#### Asymptotic Analysis

CSE 373
Data Structures & Algorithms
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Spring 2007

## Today's Outline

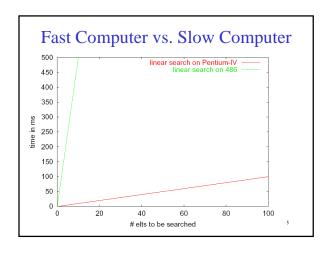
- **Admin**: Assignment #1 due next thurs. at 11:59pm
- · Asymptotic analysis

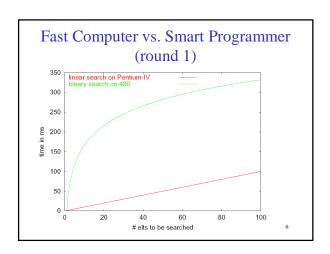
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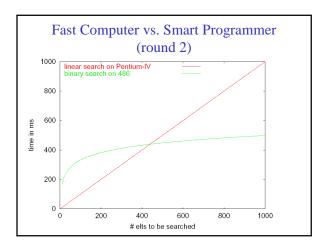
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# **Asymptotic Analysis**

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#### Asymptotic Analysis

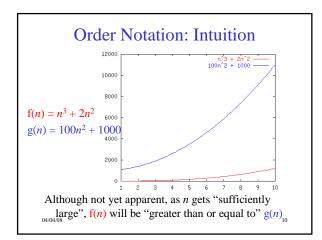
- · Asymptotic analysis looks at the order of the running time of the algorithm
  - A valuable tool when the input gets "large"
  - Ignores the effects of different machines or different implementations of the same algorithm
- Intuitively, to find the asymptotic runtime, throw away the constants and low-order terms
  - Linear search is  $T(n) = 3n + 2 \in O(n)$
  - Binary search is  $T(n) = 4 \log_2 n + 4$  ∈  $O(\log n)$

Remember: the fastest algorithm has the slowest growing function for its runtime

#### Asymptotic Analysis

- · Eliminate low order terms
  - $-4n+5 \Rightarrow$
  - $-0.5 \text{ n log n} + 2\text{n} + 7 \Rightarrow$
  - $-n^3 + 2^n + 3n \Longrightarrow$
- Eliminate coefficients
  - 4n ⇒
  - $-0.5 \text{ n log n} \Rightarrow$
  - $-\ n\ log\ n^2 =>$

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#### **Definition of Order Notation**

- Upper bound: T(n) = O(f(n))Big-O Exist constants c and n' such that

 $T(n) \le c f(n)$  for all  $n \ge n$ 

Lower bound:  $T(n) = \Omega(g(n))$ Omega

Exist constants c and n' such that

 $T(n) \ge c g(n)$  for all  $n \ge n$ 

Tight bound:  $T(n) = \theta(f(n))$ Theta

When both hold:

T(n) = O(f(n))

 $T(n) = \Omega(f(n))$ 

#### Order Notation: Definition

O(f(n)): a set or class of functions

 $g(n) \in O(f(n))$  iff there exist consts c and  $n_0$  such that:

 $g(n) \le c f(n)$  for all  $n \ge n_0$ 

Example:  $g(n) = 1000n \text{ vs. } f(n) = n^2$ 

Is  $g(n) \in O(f(n))$ ?

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Pick: n0 = 1000, c = 1

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#### **Notation Notes**

Note: Sometimes, you'll see the notation:

$$g(n) = O(f(n)).$$

This is equivalent to:

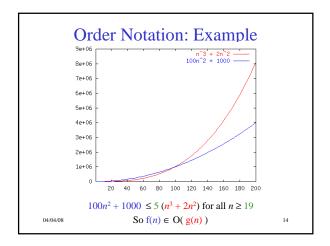
 $g(n) \in O(f(n)).$ 

However: The notation

O(f(n)) = g(n)is meaningless!

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(in other words big-O is not symmetric)



# **Big-O: Common Names**

- constant: O(1)

- logarithmic: O(log n)  $(\log_k n, \log n^2 \in O(\log n))$ 

- linear: O(n) - log-linear: O(n log n)

- quadratic:  $O(n^2)$ - cubic:  $O(n^3)$ 

– polynomial:  $O(n^k)$ (k is a constant) - exponential: O(cn) (c is a constant > 1)

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#### Meet the Family

- O(f(n)) is the set of all functions asymptotically less than or equal to f(n)
  - o( f(n) ) is the set of all functions asymptotically strictly less than f(n)
- $\Omega(f(n))$  is the set of all functions asymptotically greater than or equal to f(n)
  - $-\omega(f(n))$  is the set of all functions asymptotically strictly greater than f(n)
- $\theta(f(n))$  is the set of all functions asymptotically equal to f(n)

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#### Meet the Family, Formally

•  $g(n) \in O(f(n))$  iff

There exist c and  $n_0$  such that  $g(n) \le c f(n)$  for all  $n \ge n_0$ 

 $g(n) \in o(f(n))$  iff There exists a  $n_0$  such that g(n) < c f(n) for all c and  $n \ge n_0$ Equivalent to:  $\lim_{n\to\infty} g(n)/f(n) = 0$ 

 $g(n) \in \Omega(f(n))$  iff

There exist c and  $n_0$  such that  $g(n) \ge c$  f(n) for all  $n \ge n_0$ 

 $g(n) \in \omega(f(n))$  iff

There exists a  $n_0$  such that g(n) > c f(n) for all c and  $n \ge n_0$ 

Equivalent to:  $\lim_{n\to\infty} g(n)/f(n) = \infty$ 

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 $g(n) \in \Theta(f(n))$  iff

 $g(n) \in O(f(n))$  and  $g(n) \in \Omega(f(n))$ 

#### Big-Omega et al. Intuitively

Asymptotic Notation	Mathematics Relation
0	≤
Ω	≥
θ	=
0	<
ω	>

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# Pros and Cons of Asymptotic Analysis

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# Types of Analysis

Two orthogonal axes:

- bound flavor
  - upper bound (O, o)
  - lower bound (Ω, ω)
  - asymptotically tight  $(\theta)$

#### - analysis case

- · worst case (adversary)
- · average case
- best case
- · "amortized"

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## Algorithm Analysis Examples

• Consider the following program segment:

x:= 0; for i = 1 to N do for j = 1 to i do x := x + 1;

• What is the value of x at the end?

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## Analyzing the Loop

 Total number of times x is incremented is executed =

$$1+2+3+...=\sum_{i=1}^{N}i=\frac{N(N+1)}{2}$$

- Congratulations You've just analyzed your first program!
  - Running time of the program is proportional to N(N+1)/2 for all N
  - Big-O ??

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#### Which Function Grows Faster?

$$n^3 + 2n^2$$
 vs.  $100n^2 + 1000$ 

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# Which Function Grows Faster? $n^3 + 2n^2$ vs. $100n^2 + 1000$