## Data Structures and Algorithms

Spring 2009-2010

### Outline

Algorithm Analysis

## When the Inputs Get Bigger...

- Chief concern in algorithm analysis is determining the underlying running time and space requirements in the long term, when the input size grows larger
- Running-time is highly dependent on machine architecture, so analysis cannot be machine-specific
- To do this we introduce the order notation that describes the time and storage requirements only in terms of the algorithm and its parameters

### Order of Functions

- Which function is bigger,  $f(n) = 20 \cdot n$  or  $g(n) = 0.002 \cdot n^2$ ?
- Although f(n) is larger initially, the  $n^2$  of g(n) dominates eventually
- To help us quantify functions and to make the notion of comparison more precise we introduce four definitions

### O(n) – Big-Oh

#### Definition

T(n) = O(f(n)) if there are *constants c* and  $n_0$  so that  $T(n) \le cf(n)$  when  $n \ge n_0$ 

### T(n) is "less than or equal to" f(n)

That is, once  $n \ge n_0$ , T(n) is always less than or equal to some constant, c, times f(n).

# $\Omega(n)$ – Big-Omega

### **Definition**

 $T(n) = \Omega(f(n))$  if there are *constants c* and  $n_0$  so that  $T(n) \ge cf(n)$  when  $n \ge n_0$ 

T(n) is "greater than or equal to" g(n)Therefore, if f(n) = O(g(n)) then  $g(n) = \Omega(f(n))$ 

# $\Theta(n)$ – Big-Theta

#### **Definition**

$$T(n) = \Theta(h(n))$$
 if and only if  $T(n) = O(h(n))$  and  $T(n) = \Omega(h(n))$ 

T(n) is "behaves like" h(n); Alternative view of  $\Theta(h(n))$ :

$$\lim_{n\to\infty}\frac{T(n)}{h(n)}\to C$$

## o(n) — Little-oh

### Definition

$$T(n) = o(p(n))$$
 if  $T(n) = O(p(n))$  and  $T(n) \neq \Theta(p(n))$ 

T(n) is "is smaller than" p(n); Alternative view of o(p(n)):

$$\lim_{n\to\infty}\frac{T(n)}{p(n)}\to 0$$

# **Function Comparisons**

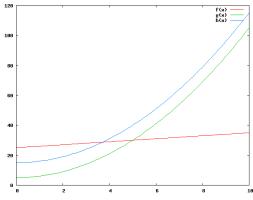
### Let

• 
$$f(n) = n + 25$$

• 
$$g(n) = n^2 + 5$$
, and

• 
$$h(n) = n^2 + 15$$

Then, we can say:



• 
$$f(n) = O(g(n))$$
 and  $f(n) = O(h(n))$ 

• 
$$g(n) = \Omega(f(n))$$
 and  $h(n) = \Omega(f(n))$ 

• 
$$g(n) = \Theta(h(n))$$
 and therefore,  $h(n) = \Theta(g(n))$ 

• 
$$f(n) = o(g(n))$$



# More on $\Theta(n)$

### With

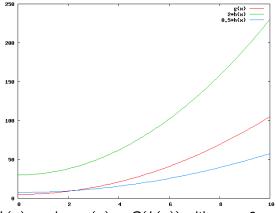
• 
$$g(n) = n^2 + 5$$

• 
$$h(n) = n^2 + 15$$

- $\bullet$  2 · h(n)
- $\bullet$   $\frac{1}{2} \cdot h(n)$
- so we can say thata(n) -

$$g(n) =$$

$$\Theta(h(n))$$



For 
$$n > 0$$
,  $g(n) \le 2 \cdot h(n)$ , and so  $g(n) = O(h(n))$  with  $n_0 = 0$  and  $c = 2$ :

similarly, for 
$$n \ge 3$$
,  $g(n) \ge 0.5 \cdot h(n)$ , and so  $g(n) = \Omega(h(n))$ 

with  $n_0 = 3$  and c = 0.5;



# Even more on $\Theta(n)$

Likewise, from below we can argue that  $g(n) = \Theta(n^2)$ 

• 
$$g(n) = n^2 + 5$$

• 
$$h(n) = n^2$$

- 2 · h(n)
- $\bullet$   $\frac{1}{2} \cdot h(n)$
- so we can say that

$$g(n) = \Theta(n^2)$$

