Outline

- Definition of performance
- Execution time
- What affects your performance

Performance

What do you want for a computer?

- Latency/Execution time
- Frame rate
- Responsiveness
- Real-time
- Throughput
- Cost
- Volume
- Weight
- Battery life
- Low power/low temperature
- Reliability

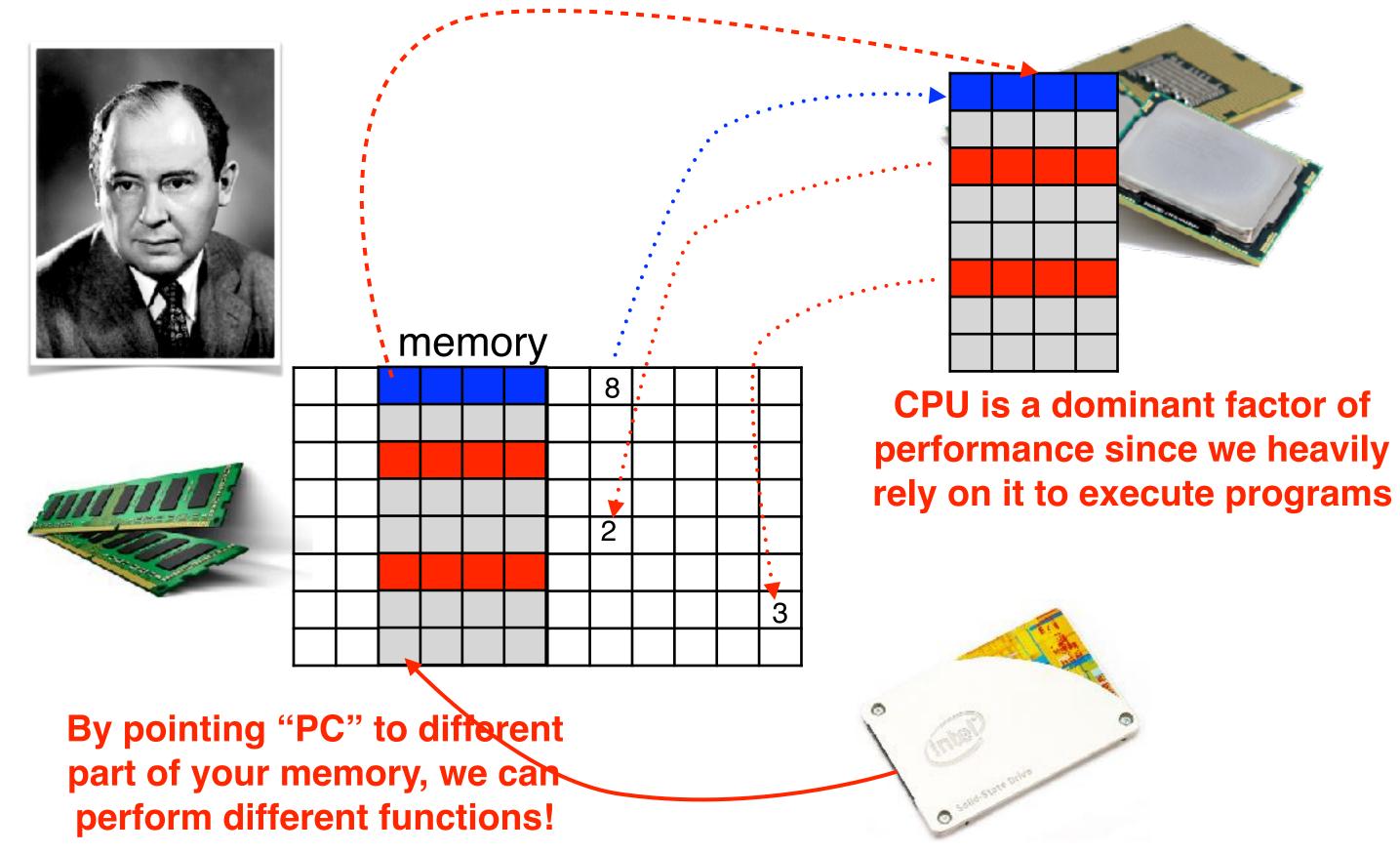
How about running a single program

- Latency/Execution time
- Frame rate
- Responsiveness
- · Real-time
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The most direct measurement of performance

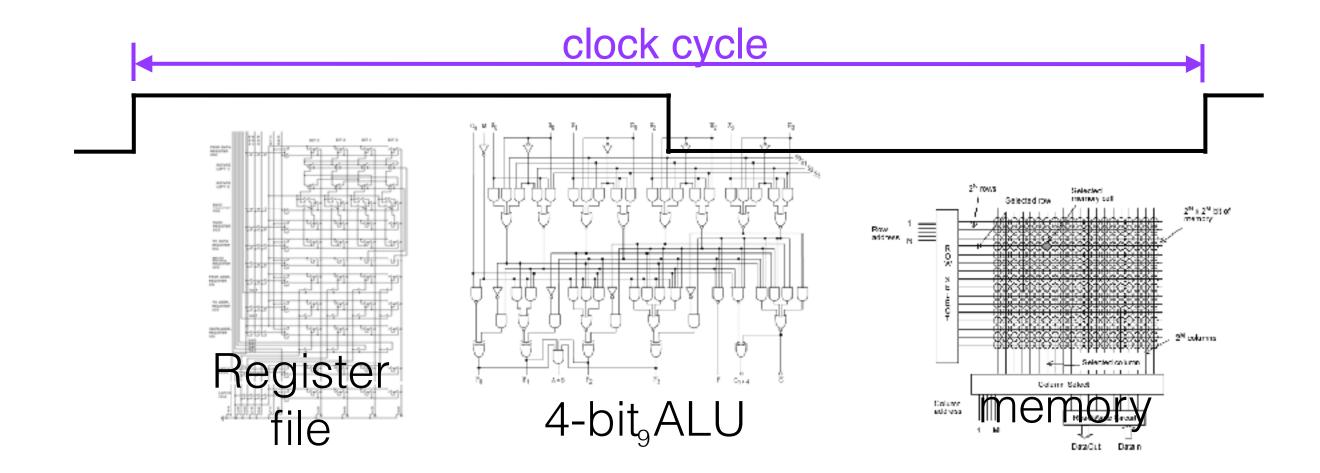
Evaluating the execution time of a program

Recap: Von Neumann architecture



Recap: Clock — synchronizing hardware components

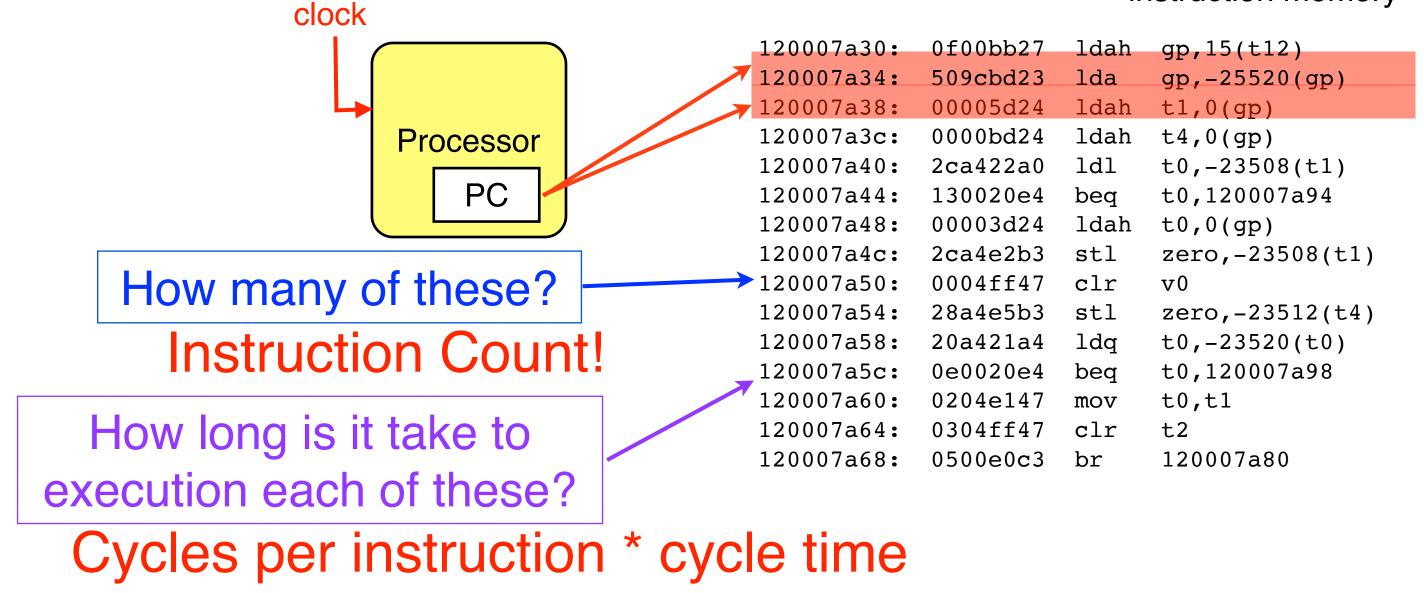
- A hardware signal defines when data for any specific component is ready to use by others
 - Think about the clock in real life!
- We use edge-triggered clocking
 - Values stored in the sequential logic is updated only on a clock edge



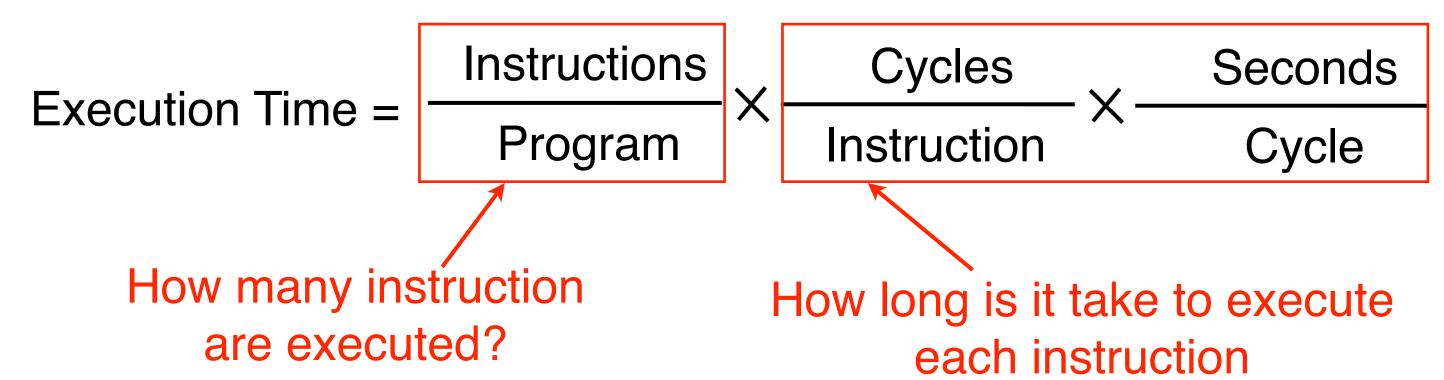
Execution Time

- The simplest kind of performance
- Shorter execution time means better performance
- Usually measured in seconds

instruction memory



Performance Equation



- ET = IC * CPI * CT
- IC (Instruction Count)
- CPI (Cycles Per Instruction)
- CT (Seconds Per Cycle)
 - 1 Hz = 1 second per cycle; 1 GHz = 1 ns per cycle



Relative performance

- Can be confusing
 - A runs in 12 seconds
 - B runs in 20 seconds
 - We know A is faster, but
 - A/B = .6, so A is 40% faster, or 1.4X faster, or B is .40% slower
 - B/A = 1.67, so A is 67% faster, or 1.67X faster, or B is 67% slower
- Needs a precise definition

Speedup

- Compare the relative performance of the baseline system and the improved system
- Definition

What affects performance

Demo: programmer & performance

- Row-major, column major
 - How do you know this?
- Let's identify where the performance gain is from!
 - Using "performance counters"
 - You may use "perf stat" in linux
 - You can also create your own functions to obtain counter values
 - https://github.ncsu.edu/htseng3/CSC456/tree/master/performance

Applications

· Different applications can have different CPIs on the same machine

Compiler

 Compiler can change the combination of instructions and lead to different CPIs, instruction counts.

Summary: Performance Equation

Execution Time =
$$\frac{Instructions}{Program} \times \frac{Cycles}{Instruction} \times \frac{Seconds}{Cycle}$$

- ET = IC * CPI * Cycle Time
- IC (Instruction Count)
 - · ISA, Compiler, algorithm, programming language
- CPI (Cycles Per Instruction)
 - Machine Implementation, microarchitecture, compiler, application, algorithm, programming language
- Cycle Time (Seconds Per Cycle)
 - · Process Technology, microarchitecture, programmer

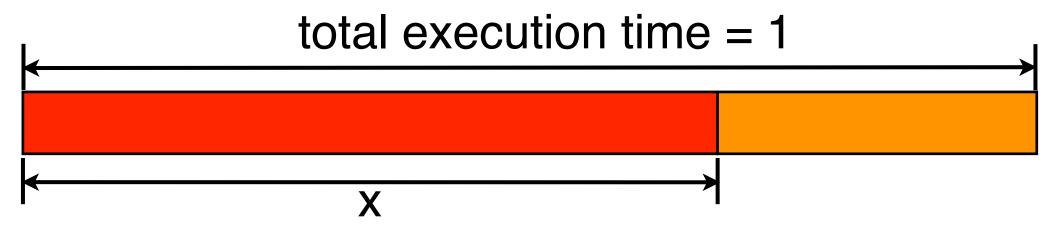
Amdahl's Law

Amdahl's Law

Speedup =
$$\frac{1}{(\frac{x}{S})+(1-x)}$$



- x: the fraction of "execution time" that we can speed up in the target application
- S: by how many times we can speedup x



total execution time =
$$((\frac{x}{S})+(1-x))$$

Performance Example

- Assume that we have an application composed with a total of 500000 instructions, in which 20% of them are the load/store instructions with an average CPI of 6 cycles, and the rest instructions are integer instructions with average CPI of 1 cycle.
 - If we double the CPU clock rate to 4GHz but keep using the same memory module, the average CPI for load/store instruction will become 12 cycles. What's the performance improvement after this change?

How much time in load/store?

500000 * (0.2*6) * 0.5 ns = 300000 ns 60%

How much time in the rest?

500000 * (0.8*1) * 0.5 ns = 200000 ns 40%

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Speedup =
$$\frac{1}{\frac{0.4}{2} + (1-0.4)}$$

Speedup = $\frac{1}{0.8}$ = 1.25

Amdahl's Law: Revisited

Performance Example

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Amdahl's Corollary #1

Maximum possible speedup Smax, if we are targeting x of the program.

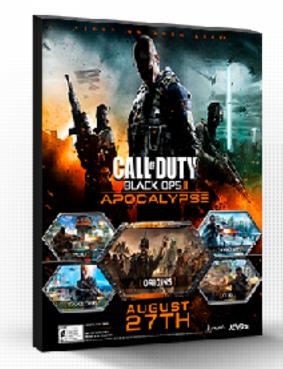
$$S = infinity$$

$$S_{max} = \frac{1}{0 \left(\frac{x}{\inf} + (1-x)\right)}$$

$$S_{max} = \frac{1}{(1-x)}$$

Maximum of speedup

 Call of Duty Black Ops II loads a zombie map for 10 minutes on my current machine, and spends 20% of this time in integer instructions



How much faster must you make the integer unit to make the map loading 5 minutes faster?

$$S_{max} = \frac{1}{(1-x)}$$

$$1.25 = \frac{1}{(1-20\%)}$$

2x is not possible.

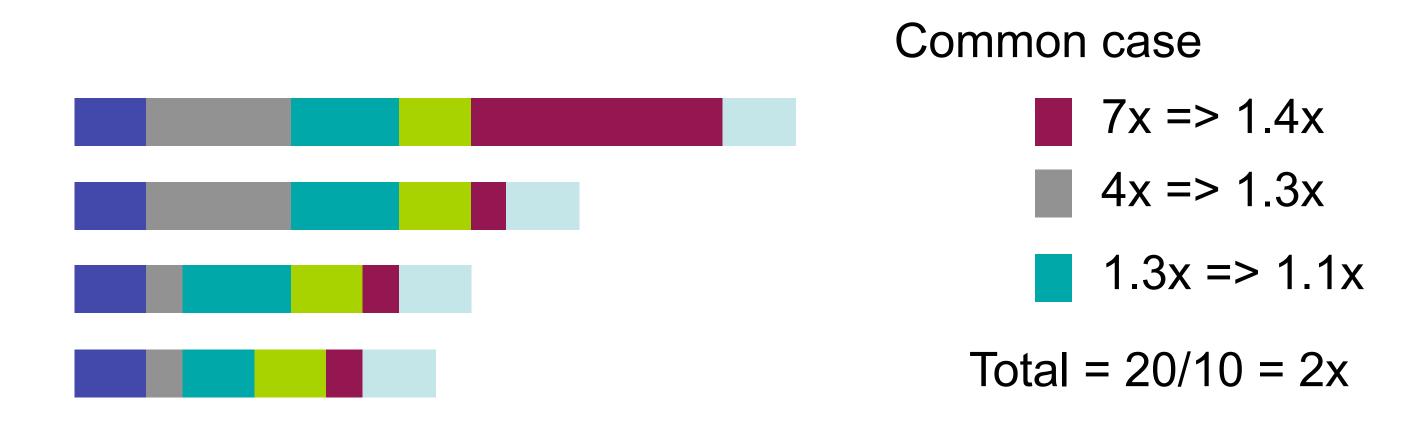
Amdahl's Corollary #2

- Make the common case fast (i.e., x should be large)!
- Common == most time consuming not necessarily the most frequent
- The uncommon case doesn't make much difference
- Be sure of what the common case is
- The common case can change based on inputs, compiler options, optimizations you've applied, etc.

Identify the most time consuming part

- Compile your program with -pg flag
- Run the program
 - It will generate a gmon.out
 - gprof your_program gmon.out > your_program.prof
- It will give you the profiled result in your_program.prof

If we repeatedly optimizing our design based on Amdahl's law...



- With optimization, the common becomes uncommon.
- An uncommon case will (hopefully) become the new common case.
- Now you have a new target for optimization.

Demo

- Quicksort takes a lot of time if we want to sort a 300M array
- GPU gives you 10x speed up!
- New bottleneck emerges!

Don't hurt non-common part too mach

- If the program spend 90% in A, 10% in B. Assume that an optimization can accelerate A by 9x, by hurts B by 10x...
- Assume the original execution time is T. The new execution time

$$T_{new} = \frac{T \times 0.9}{9} + T \times 0.1 \times 10$$

$$T_{new} = 1.1T$$

$$Speedup = \frac{T}{1.1T} = 0.91$$

Amdahl's Corollary #3

• Assume that we have an application, in which x of the execution time in this application can be fully parallelized with S processors. What's the speedup if we use a S-core processor instead of a single-core processor?

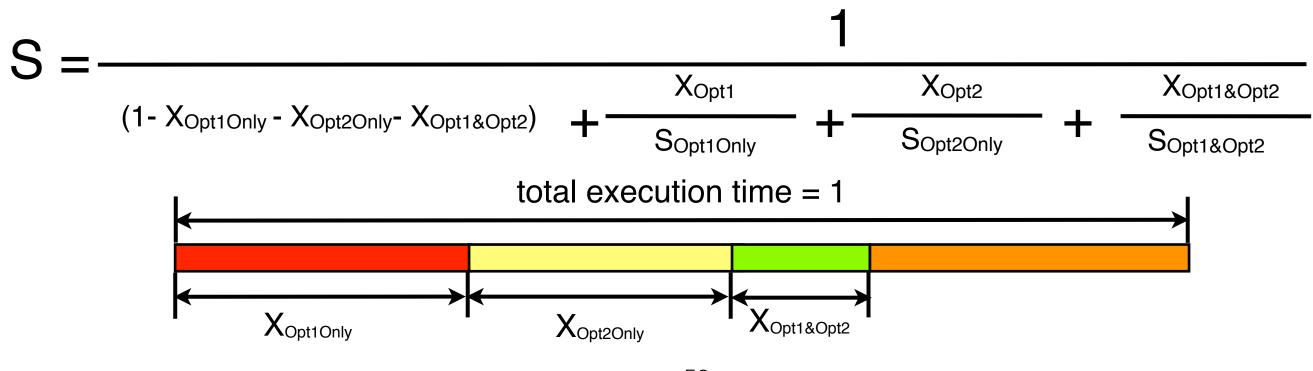
$$S_{par} = \frac{1}{\frac{X}{S} + (1-X)}$$

Multiple optimizations

- We can apply Amdahl's law for multiple optimizations
- These optimizations must be dis-joint!
 - If optimization #1 and optimization #2 are dis-joint:

Speedup =
$$\frac{1}{(1-X_{Opt1}-X_{Opt2})} + \frac{X_{Opt1}}{S_{Opt1}} + \frac{X_{Opt2}}{S_{Opt2}}$$

If optimization #1 and optimization #2 are not dis-joint:



Amdahl's Law for multicore processors

 Assume that we have an application, in which 50% of the application can be fully parallelized with 2 processors. Assuming 80% of the parallelized part can be further parallelized with 4 processors, what's the speed up of the application running on a 4-core processor?

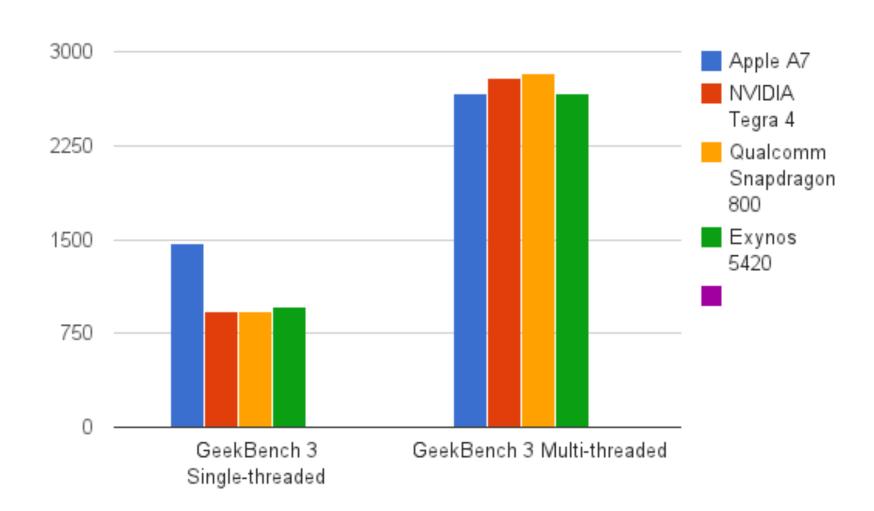
Code can be optimized for 2-core = 50%*(1-80%) = 10%

Code can be optimized for 4-core = 50%*80% = 40%

Speedup_{quad} =
$$\frac{1}{(1-0.5) + \frac{0.10}{2} + \frac{0.40}{4}} = 1.54$$

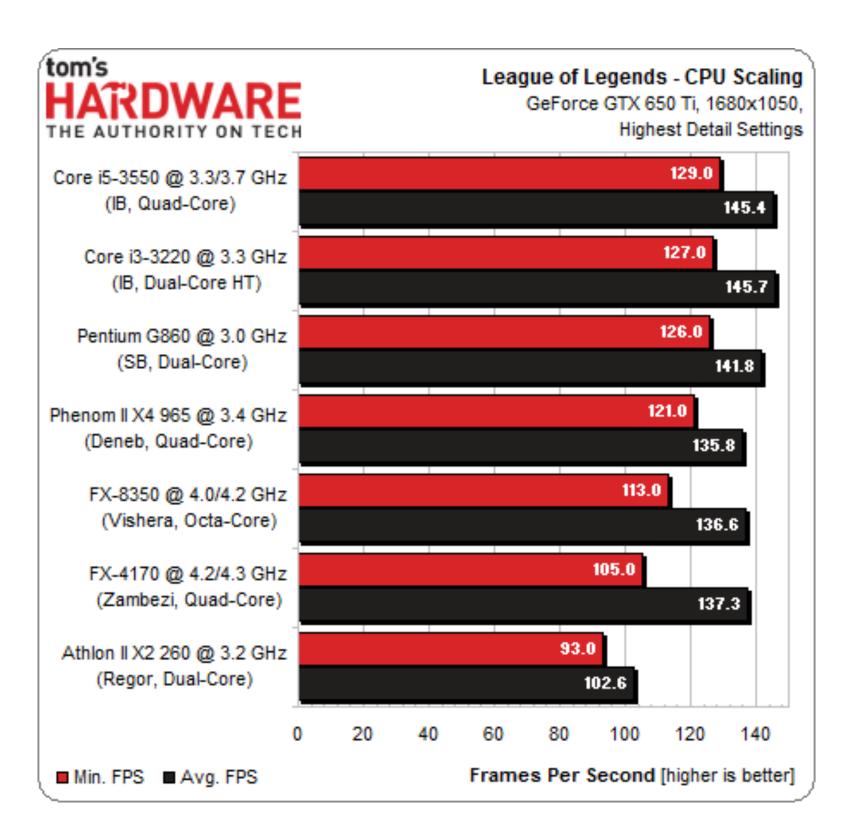
Case study: more cores?

If you cannot make your mobile Apps multithreaded, Apple A7 is the best



Case study: LOL

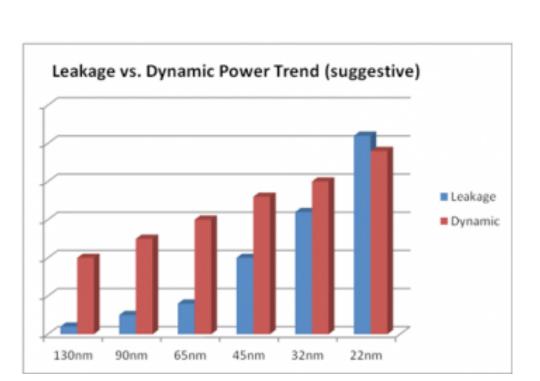
- Corollary #2
- The CPU is not the main performance bottleneck
- CPU parallelism doesn't help, either
- You might consider
 - · GPU
 - network
 - storage (loading maps)



Power & Energy

Power

- Dynamic power: P=aCV²f
 - a: switches per cycle
 - · C: capacitance
 - V: voltage
 - f: frequency, usually linear with V
 - Doubling the clock rate consumes more power than a quad-core processor!
- Static/Leakage power becomes the dominant factor in the most advanced process technologies.
- Power is the direct contributor of "heat"
 - Packaging of the chip
 - Heat dissipation cost



Energy

- Energy = P * ET
- The electricity bill and battery life is related to energy!
- Lower power does not necessary means better battery life if the processor slow down the application too much

Double Clock Rate or Double the Processors?

 Assume 60% of the application can be fully parallelized with 2-core or speedup linearly with clock rate. Should we double the clock rate or duplicate a core?

Speedup_{2-core} =
$$\frac{1}{(1-0.6)+\frac{0.6}{2}}$$
 = 1.43

Power_{2-core} = 2x

Energy_{2-core} =
$$2 * [1/(1.43)] = 1.39$$

$$Speedup_{2XClock} = 2$$

$$Power_{2XClock} = 8x$$

Energy₂
$$\times$$
Clock = 8 / 2 = 4

Other important metrics

Bandwidth

- The amount of work (or data) during a period of time
 - Network/Disks: MB/sec, GB/sec, Gbps, Mbps
 - Game/Video: Frames per second
- Also called "throughput"
- "Work done" / "execution time"

Response time and BW trade-off

- Increase bandwidth can hurt the response time of a single task
- If you want to transfer a 2 Peta-Byte video from UNC
 - 25 miles from NCSU
 - Assume that you have a 100Gbps ethernet
 - 2 Peta-byte over 167772 seconds = 1.94 Days
 - · 22.5TB in 30 minutes
 - Bandwidth: 100 Gbps

Or ...

	Toyota Prius	10Gb Ethernet	
	 25 miles from UNC 75 MPH on highway! 50 MPG Max load: 374 kg = 2,770 hard drives (2TB per drive) 		
bandwidth	1.53TB/sec	100 Gb/s or 12.5GB/sec	
latency	1 hour	2 Peta-byte over 167772 seconds = 1.94 Days	
response time	You see nothing in the first hour	You can start watching the movie as soon as you get a frame!	

Reliability

- Mean time to failure (MTTF)
 - Average time before a system stops working
 - Very complicated to calculate for complex systems
- Hardware can fail because of
 - Electromigration
 - Temperature
 - High-energy particle strikes

GFLOPS (Giga FLoating-point Operations Per Second)

- MIPS does not include instruction count!
 - Cannot compare different ISA/compiler
 - Different CPI of applications, for example, I/O bound or computation bound
 - If new architecture has more IC but also lower CPI?

	GFLOPS	clock rate
XBOX One	1310	1.75 GHz
PS4	1843	1.6 GHz
Core i7 EE 3970X + AMD Radeon 6990	5099	3.5 GHz

Is GFLOPS (Giga FLoating-point Operations Per Second) a good metric?

- Cannot compare different ISA/compiler
 - What if the compiler can generate code with fewer instructions?
 - What if new architecture has more IC but also lower CPI?
- Does not make sense if the application is not floating point intensive

$$\frac{\text{GFLOPS} = \frac{\text{\# of floating point instructions / 10}^9}{\text{Execution Time}} \\
= \frac{\text{ICX \% of floating point instructions}}{\text{ICX CPIXCycleTime } \times 10^9} = \frac{\text{Clock Rate X\% FP ins.}}{\text{CPI X 10}^9}$$