CS425 Computer Systems Architecture

Fall 2017

Metrics

Previous Lecture

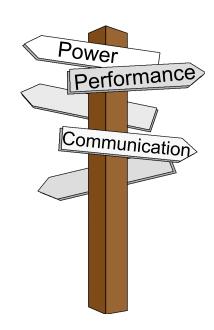
- CPU Evolution
- What is Computer Architecture?

Outline

- Measurements and metrics:
 - Performance, Cost, Dependability, Power
- Guidelines and principles in the design of computers
- CPU Performance

Major Design Challenges

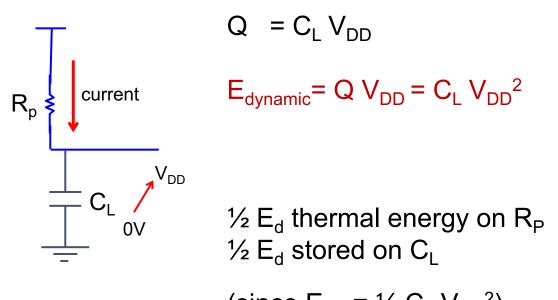
- Power
- CPU time
- Memory latency/bandwidth
- Storage latency/bandwidth
- Transactions per second
- Intercommunication
- Dependability



Everything Looks a Little Different

Power Consumption

Charge external capacitance

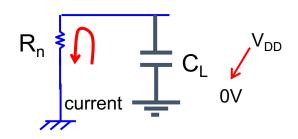


$$Q = C_L V_{DD}$$

$$E_{dynamic} = Q V_{DD} = C_L V_{DD}^2$$

(since $E_{CI} = \frac{1}{2} C_I V_{DD}^2$)

Discharge external capacitance



½ E_{dynamic} stored on C_L becomes thermal energy on R_N

$$P_{dynamic} = \frac{1}{2} C_L V_{DD}^2$$
 frequency

Power Equations

 $Power_{dynamic} = \frac{1}{2} \times Capacitive load \times Voltage^2 \times Frequency$

Energy_{dynamic} = Capacitive load x Voltage²

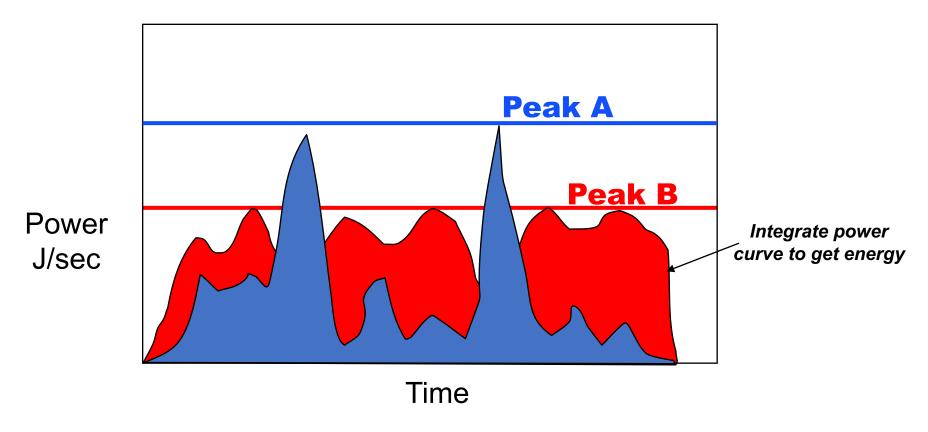
 $Power_{static} = Current_{static} \times Voltage$

- Power due to switching more transistors increases
- Static power due to leakage current increasing

Power and Energy

- Energy to complete operation (Joules)
 - Corresponds approximately to battery life
 - (Battery energy capacity actually depends on rate of discharge)
- Peak power dissipation (Watts = Joules/second)
 - Affects packaging (power and ground pins, thermal design)
- d_i/d_t, peak change in supply current (Amps/second)
 - Affects power supply noise (power and ground pins, decoupling capacitors)

Peak Power versus Lower Energy



- System A has higher peak power, but lower total energy
- System B has lower peak power, but higher total energy

Measuring Reliability (Dependability)

Reliability equations

$$MTTF = \textit{Mean Time To Failure}$$

$$FIT = \textit{Failures In Time (per billion hours)} = \frac{10^9}{\textit{MTTF}}$$

$$MTTR = \textit{Mean Time to Repair (MTBF = MTTF + MTTR)}$$

$$\textit{Module availability} = \frac{\textit{MTTF}}{\textit{MTTF} + \textit{MTTR}}$$

$$\textit{FIT}_{\textit{system}} = \sum_{i=1}^{\#\textit{components}} \textit{FIT}_i$$

MTTF = 1,000,000 hours → FIT = ?

Comparing Design Alternatives

Design X is n times faster than design Y

$$n = \frac{\text{Execution time}_{Y}}{\text{Execution time}_{X}} = \frac{\frac{1}{\text{Performance}_{Y}}}{\frac{1}{\text{Performance}_{X}}} = \frac{\text{Performance}_{X}}{\text{Performance}_{Y}}$$

- Wall-clock time: time to complete a task
- CPU time: time CPU is busy
- Workload: Mixture of programs (including OS) on a system
- Kernels: Common, important functions in applications
- Microbenchmarks: Synthetic programs trying to:
 - Isolate components and measure performance
 - Imitate workloads of real world in a controlled setting

Benchmark Suites

Desktop (SPEC = Standard Performance Evaluation Corporation, 12 INT, 17 FP, 1980)

- SPECCPU (revised every few years)
- Real programs measuring processor-memory activity

Multi-core desktop/server

- SPECOMP, SPECMPI (scientific), SPECapc (graphics)
- Focus on parallelism, synchronization, communication

Client/Server

- SPECjbb, SPECjms, SPECjvm, SPECsfs, SPECmail SPECrate, SPECWeb ...
- Measuring throughput (how many tasks per unit of time)
- Measuring latency (how quickly does client get response)

Embedded systems

- ► EEMBC, MiBench
- Measuring performance, throughput, latency

The weakness of one benchmark is covered by the other benchmarks

Summarizing performance

Arithmetic mean of wall-clock time

- Biased by long-running programs
- May rank designs in non-intuitive ways:
 - ▶ Machine A: Program $P_1 \rightarrow 1000$ secs., $P_2 \rightarrow 1$ secs.
 - ▶ Machine B: Program $P_1 \rightarrow 800$ secs., $P_2 \rightarrow 100$ secs.
 - ▶ What if machine runs P_2 most of the time?

Means

- Total time ignores program contribution to total workload
- Arithmetic mean biased by long programs
- Weighted arithmetic mean a better choice?
- How do we calculate weights?

Weighted arithmetic mean

$$\sum_{i=1}^{n} Weight_i \times Time_i$$

Example, W(1) = W(2) = 50

	Computer A	Computer B	Computer C
Program P1 (secs)	1	10	20
Program P2 (secs)	1000	100	20
Total time (secs)	1001	110	40
Weighted mean	500.50	55.00	20.00

Weighted arithmetic mean

$$\sum_{i=1}^{n} Weight_i \times Time_i$$

Example, W(1) = 0.909 W(2) = 0.091

	Computer A	Computer B	Computer C
Program P1 (secs)	1	10	20
Program P2 (secs)	1000	100	20
Total time (secs)	1001	110	40
Weighted mean	91.91	18.19	20.00

Weighted arithmetic mean

$$\sum_{i=1}^{n} Weight_i \times Time_i$$

Example, W(1) = 0.999 W(2) = 0.001

	Computer A	Computer B	Computer C
Program P1 (secs)	1	10	20
Program P2 (secs)	1000	100	20
Total time (secs)	1001	110	40
Weighted mean	2.00	10.09	20.00

Measuring against a reference computer

$$SPEC_{ratio_A} = \frac{Execution\ time_{reference}}{Execution\ time_A} = Performance_A / Performance_{reference}$$

$$n = \frac{SPEC_{ratio_A}}{SPEC_{ratio_B}} = \frac{\frac{Execution\ time_{reference}}{Execution\ time_A}}{\frac{Execution\ time_{reference}}{Execution\ time_B}} = \frac{Execution\ time_B}{Execution\ time_A} = \frac{Performance_A}{Performance_B}$$

Using ratios

 Ratios against reference machine are independent of running time of programs

Geometric mean

$$\int_{i=1}^{n} SPEC_{ratio}(i)$$

$$\frac{Geometric\ mean_A}{Geometric\ mean_B} = Geometric\ mean(\frac{A}{B})$$

Used by SPEC98, SPEC92, SPEC95, ..., SPEC2006

Pros and cons of geometric means

Pros

- Consistent rankings, independent of program frequencies
- Not influenced by peculiarities of any single machine

Cons

- Geometric mean does not predict execution time
 - Sensitivity to benchmark vs. machine remains
 - Encourages machine tuning for specific benchmarks
 - Benchmarks can not be touched, but compilers can!
- Any "averaging" metric loses information

Qualitative principles of design

Taking advantage of parallelism

- Use pipelining to overlap instructions
- Use multiple execution units
- Use multiple cores
- Use multiple processors to increase throughput (system level: scalability)

Locality (spatial and temporal locality)

- Programs reuse instructions and data
- 90-10 rule
 - 90% of execution time spent running 10% of instructions
- Programs access data in nearby addresses (spatial)

Qualitative principles of design (cont.)

Make the common case fast

- Trade-off's in design (e.g. performance vs. power/area)
- Provide efficient design for the common case
- Amdahl's Law

Example:

First optimize instruction fetch and decode unit instead of multiplier

Amdahl's Law

 $Speedup = \frac{Execution time for entire task without using the enhancement}{Execution time for entire task using the enhancement when possible}$

$$execution time_{new} = execution time_{old} \times \\ \left((1 - fraction_{enhanced}) + \frac{fraction_{enhanced}}{speedup_{enhanced}} \right)$$

$$speedup_{overall} = rac{execution \ time_{old}}{execution \ time_{new}} = rac{1}{(1 - fraction_{enhanced}) + rac{fraction_{enhanced}}{speedup_{enhanced}}} \Rightarrow rac{1}{1 - fraction_{enhanced}}$$
Upper Limit: $speedup_{overall} o rac{1}{1 - fraction_{enhanced}}$

$$\begin{array}{c|c}
 & T_{\text{old}} \\
\hline
 & (1 - f) \\
\hline
\end{array} \longrightarrow \begin{array}{c}
 & T_{\text{new}} \\
\hline
 & (1 - f) \\
\hline
\end{array}$$

Amdahl's Law example

- New CPU 10X faster
- I/O bound server, so 60% time waiting for I/O

Speedup_{overall} =
$$\frac{1}{(1 - \text{Fraction}_{\text{enhanced}}) + \frac{\text{Fraction}_{\text{enhanced}}}{\text{Speedup}_{\text{enhanced}}}}$$
$$= \frac{1}{(1 - 0.4) + \frac{0.4}{10}} = \frac{1}{0.64} = 1.56$$

 Apparently, its human nature to be attracted by 10X faster, vs. keeping in perspective its just 1.6X faster

Processor Performance

CPU time

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CPU \ time = CPU \ clock \ cycles \times Clock \ cycles \ time
CPI = \frac{CPU \ clock \ cycles}{instruction \ count} \Rightarrow
CPU \ time = instruction \ count \times CPI \times cycle \ time \Rightarrow
CPU \ time = \frac{instructions}{program} \times \frac{clock \ cycles}{instructions} \times \frac{seconds}{clock \ cycles}
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Cycles Per Instruction (CPI)

"Average Cycles per Instruction"

CPU time = Cycle Time
$$\times \sum_{j=1}^{n} CPI_{j} \times IC_{j}$$

$$CPI = \sum_{j=1}^{n} CPI_{j} \times F_{j} \quad \text{where } F_{j} = \frac{IC_{j}}{Instruction Count}$$

"Instruction Frequency"

Example: Calculating CPI bottom up

Run benchmark and collect workload characterization (simulate, machine counters, or sampling)

Base Machine (Reg / Reg)

Ор	Freq	CPI_i	F*CPI _i	(% Time)
ALU	50%	1	0.5	(33%)
Load	20%	2	0.4	(27%)
Store	10%	2	0.2	(13%)
Branch	20%	2	0.4	(27%)
			1.5	
	Typical Mix			
instruction types				
	in program			

Design guideline: Make the common case fast

<u>MIPS 1% rule:</u> only consider adding an instruction if it is shown to add 1% performance improvement on reasonable benchmarks.

Processor Performance

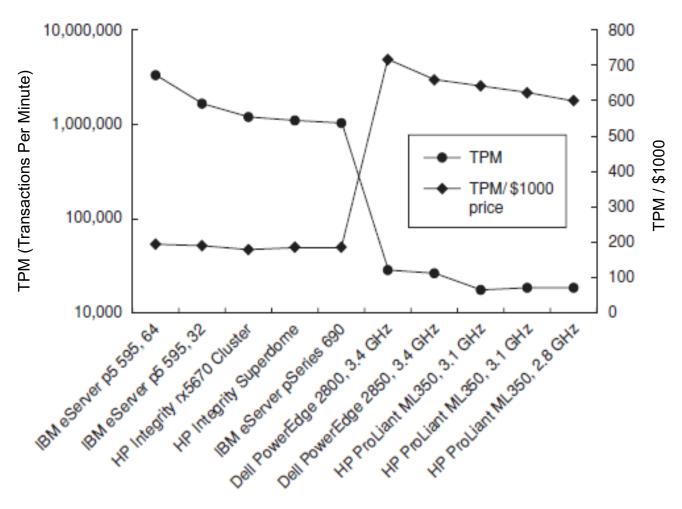
CPU time = instruction count \times *CPI* \times cycle time

How can CA help?

- Technology has been providing faster clock speeds
 - Main performance factor for almost 20 years
 - Trend seems to reverse
 - Limitations due to power consumption, reliability
- Architecture can pack more computing power in same area
- Architecture can improve CPI
- Algorithms and compilers can reduce instruction count

Price/performance

benchmark for online transaction processing (OLTP) is TPC-C



Next Lecture : Pipelining

