

# CS206 Data Structures

## Analysis Tools: Part 2

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# Class Objectives

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- Know the definition of big-Oh notation with big-Omega and big-Theta
- Can perform asymptotic analysis

# Big-Oh and Growth Rate

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- The big-Oh notation gives an upper bound on the growth rate of a function
- The statement " $f(n)$  is  $O(g(n))$ " means that the growth rate of  $f(n)$  is no more than the growth rate of  $g(n)$
- We can use the big-Oh notation to rank functions according to their growth rate

	$f(n)$ is $O(g(n))$	$g(n)$ is $O(f(n))$
$g(n)$ grows more	Yes	No
$f(n)$ grows more	No	Yes
Same growth	Yes	Yes

# Big-Oh Rules

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- If  $f(n)$  is a polynomial of degree  $d$ , then  $f(n)$  is  $O(n^d)$ , i.e.,
  1. Drop lower-order terms
  2. Drop constant factors
- Use the smallest possible class of functions
  - Say " $2n$  is  $O(n)$ " instead of " $2n$  is  $O(n^2)$ " (although the latter is still correct)
- Use the simplest expression of the class
  - Say " $3n + 5$  is  $O(n)$ " instead of " $3n + 5$  is  $O(3n)$ "

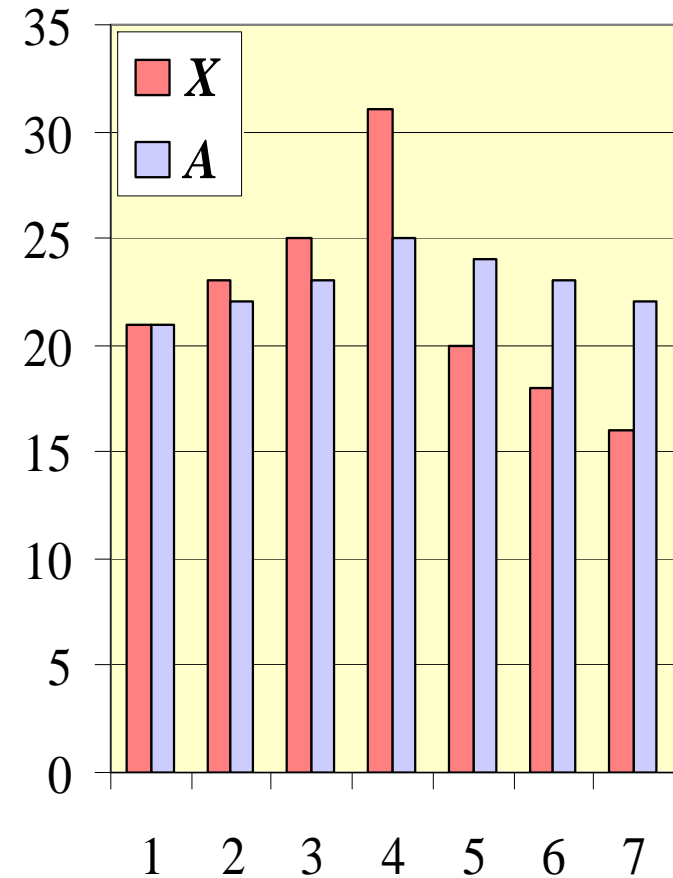
# Asymptotic Algorithm Analysis

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- The asymptotic analysis of an algorithm determines the running time in big-Oh notation
- To perform the asymptotic analysis
  - We find the worst-case number of primitive operations executed as a function of the input size
  - We express this function with big-Oh notation
- Example:
  - We determine that algorithm `arrayMax` executes at most  $8n - 2$  primitive operations
  - We say that algorithm `arrayMax` “runs in  $O(n)$  time”
- Since constant factors and lower-order terms are eventually dropped anyhow, we can disregard them when counting primitive operations

# Computing Prefix Averages

- We further illustrate asymptotic analysis with two algorithms for prefix averages
- The  $i$ -th prefix average of an array  $X$  is average of the first  $(i + 1)$  elements of  $X$ :  
$$A[i] = (X[0] + X[1] + \dots + X[i]) / (i+1)$$
- Computing the array  $A$  of prefix averages of another array  $X$  has applications to financial analysis



# Prefix Averages (Quadratic)

- The following algorithm computes prefix averages in quadratic time by applying the definition

Algorithm *prefixAverages1*( $X, n$ )

Input array  $X$  of  $n$  integers

Output array  $A$  of prefix averages of  $X$

$A \leftarrow$  new array of  $n$  integers

for  $i \leftarrow 0$  to  $n - 1$  do

$s \leftarrow X[0]$

    for  $j \leftarrow 1$  to  $i$  do

$s \leftarrow s + X[j]$

$A[i] \leftarrow s / (i + 1)$

return  $A$

#operations

$n$

$n$

$n$

$1 + 2 + \dots + (n - 1)$

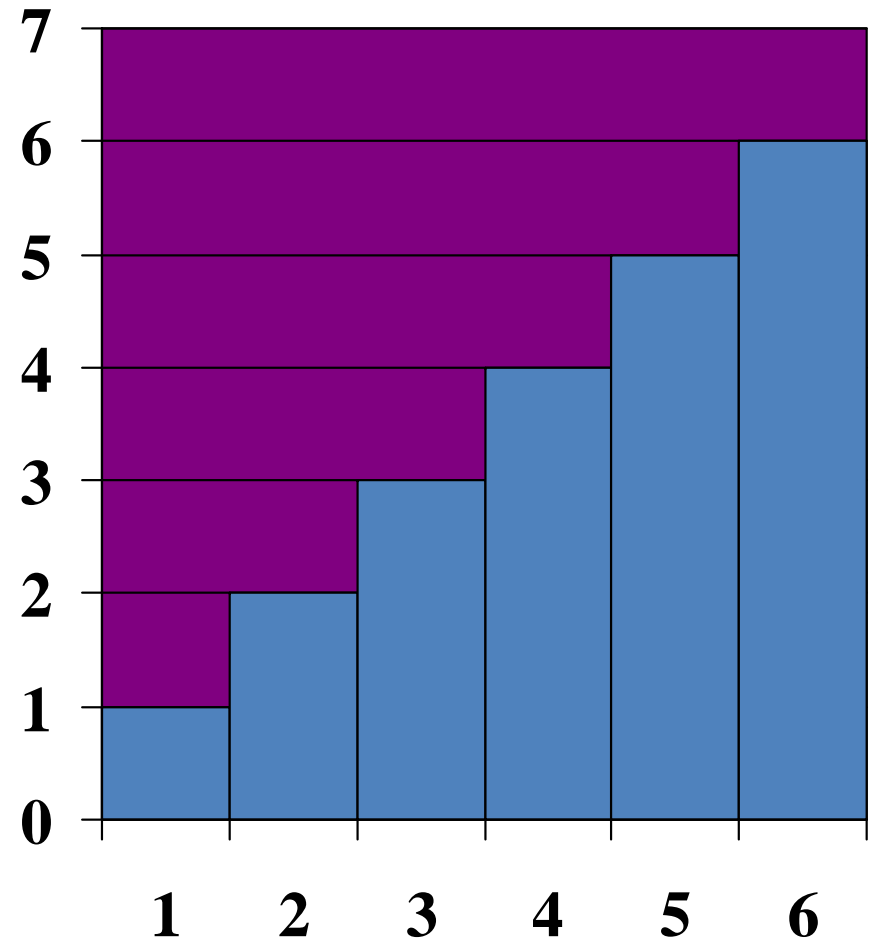
$1 + 2 + \dots + (n - 1)$

$n$

1

# Arithmetic Progression

- The running time of *prefixAverages1* is  $O(1 + 2 + \dots + n)$
- The sum of the first  $n$  integers is  $n(n + 1) / 2$ 
  - There is a simple visual proof of this fact
- Thus, algorithm *prefixAverages1* runs in  $O(n^2)$  time





# Prefix Averages (Linear)

- The following algorithm computes prefix averages in linear time by keeping a running sum

Algorithm *prefixAverages2*( $X, n$ )

**Input** array  $X$  of  $n$  integers

**Output** array  $A$  of prefix averages of  $X$

$A \leftarrow$  new array of  $n$  integers

$s \leftarrow 0$

**for**  $i \leftarrow 0$  to  $n - 1$  **do**

$s \leftarrow s + X[i]$

$A[i] \leftarrow s / (i + 1)$

**return**  $A$

#operations

$n$

1

$n$

$n$

$n$

1

- Algorithm *prefixAverages2* runs in  $O(n)$  time

# Math Tools for Analysis of Algorithms

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Summations

Logarithms and Exponents

- **properties of logarithms:**

$$\log_b(xy) = \log_b x + \log_b y$$

$$\log_b(x/y) = \log_b x - \log_b y$$

$$\log_b x^a = a \log_b x$$

$$\log_b a = \log_x a / \log_x b$$

- **properties of exponentials:**

$$a^{(b+c)} = a^b a^c$$

$$a^{bc} = (a^b)^c$$

$$a^b / a^c = a^{(b-c)}$$

$$b = a^{\log_a b}$$

$$b^c = a^{c \cdot \log_a b}$$

Proof techniques

Basic probability

# Relatives of Big-Oh

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## □ big-Omega

- $f(n)$  is  $\Omega(g(n))$  if there is a constant  $c > 0$  and an integer constant  $n_0 \geq 1$  such that  $f(n) \geq c \cdot g(n)$  for  $n \geq n_0$

## □ big-Theta

- $f(n)$  is  $\Theta(g(n))$  if there are constants  $c' > 0$  and  $c'' > 0$  and an integer constant  $n_0 \geq 1$  such that  $c' \cdot g(n) \leq f(n) \leq c'' \cdot g(n)$  for  $n \geq n_0$

# Intuition for Asymptotic Notation

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## □ big-Oh

- $f(n)$  is  $O(g(n))$  if  $f(n)$  is asymptotically *less than or equal to*  $g(n)$

## □ big-Omega

- $f(n)$  is  $\Omega(g(n))$  if  $f(n)$  is asymptotically *greater than or equal to*  $g(n)$

## □ big-Theta

- $f(n)$  is  $\Theta(g(n))$  if  $f(n)$  is asymptotically *equal to*  $g(n)$

# Example Uses of the Relatives of Big-Oh

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## □ $5n^2$ is $\Omega(n^2)$

- $f(n)$  is  $\Omega(g(n))$  if there is a constant  $c > 0$  and an integer constant  $n_0 \geq 1$  such that  $f(n) \geq c \cdot g(n)$  for  $n \geq n_0$
- let  $c = 5$  and  $n_0 = 1$

## □ $5n^2$ is $\Omega(n)$

- $f(n)$  is  $\Omega(g(n))$  if there is a constant  $c > 0$  and an integer constant  $n_0 \geq 1$  such that  $f(n) \geq c \cdot g(n)$  for  $n \geq n_0$
- let  $c = 1$  and  $n_0 = 1$

## □ $5n^2$ is $\Theta(n^2)$

- $f(n)$  is  $\Theta(g(n))$  if it is  $\Omega(n^2)$  and  $O(n^2)$ .  
We have already seen the former, for the latter recall that  $f(n)$  is  $O(g(n))$  if there is a constant  $c > 0$  and an integer constant  $n_0 \geq 1$  such that  $f(n) < c \cdot g(n)$  for  $n \geq n_0$
- Let  $c = 5$  and  $n_0 = 1$

# Class Objectives were:

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- Know the definition of big-Oh notation with big-Omega and big-Theta
- Can perform asymptotic analysis

# Next Time

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Stacks and queues

HW:

- Go over the next lecture slides before the class
- Just 10 min ~ 20 min should be okay

# Any Questions?

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- Come up with one question on what we have discussed in the class and submit at the end of the class
  - 1 for typical questions
  - 2 for questions with thoughts or that surprised me
  
- Write questions at least 4 times
  
- You can type KLMS