Benchmarking and instrumentation

More caches and pipelines

- recall that memory is virtualized
- a virtual address \rightarrow hardware address translation is necessary for every memory access
- the translation uses a page table
- the page table is stored in memory
- hence at least two effective memory accesses per memory fetch instruction?

Solution: part of the page table is **cached** in the CPU the "Translation Lookaside Buffer" (TLB)

Caches and pipelines are used at various levels to hide access latency

typical latency

Examples of caches

- SSDs have internal RAM caches (typically 0-4 GB)
- the operating system caches files in memory
- large content providers (Google, Amazon, Netflix, Cloudflare) have caches all over the world

ping math.uwaterloo.ca

PING math.uwaterloo.ca (129.97.206.16) 56(84) bytes of data. bytes from ingress-p01.math.uwaterloo.ca (129.97.206.16): icmp_seq=1 ttl=43 time=110 ms bytes from ingress-p01.math.uwaterloo.ca (129.97.206.16): icmp_seq=2 ttl=43 time=110 ms bytes from ingress-p01.math.uwaterloo.ca (129.97.206.16): icmp_seq=3 ttl=43 time=111 ms bytes from ingress-p01.math.uwaterloo.ca (129.97.206.16): icmp_seq=4 ttl=43 time=111 ms

ping google.com.au

PING google.com.au (142.251.209.3) 56(84) bytes of data. bytes from mil04s50-in-f3.1e100.net (142.251.209.3): icmp_seq=1 ttl=115 time=12.2 ms bytes from mil04s50-in-f3.1e100.net (142.251.209.3): icmp_seq=2 ttl=115 time=14.6 ms bytes from mil04s50-in-f3.1e100.net (142.251.209.3): icmp_seq=3 ttl=115 time=12.9 ms bytes from mil04s50-in-f3.1e100.net (142.251.209.3): icmp_seq=4 ttl=115 time=11.8 ms

Examples of pipelines

- Storage devices:
	- SSDs typically access data in "pages" of 4096 bytes
	- **0.2ms** SSD latency would imply a max speed of 20 MB / s
	- instead SSDs routinely read and write 500 MB / s
- Networks:
	- Network packets are typically 1500 bytes
	- **10ms** WiFi latency would imply 150 KB / s
	- instead most WiFi networks do at least 10,000 KB / s
- Browsers:
	- Google Chrome maintains up to **6** connections per domain

Benchmarking

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- real: elapsed "real" (wall-clock) time
- user: time spent in user mode (running . /application code)
- sys: time spent in system mode (running OS kernel code)
- user + sys \lesssim real (there may be other applications running)

time head -n 1000000 /dev/random > /dev/null

time head -n 1000000 /dev/random > /dev/null

r e a l 0 m 0.4 4 2 s user 0 m 0.0 7 0 s s y s 0 m 0.3 7 1 s

time head -n 1000000 /dev/random > /dev/null

time head -n 1000000 /dev/random > /dev/null

Reasons for variance

- throttling for temperature and power limits (CPU adapts speed to avoid overheating or exceeding power supply capabilities)
- interactions with devices

(OS has in-memory caches for files, storage devices have internal memory caches)

• other processes

(must share resources)

top htop ps aux

(271 processes)

Effect of file caches

time md5sum 2GB_file

860a0023a913fd3fa4b6ad8bfbdd2c62 2GB_file

real 0m5.904s user 0m4.062s sys 0m0.560s

time md5sum 2GB_file

860a0023a913fd3fa4b6ad8bfbdd2c62 2GB_file

real 0m4.029s user 0m3.674s sys 0m0.331s

Inaccuracies

- executable startup is slow
- initialization adds overhead
- input and output are slow

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Executable startup is slow

int main() { $return 0;$ }

clang -O3 -o main main.c time ./main

time python -c 'exit(0)'

 \rightarrow we cannot accurately benchmark application that only take a few milliseconds.

Initialization adds overhead

sys 0m0.008s

What are we really measuring?

The speed of the MPS file parser, not the simplex algorithm.

Input and output are slow

```
def riemann_zeta(s):
    r = 0.0for i in range(1, 1000000):
       r = 1 / (i * s)return r
\# \zeta(2) = (pi^{**} 2) / 6print('pi ≈ ', (riemann_zeta(2) * 6) ** 0.5)
```
time python zeta.py

pi [≈] 3.141591698659554


```
def riemann_zeta(s):
    r = 0.0for i in range(1, 1000000):
         r = r + 1 / (i * s)print('r = ', r)
    return r
\# \zeta(2) = (pi \ * \ * 2) / 6\lvert \text{print('pi \approx ' , (riemann_zeta(2) * 6) ** 0.5)} \rvert
```
time python zeta.py

 $r = 1.0$ $r = 1.25$ $r = 1.3611111111111112$ $r = 1.4236111111111112$ $r = 1.4636111111111112$ r = 1.4913888888888889 $r = 1.511797052154195$ [...] $r = 1.6449330668467699$ $r = 1.64493306684777$ $pi \approx 3.141591698659554$ real 0m3.768s user 0m2.516s sys 0m0.999s

Aggregate measures

- if we benchmark our code on different inputs, we may want to use
	- total time / average time
	- **geometric mean**
	- or other aggregate measures
	- or some visualization (bar graphs, performance profiles, etc.)
- but beware: all aggregate measures are biased

Static instrumentation

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- we may want to benchmark specific parts of our code
	- to circumvent executable startup, initialization, and input/output
	- to benchmark parts of the code that run quickly
	- **to find bottlenecks**
- for that, we need to add timing instrumentation to our code

About bottlenecks

"Premature optimization is the root of all evil"

– Donald Knuth "Structured Programming With GoTo Statements" 1974

nes of code

les of code

les of code

lines of code

t i m e . t i m e ()

initialize() function_A() function_B() function_C() cleanup()

```
import time
t0 = time.time()initialize()
t1 = time.time()function_A()
t2 = time.time()function_B()
t3 = time.time()function_C()
t4 = time.time()cleanup()
t5 = time.time()print(f'total time: {t5 - t0:16.6f}')
print(f'function_A: {t2 - t1:16.6f}')
print(f'function_B: {t3 - t2:16.6f}')
print(f'function_C: {t4 - t3:16.6f}')
print(f' rest: {(t5 - t0) - (t4 - t1):16.6f}')
```
c l o c k _ g e t t i m e ()

```
int main()
{
    initialize();
    function_A();
    function_B();
    function_C();
    cleanup();
    return 0;
}
```


clock_gettime(CLOCK_MONOTONIC, &t1); function_A();

clock_gettime(CLOCK_MONOTONIC, &t2); function_B();

}

 $\left(\ \right)$

uct timespec t0, t1, t2, t3, t4, t5;

clock_gettime(CLOCK_MONOTONIC, &t0); initialize();

clock_gettime(CLOCK_MONOTONIC, &t3); function_C();

clock_gettime(CLOCK_MONOTONIC, &t4); cleanup();

clock_gettime(CLOCK_MONOTONIC, &t5);

print_all_clocks(&t0, &t1, &t2, &t3, &t4, &t5); return 0;

Cumulative time

```
initialize()
tA, tB, tC = \emptysetfor i in range(1000000):
   t0 = time.time()function_A()
   t1 = time.time()function_B()
   t2 = time.time()function_C()
   t3 = time.time()tA = (t1 - t0)tB == (t2 - t1)tC = (t3 - t2)
```


cleanup()

Caveat: measuring time takes time!

time.time(): ~40 ns (and this value fluctuates!)

Microbenchmarks

What do we do if function_A() takes much less time than time.time()?

nchmark for function_A():

```
me.timee()time()range(50000000):
ion_A()time()
```
Microbenchmarks limitations

- It may not make sense to call function A() in isolation
	- \blacksquare Take sin(x) for example: which value of x do we choose?
	- Always the same?
		- \circ Are we sure sin(0) takes as much time as sin(0.1)?
	- A random value for x ?
		- \circ What if generating pseudo-random values takes more time than $\sin($)?
- What about caches?
	- Caches will be "hot" (already filled with relevant data)
	- Microbenchmarking presents an over-optimistic picture of memory access times

Automated instrumentation: Profilers

Add "-pg" to gcc/clang parameters

gcc -O3 -o app app.c -pg

Run the application

./app

Generate report

gprof app

- Pros
	- Easy to use
	- **Exhaustive profile information**
	- Generally low overhead
- Cons
	- Overhead increases when bottlenecks are in small, short functions (up to 2x runtime)
	- **ELimited accuracy**

Hardware performance counters

The simplest hardware-aided performance-measuring tool is: the time stamp counter (TSC)

- Introduced by Intel with the Pentium architecture (1993)
- Similar feature available on ARM since ARMv7 (1996)
- Special integer register
- Incremented by one at a constant rate (e.g. every clock cycle)
- Reading this register has high latency $(>10$ cycles)
- Useful for microbenchmarks and instrumentation
- time.time() / clock_gettime() use this internally

More complex performance counters

Since then, Intel and ARM have added many more performance counters:

 \mathcal{L}^{max}

- lys measured
- erformance penalty
- nterference with normal execution

an aggregate measure (totals)

Linux perf

perf stat ./application

Performance counter stats for './application': 3,216.90 msec task-clock # 1.000 CPUs utilized 8 context-switches # 2.487 /sec 1 cpu-migrations # 0.311 /sec 6,205 page-faults # 1.929 K/sec 9,442,508,623 cycles # 2.935 GHz 7,596,331,032 instructions # 0.80 insn per cycle (58.81%) 1,086,117,213 branches # 337.629 M/sec 1,085,287 branch-misses $\qquad \qquad # \qquad 0.10\%$ of all br 2,162,685,901 L1-dcache-loads # 672.289 M/sec 1,079,393,101 L1-dcache-load-misses # 49.91% of all L1 1,069,062,732 LLC-loads # 332.327 M/sec 6,537,301 LLC-load-misses $#$ 0.61% of all L1 2,161,850,109 dTLB-loads # 672.029 M/sec 896,301 dTLB-load-misses # 0.04% of all dT 9,051,173 dTLB-stores # 2.814 M/sec 81,624 dTLB-store-misses # 25.374 K/sec (23.50%) 3.217829387 seconds time elapsed

3.167788000 seconds user 0.022723000 seconds sys

