# Performance Analysis

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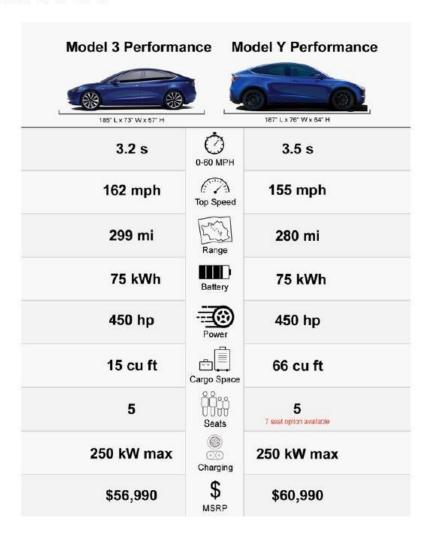
### Performance?

 To measure improvement in computer architectures, it is necessary to compare alternative designs

 A <u>better</u> <u>system</u> has <u>better</u> <u>performance</u>, but <u>what exactly is the</u> <u>performance</u>?

## Performance?

#### Performance?



# Performance Metrics – Sequential Systems

#### Performance?

- For the computer systems and programs:
  - one main performance metric is Time
  - Or just wall-clock time

### Performance?

#### The execution time of a program A can be split into:

- User CPU time: capturing the time that the CPU spends for executing A
- System CPU time: capturing the time that the CPU spends for the execution of routines of the operating system issued by A
- Waiting time: caused by waiting for the completion of I/O operations and by the execution of other programs because of <u>time sharing</u>

Here we concentrate on user CPU

# Computer

- Measuring coff companies to the control of the co
  - Clock Speed
  - MIPS
  - FLOPS
  - Benchmark Tests



- Factors affecting Computer
   Performance
  - Processor Speed
  - Data Bus width
  - Amount of cache
  - Faster interfaces

## Measuring

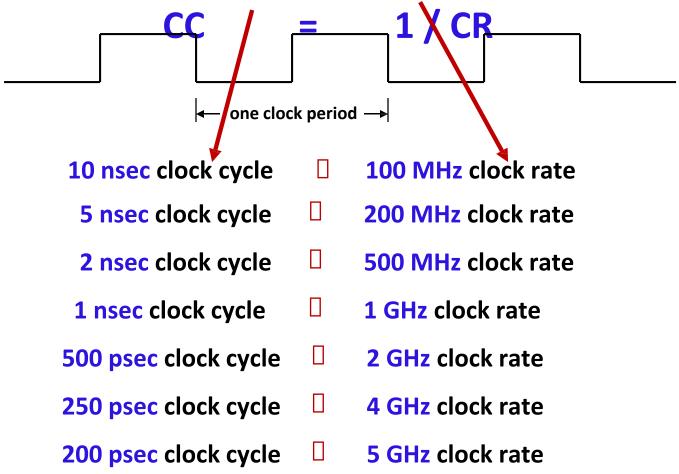
- Every processor has a clock which ticks continuously at a regular rate
- Clock synchronises all the digital components

Cycle time measured in GHz

200 MHz (megahertz) means the clock ticks
 200,000,000 times a second (*Pentium1* - 1995)

#### **Machine Clock Rate**

 Clock Rate (CR) in MHz, GHz, etc. is inverse of Clock Cycle (CC) time (time of a single clock period)



## **Measuring Performance**

- Clock Speed
  - Generally the faster the clock speed the faster the processor 3.2 GHz is faster than 1.2 GHz
- MIPS Millions of Instructions per Second
  - Better comparison
  - But beware of false claims:
    - Such as, only using the simplest & fastest instructions and different processor families (having different ISA).
- Flops Floating Point Operations per sec.
  - Best measure as FP operations are the same in every processor and provide best basis

# **Units of High Performance Computing**

#### **Basic Unit**

#### **Speed**

#### **Capacity**

Kilo
Mega
Giga
Tera
Peta
Exa
Zeta

•	
1 Kflop/s	10 <sup>3</sup> Flop/second
1 Mflop/s	10 <sup>6</sup> Flop/second
1 Gflop/s	10 <sup>9</sup> Flop/second
1 Tflop/s	10 <sup>12</sup> Flop/second
1 Pflop/s	10 <sup>15</sup> Flop/second
1 Eflop/s	10 <sup>18</sup> Flop/second
1 Zflop/s	10 <sup>21</sup> Flop/second

10 <sup>3</sup> Bytes
10 <sup>6</sup> Bytes
10 <sup>9</sup> Bytes
10 <sup>12</sup> Bytes
10 <sup>15</sup> Bytes
10 <sup>18</sup> Bytes
10 <sup>21</sup> Bytes

# Measuring

• When we disadire a the computer carries out instructions

• The measure we use is **MIPS** (*Millions of Instructions Per Second*).

#### MIPS affected by:

- The clock speed of the processor
- The speed of the buses
- The speed of memory access.

## **MIPS**

▶ A performance measure often used in practice to evaluate the performance of a computer system is the MIPS rate for a program A:

$$MIPS(A) = \frac{n_{instr}(A)}{T_{U\_CPU}(A) \cdot 10^6} . \tag{1}$$

 $n_{instr}(A)$ : number of instructions of program A  $T_{U\_CPU}(A)$ : user CPU time of program A

#### **Example:**

$$n_{instr}(A) = 4$$
 Millions  
 $Tu_{CPU}(A) = 0.05$  seconds

$$4 / 0.05 = 80 \text{ Millions}$$
  $/ 10^6 = 80 \text{ MIPS}$ 

## **MIPS**

modification:

$$MIPS(A) = \frac{r_{cycle}}{CPI(A) \cdot 10^6} ,$$

where  $r_{cycle} = 1/t_{cycle}$  is the clock rate of the processor. CPI(A): Clock cycles Per Instruction: average number of CPU cycles used for instructions of program A

► Faster processors lead to larger MIPS rates than slower processors.

#### **Example:**

$$r_{\text{cycle}} = 600 \text{ MHz } (Mega == 10^6)$$
  
 $CPI(A) = 3$ 

$$600 * 10^6 / 3 = 200 * 10^6 / 10^6 = 200$$

**MIPS** 

## **FLOP**

► For program with scientific computations, the MFLOPS rate (Million Floating-point Operations Per Second) is sometimes used. The MFLOPS rate of a program A is defined by

$$MFLOPS(A) = \frac{n_{f|p\_op}(A)}{T_{U\_CPU}(A) \cdot 10^6}$$
 (2)

 $n_{f|p\_op}(A)$ : number of floating-point operations executed by A.  $T_{U\_CPU}(A)$ : user CPU time of program A

► The effective number of operations performed is used for MFLOPS: the MFLOPS rate provides a fair comparison of different program versions performing the same operations.

#### **Example:**

 $n_{flp\_op}(A) = 90$  Millions (floating-point operations)  $Tu\_cpu(A) = 3.5$  seconds

$$(90 * 10^6)/(3.5 * 10^6) = 25.71 \text{ MFLOPS(A)}$$

# **Benchmarks**

# Why Do

- · How we Beingbmerkses?
  - –Different systems
  - -Changes to a single system

 Benchmarks represent large class of important programs

## **Benchmarks**

#### Microbenchmarks

- Measure one performance dimension or aspect
  - Cache bandwidth
  - Memory bandwidth
  - Procedure call overhead
  - FP performance
- Insight into the underlying performance factors
- Not a good predictor of overall application performance

#### Macrobenchmarks

- Application execution time
  - Measures overall performance, using one application
  - Need application suite

# **Popular Benchmark Suites**

#### Desktop

- SPEC CPU2000 CPU intensive, integer & floating-point applications
- SPECviewperf, SPECapc Graphics benchmarks
- SysMark, Winstone, Winbench

#### Embedded

- EEMBC Collection of kernels from 6 application areas
- Dhrystone Old synthetic benchmark

#### Servers

- SPECweb, SPECfs
- TPC-C Transaction processing system
- TPC-H, TPC-R Decision support system
- TPC-W Transactional web benchmark

#### Parallel Computers

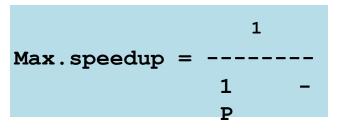
- SPLASH Scientific applications & kernels
- Linpack

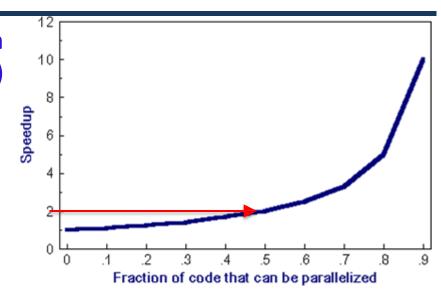
# <u>Performance Metrics – Parallel</u> <u>Systems</u>

# Amdahl's Law & Speedup Factor

# **Amdahl's Law**

□ Amdahl's Law states that potential program speedup is defined by the fraction of code (P) that can be parallelized:





- If none of the code can be parallelized, P = 0 and the speedup
  - = 1 (no speedup). If all of the code is parallelized, P = 1 and the speedup is infinite (in theory).
- If 50% of the code can be parallelized, maximum speedup

# **Amdahl's Law**

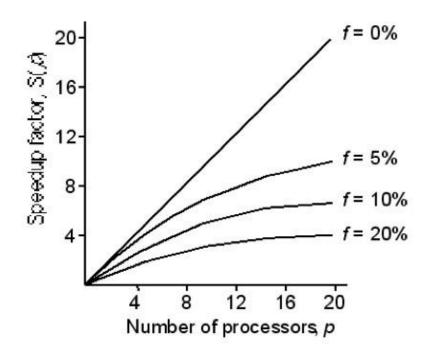
 It soon becomes obvious that there are limits to the scalability of parallelism

For example, at P = .50, .90 and .99 (50%, 90% and 99% of the code is parallelizable)

	speedup		
N	P = .50	P = .90	P = .99
10	1.82	5.26	9.17
100	1.98	9.17	50.25
1000	1.99	9.91	90.99
10000	1.99	9.91	99.02

# Maximum Speedup (Amdahl's Law)

### Speedup against number of processors



Even with infinite number of processors, maximum speedup limited to 1/f.

Example: With only 5% of computation being serial, maximum speedup is 20,

irrespective of number of processors.

f = serial fraction

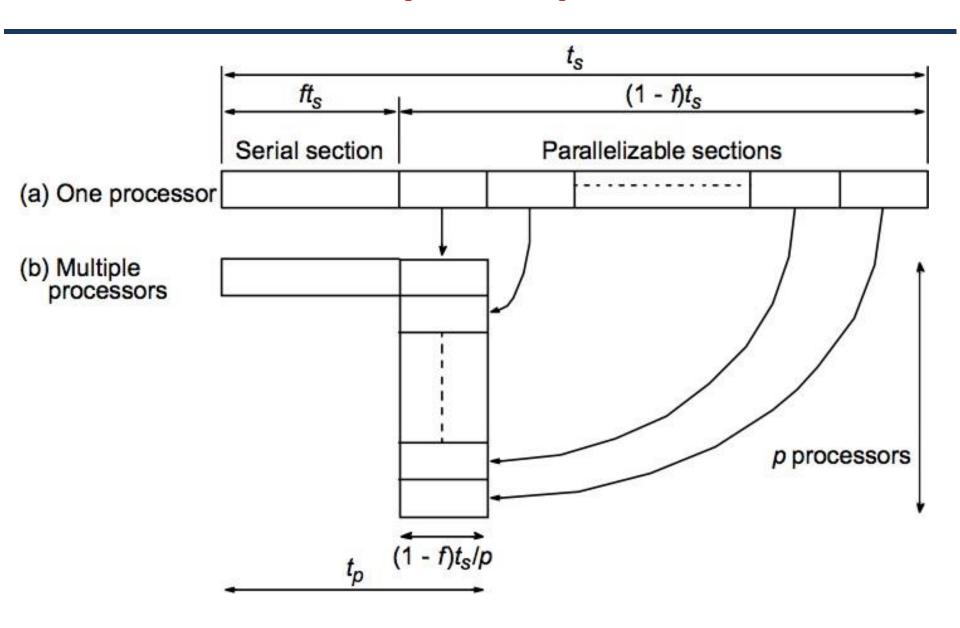
E.g., 1/0.05 (5% serial) = 20 speedup (maximum)

## Maximum Speedup (Amdahl's Law)

Maximum speedup is usually p with p processors (linear speedup).

Possible to get super-linear speedup (greater than

- **p**) but usually a specific reason such as:
  - Extra memory in multiprocessor system
  - Nondeterministic algorithm



$$S(p) = \frac{\text{Execution time using one processor (best sequential algorithm)}}{\text{Execution time using a multiprocessor with } p \text{ processors}} = \frac{t_s}{t_p}$$

where  $t_s$  is execution time on a single processor and  $t_p$  is execution time on a multiprocessor.

- S(p) gives increase in speed by using multiprocessor
- Use best sequential algorithm with single processor system instead of parallel program run with 1 processor for t<sub>s</sub>. Underlying algorithm for parallel implementation might be (and is usually) different.

**Speedup** can also be used in termsof computational steps:

 $S(p) = \frac{\text{Number of computational steps using one processor}}{\text{Number of parallel computational steps with } p \text{ processors}}$ 

Speedup factor is given by:

$$S(p) = \frac{t_s}{f + (1 - f)t_s/p} = \frac{p}{1 + (p - 1)f}$$

Here **f** is the part of the code that is serial:

e.g. if f==1 (all the code is serial, then the speedup will be 1 no matter how may processors are used

# Speedup (with N CPUs or Machines)

 Introducing the number of processors performing the parallel fraction of work, the relationship can be modelled by:

```
1
speedup = -----
---
fS + --fP-
Proc
```

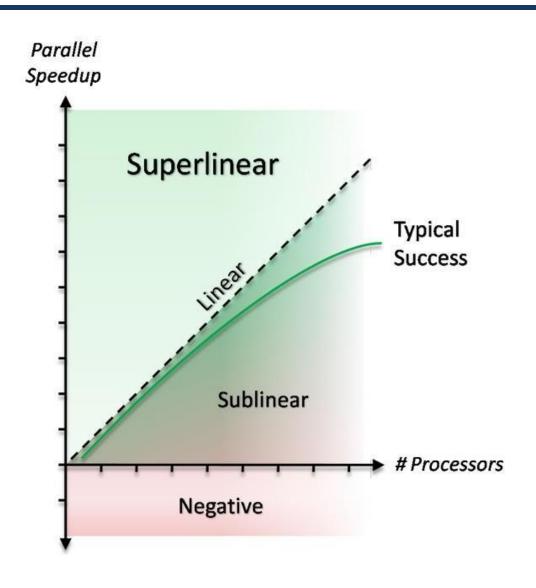
```
    where fP = parallel fraction,
    Proc = number of processors and
    fS = serial fraction
```

# Linear and Superlinear Speedup

- Linear speedup
  - Speedup of N, for N processors
  - Parallel program is perfectly scalable
  - Rarely achieved in practice

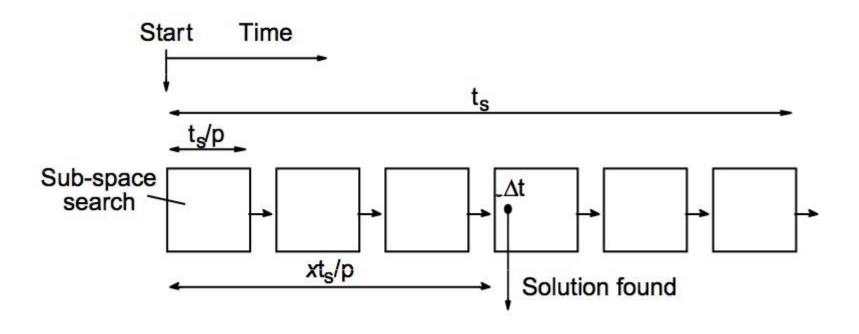
- Superlinear Speedup
  - Speedup of >N, for N processors
    - Theoretically not possible
    - How is this achievable on real machines?
      - Think about physical resources (cache, memory etc) of N processors

# **Super-linear Speedup**



## Super-linear Speedup Example -

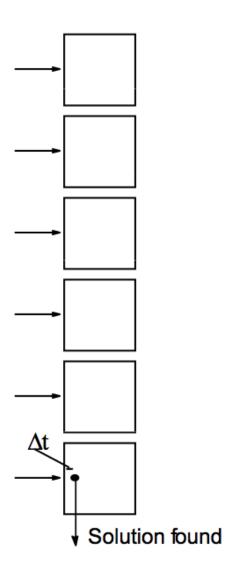
Searching
(a) Searching each sub-space sequentially



x indeterminate

## Super-linear Speedup Example -

Searching (b) Searching each sub-space in parallel



## Efficienc

- Efficiency is the ability to avoid wasting materials, energy, efforts, money, and time in doing something or in producing a desired result
- The ability to do things well, successfully, and without waste

# **Efficienc**

Y

### ° Efficiency:

$$E = \frac{\text{Speedup}}{\text{Number of Processors}}$$

# Speedups and Efficiencies of parallel program on different problem sizes

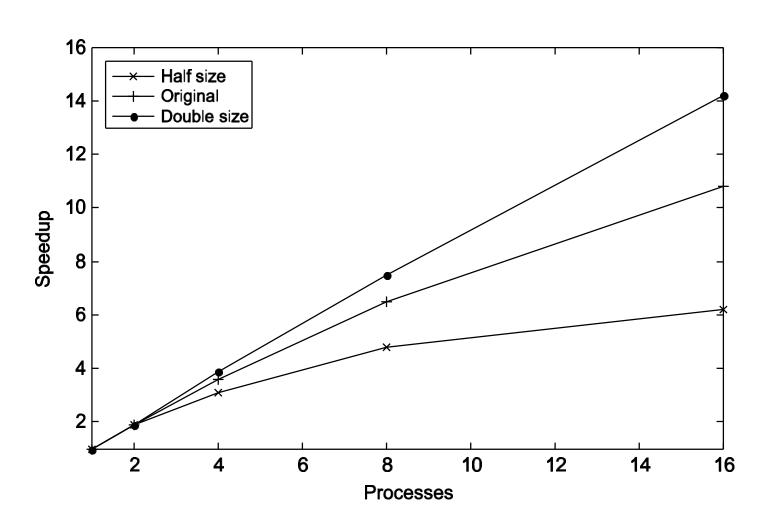
#### **Machine size (Processors)** 8 16 Half 3.1 4.8 S 1.9 6.2 1.0 $\boldsymbol{E}$ 0.95 0.78 0.390.60 1.0Original S 1.9 3.6 6.5 10.8**Problem size** 1.0E0.950.900.810.68Double S 1.0 3.9 7.5 1.9E0.950.94 0.98 0.891.0

S: Speedup

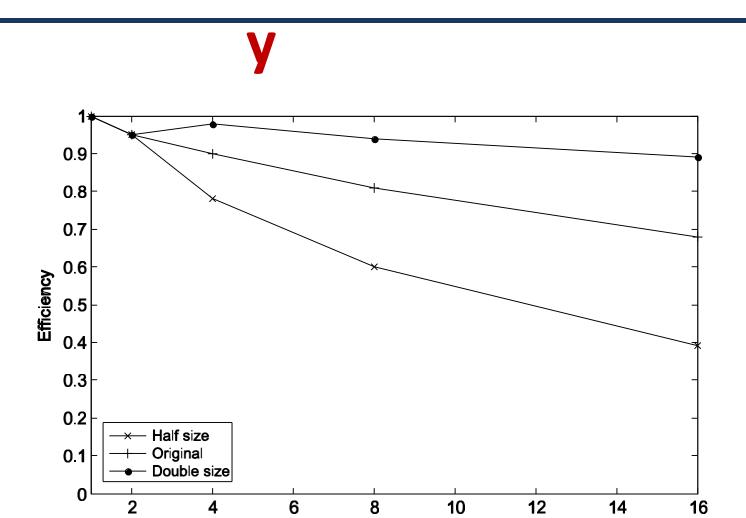
E:

Efficiency

## Speedup



## **Efficienc**



**Processes** 

# **Class Tasks**

#### Task-

Mr. Mohib Ullah is given a job in a software company as a senior software engineer. His boss has assigned him a project to parallelize the application named "Barnes-Hut". If he could successfully achieve his goals, he will get a bonus from the company. The objective set by his boss is that he must enhance the "Barnes-hut" application's performance 3.555 times as compared to its serial version. After analyzing the "Barnes-hut" application, he figured out that the size of the parallelizable code in the application is three-fourth (i.e., 3/4) of the application size.

#### Task:

Kindly suggest Mr. Mohib Ullah, How many processors based multi-core machine should be purchased to achieve the above-mentioned performance?

NOTE: You should justify (by providing arguments) with each step of the solution

#### Task-

Assume that Mr. Daud is working in a software company as a senior software developer. To get promoted as Project Manager, his boss has assigned him a code to parallelize and asked him to attain exactly the Speedup of 7.804 (from the parallel version of the code executed on a SMP machine having 32 processors). Now, help him to parallelize the code.

Mr Daud is analyzing different algorithms/versions of the code. Help him to select appropriate version of the code from which the required performance can be attained. Guide him about:

"How much parallelism (percentage of the parallelizable code) should be there in the program to attain the above mentioned speedup (i.e., 7.804) executed using 32 processors of the SMP machine?"

#### Task-

- (A) Consider a serial program with total execution time of 632 seconds. The program contains a portion of 95% that can be executed in parallel. If there are 16 processors available what will be the speedup of the parallel version of the program.
- (A) Calculate the efficiency of the parallel program mentioned in part (A)

# Gustafson's Law

#### Amdahl's law

- Amdah Syfficient on a fixed problem size
  - Shows how execution time decreases as number of processors increases
  - Limits maximum speedup achievable
  - So, does it mean large parallel machines are not useful?
  - Ignores performance overhead (e.g. communication, load imbalance)
- Gustafson's Law says that increase of problem size for large machines can retain scalability with respect to the number of processors

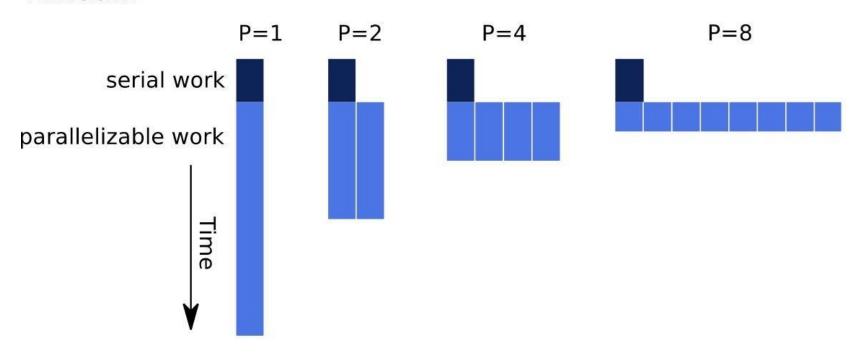
#### Gustafson's

- Time-constrained scaling (i.e., we have fixedtime to do performance analysis or execution)
- Example: a user wants more accurate results within a time limit
- Execution time is fixed as system scales

#### **Amdahl versus Gustafson's**

#### Law

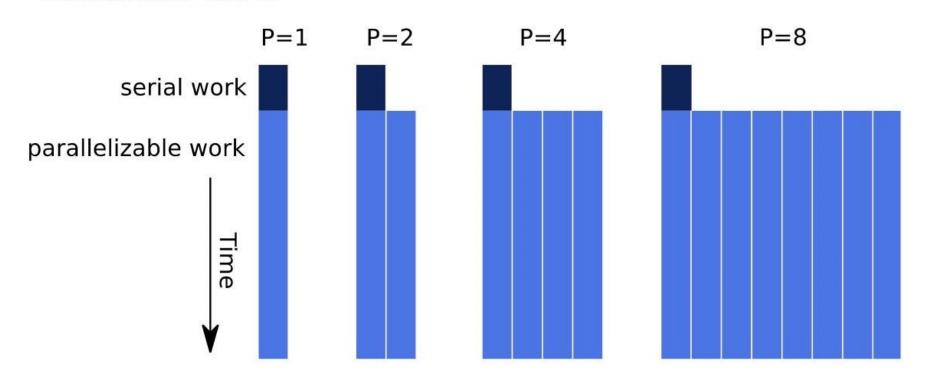
#### **Amdahl**



#### **Amdahl versus Gustafson's**

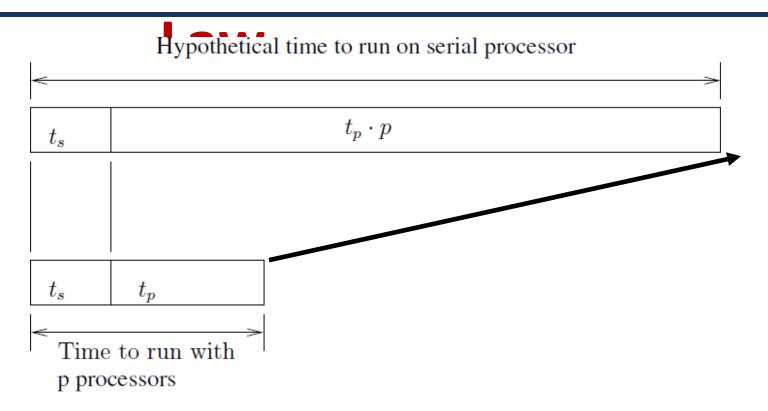
#### Law

**Gustafson-Baris** 



**Credits: University of Oregon** 

#### Gustafson's



- P processors, with increased number of processors the
  - problem size will also be increased
    - Importantly parallel part will be increased

#### Gustafson's

$$Speedup = \frac{Sequential\ execution\ time}{Parallel\ execution\ time}$$

$$\label{eq:Scaled Speedup} \begin{aligned} \text{Scaled speedup} &= \frac{\text{Hypothetical time to solve problem on sequential computer}}{\text{Actual parallel execution time}} \end{aligned}$$

$$S(p) = \frac{t_s + t_p p}{t_s + t_p} = \frac{t_s}{t_s + t_p} + \frac{t_p}{t_s + t_p} p$$

$$= s + (1 - s)p = s + p - ps = p + (1 - p)s$$

$$S(p) = p + (1 - p)s$$

where, S(p) Scaled Speedup, using P processors

s [] fraction of program that is serial (cannot be parallelized)

#### **Gustafson's Law-**

 An application running on 10 processors spends 3% of its time in serial code. What is the scaled speedup of the application?

$$S(p) = p + (1 - p)s$$

$$S(p) = 10 + (1 - 10) * (0.03) = 10 - 0.27 = 9.73$$
 (Scaled Speedup)

**Speedup Using Amdahl's Law?** 

$$S(p) = \frac{t_s}{f + (1 - f)t_s/p} = \frac{p}{1 + (p - 1)f}$$

#### **Gustafson's Law-**

#### Speedup Example ew

$$S(p) = \frac{t_s}{f + (1 - f)t_s/p}$$

$$OR \quad \frac{p}{1 + (p-1)f}$$

### **Summary Gustafson's**

 Derived by fixing the parallel execution time (Amdahl fixes the problem size -> fixed serial execution time)

- For many practical situations, Gustafson's law makes more sense
  - Have a bigger computer, solve a bigger problem

## Scalability

 In general, a problem is scalable if it can handle ever increasing problem sizes

 If we increase the number of processes/threads and keep the efficiency fixed without increasing problem size, the problem is strongly scalable.

If we keep the efficiency fixed by increasing the problem size at the same rate as we increase the number of processes/threads, the problem is weakly scalable.

# Any Questions