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Review & Background

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Measuring & Reporting Performance



Performance Metrics

- Latency (execution/response time): time to finish one task
- **Throughput** (bandwidth): number of tasks finished per unit of time
 - Throughput can exploit parallelism, latency can't
 - Sometimes complimentary, often contradictory
- Example: move people from A to B, 10 miles
 - Car: capacity = 5, speed = 60 miles/hour
 - Bus: capacity = 60, speed = 20 miles/hour
 - Latency: car = 10 min, bus = 30 min
 - Throughput: car = 15 PPH (w/ return trip), bus = 60 PPH

Pick the *right* metric for *your* goals



Performance Comparison

- "Processor A is X times faster than processor B" if
 - Latency(P, A) = Latency(P, B) / X
 - Throughput(P, A) = Throughput(P, B) * X
- "Processor A is X% faster than processor B" if
 - -Latency(P, A) = Latency(P, B) / (1+X/100)
 - Throughput(P, A) = Throughput(P, B) * (1+X/100)
- Car/bus example
 - Latency? Car is 3 times (200%) faster than bus
 - Throughput? Bus is 4 times (300%) faster than car



Latency/throughput of What Program?

- Very difficult question!
- Best case: you always run the same set of programs
 - Just measure the execution time of those programs
 - Too idealistic

• Use *benchmarks*

- *Representative* programs chosen to measure performance
- (Hopefully) predict performance of actual workload
- Prone to Benchmarketing:

"The misleading use of unrepresentative benchmark software results in marketing a computer system"

-- wikitionary.com



Types of Benchmarks

- Real programs
 - Example: CAD, text processing, business apps, scientific apps
 - Need to know program inputs and options (not just code)
 - May not know what programs users will run
 - Require a lot of effort to port
- Kernels
 - Small key pieces (inner loops) of scientific programs where program spends most of its time
 - Example: Livermore loops, LINPACK
- Toy Benchmarks
 - e.g. Quicksort, Puzzle
 - Easy to develop, predictable results, may use to check correctness of machine but not as performance benchmark



SPEC Benchmarks

• <u>System Performance Evaluation Corporation</u>

"non-profit corporation formed to establish, maintain and endorse a <u>standardized set of relevant benchmarks</u>..."

- Different set of benchmarks for different domains:
 - CPU performance (SPEC CINT and SPEC CFP)
 - High Performance Computing (SPEC MPI, SPEC OpenMP)
 - Java Client Server (SPECjAppServer, SPECjbb, SPECjEnterprise, SPECjvm)
 - Web Servers (SPECWeb)
 - Virtualization (SPECvirt)



Example: SPEC CINT2006

Program	Language	Description
400.perlbench	С	Programming Language
401.bzip2	С	Compression
<u>403.gcc</u>	С	C Compiler
<u>429.mcf</u>	С	Combinatorial Optimization
445.gobmk	С	Artificial Intelligence: Go
<u>456.hmmer</u>	С	Search Gene Sequence
458.sjeng	С	Artificial Intelligence: chess
462.libquantum	С	Physics / Quantum Computing
464.h264ref	С	Video Compression
471.omnetpp	C++	Discrete Event Simulation
473.astar	C++	Path-finding Algorithms
483.xalancbmk	C++	XML Processing



Example: SPEC CFP2006

Program	Language	Description
410.bwaves	Fortran	Fluid Dynamics
416.gamess	Fortran	Quantum Chemistry.
<u>433.milc</u>	С	Physics / Quantum Chromodynamics
434.zeusmp	Fortran	Physics / CFD
435.gromacs	C, Fortran	Biochemistry / Molecular Dynamics
436.cactusADM	C, Fortran	Physics / General Relativity
<u>437.leslie3d</u>	Fortran	Fluid Dynamics
<u>444.namd</u>	C++	Biology / Molecular Dynamics
447.dealII	C++	Finite Element Analysis
450.soplex	C++	Linear Programming, Optimization
453.povray	C++	Image Ray-tracing
454.calculix	C, Fortran	Structural Mechanics
459.GemsFDTD	Fortran	Computational Electromagnetics
<u>465.tonto</u>	Fortran	Quantum Chemistry
<u>470.lbm</u>	С	Fluid Dynamics
<u>481.wrf</u>	C, Fortran	Weather
482.sphinx3	С	Speech recognition



Benchmark Pitfalls

- Benchmark not representative
 - Your workload is I/O bound \rightarrow SPECint is useless
 - Benchmarketing pressure causes vendors to optimize compiler/hardware/software to benchmarks
- Benchmark too old
 - Benchmarks age poorly
 - \rightarrow Need to be periodically refreshed



Summarizing Performance Numbers (1)

- Latency is additive, throughput is not
 - Latency(P1+P2, A) = Latency(P1, A) + Latency(P2, A)
 - Throughput(P1+P2, A) !=

Throughput(P1, A) + Throughput(P2,A)

- Example:
 - 180 miles @ 30 miles/hour + 180 miles @ 90 miles/hour
 - 6 hours at 30 miles/hour + 2 hours at 90 miles/hour
 - Total latency is 6 + 2 = 8 hours
 - Total throughput is *not 60* miles/hour
 - Total throughput is *only 45* miles/hour! (360 miles / (6 + 2 hours))

Arithmetic Mean is Not Always the Answer!

Summarizing Performance Numbers (2)

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• Any of these can be *weighted*

Memorize these to avoid looking them up later

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Improving Performance



Principles of Computer Design

- Take Advantage of Parallelism
 - E.g., multiple processors, disks, memory banks, pipelining, multiple functional units
 - Speculate to create (even more) parallelism
- Principle of Locality
 - Reuse of data and instructions
- Focus on the Common Case
 - Amdahl's Law



- *Parallelism*: number of independent tasks available
- <u>Work</u> (T₁): time on sequential system
- <u>Critical Path</u> (T_{∞}): time on infinitely-parallel system
- <u>Average Parallelism</u>: $P_{avg} = T_1 / T_{\infty}$ • For a p-wide system: $T_p \ge max\{T_1/p, T_{\infty}\}$ $P_{avg} >> p \implies T_p \approx T_1/p$

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Principle of Locality

• Recent past is a good indication of near future

<u>Temporal Locality</u>: If you looked something up, it is very likely that you will look it up again soon

<u>Spatial Locality</u>: If you looked something up, it is very likely you will look up something nearby soon



Amdahl's Law

Speedup = time_{without enhancement} / time_{with enhancement}

An enhancement speeds up fraction f of a task by factor S

$$time_{new} = time_{orig} \cdot ((1-f) + f/S)$$
$$S_{overall} = 1 / ((1-f) + f/S)$$



Make the common case fast!



The *Iron Law* of Processor Performance (1)



Architects target CPI, but *must* understand the others



The *Iron Law* of Processor Performance (2)

- The three components of Iron Law are interdependent
 - Because of the factors they depend upon
 - Trying to change one will imply changes in the others
- Processor architects mostly target CPI but *must* understand the others extremely well
 - Architects are the interface between software people (compiler, OS, etc.) and those who build the physical hardware



Another View of CPU Performance (1)

• Instruction frequencies for a given program on a given machine

Instruction Type	Frequency	Avg. CPI
Load	25%	2
Store	15%	2
Branch	20%	2
ALU	40%	1

• What is the average *CPI* (cycles per instruction)?

Average CPI =
$$\frac{\sum_{i=1}^{n} InstFrequency_i \times CPI_i}{\sum_{i=1}^{n} InstFrequency_i}$$
$$= \frac{0.25 \times 2 + 0.15 \times 2 + 0.2 \times 2 + 0.4 \times 1}{1} = 1.6$$



Another View of CPU Performance (2)

- Assume all conditional branches in previous machine use simple tests of equality with zero (BEQZ, BNEZ)
- Consider adding complex comparisons to conditional branches
 - 25% of branches can use complex scheme \rightarrow no need for preceding ALU instruction
- Because of added complexity, CPU cycle time of original machine is 10% faster
- Will this change increase CPU performance?

New CPU CPI = $\frac{0.25 \times 2 + 0.15 \times 2 + 0.2 \times 2 + (0.4 - 0.25 \times 0.2) \times 1}{1 - 0.25 \times 0.2} = 1.63$ Hmm... Both slower clock and increased CPI? Something smells fishy !!!



Another View of CPU Performance (3)

- Recall the Iron Law
- The two programs have different number of instructions

Old CPU Time = $InstCount_{old} \times CPI_{old} \times cycle_time_{old} = N \times 1.6 \times ct$

New CPU Time =

 $InstCount_{new} \times CPI_{new} \times cycle_time_{new} = (1 - 0.25 \times 0.2)N \times 1.63 \times 1.1ct$

Speedup = $\frac{1.6}{(1-0.25 \times 0.2) \times 1.63 \times 1.1} = 0.94$

The new CPU is *slower* for this instruction mix



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- Which processor would you buy?
 - Processor A: CPI = 2, clock = 2.8 GHz
 - Processor B: CPI = 1, clock = 1.8 GHz
 - Probably A, but B is faster (assuming same ISA/compiler)
- Classic example
 - 800 MHz Pentium III faster than 1 GHz Pentium 4
 - Same ISA and compiler
- Some Famous Partial Performance Metrics
 - MIPS: Million Instruction Per Second
 - MFLOPS: Million Floating-Point Operations Per Second



A simplified review of **Trends in Computing Technology**



Early days: 60s &70s

- Focus on instruction set designs
- A lot of programming done in assembly, and memory was scarce
 - CISC instruction sets popular
- **CISC**: Complex Instruction Set Computing)
 - Improve "instructions/program" with "complex" instructions
 - Easy for assembly-level programmers, good code density
- Example: x86 (Intel and AMD processors)



Age of RISC: 80s & 90s (1)

A combination of multiple effects led to advent of fast processors:

- 1) Compilers became powerful and popular in late 70s
 - Compilers are not good at using complex instructions effectively; they would mostly use a simple subset of instructions in a CISC ISA
- 2) It is not easy to build optimized, high-performance pipelines for a CISC ISA
- \rightarrow Simple RISC instruction sets became popular



Interlude: RISC vs. CISC

- **RISC**: Reduced Instruction Set Computing)
 - Improve "cycles/instruction" with many single-cycle instructions
 - Increases "instruction/program", but hopefully not as much
 - Help from smart compiler
 - Perhaps improve clock cycle time (seconds/cycle)
 - via aggressive implementation allowed by simpler instructions
- Example: MIPS, SPARC, ARM, ...
- Modern x86 processors translate CISC code to RISC internally
 - Called " μ -ops" by Intel and "ROPs" (RISC-ops) by AMD
 - And then execute the RISC code



Age of RISC: 80s & 90s (2)

- 3) Exponential growth in transistor count and speed
 - Thanks to trends called "Moore's Law" and "Dennard Scaling"
 - Moore's law: more transistors
 - **Dennard Scaling**: smaller, faster, more power-efficient transistors
- → More transistors + simple RISC ISA led to many architectural innovations
 - Super-scalar (wide) pipelines: ability to execute multiple instructions in parallel
 - **Out-of-order executions**: better utilization of wide pipelines
 - Branch prediction and speculation: to find even more parallel work to do
 - multi-level on-chip caches: to hide memory latency
 - Super pipelines: deep pipelines to allow faster clock speed

Faster transistor + better architecture \rightarrow 50% per year perf. improvement

Interlude: Moore's Law

• 1st Moore's Law (1965)

"The complexity for minimum component costs has increased at a rate of roughly a factor of two per year. Certainly over the short term this rate can be expected to continue, if not to increase."

• 2nd Moore's Law (1975)

"The new slope might approximate a doubling every two years, rather than every year"

- Nowadays, Moore's law is a general term for any exponential change in technology (with different slopes)
 - E.g., transistor size, transistor speed, processor performance, etc.





Age of RISC: 80s & 90s (3)

- ILP wall hit in 90s
- Even with very wide pipelines, it is very hard to find many independent instructions to execute in parallel
 - No point in building very wide pipelines
- Impacts on processor design:
 - Very large on-chip caches: spend transistors on cache instead of pipeline
 - Hardware multi-threading: let multiple threads share the pipeline
 - Multi-core processors: instead of building one beefy, ultra-wide processor, build multiple less-wide processor cores on the same chip



Age of Many cores: 2000s

- Power wall was hit around 2004
- Dennard scaling was no longer true
 - Transistors were getting smaller but not much faster or more power efficient
- We could not increase transistor count and push frequency higher at the same time
- Many cores: push for even more cores
 - Pipelines became simpler freeing up transistors
 - More transistors used for more cores
- GPU: extreme example of many-core processor



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- Moore's law is no more
 - Transistor density increasing much more slowly today
 - And soon will die out without new technological breakthroughs
- So, how to get more performance with the same transistor count, power budget and frequency?
- Answer: specialized hardware
 - Hardware that is good for one for a few tasks but much less powerhungry and less complicated than general purpose processors
 - Examples: Google TPU (for DNNs), Microsoft Catapult (for Bing search), Pixel Visual Core (for mobile image processing), Intel Crest (for DNN training), ...



Power Basics



Power vs. Energy (1)

- Energy: capacity to do work or amount of work done
 - Expressed in joules
 - Energy(OP1+OP2)=Energy(OP1)+Energy(OP2)
- **Power:** instantaneous rate of energy transfer
 - Expressed in watts
 - energy / time (watts = joules / seconds)
 - Power(Comp1+Comp2)=Power(Comp1)+Power(Comp2)
- In processors, all consumed energy is converted to heat
 → power consumption = <u>rate of heat generation</u>



Power vs. Energy (2)





Why is Energy Important?

- Impacts battery life for mobile devices
- Impacts electricity costs for tethered (plugged) machines
 - You have to buy electricity
 - It costs to produce and deliver electricity
 - You have to remove generated heat
 - It costs to buy and operate cooling systems
- Gets worse as data centers grow
 - \$7M for 1000 server racks
 - 2% of US electricity used by DCs in 2010 (Koomey'11)





Why is Power Important?

- Because power delivery has a peak
- Power is also heat generation rate
 - Must dissipate the heat
 - Need heat sinks and fans and ...
- What if fans not fast enough?
 - Chip powers off (if it's smart enough)
 - Otherwise, it burns (or melts)
- Thermal failures even when fans OK
 - 50% server reliability degradation for +10°C
 - 50% decrease in hard disk lifetime for +15°C





Power: The Basics (1)

• Dynamic Power

- Related to switching activity of transistors (from $0 \rightarrow 1$ and $1 \rightarrow 0$)



- Dynamic Power $\propto C V_{dd}^2 A f$
 - C: capacitance, function of transistor size and wire length
 - V_{dd}: Supply voltage
 - A: Activity factor (average fraction of transistors switching)
 - f: clock frequency
 - About 50-70% of processor power



Power: The Basics (2)

• Static Power

 Current leaking from a transistor even if doing nothing (steady, constant energy cost)



- Static Power $\propto V_{dd}$ and $\propto e^{-c_1 V_{th}}$ and $\propto e^{c_2 T}$
 - This is a first-order model
 - $-c_1, c_2$: some positive constants
 - *V*_{th}: Threshold Voltage
 - T: Temperature
 - About 30-50% of processor power



Thermal Runaway

- Leakage is an exponential function of temperature
- Temp leads to A Leakage
- Which burns more power
- Which leads to **↑** Temp, which leads to...

Positive feedback loop will melt your chip

Why Power Became an Issue (1)

- Good old days of ideal scaling (a.k.a. Dennard scaling)
 - Every new semiconductor generation:
 - Transistor dimension: x 0.7
 - Transistor area: x 0.49
 - C and V_{dd}: x 0.7
 - Frequency: 1 / 0.7 = 1.4
 - →Constant dynamic power density

- In those good old days, leakage was not a big deal

→ Faster and more transistors with constant power density ⁽³⁾

Dynamic Power $\propto CV_{dd}^2 A f$

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Why Power Became an Issue (2)

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- Recent reality: V_{dd} does not decrease much
 - Switching speed is roughly proportional to V_{dd} V_{th}
 - If too close to threshold voltage (V_{th}) \rightarrow slow transistor
 - Fast transistor & low $V_{dd} \rightarrow low V_{th} \rightarrow exponential increase in leakage <math>\times$
 - →Dynamic power density keeps increasing
 - Leakage power has also become a big deal today
 - Due to lower V_{th}, smaller transistors, higher temperatures, etc.

\rightarrow We hit the **power wall** \otimes

- Example: power consumption in Intel processors
 - Intel 80386 consumed 2 W
 - 4 GHz Intel Core i7-6700K consumes 95 W
 - Heat must be dissipated from 1.5 x 1.5 cm² chip
 - This is the limit of what can be cooled by air

How to Reduce Processor Power (1)

- **Clock gating**: Stop switching in unused components
 - reduces dynamic power
 - Done automatically in most designs
 - Near instantaneous on/off behavior
- Power gating: Turn off power to unused cores/caches
 - reduces both static and dynamic power
 - High latency for on/off
 - Saving SW state, flushing dirty cache lines, turning off clock tree
 - Carefully done to avoid voltage spikes or memory bottlenecks
 - Opportunity: use thermal headroom for other cores









How to Reduce Processor Power (2)

- Reduce Voltage (V_{dd}): quadratic effect on dyn. power
 Negative (~linear) effect on frequency
- Dynamic Voltage/Frequency Scaling (DVFS): set frequency to the lowest needed

– Execution time = IC * CPI * f

- Scale back V_{dd} to lowest for that frequency
 - Lower voltage \rightarrow slower transistors
 - Dynamic Power proportional to C * V_{dd}^2 * F

Not Enough! Need Much More!



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- Design for E & P efficiency rather than speed
- New architectural designs:
 - Simplify the processor, shallow pipeline, less speculation
 - Efficient support for high concurrency (think GPUs)
 - Augment processing nodes with accelerators
 - New memory architectures and layouts
 - Data transfer minimization
 - ...
- New technologies:
 - Low supply voltage (V_{dd}) operation: Near-Threshold Voltage Computing
 - Non-volatile memory (Resistive memory, STT-MRAM, ...)
 - 3D die stacking
 - Efficient on-chip voltage conversion
 - Photonic interconnects
 - ...



Voltage/Frequency Scaling Example

- Example: say you reduce processor frequency by 20%, allowing 20% reduction in $\rm V_{\rm dd}$
- What is the resulting power impact (considering only dynamic power)?
 0.8 × 0.8 × 0.8 = 0.512 (half the power)
- What is the resulting energy impact (considering only dynamic power)?
 0.8 × 0.8 = 0.8 energy
- What if I turned on two processor cores with 0.8 freq. and voltage?
 Almost same power, with 1.6x performance <u>if job is parallelizable</u>
- → In many cases you can get better "performance per watt" (as well as "performance per joule") with more parallel systems



Processor Is Not Alone



No single component dominates power consumption Need whole-system approaches to save energy Spring 2018 :: CSE 502



Instruction Set Architecture (ISA)



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- **ISA:** Instruction Set Architecture
 - A well-defined hardware/software interface
 - Old days: target language for human programmers
 - More recently: target language for compilers
- The "contract" between software and hardware
 - Functional definition of operations supported by hardware
 - Precise description of how to invoke all features
- No guarantees regarding
 - How operations are implemented
 - Which operations are fast and which are slow (and when)
 - Which operations take more energy (and which take less)



Components of an ISA (1)

- 1) Programmer-visible machine states
 - Program counter, general purpose registers, control registers, etc.
 - Memory
 - Page table, interrupt descriptor table, etc.
- 2) Programmer-visible operations
 - Operations: ALU ops, floating-point ops, control-flow ops, string ops, etc.
 - Type and size of operands for each op: byte, half-word, word, double word, single precision, double precision, etc.
- 3) Addressing modes for each operand of an instruction
 - Immediate mode (for immediate operands)
 - Register addressing modes: stack-based, accumulator-based, generalpurpose registers, etc.
 - Memory addressing modes: displacement, register indirect, indexed, direct, memory-indirect, auto-increment(decrement), scaled, etc.

ISAs last forever, don't add stuff you don't need



Components of an ISA (2)

- 4) Programmer-visible behaviors
 - What to do, when to do it

if imem[rip]=="add rd, rs, rt" then rip ⇐ rip+1 gpr[rd]=gpr[rs]+gpr[rt]

Example "register-transferlevel" description of an instruction

5) A binary encoding

ISAs last forever, don't add stuff you don't need



RISC vs. CISC

- Recall Iron Law:
 - (instructions/program) * (cycles/instruction) * (seconds/cycle)
- <u>CISC</u> (Complex Instruction Set Computing)
 - Improve "instructions/program" with "complex" instructions
 - Easy for assembly-level programmers, good code density
- <u>*RISC*</u> (Reduced Instruction Set Computing)
 - Improve "cycles/instruction" with many single-cycle instructions
 - Increases "instruction/program", but hopefully not as much
 - Help from smart compiler
 - Perhaps improve clock cycle time (seconds/cycle)
 - via aggressive implementation allowed by simpler instructions

Today's x86 chips translate CISC into ~RISC



RISC ISA

- Focus on simple instructions
 - Easy to use for compilers
 - Simple (basic) operations, many registers
 - Easy to design high-performance implementations
 - Easy to fetch and decode, simpler pipeline control, faster caches
- Fixed-length
 - MIPS and SPARCv8 all insts are 32 bits (4 bytes)
 - Especially useful when decoding multiple instructions simultaneously
- Few instruction formats
 - MIPS has 3: R (reg, reg, reg), I (reg, reg, imm), J (addr)
 - Alpha has 5: Operate, Op w/ Imm, Mem, Branch, FP
- Regularity across formats (when possible/practical)
 - MIPS & Alpha opcode in same bit-position for all formats
 - MIPS rs & rt fields in same bit-position for R and I formats
 - Alpha ra/fa field in same bit-position for all 5 formats



CISC ISA

- Focus on max expressiveness per min space
 - Designed in era with fewer transistors
 - Each memory access very expensive
 - Pack as much work into as few bytes as possible
- Difficult to use for compilers
 - Complex instructions are not compiler friendly \rightarrow many instructions remain unused
 - Fewer registers: register IDs take space in instructions
 - For fun: compare x86 vs. MIPS backend in LLVM
- Difficult to build high-performance processor pipelines
 - Difficult to decode: Variable length (1-18 bytes in x86), many formats
 - Complex pipeline control logic
 - Deeper pipelines
- Modern x86 processors translate CISC code to RISC first
 - Called "μ-ops" by Intel and "ROPs" (RISC-ops) by AMD
 - And then execute the RISC code