#### CSE 250 Data Structures

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### Lec 05: Asymptotic Analysis

### **Announcements and Feedback**

- Normal recitations begin next week
- PA0 due Sunday @ 11:59PM
- WA1 due Sunday @ 11:59PM
- Be mindful of Office Hour changes for next week

### **Analysis Checklist**

- 1. Don't think in terms of wall-time, think in terms of "number of steps"
- 2. To give a useful solution, we should take "scale" into account
  - How does the runtime change as we change the size of the input?
- 3. Focus on "large" inputs
  - Rank functions based on how they behave at large scales
- 4. Decouple algorithm from infrastructure/implementation

### Attempt #1: Wall-clock time?

- What is fast?
  - o 10s? 100ms? 10ns?
  - ...it depends on the task
- Algorithm vs Implementation
  - Compare Grace Hopper's implementation to yours
- What machine are you running on?
  - Your old laptop? A lab machine? The newest, shiniest processor?
- What bottlenecks exist? CPU vs IO vs Memory vs Network...

#### Wall-clock time is not terribly useful... 4

#### **Attempt #2: Growth Functions**

Not a function in code...but a mathematical function:

#### **T(n)**

#### n: The "size" of the input

ie: number of users,rows, pixels, etc

T(n): The number of "steps" taken for input of size n

ie: 20 steps per user, where n = |Users|, is 20 x n

#### Attempt #3: Asymptotic Analysis

# We want to organize runtimes (growth functions) into different *Complexity Classes*

Within the same complexity class, runtimes "behave the same"/"have the same shape"

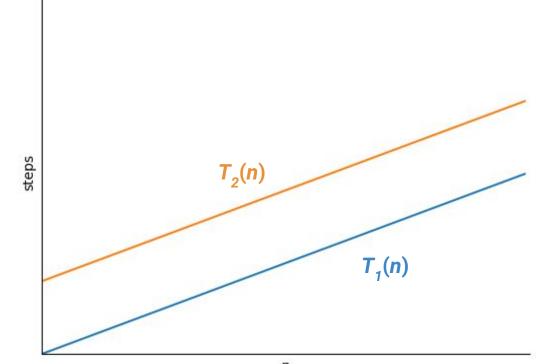
### **Getting More Formal**

When do we consider two functions to have the same shape?

# Additive Factors

Consider two growth functions:

 $T_1(n) = 3n$  $T_2(n) = 3n + 3$ 

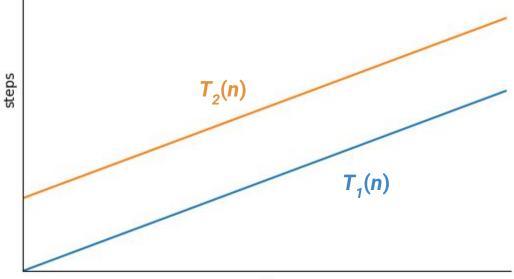


n

# Additive Factors

Consider two growth functions:

 $T_1(n) = 3n$  $T_2(n) = 3n + 3$  These functions still have the same shape...the same complexity

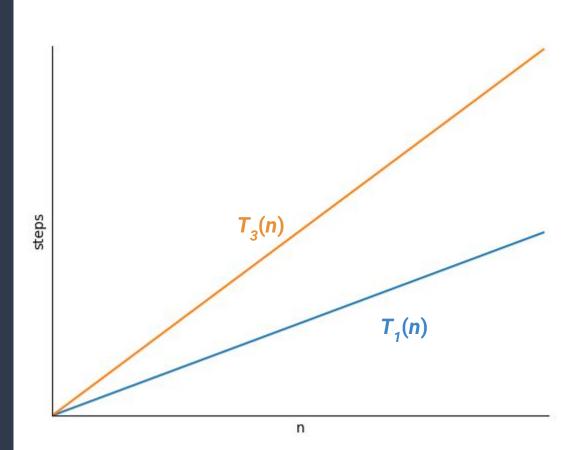


n

# Multiplicative Factors

Consider two growth functions:

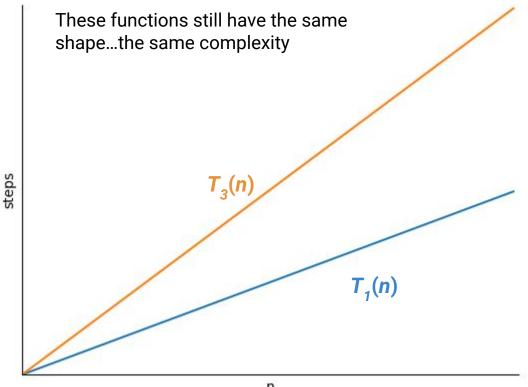
 $T_{1}(n) = 3n$  $T_{3}(n) = 6n$ 



# Multiplicative Factors

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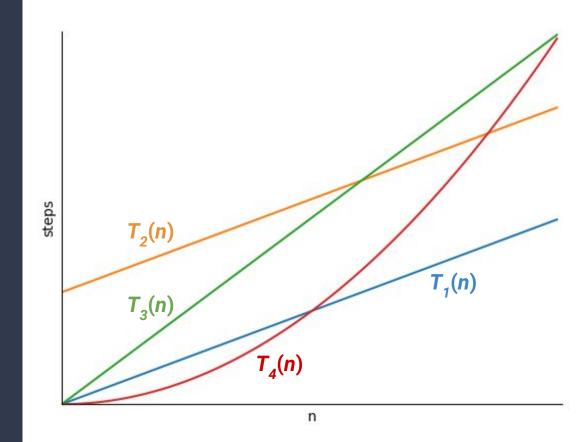


n

# A Counter Example

Now consider:

 $\overline{T}_4(n) = n^2$ 



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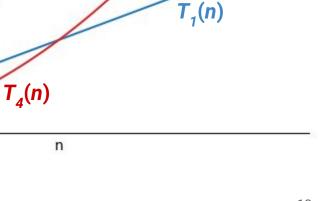
 $T_4(n) = n^2$ 

 $T_4$  is a distinctly different shape. Notice that no matter what constant factors we add or multiply by,  $T_4$  will **always** outgrow  $T_1$ ,  $T_2$ ,  $T_3$ 

steps

 $T_2(n)$ 

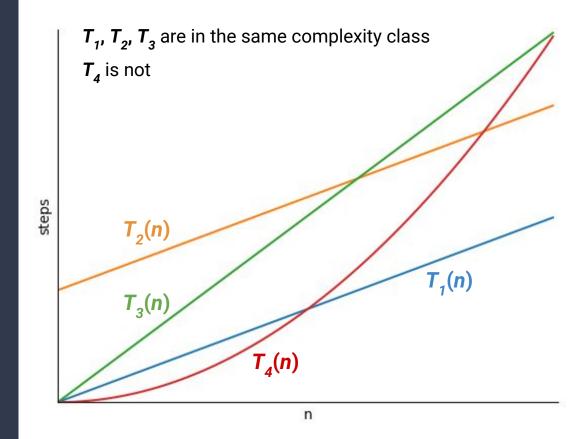
 $T_3(n)$ 



# A Counter Example

Now consider:

 $T_4(n) = n^2$ 



### Complexity (so far...)

If there are constants  $c_1$  and  $c_2$  such that:

$$T_1(n) = c_1 + c_2 T_2(n)$$

then we say  $T_1$  and  $T_2$  are in the same complexity class\*

\* not a complete definition...but we are getting there

#### **Back To Growth Functions**

So what exactly counts as a step?

### **Back To Growth Functions**

So what exactly counts as a step?

- An arithmetic operation
- Accessing a variable
- Printing to the screen
- etc

but...

#### How many steps in each of these snippets?

|--|

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1	x = 10;
<b>T</b> <sub>1</sub> ( <b>n</b> )	= 1
1 2	x = 10; y = 20;

#### How many steps in each of these snippets?

1	x = 10;			
$T_{1}(n) = 1$				
1	x = 10;			
2	x = 10; y = 20;			

 $T_2(n) = 2$ 

#### How many steps in each of these snippets?

1	x = 10;
$T_1(n)$	= 1
1	x = 10; y = 20;
2	y = 20;

 $T_2(n) = 2$ 

$$T_2(n) = T_1(n) + 1$$

They are in the same complexity class...in 250 we treat them as the same 21

A step therefore is any code that always has the same runtime

### **Notation - Big Theta**

 $\Theta(f(n))$  is the set of all functions in the same complexity class as f

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Example: \Theta(3n + 4) = \{

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15n,

...
```

 $g(n) \in \Theta(f(n))$  means g and f are in the same complexity class

 $g(n) = \Theta(f(n))$  is common shorthand for  $g(n) \in \Theta(f(n))$ 

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Algorithm Foo is in  $\Theta(f(n))$  is common shorthand for  $T(n) \in \Theta(f(n))$  where T(n) is the growth function describing the runtime of Foo

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Algorithm Foo is in  $\Theta(f(n))$  is common shorthand for  $T(n) \in \Theta(f(n))$  where T(n) is the growth function describing the runtime of Foo

Moving forward: **f**(**n**), **g**(**n**), **f**<sub>1</sub>(**n**), etc will be used to name any mathematical function that's a growth function T(n),  $T_1(n)$ , etc will be used for growth functions for specific algorithms

## **Complexity Class** Names

**Θ(1):** Constant

**O(log(n)):** Logarithmic

**Θ(***n***)**: Linear

**⊖**(*n* log(*n*)): Log-Linear

**Θ**(*n*<sup>2</sup>): Quadratic

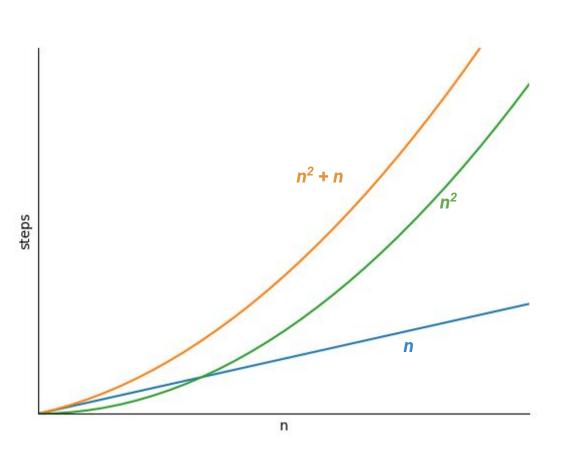
**⊖**(*n<sup>k</sup>*): Polynomial

**\Theta(2^n):** Exponential

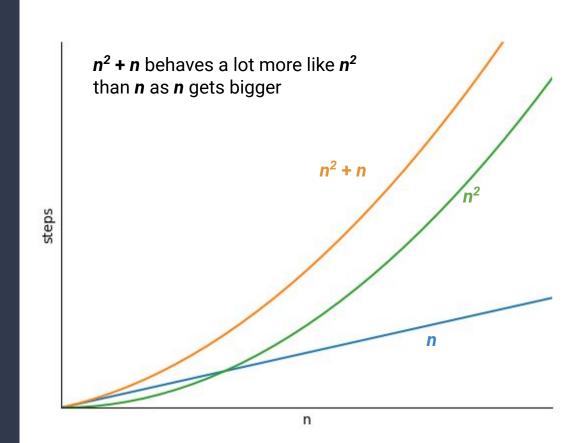
#### **Combining Classes**

#### What complexity class is $g(n) = n + n^2$ in?

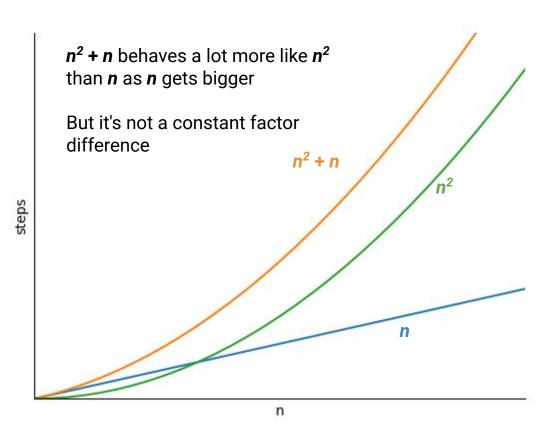
# Combining Classes



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# Combining Classes



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### **Combining Classes**

Consider the fact that  $n^2$  and  $2n^2$  are in the same complexity class... How does  $n^2 + n$  relate to these two functions?

### **Combining Classes**

#### Consider the fact that $n^2$ and $2n^2$ are in the same complexity class...

#### 1 ≤ *n*

remember, we only care about problems with non-negative input sizes

#### Consider the fact that $n^2$ and $2n^2$ are in the same complexity class...

#### 1 ≤ *n*

### $n \le n^2$ multiply both sides by n

#### Consider the fact that $n^2$ and $2n^2$ are in the same complexity class...

#### 1 ≤ n

 $n \le n^2$  $n + n^2 \le 2n^2$  add  $n^2$  to both sides

### Consider the fact that $n^2$ and $2n^2$ are in the same complexity class...

#### $0 \le n$ obviously true

## Consider the fact that $n^2$ and $2n^2$ are in the same complexity class...

### 0 ≤ n

 $n^2 \le n + n^2$  add  $n^2$  to both sides

#### Consider the fact that $n^2$ and $2n^2$ are in the same complexity class...

#### $n^2 \leq n + n^2 \leq 2n^2$

#### So $n^2 + n$ should probably be in $\Theta(n^2)$ too...

**f** and **g** are in the same complexity class iff:

**g** is bounded from above by something **f**-shaped

and

**f** and **g** are in the same complexity class iff:

**g** is bounded from above by something **f**-shaped

and

g is bounded from below by something f-shaped

**f** shifted or stretched by a constant factor

**f** and **g** are in the same complexity class iff:

g is **bounded from above** by something **f**-shaped

and

g is **bounded from below** by something **f**-shaped

What do we mean by bounded from above/below?

# Bounding from Above: Big O

**g**(**n**) is bounded from above by **f**(**n**) if:

There exists a constant **n**<sub>o</sub> **> 0** and a constant **c > 0** such that:

For all  $n > n_0, g(n) \le c \cdot f(n)$ 

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In this case, we say that  $g(n) \in O(f(n))$ 

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**f** and **g** are in the same complexity class iff:

**g** is bounded from above by something **f**-shaped

and

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

**g** is bounded from above by something **f**-shaped

and

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

and

 $g(n) \in \Theta(f(n))$  iff:  $g(n) \in O(f(n))$ and  $g(n) \in \Omega(f(n))$ 

#### $\Theta(1) < \Theta(\log(n)) < \Theta(n) < \Theta(n \log(n)) < \Theta(n^2) < \Theta(n^3) < \Theta(2^n)$

 $O(1) \subset O(\log(n)) \subset O(n) \subset O(n \log(n)) \subset O(n^2) \subset O(n^3) \subset O(2^n)$ 

 $\Omega(2^n) \subset \Omega(n^3) \subset \Omega(n^2) \subset \Omega(n \log(n)) \subset \Omega(n) \subset \Omega(\log(n)) \subset \Omega(1)$ 

If something is bounded from above by  $\log(n)$ , it's also bounded from above by n $O(1) \subset O(\log(n)) \subset O(n) \subset O(n \log(n)) \subset O(n^2) \subset O(n^3) \subset O(2^n)$ 

### $\Omega(2^n) \subset \Omega(n^3) \subset \Omega(n^2) \subset \Omega(n \log(n)) \subset \Omega(n) \subset \Omega(\log(n)) \subset \Omega(1)$

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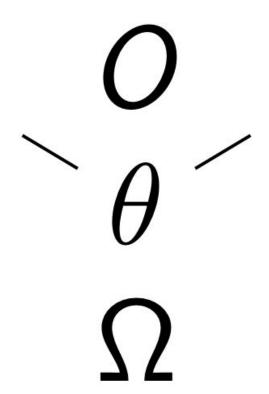
 $\Omega(2^n) \subset \Omega(n^3) \subset \Omega(n^2) \subset \Omega(n \log(n)) \subset \Omega(n) \subset \Omega(\log(n)) \subset \Omega(1)$ 

If something is bounded from below by  $n^2$ , it's also bounded from below by n

O(f(n)) (Big-O): The complexity class of f(n) and every lesser class

**Θ**(*f*(*n*)) (**Big-**Θ): The complexity class of *f*(*n*)

 $\Omega(f(n))$  (**Big-** $\Omega$ ): The complexity class of f(n) and every greater class



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