TCSS 462/562: (SOFTWARE ENGINEERING FOR) CLOUD COMPUTING

Introduction

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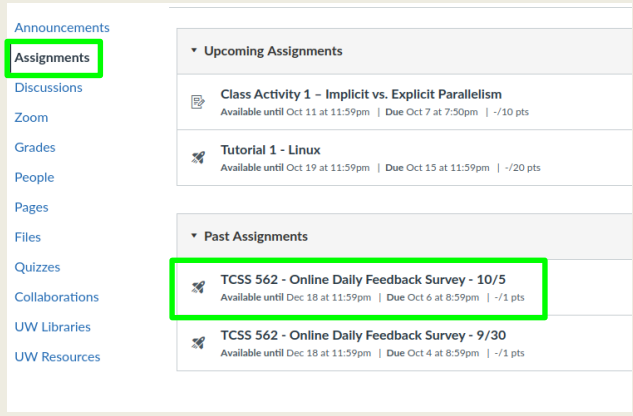
OBJECTIVES – 10/2

- Daily Feedback Surveys
- Questions from Course Introduction
- Demographics Survey
- AWS Cloud Credits Survey
- Tutorial 0 - Getting Started with AWS
- Tutorial 1 – Intro to Linux
- Cloud Computing – How did we get here? (10/4)
  Chapter 4 Marinescu 2nd edition:
  Introduction to parallel and distributed systems

2

DAILY FEEDBACK SURVEY

- Online Daily Feedback Quiz in Canvas – Take After Each Class
- **1-point Extra Credit** for completing online
- **2-points Extra Credit** for completing in-person in class
- 36 points possible
- 2.5% added to final course grade for (36/36)



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WARNING

- DO NOT SUBMIT BOTH A PAPER AND AN ONLINE SURVEY OR YOU WILL LOOSE POINTS
- CANVAS WILL AUTOMATICALLY REPLACE THE PAPER SURVEY SCORE (2 PTS) WITH THE ONLINE SURVEY (1 PT)
- **<u>* COMPLETE ONLY ONE SURVEY FOR EACH CLASS SESSION *</u>**
- WE WILL NOT BE ABLE TO DUPLICATE CHECK SURVEYS FOR EACH CLASS SESSION AND MAKE CORRECTIONS

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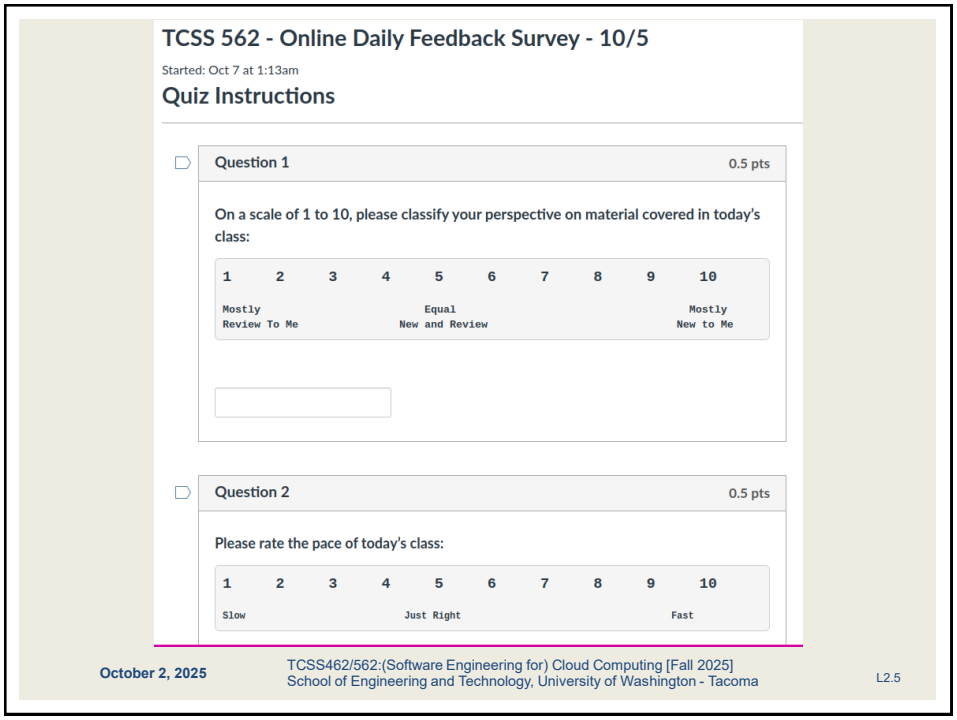
TCSS 562 - Online Daily Feedback Survey - 10/5
Started: Oct 7 at 1:13am
Quiz Instructions
Question 1
0.5 pts
On a scale of 1 to 10, please classify your perspective on material covered in today's class:
1 2 3 4 5 6 7 8 9 10
Mostly Review To Me
Equal New and Review
Mostly New to Me
Question 2
0.5 pts
Please rate the pace of today's class:
1 2 3 4 5 6 7 8 9 10
Slow
Just Right
Fast
October 2, 2025
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L2.5

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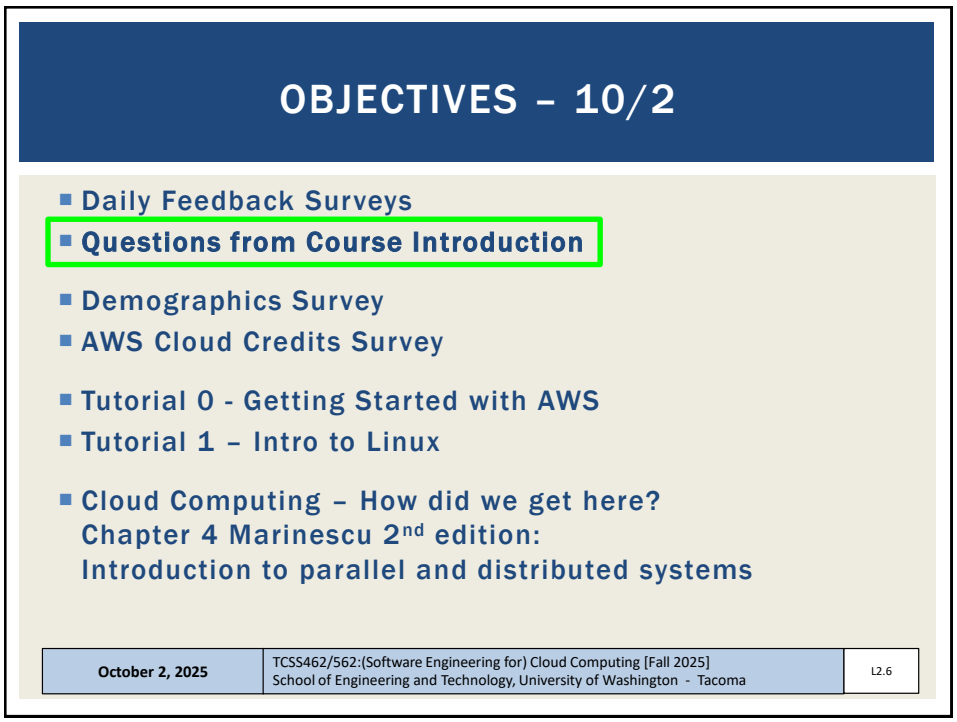
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L2.6

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MATERIAL / PACE

- Please classify your perspective on material covered in today's class (49 respondents, 38 in-person, 11 online):
- 1-mostly review, 5-equal new/review, 10-mostly new
- **Average – 7.20** (↑ - *first day f2024 - 6.16*)

- Please rate the pace of today's class:
- 1-slow, 5-just right, 10-fast
- **Average – 5.16** (↓ - *first day f2024 - 5.55*)

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FEEDBACK FROM 9/30

- ***What is the difference between the TCSS 462 and TCSS 562 term project ?***
  - TCSS 562 teams and hybrid-teams (462+562) submit a 4 to 6 page term paper
  - TCSS 462-only teams submit a 10-15 minute presentation recording with the option of submitting a term paper instead
  - The content of the paper and presentation are the same
  - A template is provided for both the term paper and presentation
  - The sections within are the same
- ***Is TCSS 562 credit transferrable ?***
  - Yes – if taking TCSS 562 as an undergraduate senior, the credit will transfer to the UWT MS CSS program. There is a good chance the credit can transfer other MS Computer Science programs besides UW. This can save ~$8,000+.

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FEEDBACK - 2

- ***How can the term project contribute to a thesis or capstone project ?***
- If pursuing a thesis/capstone related to cloud computing or distributed systems, then the skills learned in the course will be important.
- If having an idea for a capstone/thesis project, it is possible to explore the idea by proposing and conducting preliminary research in the term project.
- Performance investigation projects may be extendable into a capstone project
  - Thesis is more difficult as there needs to be an original (novel) research contribution

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FEEDBACK - 3

- ***I am unclear regarding the workload in the class***
  - Expect 1 tutorial per week throughout the quarter, 2 quizzes, plus the term project. The term project become a heavy focus after ~ week 6.
  - In general, this course is easier for students who have strong computing skills including the ability to learn how to use new services, troubleshoot issues/challenges, and solve problems.
  - 462/562 is not a programming course, but a systems course.
- ***Is the project 100% from the slides, or will I need to do additional research?***
  - The term project involves applying and synthesizing what is learned across the entire course into a final project

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FEEDBACK - 4

- ***What is the difference in quizzes for TCSS 462 vs. 562 ?***
  - Each section (462 or 562) is graded using a separate curve
  - The 462 curve is usually more generous
- ***Term Project ? Jargon ?***
- It is normal for the term project to be confusing on day 1, as we are just getting started

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DEMOGRAPHICS SURVEY

- Please complete the ONLINE demographics survey:
- https://forms.gle/QNUW2hUV7fR7BDmv7
- Linked from course webpage in Canvas:
- http://faculty.washington.edu/wlloyd/courses/tcss562/announcements.html
- Random drawing based on survey participants for two $20 Tango gift cards – October 9th in class
  - select from various options, i.e. Amazon, Starbucks, pizza, etc.

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OBJECTIVES – 10/2

- Daily Feedback Surveys
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- **AWS Cloud Credits Survey**
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AWS CLOUD CREDITS SURVEY

- Please complete the AWS CLOUD CREDITS survey as part of Tutorial 0:
- **https://forms.gle/Y4iWvBRFVLRPnPX37**
- Linked from course webpage in Canvas:
- http://faculty.washington.edu/wlloyd/courses/tcss562/announcements.html

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AWS CREDITS & BILLING

- CLOUD CREDITS are being dispersed on request
- AWS bills monthly, with charges applied to the credit card (or credit balance) on the last day of the month, for the month's charges
- END OF MONTH:– ****CHECK YOUR CLOUD BILL****
  at least a few days before the end of the month
- Billing Alarms – can be configured to generate email when there is a charge – can generate email if charges exceed $0.01
  - With cloud credits, there should be no monthly charges
  - **PROBLEM**: when accounts have a credit balance, they do not generate billing alarms for high service usage – it is necessary to manually inspect service usage

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OBJECTIVES – 10/2

- Questions from Course Introduction
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- Tutorial 0 - Getting Started with AWS
- **Tutorial 1 – Intro to Linux**
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OBJECTIVES – 10/2

- **Daily Feedback Surveys**
- Questions from Course Introduction
- Demographics Survey
- AWS Cloud Credits Survey
- Tutorial 0 - Getting Started with AWS
- Tutorial 1 – Intro to Linux
- **Cloud Computing – How did we get here?**
  Chapter 4 Marinescu 2nd edition:
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OBJECTIVES

- **Cloud Computing: How did we get here?**
  - *Parallel and distributed systems (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)*
  - Data, thread-level, task-level parallelism
  - Parallel architectures
  - SIMD architectures, vector processing, multimedia extensions
  - Graphics processing units
  - Speed-up, Amdahl's Law, Scaled Speedup
  - Properties of distributed systems
  - Modularity

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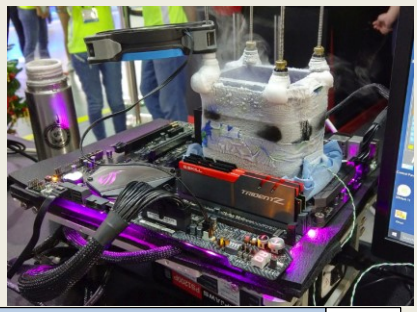
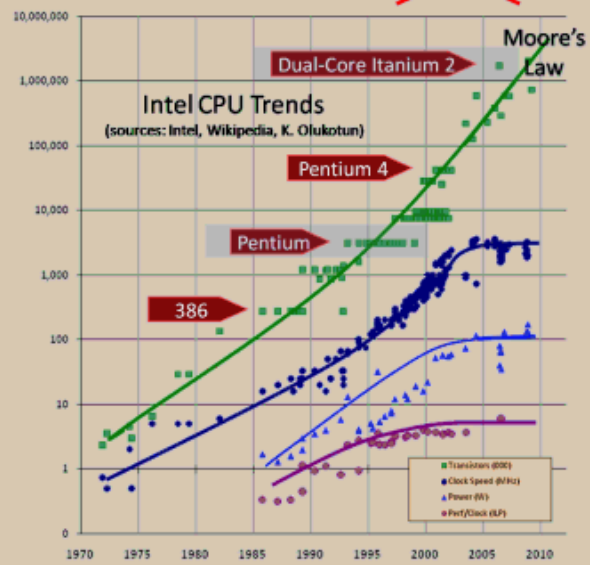
CLOUD COMPUTING: HOW DID WE GET HERE?

- General interest in parallel computing
  - Moore's Law - # of transistors doubles every 18 months
  - Post 2004: heat dissipation challenges: can no longer easily increase cloud speed
  - Overclocking to 7GHz takes more than just liquid nitrogen:
    - https://tinyurl.com/y93s2yz2
- Solutions:
  - Vary CPU clock speed
  - Add CPU cores
  - **Multi-core technology**



Each Year We Get ~~Faster~~ More Processors



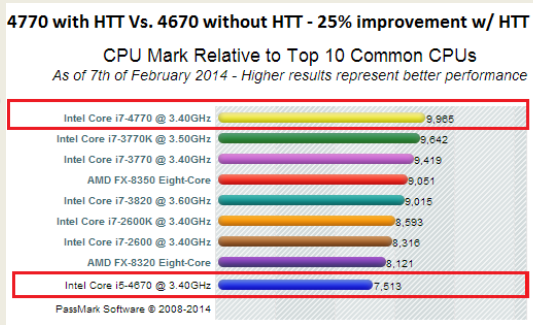
Historically: Boost single-stream performance via more complex chips.

Now: Deliver more cores per chip (+ GPU, NIC, SoC).

The free lunch is over for today's sequential apps and many concurrent apps. We need killer apps with lots of latent parallelism.

HYPER THREADING

- Modern CPUs provide multiple instruction pipelines, supporting multiple execution threads, usually 2 to feed instructions to a single CPU core…
- Two hyper-threads are not equivalent to (2) CPU cores
- i7-4770 and i5-4760 same CPU, with and without HTT
- Example: → hyperthreads add +32.9%



AMD'S 64-CORE 7NM CPUS

- Epyc Rome CPUs
- Announced August 2019
- EPYC 7H12 requires liquid cooling

AMD EPYC 7002 Processors (2P)

	Cores Threads	Frequency (GHz)		L3*	TDP	Price
		Base	Max			
EPYC 7H12	**64 / 128**	**2.60**	**3.30**	**256 MB**	**280 W**	**?**
EPYC 7742	64 / 128	2.25	3.40	256 MB	225 W	$6950
EPYC 7702	64 / 128	2.00	3.35	256 MB	200 W	$6450
EPYC 7642	48 / 96	2.30	3.20	256 MB	225 W	$4775
EPYC 7552	48 / 96	2.20	3.30	192 MB	200 W	$4025

AMD'S 64-CORE <14NM CPUS

AMD EPYC 9654/9654P/9684X/9R14 (AWS OEM):

- June 2023: ***96 cores***, 192 hyper-threads CPUs
- Mixes 4nm:APU (combines CPUs+GPU), 5nm:L3 cache (8 CPU-chiplet), and 6nm:I/O dies, 2.25 to 3.7 burst GHz, up to 400 watts
- $10,625 to $14,756

AMD EPYC 9754: *128 cores*, 256 hyperthreads !

- 2.25 to 3.1 burst GHz, 360 watts
- $11,900

AMD EPYC 9005: *192 cores*, 384 threads, 3nm (in dev)

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X86_64 HOST SERVER VCPUS – AMAZON EC2 INFRASTRUCTURE-AS-A-SERVICE CLOUD

- Cloud server virtual CPUs/host (x86_64)
- Growth since 2006 - Amazon Compute Cloud (EC2)

1st generation Intel: m1 – 8 vCPUs / host	(Aug 2006)
2nd generation Intel: m2 – 16 vCPUs / host	(Oct 2009)
3rd generation Intel: m3 - 32 vCPUs / host	(Oct 2012)
4th generation Intel: m4 – 48 vCPUs / host	(June 2015)
5th generation Intel: m5 – 96 vCPUs / host	(Nov 2017)
6th generation Intel: m6i – 128 vCPUs / host	(Aug 2021)
6th generation AMD: m6a – 192 vCPUs / host	(Nov 2021)
7th generation Intel: m7i – 192 vCPUs / host	(Aug 2023)
7th generation AMD: m7a – 192 vCPUs / host	(Aug 2023)

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ARM64 HOST SERVER VCPUS – AMAZON EC2 INFRASTRUCTURE-AS-A-SERVICE CLOUD

- Cloud server virtual CPUs/host (ARM64)
- Launched in 2018 on the Amazon Compute Cloud (EC2)
- 64-bit ARM CPUs designed by AWS subsidiary Annapurna Labs
- Lower energy consumption compared to x86-64
- Fixed (non-variable) clock rates, No hyperthreading
- Each new release – performance boost of ~ 30%
- Cost savings of ~20% less for ARM resources on AWS

- 1st generation Graviton: a1– 16 vCPUs / host (Nov 2018)
- 2nd generation Graviton2: m6g– 64 vCPUs/host (Dec 2019)
  - AWS Lambda limited to Graviton2
- 3rd generation Graviton3: m7g- 64 vCPUs/host (May 2022)
- 4th generation Graviton4: m8g– 192 vCPUs/host (Sept 2024)

CLOUD COMPUTING: HOW DID WE GET HERE? - 2

- To make computing faster, we must go "parallel"
- Difficult to expose parallelism in scientific applications
- Not every problem solution has a parallel algorithm
  - Chicken and egg problem…
- Many commercial efforts promoting pure parallel programming efforts have failed
- Enterprise computing world has been *skeptical* and less involved in parallel programming

CLOUD COMPUTING: HOW DID WE GET HERE? - 3

- **Cloud computing** provides access to "infinite" scalable compute infrastructure on demand
- Infrastructure availability is key to exploiting parallelism
- **Cloud applications**
  - Based on **client-server** paradigm
  - **Thin clients** leverage compute hosted on the cloud
  - Applications run many web service instances
  - Employ load balancing

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CLOUD COMPUTING: HOW DID WE GET HERE? - 4

- **Big Data** requires massive amounts of compute resources
- MAP – REDUCE
  - Single instruction, multiple data (SIMD)
  - Exploit data level parallelism
- Bioinformatics example

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SMITH WATERMAN USE CASE

- Applies dynamic programming to find best local alignment of two protein sequences
  - Embarrassingly parallel, each task can run in isolation
  - Use case for GPU acceleration
- **AWS Lambda Serverless Computing Use Case: Goal:** Pair-wise comparison of all unique human protein sequences (20,336)
  - Python client as scheduler
  - C Striped Smith-Waterman (SSW) execution engine

From: Zhao M, Lee WP, Garrison EP, Marth GT: SSW library: an SIMD Smith-Waterman C/C++ library for use in genomic applications. PLoS One 2013, 8:e82138

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SMITH WATERMAN RUNTIME

- Laptop server and client (2-core, 4-HT): 8.7 hours
- AWS Lambda FaaS, laptop as client: 2.2 minutes
  - Partitions 20,336 sequences into 41 sets
  - Execution cost: **~ 82¢ (~237x speed-up)**
- AWS Lambda server, EC2 instance as client: 1.28 minutes
  - Execution cost: **~ 87¢ (~408x speed-up)**
- Hardware
  - Laptop client: Intel i5-7200U 2.5 GHz :4 HT, 2 CPU
  - Cloud client: EC2 Virtual Machine - m5.24xlarge: 96 vCPUs
  - Cloud server: Lambda ~1000 Intel E5-2666v3 2.9GHz CPUs

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CLOUD COMPUTING: HOW DID WE GET HERE? - 5

- Compute clouds are large-scale distributed systems
  - **Heterogeneous systems**
    - Many services/platforms w/ diverse hw + capabilities
  - **Homogeneous systems**
    - Within a platform – illusion of identical hardware
  - **Autonomous**
    - Automatic management and maintenance- largely with little human intervention
  - **Self organizing**
    - User requested resources organize themselves to satisfy requests on-demand

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CLOUD COMPUTING: HOW DID WE GET HERE? - 6

- Compute clouds are large-scale distributed systems
- Infrastructure-as-a-Service (IaaS) Cloud
  - Provide VMs on demand to users
  - ***ec2instances.info*** (AWS EC2)
- Clouds can consist of
  - **Homogeneous** hardware (servers, etc.)
  - **Heterogeneous** hardware (servers, etc.)
- Which is preferable?

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HARDWARE HETEROGENEITY

- **If providing IaaS, what are advantages/ disadvantages of using homogeneous hardware?**
  - Easier to provide same quality of service to end users
    - Less performance variance
    - Components with variable performance: CPUs, memory (speed differences), disks (SSDs, HDDs), network interfaces (caches?)
  - Homogeneous hardware (servers): components are interchangeable
    - As components fail, identical backups are immediately available
    - Example: **blade servers**
  - As clouds grow, why is HW homogeneity difficult to maintain?
- **What are some advantages of using heterogeneous HW?**

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OBJECTIVES

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  - *Parallel and distributed systems (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)*
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  - SIMD architectures, vector processing, multimedia extensions
  - Graphics processing units
  - Speed-up, Amdahl's Law, Scaled Speedup
  - Properties of distributed systems
  - Modularity

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PARALLELISM

- Discovering parallelism and development of parallel algorithms requires considerable effort
- Example: numerical analysis problems, such as solving large systems of linear equations or solving systems of Partial Differential Equations (PDEs), require algorithms based on domain decomposition methods.
- ***How can problems be split into independent chunks?***
- **Fine-grained parallelism**
  - Only small bits of code can run in parallel without coordination
  - Communication is required to synchronize state across nodes
- **Coarse-grained parallelism**
  - Large blocks of code can run without coordination

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PARALLELISM - 2

- **Coordination of nodes**
- Requires ***message passing*** or ***shared memory***
- Debugging parallel ***message passing*** code is easier than parallel ***shared memory*** code
- ***Message passing***: all of the interactions are clear
  - Coordination via specific programming API (MPI)
- ***Shared memory***: interactions can be implicit – *must read the code!!*
- Processing speed is orders of magnitude faster than communication speed (CPU > memory bus speed)
- Avoiding coordination achieves the best speed-up

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TYPES OF PARALLELISM

- Parallelism:
  - Goal: Perform multiple operations at the same time to achieve a speed-up
- Thread-level parallelism (TLP)
  - Control flow architecture (Von Neumann architecture)
- Data-level parallelism
  - Data flow architecture
- Bit-level parallelism
- Instruction-level parallelism (ILP)

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THREAD LEVEL PARALLELISM (TLP)

- Number of threads an application runs at any one time
- Varies throughout program execution
- As a metric:
- **Minimum**: 1 thread
- Can measure **average**, **maximum (peak)**
- **QUESTION: What are the consequences of average (TLP) for scheduling an application to run on a computer with a fixed number of CPU cores and hyperthreads?**
- Let's say there are 4 cores, or 8 hyper-threads…
- ***Key to avoiding waste of computing resources is knowing your application's TLP…***

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TLP – PRIMES EXAMPLE

- Multi-threaded prime number generation
- Compute-bound workload
- Can use variable # of threads
- Generates n prime numbers

- Runtimes: 100,000 primes
- 1 thread: 59.15 s
- 2 threads: 30.957 s
- 4 threads: 15.539 s
- 8 threads: 12.112 s

- Observe TLP with top

```
time ./primes8 30000 >/dev/null
```


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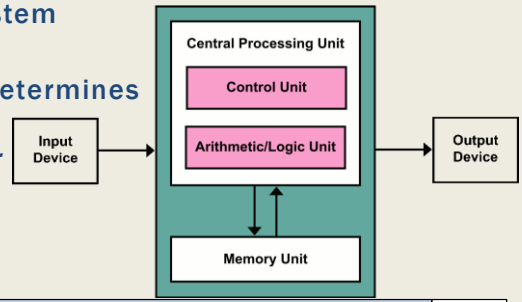
CONTROL-FLOW ARCHITECTURE

- Typical architecture used today – w/ multiple threads
  - Each thread runs a sequential program sequence
- By John von Neumann (1945), also called the Von Neumann architecture
- Dominant computer system architecture
- Program counter (PC) determines next instruction to load into *instruction register*
- Program execution is sequential



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DATA-LEVEL PARALLELISM

- Partition data into big chunks, run separate copies of the program on them with little or no communication
- Problems are considered to be ***embarrassingly parallel***
- Also perfectly parallel or pleasingly parallel…
- Little or no effort needed to separate problem into a number of parallel tasks
- MapReduce programming model is an example

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DATA FLOW ARCHITECTURE

- ***Alternate architecture*** used by network routers, digital signal processors, special purpose systems
- Operations performed when input (data) becomes available
- Envisioned to provide much higher parallelism
- Multiple problems has prevented wide-scale adoption
  - Efficiently broadcasting data tokens in a massively parallel system
  - Efficiently dispatching instruction tokens in a massively parallel system
  - Building content addressable memory large enough to hold all of the dependencies of a real program

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DATA FLOW ARCHITECTURE - 2

- Architecture not as popular as control-flow
- Modern CPUs emulate data flow architecture for dynamic instruction scheduling since the 1990s
  - Out-of-order execution – reduces CPU idle time by not blocking for instructions requiring data by defining execution windows
  - Execution windows: identify instructions that can be run by data dependency
  - Instructions are completed in data dependency order within execution window
    - Execution window size typically 32 to 200 instructions

Utility of data flow architectures has been much less than envisioned

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WE WILL RETURN AT 4:50PM

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BIT-LEVEL PARALLELISM

- Computations on large words (e.g. 64-bit integer) are performed as a single instruction
- Fewer instructions are required on 64-bit CPUs to process larger operands (A+B) providing dramatic performance improvements
- Processors have evolved: 4-bit, 8-bit, 16-bit, 32-bit, 64-bit

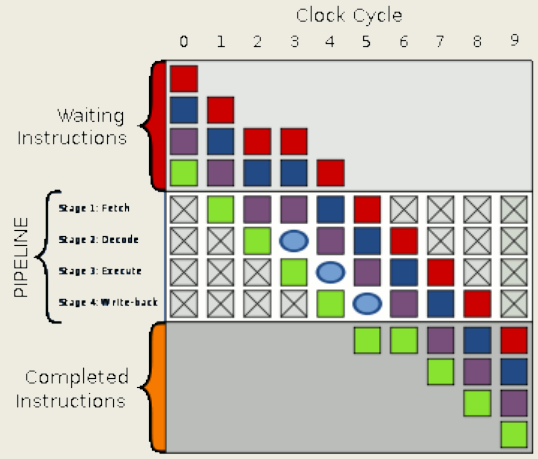
***QUESTION:* How many instructions are required to add two 64-bit numbers on a 16-bit CPU? (Intel 8088)**

- 64-bit MAX int = 9,223,372,036,854,775,807 (signed)
- 16-bit MAX int = 32,767 (signed)
- Intel 8088 – limited to 16-bit registers

INSTRUCTION-LEVEL PARALLELISM (ILP)

- CPU pipelining architectures enable ILP
- CPUs have multi-stage processing pipelines
- Pipelining: split instructions into sequence of steps that can execute concurrently on different CPU circuitry
- Basic RISC CPU - Each instruction has 5 pipeline stages:
- **IF** – *instruction fetch*
- **ID**- *instruction decode*
- **EX** – *instruction execution*
- **MEM** – *memory access*
- **WB** – *write back*

CPU PIPELINING



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INSTRUCTION LEVEL PARALLELISM - 2

- RISC CPU:
- After 5 clock cycles, all 5 stages of an instruction are loaded
- Starting with 6th clock cycle, one full instruction completes each cycle
- The CPU performs 5 tasks per clock cycle!
  Fetch, decode, execute, memory read, memory write back
- Pentium 4 (CISC CPU) – processing pipeline w/ 35 stages!

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OBJECTIVES

- **Cloud Computing: How did we get here?**
  - *Parallel and distributed systems (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)*
  - Data, thread-level, task-level parallelism
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  - Properties of distributed systems
  - Modularity

MICHAEL FLYNN'S COMPUTER ARCHITECTURE TAXONOMY

- Michael Flynn's proposed taxonomy of computer architectures based on concurrent instructions and number of data streams (1966)
- **SISD (Single Instruction Single Data)**
- **SIMD (Single Instruction, Multiple Data)**
- **MIMD (Multiple Instructions, Multiple Data)**
- *LESS COMMON*: MISD (Multiple Instructions, Single Data)
- Pipeline architectures: functional units perform different operations on the same data
- For fault tolerance, may want to execute same instructions redundantly to detect and mask errors – for task replication

FLYNN'S TAXONOMY

- **SISD (Single Instruction Single Data)**
  Scalar architecture with one processor/core.
  - Individual cores of modern multicore processors are "SISD"
- **SIMD (Single Instruction, Multiple Data)**
  Supports vector processing
  - When SIMD instructions are issued, operations on individual vector components are carried out concurrently
  - Two 64-element vectors can be added in parallel
  - Vector processing instructions added to modern CPUs
  - Example: Intel MMX (multimedia) instructions

(SIMD): VECTOR PROCESSING ADVANTAGES

- Exploit data-parallelism: vector operations enable speedups
- Vectors architecture provide vector registers that can store entire matrices into a CPU register
- SIMD CPU extension (e.g. MMX) add support for vector operations on traditional CPUs
- Vector operations reduce total number of instructions for large vector operations
- Provides higher potential speedup vs. MIMD architecture
- Developers can think sequentially; not worry about parallelism

FLYNN'S TAXONOMY - 2

- **MIMD (Multiple Instructions, Multiple Data)** - system with several processors and/or cores that function asynchronously and independently
- At any time, different processors/cores may execute different instructions on different data
- Multi-core CPUs are MIMD
- Processors share memory via interconnection networks
  - Hypercube, 2D torus, 3D torus, omega network, other topologies
- MIMD systems have different methods of sharing memory
  - Uniform Memory Access (UMA)
  - Cache Only Memory Access (COMA)
  - Non-Uniform Memory Access (NUMA)

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ARITHMETIC INTENSITY

- **Arithmetic intensity**: Ratio of work (W) to memory traffic r/w (Q)

$$I = \frac{W}{Q}$$

  Example: # of floating-point ops per byte of data read
- Characterizes application scalability with SIMD support
  - *SIMD can perform many fast matrix operations in parallel*
- ***High arithmetic Intensity:***
  Programs with dense matrix operations scale up nicely (many calcs vs memory RW, supports lots of parallelism)
- ***Low arithmetic Intensity:***
  Programs with sparse matrix operations do not scale well with problem size
  (memory RW becomes bottleneck, not enough ops!)

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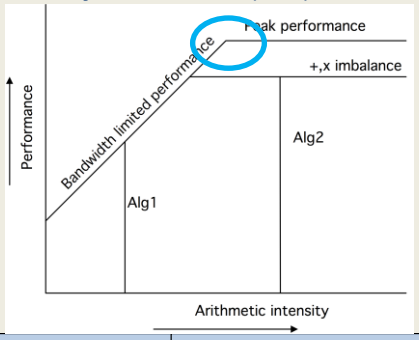
ROOFLINE MODEL

- When program reaches a given arithmetic intensity performance of code running on CPU hits a "roof"
- CPU performance bottleneck changes from: **memory bandwidth** (left) → **floating point performance** (right)



Key take-aways:
When a program's has **low** Arithmetic Intensity, memory bandwidth limits performance..

With **high** Arithmetic intensity, the system has peak parallel performance…
→ *performance is limited by??*

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OBJECTIVES

- **Cloud Computing: How did we get here?**
  - *Parallel and distributed systems (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)*
  - Data, thread-level, task-level parallelism
  - Parallel architectures
  - SIMD architectures, vector processing, multimedia extensions
  - Graphics processing units
  - Speed-up, Amdahl's Law, Scaled Speedup
  - Properties of distributed systems
  - Modularity

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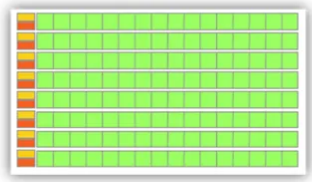
CPU

- Low compute density
- Complex control logic
- Large caches (L1$/L2$, etc.)
- Optimized for serial operations
  - Fewer execution units (ALUs)
  - Higher clock speeds
- Shallow pipelines (<30 stages)
- Low Latency Tolerance
- Newer CPUs have more parallelism

GPU



- High compute density
- High Computations per Memory Access
- Built for parallel operations
  - Many parallel execution units (ALUs)
  - Graphics is the best known case of parallelism
- Deep pipelines (hundreds of stages)
- High Throughput
- High Latency Tolerance
- Newer GPUs:
  - Better flow control logic (becoming more CPU-like)
  - Scatter/Gather Memory Access
  - Don't have one-way pipelines anymore

From: https://hypertec.com/blog/gpus-taking-over-cpus/

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GRAPHICAL PROCESSING UNITS (GPUs)

- GPU provides multiple SIMD processors
- Typically 7 to 15 SIMD processors each
- 32,768 total registers, divided into 16 lanes (2048 registers each)
- GPU programming model: single instruction, multiple thread
- Programmed using CUDA- C like programming language by NVIDIA for GPUs
- CUDA threads – single thread associated with each data element (e.g. vector or matrix)
- Thousands of threads run concurrently

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PARALLEL COMPUTING

- Parallel hardware and software systems allow:
  - Solve problems demanding resources not available on single system.
  - Reduce time required to obtain solution
- The *speed-up* (S) measures effectiveness of parallelization:

$$S(N) = T(1) / T(N)$$

T(1) → execution time of total sequential computation

T(N) → execution time for performing N parallel computations in parallel

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SPEED-UP EXAMPLE

- Consider embarrassingly parallel image processing
- Eight images (multiple data)
- Apply image transformation (greyscale) in parallel
- 8-core CPU, 16 hyper threads
- Sequential processing: perform transformations one at a time using a single program thread
  - 8 images, 3 seconds each: `T(1) = 24 seconds`
- Parallel processing
  - 8 images, 3 seconds each: `T(N) = 3 seconds`
- Speedup: `S(N) = 24 / 3 = 8x speedup`
- Called "**perfect scaling**"
- Must consider data transfer and computation setup time

AMDAHL'S LAW

- Amdahl's law is used to estimate the speed-up of a job using parallel computing

1. Divide job into two parts
2. Part A that will still be sequential
3. Part B that will be sped-up with parallel computing

- Portion of computation which cannot be parallelized will determine (i.e. limit) the overall speedup
- Amdahl's law assumes jobs are of a fixed size
- Also, Amdahl's assumes no overhead for distributing the work, and a perfectly even work distribution

AMDAHL'S LAW

$$S = \frac{1}{(1-f) + \frac{f}{N}}$$

- S = theoretical speedup of the whole task
- f= fraction of work that is parallel (ex. 25% or 0.25)
- N= proposed speed up of the parallel part (ex. 5 times speedup)

- % improvement of task execution = 100 * (1 – (1 / S))

- **Using Amdahl's law, what is the maximum possible speed-up?**

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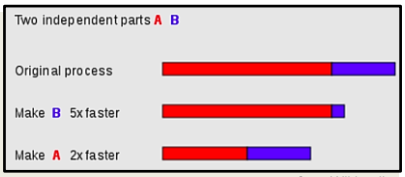
AMDAHL'S LAW EXAMPLE

- Program with two independent parts:
  - Part A is 75% of the execution time
  - Part B is 25% of the execution time
- Part B is made 5 times faster with parallel computing
- Estimate the percent improvement of task execution
- Original Part A is 3 seconds, Part B is 1 second



from Wikipedia

- N=5 (speedup of part B)
- f=.25 (only 25% of the whole job (A+B) will be sped-up)
- S=1 / ((1-f) + f/S)
- S=1 / ((.75) + .25/5)
- S=1.25
- % improvement = 100 * (1 – 1/1.25) = **20%**

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GUSTAFSON'S LAW

- Calculates the ***scaled speed-up*** using "N" processors

$$S(N) = N + (1 - N)\,\alpha$$

N: Number of processors

α: fraction of program run time which can't be parallelized (e.g. must run sequentially)

- *Can be used to estimate runtime of parallel portion of program*

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GUSTAFSON'S LAW

- Calculates the ***scaled speed-up*** using "N" processors

$$S(N) = N + (1 - N)\,\alpha$$

N: Number of processors

α: fraction of program run time which can't be parallelized (e.g. must run sequentially)

- *Can be used to estimate runtime of parallel portion of program*
- Where $\alpha = \sigma / (\pi + \sigma)$
- Where σ= sequential time, π =parallel time
- Our Amdahl's example: σ= 3s, π =1s, α =.75

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GUSTAFSON'S LAW

- Calculates the ***scaled speed-up*** using "N" processors

$$S(N) = N + (1 - N)\,\alpha$$

N: Number of processors

α: fraction of program run time which can't be parallelized (e.g. must run sequentially)

- Example:
  Consider a program that is embarrassingly parallel, but 75% cannot be parallelized. α=.75
  QUESTION: ***If deploying the job on a 2-core CPU, what scaled speedup is possible assuming the use of two processes that run in parallel?***

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GUSTAFSON'S EXAMPLE

- **QUESTION:**
  What is the maximum theoretical speed-up on a **2-core CPU** ?
  $S(N) = N + (1 - N)\,\alpha$
  $N=2,\ \alpha=.75$
  $S(N) = 2 + (1 - 2)\,.75$
  $S(N) = ?$

- What is the maximum theoretical speed-up on a 16-**core CPU**?
  $S(N) = N + (1 - N)\,\alpha$
  $N=16,\ \alpha=.75$
  $S(N) = 16 + (1 - 16)\,.75$
  $S(N) = ?$

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GUSTAFSON'S EXAMPLE

- **QUESTION:**
  What is the maximum theoretical speed-up on a **2-core CPU** ?
  $S(N) = N + (1 - N)\ \alpha$
  $N=2,\ \alpha=$
  $S(N) = 2$
  $S(N) = ?$

For 2 CPUs, speed up is 1.25x

For 16 CPUs, speed up is 4.75x

- What is the maximum theoretical speed-up on a **16-core CPU**?
  $S(N) = N + (1 - N)\ \alpha$
  $N=16,\ \alpha=.75$
  $S(N) = 16 + (1 - 16)\ .75$
  $S(N) = ?$

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MOORE'S LAW

- Transistors on a chip doubles approximately every 1.5 years
- CPUs now have billions of transistors
- Power dissipation issues at faster clock rates leads to heat removal challenges
  - Transition from: increasing clock rates → to adding CPU cores
- ***Symmetric core processor*** –multi-core CPU, all cores have the same computational resources and speed
- ***Asymmetric core processor*** – on a multi-core CPU, some cores have more resources and speed
- ***Dynamic core processor*** – processing resources and speed can be dynamically configured among cores
- ***Observation: asymmetric processors offer a higher speedup***

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DISTRIBUTED SYSTEMS

- Collection of autonomous computers, connected through a network with distribution software called "middleware" that enables coordination of activities and sharing of resources
- **Key characteristics:**
- Users perceive system as a single, integrated computing facility.
- Compute nodes are autonomous
- Scheduling, resource management, and security implemented by every node
- Multiple points of control and failure
- Nodes may not be accessible at all times
- System can be scaled by adding additional nodes
- Availability at low levels of HW/software/network reliability

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DISTRIBUTED SYSTEMS - 2

- Key non-functional attributes
  - Known as "ilities" in software engineering
- Availability – 24/7 access?
- Reliability - Fault tolerance
- Accessibility – reachable?
- Usability – user friendly
- Understandability – can under
- Scalability – responds to variable demand
- Extensibility – can be easily modified, extended
- Maintainability – can be easily fixed
- Consistency – data is replicated correctly in timely manner

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TRANSPARENCY PROPERTIES OF DISTRIBUTED SYSTEMS

- **Access transparency**: local and remote objects accessed using identical operations
- **Location transparency**: objects accessed w/o knowledge of their location.
- **Concurrency transparency**: several processes run concurrently using shared objects w/o interference among them
- **Replication transparency**: multiple instances of objects are used to increase reliability
  - users are unaware if and how the system is replicated
- **Failure transparency**: concealment of faults
- **Migration transparency**: objects are moved w/o affecting operations performed on them
- **Performance transparency**: system can be reconfigured based on load and quality of service requirements
- **Scaling transparency**: system and applications can scale w/o change in system structure and w/o affecting applications

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TYPES OF MODULARITY

- ***Soft modularity:*** TRADITIONAL
- Divide a program into modules (classes) that call each other and communicate with shared-memory
- A procedure calling convention is used (or method invocation)

- ***Enforced modularity:*** CLOUD COMPUTING
- Program is divided into modules that communicate only through message passing
- The ubiquitous client-server paradigm
- Clients and servers are independent decoupled modules
- System is more robust if servers are stateless
- May be scaled and deployed separately
- May also FAIL separately!

CLOUD COMPUTING – HOW DID WE GET HERE? SUMMARY OF KEY POINTS

- Multi-core CPU technology and hyper-threading
- What is a
  - Heterogeneous system?
  - Homogeneous system?
  - Autonomous or self-organizing system?
- **Fine grained vs. coarse grained parallelism**
- Parallel message passing code is easier to debug than shared memory (e.g. p-threads)
- Know your application's max/avg **Thread Level Parallelism** (***TLP***)
- **Data-level parallelism**: Map-Reduce, (SIMD) Single Instruction Multiple Data, Vector processing & GPUs

CLOUD COMPUTING – HOW DID WE GET HERE? SUMMARY OF KEY POINTS - 2

- **Bit-level parallelism**
- **Instruction-level parallelism** (CPU pipelining)
- **Flynn's taxonomy**: computer system architecture classification
  - **SISD** – Single Instruction, Single Data (modern core of a CPU)
  - **SIMD** – Single Instruction, Multiple Data (Data parallelism)
  - **MIMD** – Multiple Instruction, Multiple Data
  - MISD is RARE; application for fault tolerance…
- **Arithmetic Intensity**: ratio of calculations vs memory RW
- **Roofline model:**
  Memory bottleneck with low arithmetic intensity
- **GPUs**: ideal for programs with high arithmetic intensity
  - SIMD and Vector processing supported by many large registers

CLOUD COMPUTING – HOW DID WE GET HERE? SUMMARY OF KEY POINTS - 3

- **Speed-up (S)**
  S(N) = T(1) / T(N)
- **Amdahl's law:**
  S=1 / ((1-f) + f/N),s=latency, f=parallel fraction, N=speed-up
- α = percent of program that must be sequential
- **Scaled speedup with N processes:**
  S(N) = N – α(N-1)
- Moore's Law
- Symmetric core, Asymmetric core, Dynamic core CPU
- Distributed Systems Non-function quality attributes
- Distributed Systems – Types of Transparency
- Types of modularity- Soft, Enforced

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QUESTIONS

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