

# Algorithm Analysis I: Intro

## CSE 373 Winter 2020

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# Announcements

- ❖ Schedule for Drop-in Times (finally!!!! 🤓) posted

# Questions from Reading Quiz

- ❖ Do we care about the specific operation counts (eg calculating 2 to 50,015,001)? What if we are off by one?
- ❖ What is the conversion between the number of operations and the runtime? I had a little trouble understanding what each "step" was and what counted
- ❖ Could we go over how to count the operations for dup1?

# Lecture Outline

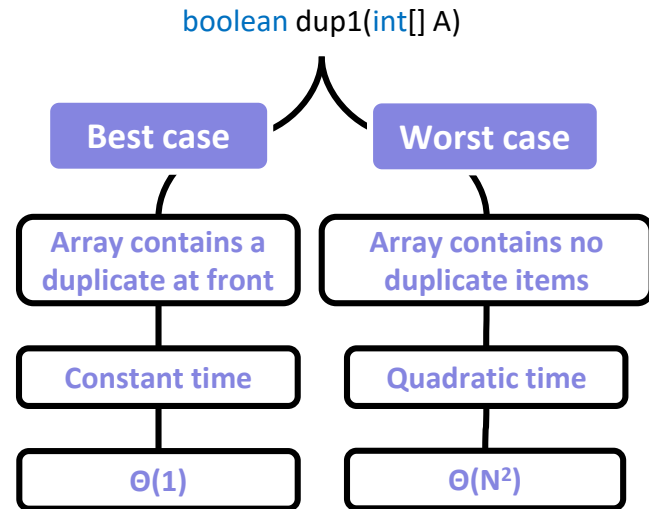
- ❖ **Algorithm Analysis Concepts**
- ❖ Asymptotic Analysis Case Study: dup1 Worst Case
- ❖ Asymptotic Analysis Simplifications
- ❖ Case Analysis vs Asymptotic Analysis
- ❖ Takeaways

# Review: What is Runtime Analysis?

- ❖ What does it mean for something to be “slow” or “fast”?
  
- ❖ Let’s run it and measure the (wallclock) time! Oh wait ...
  - Input can affect time
  - Hardware can affect time
  - Other programs running on the machine can affect time
  - ... and so much more!
  
- ❖ Count how many steps a program takes to execute on an input of size  $N$

# Algorithm Analysis: Our Destination

- ❖ **Comprehension**: Understanding the implementation
- ❖ **Cost Model**: Calculating the cost in terms of  $N$ , the size of the input
- ❖ **Analysis Types**:
  - **Case Analysis**. How certain conditions affect the program execution
  - **Asymptotic Analysis**: What happens as  $N \rightarrow \infty$
- ❖ **Formalizing**. Summarizing the final result in precise English or a mathematical notation



# Comprehension

- ❖ What is the data structure or algorithm doing?

```
public static boolean dup1(int[] A) {
    for (int i = 0; i < A.length; i += 1) {
        for (int j = i + 1; j < A.length; j += 1) {
            if (A[i] == A[j]) {
                return true;
            }
        }
    }
    return false;
}
```

# Cost Model: Time

- ❖ An algorithm's runtime is dictated by the number of CPU instructions it executes. This is why we “count steps”: each “step” is an idealized instruction
  - Example CPU instructions: comparing integers (<, ==), incrementing numbers (+, \*), accessing an array

```
public static boolean dup1(int[] A) {  
    for (int i = 0; i < A.length; i += 1) {  
        for (int j = i + 1; j < A.length; j += 1) {  
            if (A[i] == A[j]) {  
                return true;  
            }  
        }  
    }  
    return false;  
}
```

*“Computers are really dumb. They can only do a few things like shuffling around numbers, but they do them really really fast so that they appear smart.”*

– Hal Perkins, UW CSE



# Cost Model: Space

- ❖ We usually talk about time rather than space, but the principles are the same
- ❖ Similar to time, we measure “cost” in higher-level abstractions rather than machine-specific bytes
  - Examples: queue elements, graph nodes, etc.

# Asymptotic Analysis

- ❖ **Asymptotic Analysis** is how the algorithm or data structure behaves as  $N \rightarrow \infty$ 
  - Simulating billions of particles
  - Social network with billions of users
  - Logging billions of transactions
  - Encoding billions of bytes of video data
- ❖ Why  $\infty$ ?
  - We need a way to characterize data structures' and algorithms' complexity, which can vary for every finite  $N$

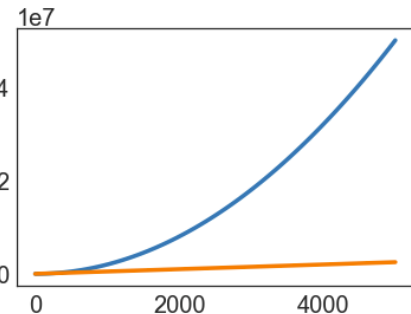
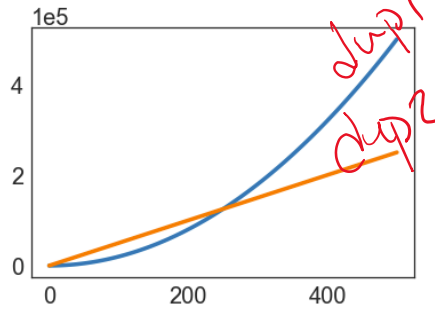
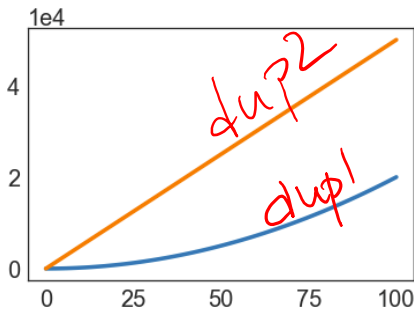
# Asymptotic Analysis Means “Infinity”: Graphically

❖ Asymptotic analysis is non-intuitive because it deals with infinity!

*because ∞ {*

*any finite value for N means that asymptotic analysis does not apply*

- What if you know your algorithm will only run on  $N < 100$  elements?
- Why do we “throw away constants”?
- How come lower-order terms “don’t matter”?



❖ Linear-time algorithms **scale better** than quadratic-time algorithms (parabolas).

# Asymptotic Analysis Means “Infinity”: Intuitively

- ❖ Since we’re dealing with infinity, constants and lower-order terms don’t meaningfully add to the final result
- ❖ The highest-order term is what matters and drives growth

**Table 2.1** The running times (rounded up) of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second. In cases where the running time exceeds  $10^{25}$  years, we simply record the algorithm as taking a very long time

*dup 2*  
*dup 2*

	$n$	$n \log_2 n$	$n^2$	$n^3$	$1.5^n$	$2^n$	$n!$
$n = 10$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
$n = 30$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	$10^{25}$ years
$n = 50$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
$n = 100$	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	$10^{17}$ years	very long
$n = 1,000$	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
$n = 10,000$	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
$n = 100,000$	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
$n = 1,000,000$	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

# Lecture Outline

- ❖ Asymptotic Analysis Concepts
- ❖ **Asymptotic Analysis Case Study: dup1 Worst Case**
- ❖ Asymptotic Analysis Simplifications
- ❖ Case Analysis vs Asymptotic Analysis
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# Counting Steps: Asymptotic Worst Case

- ❖ Assume  $N = A.length = 6$ 
  - I've broken up the for-loops to make it easier to count
  - Cheatsheet:  $1 + 2 + 3 + 4 + 5 + 6 = 21$

no dups at all!

eg

0	1	2	3	4	5
---	---	---	---	---	---

```

public static boolean dup1(int[] A) {
  1 int i = 0;
  for ( ; i < A.length; ) {
    int j = i + 1;
    for ( ; j < A.length; ) {
      if (A[i] == A[j]) {
        return true;
      }
      j += 1;
    }
    i += 1;
  }
  1 return false;
}

```

lex

num iterations + 1

# Counting Steps: Asymptotic Worst Case

❖ Cheatsheet:  $1 + 2 + 3 + 4 + 5 + 6 = 21$

Operation	Number of Executions (N = 6)	Symbolic Expression
$i = 0$		
Less-than (<)		
Increment (+=)		
Equals-to (==)		
Array accesses		

# Counting Steps: Asymptotic Worst Case

❖ Cheatsheet:  $1 + 2 + 3 + \dots + (N - 1) = ???$

Operation	Number of Executions (N = 6)	Symbolic Expression
$i = 0$	1	
Less-than (<)	28	
Increment (+=)	21	
Equals-to (==)	21	
Array accesses ([])	42	



# Counting Steps: Double-checking Our Work

❖ [Demo](#)

# Symbolic Expression

$$C = 1 + 2 + 3 + \dots + (N - 3) + (N - 2) + (N - 1)$$

$$C = (N - 1) + (N - 2) + (N - 3) + \dots + 3 + 2 + 1$$

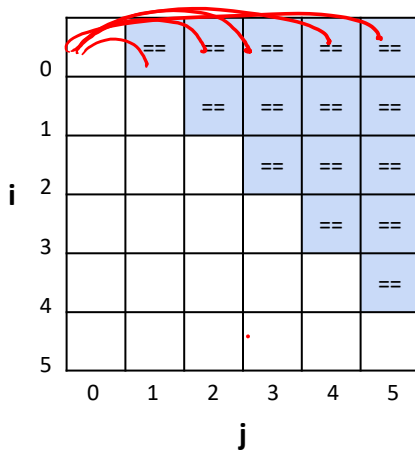
$$2C = N + N + N + \dots + N + N + N = N(N - 1)$$

$$\therefore C = N(N - 1)/2$$

$$\frac{N^2 - N}{2}$$

```
public static boolean dup1(int[] A) {
    for (int i = 0; i < A.length; i += 1) {
        for (int j = i + 1; j < A.length; j += 1) {
            if (A[i] == A[j]) {
                return true;
            }
        }
    }
    return false;
}
```

# Symbolic Expression, Intuitively


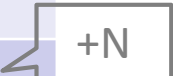



- We want the area of the right triangle
- If its side length is  $N$ , its order of growth is  $N^2$ .

```
public static boolean dup1(int[] A) {
    for (int i = 0; i < A.length; i += 1) {
        for (int j = i + 1; j < A.length; j += 1) {
            if (A[i] == A[j]) {
                return true;
            }
        }
    }
    return false;
}
```

# Counting Steps: Asymptotic Worst Case

❖ Cheatsheet:  $1 + 2 + 3 \dots N = (N^2 - N)/2$

Operation	Number of Executions (N = 6)	Symbolic Expression
$i = 0$	1	1
Less-than (<)	28	$(N^2 + 3N + 2)/2$ 
Increment (+=)	21	$(N^2 + N)/2$ 
Equals-to (==)	15	$(N^2 - N)/2$
Array accesses	30	$N^2 + N$ 

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# Simplification 1: Eliminate Lower-Order Terms

## ❖ Ignore lower-order terms

Operation	Number of Executions (N = 6)	Symbolic Expression
i = 0	1	1
Less-than (<)	28	$(N^2 + 3N + 2)/2$
Increment (+=)	21	$(N^2 + N)/2$
Equals-to (==)	15	$(N^2 - N)/2$
Array accesses	30	$N^2 + N$

$$\text{❖ } 6 + (N^2 + 3N + 2)/2 + (N^2 + N)/2 + (N^2 - N)/2 + N^2 + N$$

$$\text{❖ } 6 + \frac{1}{2} N^2 + \frac{3}{2} N + 1 + \frac{1}{2} N^2 + \frac{1}{2} N + \frac{1}{2} N^2 - \frac{1}{2} N + N^2 + N$$

$$\text{❖ } \cancel{6} + \frac{1}{2} N^2 + \cancel{\frac{3}{2} N} + \cancel{1} + \frac{1}{2} N^2 + \cancel{\frac{1}{2} N} + \frac{1}{2} N^2 - \cancel{\frac{1}{2} N} + N^2 + \cancel{N}$$

$$\text{❖ } \frac{5}{2} N^2$$

## Simplification 2: Eliminate Multiplicative Constants

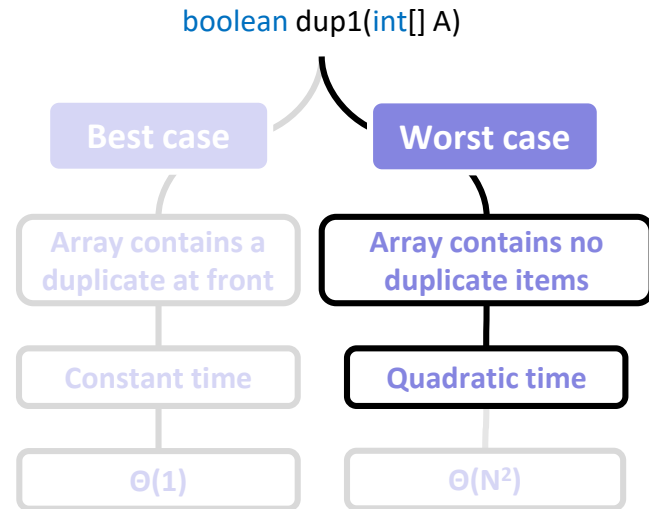
- ❖ Ignore multiplicative constants.
  - We already threw away the meaningful constant when we used a simplified cost model

- ❖  $\frac{5}{2} N^2$

- ❖   $N^2$

# Algorithm Analysis: Our Destination

- ❖ “The worst-case order-of-growth for dup1’s runtime is quadratic”





# Order of Growth

- ❖ What is the order of growth for each function?
  - (Informally, what is the shape of each function for very large  $N$ ?)

Function	Order of Growth
$N^3 + 3N^4$	$N^4$
$(1/N) + N^3$	$N^3$
$Ne^N + N$	$Ne^N$
$40 \sin(N) + 4N^2$	$N^2$

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[pollev.com/uwcse373](http://pollev.com/uwcse373)

For a very large array with billions of elements (i.e. asymptotic analysis), is it possible for dup1 to execute exactly 1 equals (==) operation? *YES! In its best case: 1 0 0 1 2 3*

Operation	Symbolic Expression (Worst Case)
Equals-to (==)	$(N^2 - N)/2$

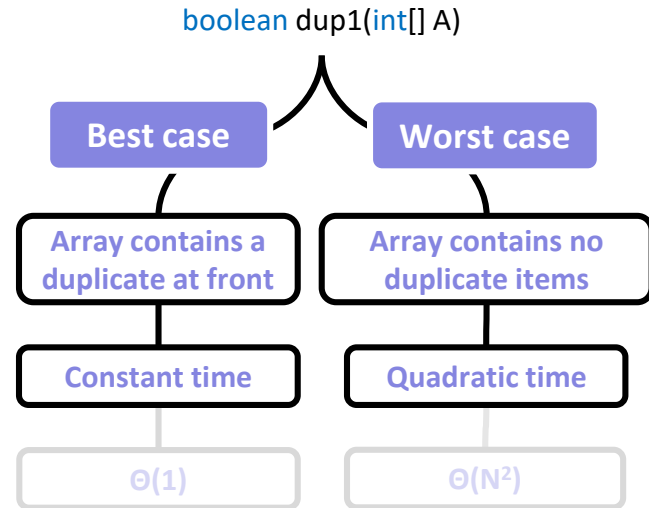
```
public static boolean dup1(int[] A) {
    for (int i = 0; i < A.length; i += 1) {
        for (int j = i + 1; j < A.length; j += 1) {
            if (A[i] == A[j]) {
                return true;
            }
        }
    }
    return false;
}
```

## Simplification 3: Consider the Worst Case

- ❖ Oftentimes, we only care about worst case
  - If a runtime statement doesn't mention, assume it's a worst-case analysis
- ❖ But dup1's best case is significantly better than its worst-case!
  - [Demo](#)

# Algorithm Analysis: Our Destination

- ❖ “The worst-case order-of-growth for dup1’s runtime is quadratic (parabolic)”
- ❖ “The best-case order-of-growth for dup1’s runtime is constant”



# Lecture Outline

- ❖ Asymptotic Analysis Concepts
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- ❖ Asymptotic Analysis Simplifications
- ❖ Case Analysis vs Asymptotic Analysis
- ❖ **Takeaways**

## tl;dr

- Runtime analysis uses a simplified cost model to represent computation
  - Space analysis also uses a simplified cost model
- Asymptotic analysis can take liberties with mathematical expressions because it deals with infinity
  - Eg, dropping lower-order terms and constants
- Asymptotic analysis gives us a common “frame of reference” with which to compare algorithms
  - “Parabolas grow faster than lines”
- Case analysis is a different axis on which to evaluate runtime and space