_me(3, 5);

Add in and in2

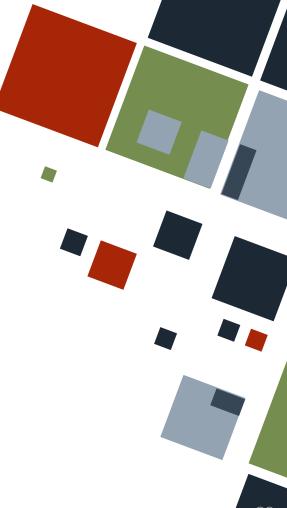
AT&T ASM

- Several binaries and software can be used to profile C/C++ Code
- Simplest is to use the built-in profiler that came with GCC



gprof

- Old profiler that uses statistical sampling to measure runtime
- Generates a decent report on the runtime per function of a program
- May be inaccurate since sampling time is usually 0.01s





gprof

- Compile code as normal with flags -pg
- Run programas normal and it will generate profile data named gmon.out in the directory where you are running the program
- Run gprof with the executable and profile data as arguments
- gprof generates a report on standard output use redirection to output into a file

\$ gcc add_me.c -o add_me -pg
\$./add me

\$ gprof add_me gmon.out



gprof - flat profile*

- Shows time spent running each function of a program
- Broken down into cumulative, number of calls, and percentage of program runtime executing such function

Flat profile:

Each sample counts as 0.01 seconds.

	-					
% C	cumulative	self		self	total	
time	seconds	seconds	calls	ms/call	ms/call	name
33.34	0.02	0.02	7208	0.00	0.00	open
16.67	0.03	0.01	244	0.04	0.12	offtime
16.67	0.04	0.01	8	1.25	1.25	memccpy
16.67	0.05	0.01	7	1.43	1.43	write
[]						



gprof - call graph*

- Shows time spent running a function and functions that it called during execution
- Useful for knowing a bit more information on where a program spends most of its runtime

granularity: each sample hit covers 2 byte(s) for 20.00% of 0.05 seconds

inde	x % time	self	children	called	name
					<spontaneous></spontaneous>
[1]	100.0	0.00	0.05		start [1]
		0.00	0.05	1/1	main [2]
		0.00	0.00	1/2	on_exit [28]
		0.00	0.00	1/1	exit [59]
		0.00	0.05	1/1	start [1]
[2]	100.0	0.00	0.05	1	main [2]
		0.00	0.05	1/1	report [3]
[1				

* not the profile of the add_me function

perf

- Newer general-purpose
 profiler for the Linux kernel
- Has a command line interface for viewing reports and even assembly code





perf

- Compile code as normal
- Run program through perf, and it will generate profile data named perf.data in the directory where you are running the program
- Run perf report to view the report

\$ gcc add_me.c -o add_me \$ perf record -g ./add_me \$ perf report



perf - call graph

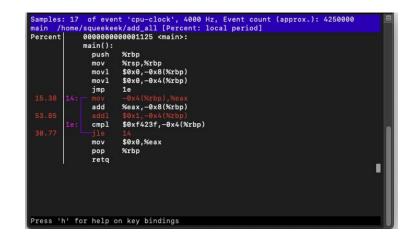
- Shows percentage time spent running a function and functions that it called during execution
- Shows the different libraries and functions called during execution

_	76.47%	Self 76,47%	Command add all		Symbol	
		9415541f6		800_811	[.] main	
		start_mai				
	main	ocur c_mar				
+	76.47%	0.00%	add_all	[unknown]	[.] 0x41fd89415541f689	
+		0.00%	add all		[.] libc start main	
+	11.76%	11.76%	add_all	[kernel.kallsyms]	[k] 0xffffffffbd06dff1	
+		0.00%	add all	[kernel.kallsvms]	[k] 0xffffffffbd80114e	
+		0.00%	add_all	[kernel.kallsyms]	[k] 0xffffffffbd066c79	
+		0.00%	add_all	[kernel.kallsyms]	[k] 0xffffffffbd800088	
+		0.00%	add_all	[kernel.kallsyms]	[k] 0xffffffffbd004183	
+		0.00%	add_all	[kernel.kallsyms]	[k] 0xffffffffbd206c06	
+			add_all	[kernel.kallsyms]	[k] 0xffffffffbd720455	
+		0.00%	add_all	[unknown]	[k] 0x0000000000000040	
+		0.00%	add_all	[unknown]	[k] 0x3d4c4c454853006c	
+		0.00%	add_all	ld-2.28.so	[.] 0x00007f0a37705a8d	
+		0.00%	add_all	ld-2.28.so	[.] 0x00007f0a37715650	
+		0.00%	add_all	[unknown]	[k] 000000000000000	
+		0.00%	add_all	ld-2.28.so	[.] 0x00007f0a377010dc	
+		0.00%	add_all	ld-2.28.so	[.] 0x00007f0a37703531	
+		0.00%	add_all	libc-2.28.so	[.] 0x00007f0a37556e56	



perf - disassembler

- Shows percentage time spent running each assembly instruction during execution
- Useful for knowing a bit more information on where a program spends most of its runtime

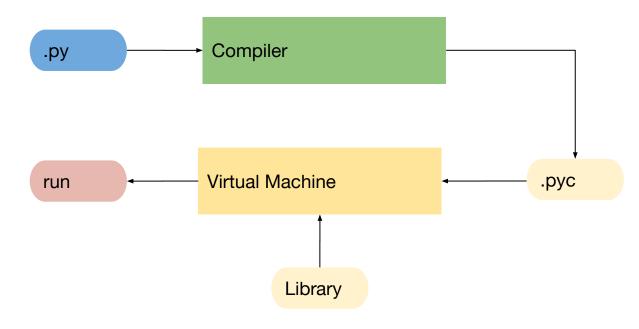


CPYTHON

- CPython is the reference implementation of Python since 1994
- It is an interpreter with an internal helper compiler
- It can either execute programs on-the-fly, or compile it into a platform-independent bytecode



CPYTHON INTERPRETATION PIPELINE



CPYTHON: "ASSEMBLY"

Since Python is interpreted, it generates a platform-independent bytecode instead of assembly code.

We can investigate how our code works at the intermediate level by reading the resulting bytecode.



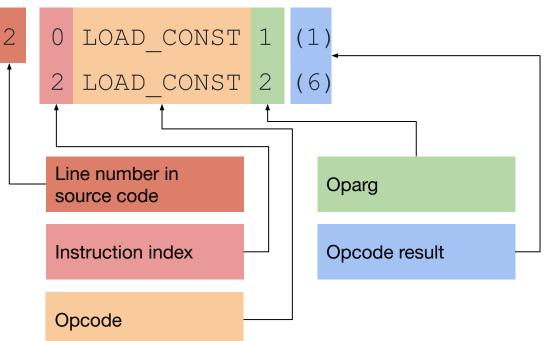


CPYTHON: BYTECODE

def	<pre>main(): x = [1, 6, 3] return x[0] + x[1]</pre>	+ x[2]		
	dis.dis(main)			
2	0 LOAD_CONST 2 LOAD_CONST 4 LOAD_CONST 6 BUILD_LIST 8 STORE FAST	2 3 3	(1) (6) (3) (x)	
3	10 LOAD_FAST 12 LOAD_CONST 14 BINARY_SUBSCR [] 32 RETURN_VALUE	0	(x) (0)	



CPYTHON: BYTECODE



CONSIDER...

Let's check the fastest way to add all the numbers in a list.







NORMAL LOOP

<pre>sum_all = 0</pre>					
<pre>for i in range(len(num_list)): sum_all += num_list[i] return sum_all</pre>					
	dis.dis(add_all)				
2 4	0 LOAD_CONST 2 STORE_FAST 4 SETUP_LOOP	1 (0) 1 (sum_all) 34 (to 40)			
5	[] 20 STORE_FAST 22 LOAD_FAST []	2 (i) 1 (sum_all)	Too many operations!		
	32 STORE_FAST 34 LOAD_FAST [] 42 RETURN VALUE	1 (sum_all) 1 (sum_all)			

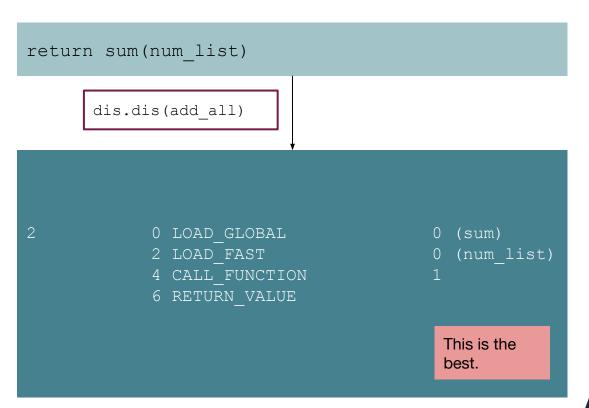


FUNCTOOLS REDUCE

retur	n reduce	e(lambda a,	b:	a +	b,	num_list)
	dis.dis	(add_all)				
2		LOAD_GLOBAL LOAD_CONST				0 (reduce) 1 ()
	6 8	4 LOAD_CONST 6 MAKE_FUNCTION 8 LOAD_FAST 10 CALL FUNCTION				2 () 0 0 (num_list) 2
		RETURN_VALU				Much better!



BUILT-IN SUM



SUMMING A LIST

- Constructing a for-loop requires set-up on the interpreter, and is relatively slow
- "Idiomatic" code, such as list comprehensions, runs faster than for-loops
- Built-in functions are the fastest due to them being globally accessible and leveraging a C backend (for CPython)



- There are several built-in modules and functions in Python for the purpose
- Simplest is to record the start and end time of executing a code section





timeit module

- Simplest timer for very small code snippets
- Runs the snippet 1 million times by default
- All code should be fed as strings

from timeit import timeit

```
timeit('''
sum(int_list)
''', setup='''
import numpy as np; int_list = np.random.randint(1, 100,
(1000,));
''')
```



time module

- Import the time library and get the time at appropriate instances of the program
- Simple and fast to use

import time
import numpy as np

```
start t = time.time()
```

```
int_list = np.random.randint(1, 100, (1000,))
add_all(int_list)
```

```
end_t = time.time()
print(f'Time elapsed: {end_t - start_t}s')
```



cProfile module

- Deterministic profiler with advanced break-down of time elapsed for each component
- Has an accuracy only up to 0.001s

import cProfile
import numpy as np

```
int_list = np.random.randint(1, 100, (1000,))
cProfile.run('''
for i in range(1000000):
        add_all(int_list)
''')
```



cProfile output

- Broken down by (sub)functions called
- Contains runtime in seconds and number of calls to that function during the whole profiling

2000003 function calls in 1.335 seconds

Ordered by: standard name

ncalls	tottime	percall	cumtime	percall	filename:lineno(function)
1000000	0.938	0.000	1.026	0.000	<ipython-input-38-94ea3e02d399>:1(add_all)</ipython-input-38-94ea3e02d399>
1	0.309	0.309	1.335	1.335	<string>:2(<module>)</module></string>
1	0.000	0.000	1.335	1.335	{built-in method builtins.exec}
1000000	0.089	0.000	0.089	0.000	{built-in method builtins.len}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>

TIPS

- Make the common case fast
- Program in assembly as a last resort
- Premature optimization is bad, but obvious optimization should be done
- Optimization takes twice time as normal programming



TIPS

- Profile different implementations to determine the fastest one
- Find the best profiler, or profiling strategy, according to your needs



RESOURCES

- Raytracer C/C++ optimization tips from <u>Clemson University</u>
- C/C++ to Assembly optimization resources from <u>Agner Fog</u>
- <u>Compiler Explorer</u> to check compilation results in C, C++, Python, and many more



RESOURCES

- Gprof resource from the <u>University of Utah</u>
- Short blog on <u>perftools</u> <u>usage</u>





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