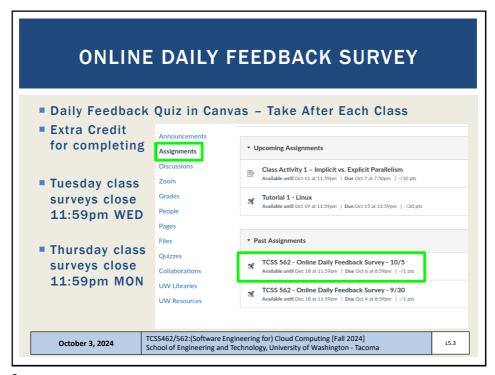


Τ

OBJECTIVES - 10/3 Questions from 10/1 Tutorial 0, Tutorial 1, Tutorial 2 Cloud Computing - How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition) Class Activity 1 - Implicit vs Explicit Parallelism SIMD architectures, vector processing, multimedia extensions Graphics processing units Speed-up, Amdahl's Law, Scaled Speedup Properties of distributed systems Modularity October 3, 2024 TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

2



	d: Oct 7 at 1:13am z Instructions			
40	2 motractions			
	Question 1		0.5 pts	
	On a scale of 1 to 10, please class:	e classify your perspective	on material covered in today's	
	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	y's class:		
	1 2 3 4	5 6 7	8 9 10	
	Slow	Just Right	Fast	

4

MATERIAL / PACE

- Please classify your perspective on material covered in today's class (40 respondents):
- 1-mostly review, 5-equal new/review, 10-mostly new
- **Average 7.03** (\downarrow previous 6.16)
- Please rate the pace of today's class:
- 1-slow, 5-just right, 10-fast
- <u>Average 5.48</u> (<u>↑ previous 5.55</u>)
- Response rates:
- TCSS 462: 27/42
- TCSS 562: 13/18

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.5

5

FEEDBACK FROM 10/3

- Difference between hyperthread and vCPU
- vCPU stands for virtual CPU
- This refers to the CPUs provided by a virtual machine
- Since a virtual machine is a virtual server, the CPUs in virtual servers are called virtual CPUs
- Instructions executed on a virtual CPU get mapped to logical CPU cores on the OS for execution (KVM)
 - The virtual to physical mapping varies based on which physical CPUs are free and available
 - 4-core server, 2-vCPU VM:
 - vCPU 0 \rightarrow (CPU 0, CPU 1, CPU 2, CPU 3) mapped based on availability
 - vCPU 1 → (CPU 0, CPU 1, CPU 2, CPU 3) mapped based on availability

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.6

6

HYPER-THREADING - 2

- How do I use hyper-threading?
- Hyper-threading is automatic
- Modern CPUs expose each physical CPU core as two CPU cores
- cat /proc/cpuinfo command lists individual cores
- Operating system schedules processes & threads to run on a hyper-thread
- On CPUs with hyper-threading, each CPU core has two hyper-threads
- To the operating system they are seen as full-featured independent CPU cores

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

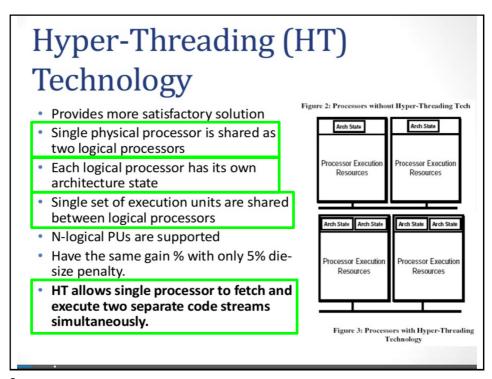
L3.7

7

CAT / PROC/CPUINFO | | LSCPU

```
wtloydødtone:-/bropbox/courses/tcss562$ cat /proc/cpuinfo | grep -C 20 ht
processor : 0
vendor_ld : GenuineIntel
cpu family : 6
model : 94
model name : Intel(R) Core(TM) i7-6700HQ CPU @ 2.60GHz
stepping : 3
microcode : 0 xdc
cpu MHz : 840.023
cache size : 6144 KB
physical id : 0
siblings : 8
core id : 0
cpu cores : 4
apicid : 0
fpu : yes
fpu_exception : yes
cpuid level : 22
wp : yes
flags : fpu wme de pse tsc msr pae mce cx8 apic sep mtrr pge mca cmov pat pse36 clflush dts acpi mmx
fxsr sse sse2 ss ht tm pbe syscall nx pdpeigb rdtscp lm constant_tsc art arch_perfmon pebs bts rep_good nopl xt
goology nonstop_tsc aperfmperf pni pclmulqd dtes64 monitor ds_cpl vmx est tm2 ssea3 sdbg fma cx16 xtpr ydcm pc
id sse4_1 sse4_2 x2apic movbe popcnt tsc_deadline_timer aes xsave avx f16c rdrand lahf_lm abm 3dnowprefetch epb
lnvpcid_single intel_pt ssbd ibrs lbpb stibp kalser tpr_shadow vnmi flexpriority ept typid fsgsbase tsc_adjust
bmin hie avx2 smep bmi2 erms invpcid rtm mpx rdseed adx smap clflushopt xsaveopt xsavecx ysetbx1 dtherm fda arat
pln pts hwp hwp_notify hwp_act_window hwp_epp md_clear flush_lid
bugs : cpu_meltdown spectre_v1 spectre_v2 spec_store_bypass litf mds swapgs taa itlb_multihit srbds
bogomips : 5184.46
clflush size : 64
address sizes : 39 bits physical, 48 bits virtual
```

8



HYPER-THREADING - 3

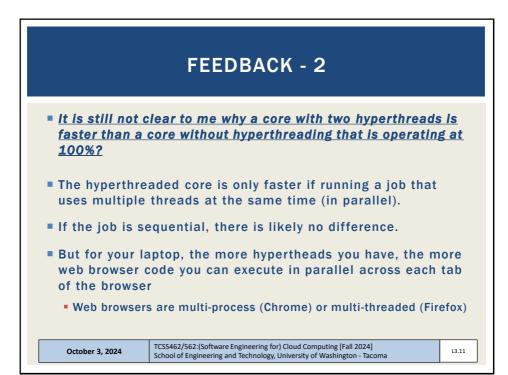
- When should we use hyper-threading, and when should not?
 - For personal computing, hyper-threading helps improve system performance when many programs use only short bursts of CPU time
 - Databases, HPC (science) applications, and others may benefit from disabling hyper-threading. Testing will help quantify performance.
 - Disabling hyper-threading (HW setting), cuts the number of CPU cores available to operating system in half
 - Can be disabled in the System BIOS or UEFI (uniform extensible firmware interface) software
 - BIOS / UEFI is a small resident program that can be accessed by pressing a function-key when rebooting the computer
 - BIOS / UEFI is used to configure hardware options
 - Making changes requires rebooting the computer

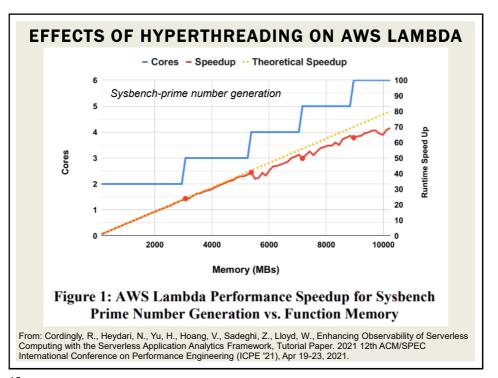
October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

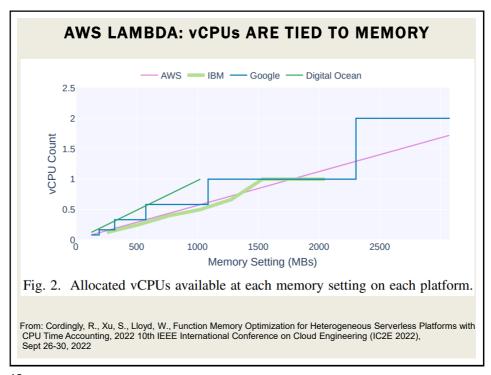
L3.10

10





12



FEEDBACK - 3 • If I use a computer with 8 cores (client) to rent a virtual machine with 128 cores through a cloud provider, the computer with less cores won't decrease the performance of the virtual machine with more cores because they are separate? CORRECT, the performance will not decrease. ■ The 8-core (laptop/desktop) is just used to access the remote computer via ssh/graphical desktop The laptop/desktop acts as a client computer used to access the powerful remote server Any applications / jobs /workloads are run on the remote server, but are launched by the client Through a terminal session (ssh), or remote graphical desktop Or by calling a web service hosted on the powerful server You may experience <u>network latency</u> between the client and server for large data transfers TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma October 3, 2024

14

FEEDBACK - 4

- Thread level parallelism is not clear
- Thread level parallelism refers to when parallelism occurs as a result of multiple threads performance operations in parallel typically on a multi-core computer
- As DevOps engineers, we often are responsible for deploying our applications in the cloud. Therefore, we need to understand the average number and peak number of threads our application requires.
- In class, I demonstrated how this can be observed in Linux using "top" and a multi-threaded prime number generation program

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.15

15

DEMOGRAPHICS SURVEY

Please complete the ONLINE demographics survey:

We have received 37 responses so far. We are waiting on ~23 responses.

- https://forms.gle/6ER7PzfP521vdxYW9
- Linked from course webpage in Canvas:
- http://faculty.washington.edu/wlloyd/courses/tcss562/ announcements.html

October 3, 2024

TCSS462/562: (Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.16

16

AWS CLOUD CREDITS SURVEY

Please complete the AWS Cloud Credits survey:

Please only complete survey after setting up AWS account or if requiring an IAM user (no-credit card option)

- https://forms.gle/fmKkLZbxZECbAay16
- Linked from course webpage in Canvas:
- http://faculty.washington.edu/wlloyd/courses/tcss562/ announcements.html

October 3, 2024

TCSS462/562: (Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.17

17

OBJECTIVES - 10/3

- Questions from 10/1
- Tutorial 0, Tutorial 1, Tutorial 2
- Cloud Computing How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)
- Class Activity 1 Implicit vs Explicit Parallelism
- SIMD architectures, vector processing, multimedia extensions
- Graphics processing units
- Speed-up, Amdahl's Law, Scaled Speedup
- Properties of distributed systems
- Modularity

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.18

18

OBJECTIVES - 10/3 Ouestions from 10/1 Tutorial 0, Tutorial 1, Tutorial 2 Cloud Computing - How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition) Class Activity 1 - Implicit vs Explicit Parallelism SIMD architectures, vector processing, multimedia extensions Graphics processing units Speed-up, Amdahl's Law, Scaled Speedup Properties of distributed systems Modularity October 3, 2024 TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

19

OBJECTIVES - 10/3 Questions from 10/1 Tutorial 0, Tutorial 1, Tutorial 2 Cloud Computing - How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition) Class Activity 1 - Implicit vs Explicit Parallelism SIMD architectures, vector processing, multimedia extensions Graphics processing units Speed-up, Amdahl's Law, Scaled Speedup Properties of distributed systems Modularity October 3, 2024 TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

20

OBJECTIVES - 10/3

- Questions from 10/1
- Tutorial 0, Tutorial 1, Tutorial 2
- Cloud Computing How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)
- Class Activity 1 Implicit vs Explicit Parallelism
- SIMD architectures, vector processing, multimedia extensions
- Graphics processing units
- Speed-up, Amdahl's Law, Scaled Speedup
- Properties of distributed systems
- Modularity

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.21

21

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024]
School of Engineering and Technology, University of Washington - Tacoma

3.22

22

CLOUD COMPUTING: HOW DID WE GET HERE? - 6

- Compute clouds are large-scale distributed systems
- Infrastructure-as-a-Service (laaS) 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?

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.23

23

HARDWARE HETEROGENEITY

- If providing laaS, 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?

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

3.24

24

OBJECTIVES - 10/3

- Questions from 10/1
- Tutorial 0, Tutorial 1, Tutorial 2
- Cloud Computing How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)
- Class Activity 1 Implicit vs Explicit Parallelism
- SIMD architectures, vector processing, multimedia extensions
- Graphics processing units
- Speed-up, Amdahl's Law, Scaled Speedup
- Properties of distributed systems
- Modularity

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.25

25

CLASS ACTIVITY 1

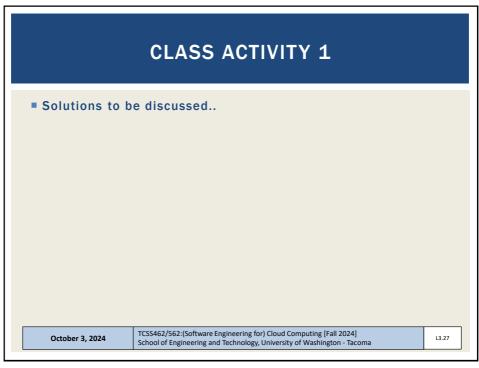
- Form groups of ~3 in class or with Zoom breakout rooms
- Each group will complete a MSWORD DOCX worksheet
- Be sure to add names at top of document as they appear in Canvas
- Activity can be completed in class or after class
- The activity can also be completed individually
- When completed, one person should submit a PDF of the documet to Canvas
- Instructor will score all group members based on the uploaded PDF file
- To get started:
 - Follow the link: (link also available in Canvas)
 https://faculty.washington.edu/wlloyd/courses/tcss562/assignments/tcss462_562_f2024_tps1.docx

October 3, 2024

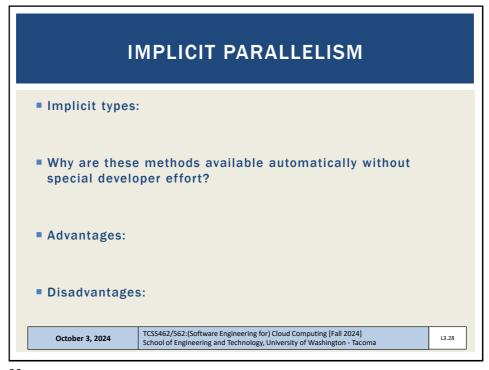
TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.26

26



27



28

EXPLICIT PARALLELISM				
■ Explicit types:				
Advantages:				
■ Disadvantages:				
October 3, 2024 TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma				

29

PARALLELISM QUESTIONS

- 7. For bit-level parallelism, should a developer be concerned with the available number of virtual CPU processing cores when choosing a cloud-based virtual machine if wanting to obtain the best possible speed-up? (Yes / No)
- 8. For instruction-level parallelism, should a developer be concerned with the physical CPU's architecture used to host a cloud-based virtual machine if wanting to obtain the best possible speed-up? (Yes / No)

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.30

30

PARALLELISM QUESTIONS - 2

- 9. An application developer measures the average and peak thread level parallelism (TLP) of an application prior to deployment on the AWS EC2. The developer measures an average TLP of 2.3, and a peak TLP of 7.3. The application is to be deployed using a compute-optimized (c-series) ec2 instance. Using resources online, such as the websites below, propose a good virtual machine (ec2 type) that satisfies average TLP, and a second for satisfying peak TLP.
- https://docs.aws.amazon.com/ec2/latest/instancetypes/ co.html
- https://instances.vantage.sh/

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.31

31

PARALLELISM QUESTIONS - 3

- What is a good ec2 c-series instance for average TLP?
- Why is this instance good/sufficient for satisfying average TLP?
- What is a good ec2 c-series instance for peak TLP?
- Why is this instance good/sufficient for satisfying peak TLP?

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.32

32

OBJECTIVES - 10/3

- Questions from 10/1
- Tutorial 0, Tutorial 1, Tutorial 2
- Cloud Computing How did we get here?
 (Marinescu Ch. 2 1st edition, Ch. 4 2nd edition)
- Class Activity 1 Implicit vs Explicit Parallelism
- SIMD architectures, vector processing, multimedia extensions
- Graphics processing units
- Speed-up, Amdahl's Law, Scaled Speedup
- Properties of distributed systems
- Modularity

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.33

33

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

October 3, 2024

TCSS462/562: (Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

3.34

34

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

October 3, 2024

TCSS462/562: (Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.35

35

(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

October 3, 2024

TCSS462/562: (Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.36

36

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)

October 3, 2024

TCSS462/562: (Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.37

37

ARITHMETIC INTENSITY

Arithmetic intensity: Ratio of work (W) to memory traffic r/w (Q) $I = \frac{V}{Q}$

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

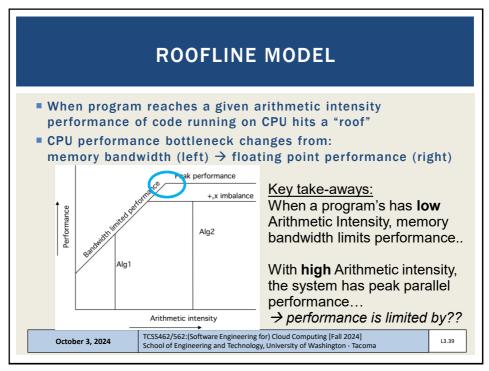
(memory RW becomes bottleneck, not enough ops!)

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.38

38



OBJECTIVES - 10/3 Questions from 10/1 Tutorial 0, Tutorial 1, Tutorial 2 Cloud Computing - How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition) Class Activity 1 - Implicit vs Explicit Parallelism SIMD architectures, vector processing, multimedia extensions Graphics processing units Speed-up, Amdahl's Law, Scaled Speedup Properties of distributed systems Modularity October 3, 2024 TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

40

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.41

41

OBJECTIVES - 10/3

- Questions from 10/1
- Tutorial 0, Tutorial 1, Tutorial 2
- Cloud Computing How did we get here?
 (Marinescu Ch. 2 1st edition, Ch. 4 2nd edition)
- Class Activity 1 Implicit vs Explicit Parallelism
- SIMD architectures, vector processing, multimedia extensions
- Graphics processing units
- Speed-up, Amdahl's Law, Scaled Speedup
- Properties of distributed systems
- Modularity

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

3.42

42

PARALLEL COMPUTING

- Parallel hardware and software systems allow:
 - Solving problems needing resources not available on a single system.
 - Reduced time required to obtain solution
- The speed-up (S) measures effectiveness of parallelization:

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

 $T(1) \rightarrow$ execution time of total sequential computation

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

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.43

43

SPEED-UP EXAMPLE

- Consider embarrassingly parallel image processing
- Eight images (multiple data)
- Apply image transformation (greyscale) in parallel
- 8-core CPU, 16 hyperthreads
- 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

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.44

44

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

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.45

45

AMDAHL'S LAW

Speed-up formula

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

- 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))

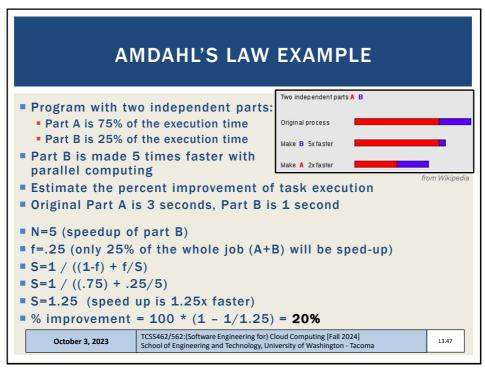
Using Amdahl's law, we can find the maximum possible speed-up (S) for a given scenario (e.g. ~8x) ...

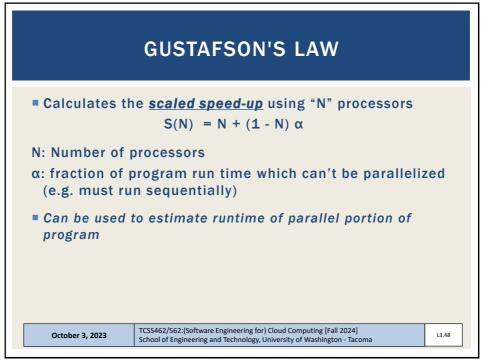
October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.46

46





48

GUSTAFSON'S LAW

■ Calculates the <u>scaled speed-up</u> 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

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.49

49

GUSTAFSON'S LAW

Calculates the <u>scaled speed-up</u> 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?

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.50

50

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■ 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) = ?

S(N) = ?

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

51

GUSTAFSON'S EXAMPLE

What is the maximum theoretical speed-up on a 2-core CPU?

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

 $N=2, \alpha=$

For 2 CPUs, speed up is 1.25x S(N) =

S(N) = ?

For 16 CPUs, speed up is 4.75x

■ What is the maximum theoretical speed-up on a 16-core CPU?

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

 $N=16, \alpha=.75$

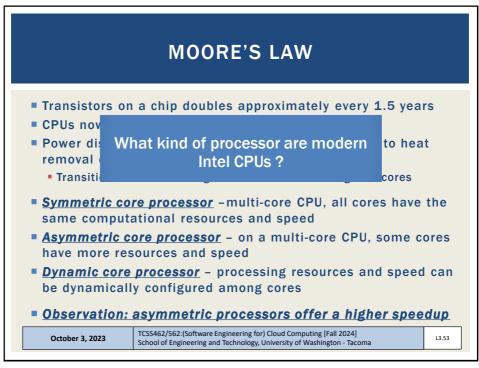
S(N) = 16 + (1 - 16).75

S(N) = ?

October 3, 2023

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

52



OBJECTIVES - 10/3 Questions from 10/1 Tutorial 0, Tutorial 1, Tutorial 2 Cloud Computing - How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition) Class Activity 1 - Implicit vs Explicit Parallelism SIMD architectures, vector processing, multimedia extensions Graphics processing units Speed-up, Amdahl's Law, Scaled Speedup Properties of distributed systems Modularity October 3, 2024 TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

54

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.55

55

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024]
School of Engineering and Technology, University of Washington - Tacoma

3.56

56

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.57

57

OBJECTIVES - 10/3

- Questions from 10/1
- Tutorial 0, Tutorial 1, Tutorial 2
- Cloud Computing How did we get here? (Marinescu Ch. 2 - 1st edition, Ch. 4 - 2nd edition)
- Class Activity 1 Implicit vs Explicit Parallelism
- SIMD architectures, vector processing, multimedia extensions
- Graphics processing units
- Speed-up, Amdahl's Law, Scaled Speedup
- Properties of distributed systems

Modularity

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.58

58

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!

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.59

59

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.60

60

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

October 3, 2024

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

L3.61

61

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

- Speed-up (S)
 S(N) = T(1) / T(N)
- Amdahl's law:

 $S = 1/\alpha$

 α = percent of program that must be sequential

Scaled speedup with N processes:

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

Moore's Law

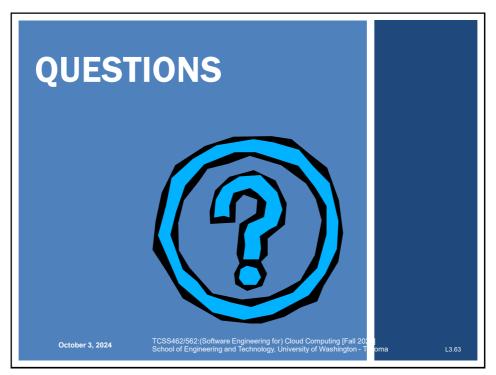
October 3, 2024

- Symmetric core, Asymmetric core, Dynamic core CPU
- Distributed Systems Non-function quality attributes
- Distributed Systems Types of Transparency
- Types of modularity- Soft, Enforced

TCSS462/562:(Software Engineering for) Cloud Computing [Fall 2024] School of Engineering and Technology, University of Washington - Tacoma

3.62

62



63