

## Welcome to

### CSE 326 Data Structures

## Staff

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- Instructor
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- TA's
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See web-page for office hours.

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## Web Page

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- All info is on the web page for CSE 326
  - › <http://www.cs.washington.edu/326>
  - › also known as
    - <http://www.cs.washington.edu/education/courses/326/04wi>

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## CSE 326 E-mail List

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- Subscribe by going to the class web page.
- E-mail list is used for posting announcements by instructor and TAs.
- It is your responsibility to subscribe. It might turn out to be very helpful for assignments hints, corrections etc.

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## Textbook

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- *Data Structures and Algorithm Analysis in Java (or in C++)*, by Weiss
- See Web page for errata and source code

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## Grading

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- Dry assignments 25% - submit in singles
- Wet assignments (programming projects) 25% - can submit in pairs.
- Midterm 20%
  - › Friday, Feb 6, 2004
- Final 30%
  - › Group I : 8:30-10:20 a.m. Thursday, Mar. 18, 2004
  - › Group II: 2:30-4:20 p.m. Tuesday, Mar. 16, 2004

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## Class Overview

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- Introduction to many of the basic data structures used in computer software
  - › Understand the data structures
  - › Analyze the algorithms that use them
  - › Know when to apply them
- Practice design and analysis of data structures.
- Practice using these data structures by writing programs.

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## Goal

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- You will understand
  - › what the tools are for storing and processing common data types
  - › which tools are appropriate for which need
- So that you will be able to
  - › make good design choices as a developer, project manager, or system customer

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## Course Topics

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- Introduction to Algorithm Analysis
- Lists, Stacks, Queues
- Search Algorithms and Trees
- Hashing and Heaps
- Sorting
- Disjoint Sets
- Graph Algorithms

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## Reading

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- Chapters 1 and 2, *Data Structures and Algorithm Analysis in Java*, by Weiss
  - › Most of Chapter 2 will be seen in class next week.

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## Data Structures: What?

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- Need to organize program data according to problem being solved
- Abstract Data Type (ADT) - A data object and a set of operations for manipulating it
  - › List ADT with operations `insert` and `delete`
  - › Stack ADT with operations `push` and `pop`
- Note similarity to Java classes
  - › private data structure and public methods

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## Data Structures: Why?

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- Program design depends crucially on how data is structured for use by the program
  - › Implementation of some operations may become easier or harder
  - › Speed of program may dramatically decrease or increase
  - › Memory used may increase or decrease
  - › Debugging may become easier or harder

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# Terminology

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- Abstract Data Type (ADT)
  - › Mathematical description of an object with set of operations on the object. Useful building block.
- Algorithm
  - › A high level, language independent, description of a step-by-step process
- Data structure
  - › A specific family of algorithms for implementing an abstract data type.
- Implementation of data structure
  - › A specific implementation in a specific language

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# Algorithm Analysis: Why?

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- Correctness:
  - › Does the algorithm do what is intended.
- Performance:
  - › What is the running time of the algorithm.
  - › How much storage does it consume.
- Different algorithms may correctly solve a given task
  - › Which should I use?

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# Evaluating an algorithm

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Mike: My algorithm can sort  $10^6$  numbers in 3 seconds.

Bill: My algorithm can sort  $10^6$  numbers in 5 seconds.

Mike: I've just tested it on my new Pentium IV processor.

Bill: I remember my result from my undergraduate studies (1985).

Mike: My input is a random permutation of  $1..10^6$ .

Bill: My input is the sorted output, so I only need to verify that it is sorted.

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# Program Evaluation / Complexity

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- Processing time is surely a bad measure!!!
- We need a 'stable' measure, independent of the implementation.
- \* A complexity function is a function  $T: N \rightarrow N$ .  
 $T(n)$  is the number of operations the algorithm does on an input of size  $n$ .
- \* We can measure three different things.
  - Worst-case complexity
  - Best-case complexity
  - Average-case complexity

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# The RAM Model of Computation

- Each simple operation takes 1 time step.
- Loops and subroutines are not simple operations.
- Each memory access takes one time step, and there is no shortage of memory.

For a given problem instance:

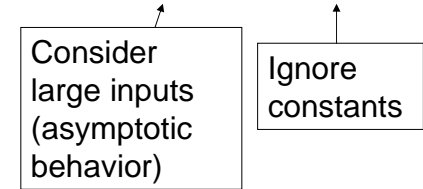
- Running time of an algorithm = # RAM steps.
- Space used by an algorithm = # RAM memory cells

useful abstraction  $\Rightarrow$  allows us to analyze algorithms in a machine independent fashion.

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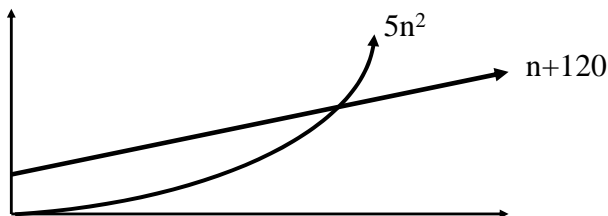
# Big O Notation

- Goal :
  - › A stable measurement independent of the machine.
- Way:
  - › ignore constant factors.
- $f(n) = O(g(n))$  if  $c \cdot g(n)$  is upper bound on  $f(n)$ 
  - $\Leftrightarrow$  There exist  $c, N$ , s.t. for any  $n \geq N$ ,  $f(n) \leq c \cdot g(n)$



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# Big O Notation



For all  $n \geq 5$  ( $N=5$ )  
 $n+120 \leq 5n^2$   
 $\Rightarrow n+120 = O(n^2)$

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# $\Omega$ , $\Theta$ Notation

- $f(n) = \Omega(g(n))$  if  $c \cdot g(n)$  is lower bound on  $f(n)$ 
  - $\Leftrightarrow$  There exist  $c, N$ , s.t. for any  $n \geq N$ ,  $f(n) \geq c \cdot g(n)$
- $f(n) = \Theta(g(n))$  if  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ 
  - $\Leftrightarrow$  There exist  $c_1, c_2, N$ , s.t. for  $n \geq N$ ,  
 $c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$

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## Ω, Θ Notation

Examples:

$$4x^2+100 = O(x^2)$$

$$4x^2+100 = \Omega(x^2)$$

$$4x^2+100 = \Theta(x^2)$$

$$4x^2 - 100 = O(x^2)$$

$$123400 = O(1)$$

$$4x^2+100 \neq \Theta(x^3)$$

$$4x^2+100 = O(x^3)$$

$$4x^2+100 = \Omega(x)$$

$$4x^2 + x \log x = O(x^2)$$

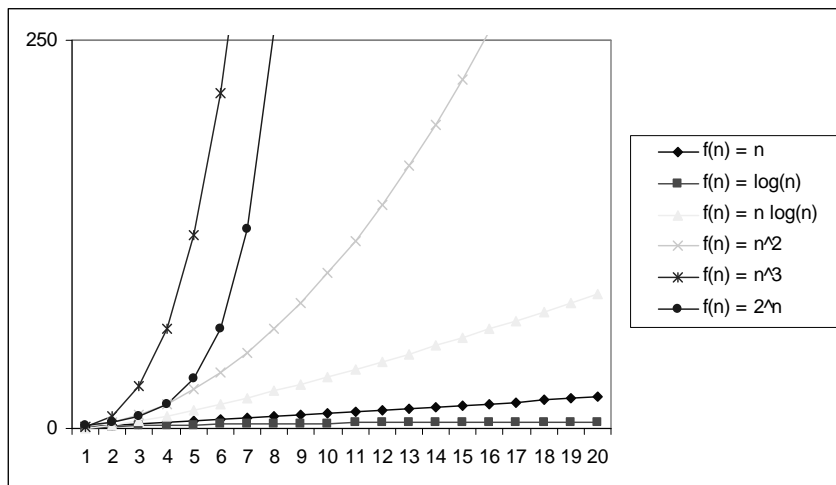
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## Growth Rates

- Even by ignoring constant factors, we can get an excellent idea of whether a given algorithm will be able to run in a reasonable amount of time on a problem of a given size.
- The “big O” notation and worst-case analysis are tools that greatly simplify our ability to compare the efficiency of algorithms.

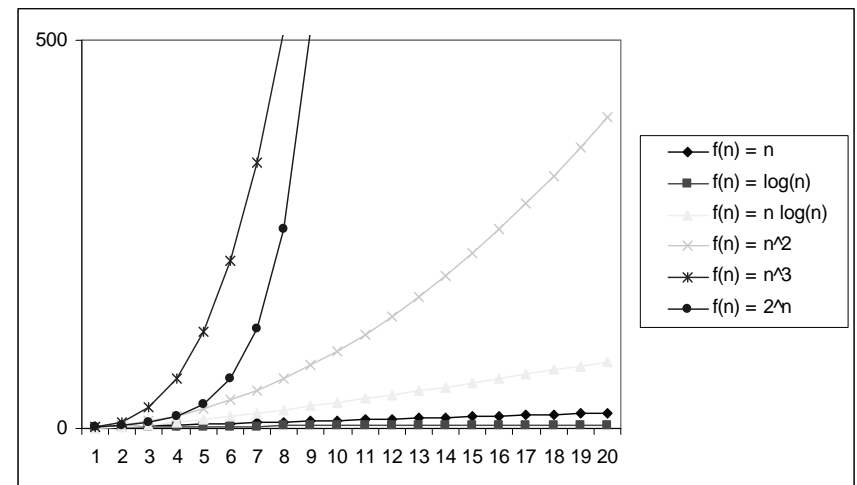
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## Practical Complexity



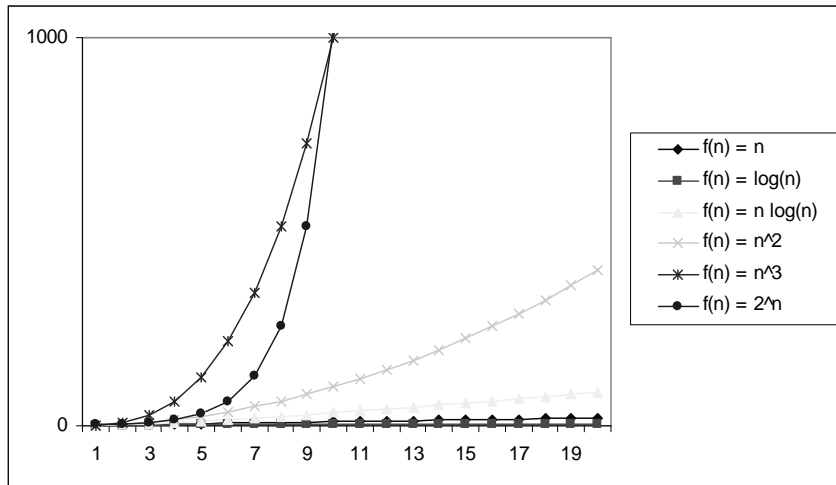
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## Practical Complexity



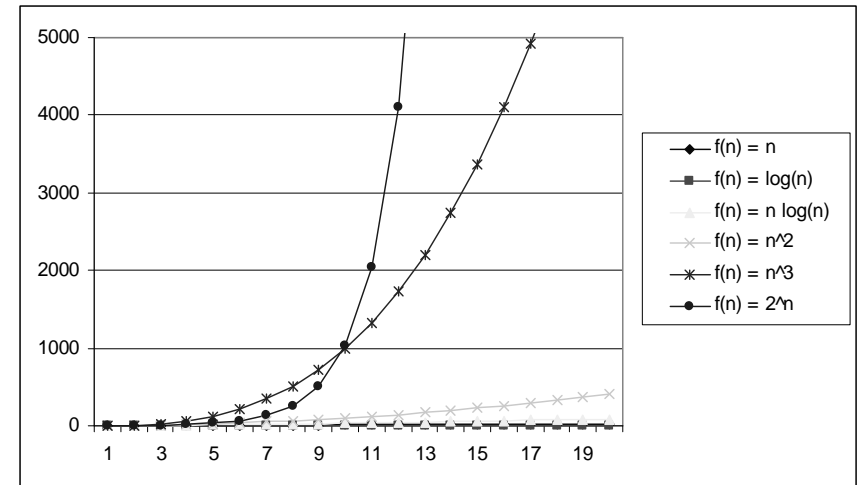
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# Practical Complexity



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# Practical Complexity



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## Big O Fact

- A polynomial of degree k is  $O(n^k)$
- Proof:
  - › Suppose  $f(n) = b_k n^k + b_{k-1} n^{k-1} + \dots + b_1 n + b_0$ 
    - Let  $a = \max_i \{b_i\}$
  - ›  $f(n) \leq a n^k + a n^{k-1} + \dots + a n + a$ 

$$\leq k a n^k \leq c n^k \quad (\text{for } c=ka).$$

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## Iterative Algorithm for Sum

- Find the sum of the first `num` integers stored in an array `v`.

```

sum(v[ ]: integer array, num: integer): integer{
    temp_sum: integer ;
    temp_sum := 0;
    for i := 0 to num - 1 do
        temp_sum := v[i] + temp_sum;
    return temp_sum;
}
    
```

Note the use of pseudocode

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## Programming via Recursion

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- Write a *recursive* function to find the sum of the first `num` integers stored in array `v`.

```
sum (v[ ]: integer array, num: integer): integer {
  if (num = 0) then
    return 0
  else
    return (v[num-1] + sum(v,num-1));
}
```

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## Pseudocode

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- In the lectures algorithms will be presented in pseudocode.
  - › This is very common in the computer science literature
  - › Pseudocode is usually easily translated to real code.
  - › This is programming language independent
- Pseudocode should also be used for homework (dry ones)

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## Review: Induction

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- Suppose
  - ›  $S(k)$  is true for fixed constant  $k$ 
    - Often  $k = 0$
  - ›  $S(n)$  implies  $S(n+1)$  for all  $n \geq k$
- Then  $S(n)$  is true for all  $n \geq k$

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## Proof By Induction

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- Claim:  $S(n)$  is true for all  $n \geq k$
- Base:
  - › Show  $S(n)$  is true for  $n = k$
- Inductive hypothesis:
  - › Assume  $S(n)$  is true for an arbitrary  $n$
- Step:
  - › Show that  $S(n)$  is then true for  $n+1$

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## Induction Example: Geometric Closed Form

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- Prove  $a^0 + a^1 + \dots + a^n = (a^{n+1} - 1)/(a - 1)$  for all  $a \neq 1$ 
  - › Basis: 1. show that  $a^0 = (a^{0+1} - 1)/(a - 1)$  :  
 $a^0 = 1 = (a^1 - 1)/(a - 1)$ . 2. Show true for  $n=2$ .
  - › Inductive hypothesis:
    - Assume  $a^0 + a^1 + \dots + a^n = (a^{n+1} - 1)/(a - 1)$
  - › Step (show true for  $n+1$ ):  
 $a^0 + a^1 + \dots + a^{n+1} = a^0 + a^1 + \dots + a^n + a^{n+1}$   
 $= (a^{n+1} - 1)/(a - 1) + a^{n+1} = (a^{n+1+1} - 1)/(a - 1)$

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## Program Correctness by Induction

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- **Basis Step:**  $\text{sum}(v,0) = 0$ . ü
- **Inductive Hypothesis (n=k):** Assume  $\text{sum}(v,k)$  correctly returns sum of first k elements of v, i.e.  $v[0]+v[1]+\dots+v[k-1]$
- **Inductive Step (n=k+1):**  $\text{sum}(v,n)$  returns  $v[k]+\text{sum}(v,k)$  which is the sum of first k+1 elements of v. ü

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## Algorithms vs Programs

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- Proving correctness of an algorithm is very important
  - › a well designed algorithm is guaranteed to work correctly and its performance can be estimated
- Proving correctness of a program (an implementation) is fraught with weird bugs
  - › Abstract Data Types are a way to bridge the gap between mathematical algorithms and programs

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