





# Lecture 5: Analyzing Complexity

CS 2110

February 3, 2026

# Today's Learning Outcomes

25. Explain the benefits of using asymptotic analysis versus performance testing to evaluate the efficiency of a piece of code.
26. Determine the asymptotic time and space complexity of a piece of code involving one or more loops and/or method calls.
27. Determine whether a given (mathematical) function belongs to a given big-O complexity class.
28. Compare and contrast *linear search* and *binary search*.

# Code Performance

Better code uses its resources more efficiently:

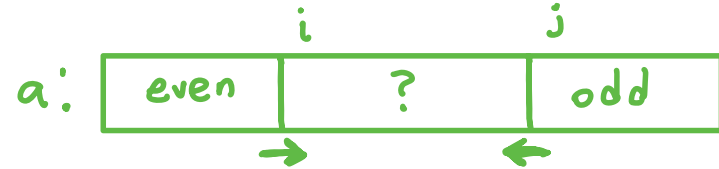
- Time user experience (lag), use more data, get better answers, time-sensitive domains (flight software)
- Memory Space embedded systems have space constraints
- Power
- CPU time, network bandwidth, cache space ....

Why does this matter?

To improve our code performance, we need a way to reliably measure time/space usage.

# Recall: Our paritySplit() Method

```
static int paritySplit(int[] a) {
    int i = 0;    int j = a.length;
    /* Loop invariant: `a[..i)` is even, `a[j..)` is odd */
    while (i < j) {
        if (a[i] % 2 == 0) {
            i++;
        } else {
            swap(a,i,j-1);
            j--;
        }
    }
    return j;
}
```



The diagram shows an array 'a' represented as a horizontal rectangle. It is divided into three sections by vertical lines. The first section is labeled 'even', the second is labeled '?', and the third is labeled 'odd'. Above the first section is the index 'i', and above the second section is the index 'j'. Below the array, there are two green arrows: one pointing to the right starting from the position of 'i', and one pointing to the left starting from the position of 'j'.

```
static void swap(int[] a, int x, int y) {
    int temp = a[x];
    a[x] = a[y];
    a[y] = temp;
}
```

How should we measure the runtime of this code?

# Measuring Runtime

## Option 1: Time its execution

```
int[] a = ... ; // initialize input array
```

```
long start = System.nanoTime();
```

```
int i = paritySplit(a);
```

```
long end = System.nanoTime();
```

```
System.out.println("paritySplit() ran in " + (end - start) / 1e6 + " ms.");
```

Output for 1000 elements:

- 0.285125 ms

- 0.222625 ms

- 0.169167 ms

### Issues:

- What about other inputs?

- Other computers may be faster/slower.

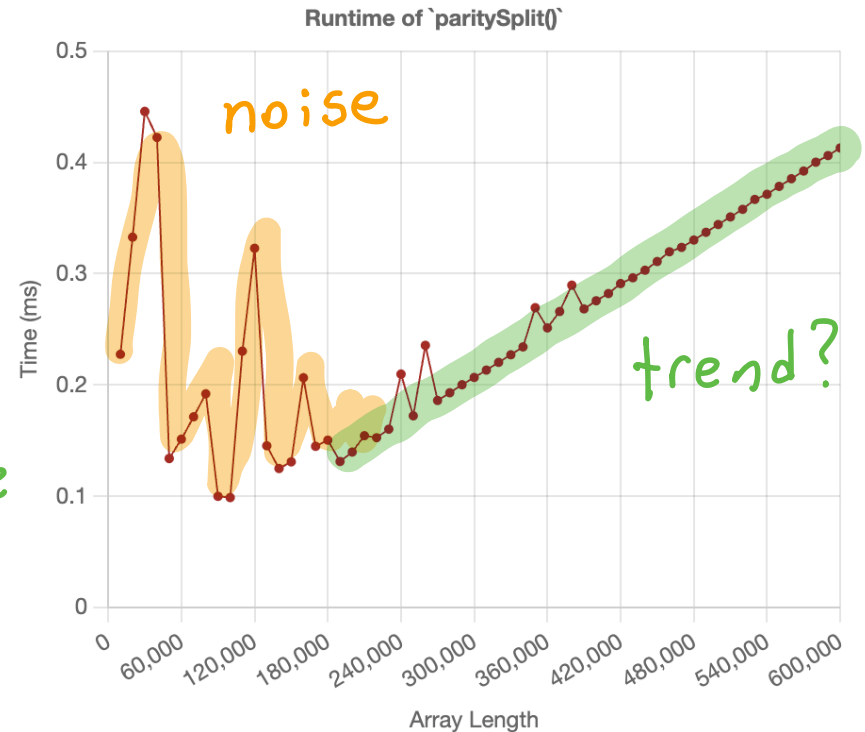
- Bigger inputs will take longer

- Different inputs (of same size) might take more/less time

# Measuring Runtime

Option 2: Time its execution across many input sizes

- Gives a better picture of how runtime depends on input size
- Picture is still a bit noisy
- Can we extract away noise and unneeded details to leave only trend summary?



# Runtime vs. Time Complexity

The "wall-clock" runtime of a method is how long it takes to execute.

Its time complexity  $T$  is the number of basic operations performed during its execution, expressed as a function of its input parameter sizes.

Basic ops: - Read value of var, write value to var, math operation, etc. (soon, we'll see details aren't crucial)

Today: Methods involving arrays,  $N = \text{array length}$   
runtimes will be functions of  $N$ ,  $T(N)$

# Counting Basic Operations

```
static void swap(int[] a, int x, int y) {  
    int temp = a[x];  
    a[x] = a[y];  
    a[y] = temp;  
}
```

- read x  
- read a[x]      3 ops  
- write to temp

- read y  
- read a[y]      4 ops  
- read x  
- write to a[x]

- read temp  
- read y  
- write to a[y]      3 ops

For swap()  $T(N) = 10$   
does not depend on input size  
or input values

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```
static int paritySplit(int[] a) {
    int i = 0;    int j = a.length;
    /* Loop invariant: `a[..i)` is even, `a[j..)` is odd */
    while (i < j) {
        if (a[i] % 2 == 0) {
            i++;
        } else {
            swap(a,i,j-1);
            j--;
        }
    }
    return j;
}
```

← even less work  
← than odd

Which paritySplit()  
input will execute more  
basic operations?

{1, 2, 3, 5, 7}

**(A)**

{2, 4, 6, 7, 8}

**(B)**

They are the same

**(C)**

# Best-Case vs. Worst-Case Inputs

Sometimes, # operations depends not only on input size, but also values (e.g., to determine if "if" body is executed)

Typically consider worst-case input, whose values force us down most complex execution path

vs. best-case input, whose values allow minimal # operations

CAUTION!!



Common misconception: worst-case  $\neq$  bigger  
best-case  $\neq$  smaller

These are descriptors of inputs of some pre-determined size.

# Counting Basic Operations (Worst-Case)

```

static int paritySplit(int[] a) {
    int i = 0;    int j = a.length;
    while (i < j) {
        if (a[i] % 2 == 0) {
            i++;
        } else {
            swap(a, i, j-1);
            j--;
        }
    }
    return j;
}
    
```

best-case  
will hit  
if branch

worst-case  
will hit  
else branch

# ops	# times executed	total # ops
$1+3 = 4$	1	4
3	$N+1$	$3N+3$
4	$N$	$4N$
3	0	0
$1+2+4+10$	$N$	$17N$
3	$N$	$3N$
1	1	1
		<hr/>
		$T(N) = 27N + 8$

# Asymptotic (Big-O) Notation

$T(N) = 27N + 8$  is a bit too specific

- Variations in hardware can change 27 (drop leading coefficients)

- For large  $N$ , "low-order" term 8 is negligible (drop)

We write  $T(N) = O(N)$  to summarize runtime complexity grows linearly with input size.

Def. We say runtime  $T(N)$  is  $O(g(N))$  for some function  $g$  if

$$\lim_{N \rightarrow \infty} \frac{T(N)}{g(N)} < \infty$$

"for large inputs, the growth of  $T(N)$  does not exceed  $g(N)$ , up to a constant factor"

For our purposes, enough to place  $T(N)$  appropriately in hierarchy:

# The Runtime Hierarchy

Better  
Performance



Worse  
Performance

Complexity Class	Name
$O(1)$	Constant Time
$O(\log N)$	Logarithmic Time
$O(N)$	Linear Time
$O(N \log N)$	Linearithmic Time
$O(N^2)$	Quadratic Time
$O(N^3)$	Cubic Time
$O(2^N)$	Exponential Time

} swap()  
binary Search()  
Really Good parity Split()  
linear Search()  
hasDuplicates()  
} Really Bad!

# Another Method: hasDuplicates()

```
/** Returns whether `a` contains duplicate entries,  
 * distinct indices `i != j` with `a[i] == a[j]`. */  
static boolean hasDuplicates(int[] a) {  
    for (int i = a.length - 1; i >= 0; i--) {  
        for (int j = 0; j < i; j++) {  
            if (a[i] == a[j]) {  
                return true;  
            }  
        }  
    }  
    return false;  
}
```

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Which of the following is a **best-case** input of length 6 to `hasDuplicates()`?

```
/** Returns whether `a` contains duplicate entries,  
 * distinct indices `i != j` with `a[i] == a[j]`.*/
```

```
static boolean hasDuplicates(int[] a) {
```

```
    for (int i = a.length - 1; i >= 0; i--) { ← i
```

```
        for (int j = 0; j < i; j++) { → j
```

```
            if (a[i] == a[j]) {
```

```
                return true;
```

```
            }  
        }  
    }
```

```
    return false;
```

```
}
```

{ 1, 2, 3, 4, 5, 6 } (A)

{ 1, 1, 2, 3, 4, 5 } (B)

{ ① 2, 3, 4, 5, ① } (C)  
*first two compared*

{ 1, 2, 3, 4, 5, 5 } (D)

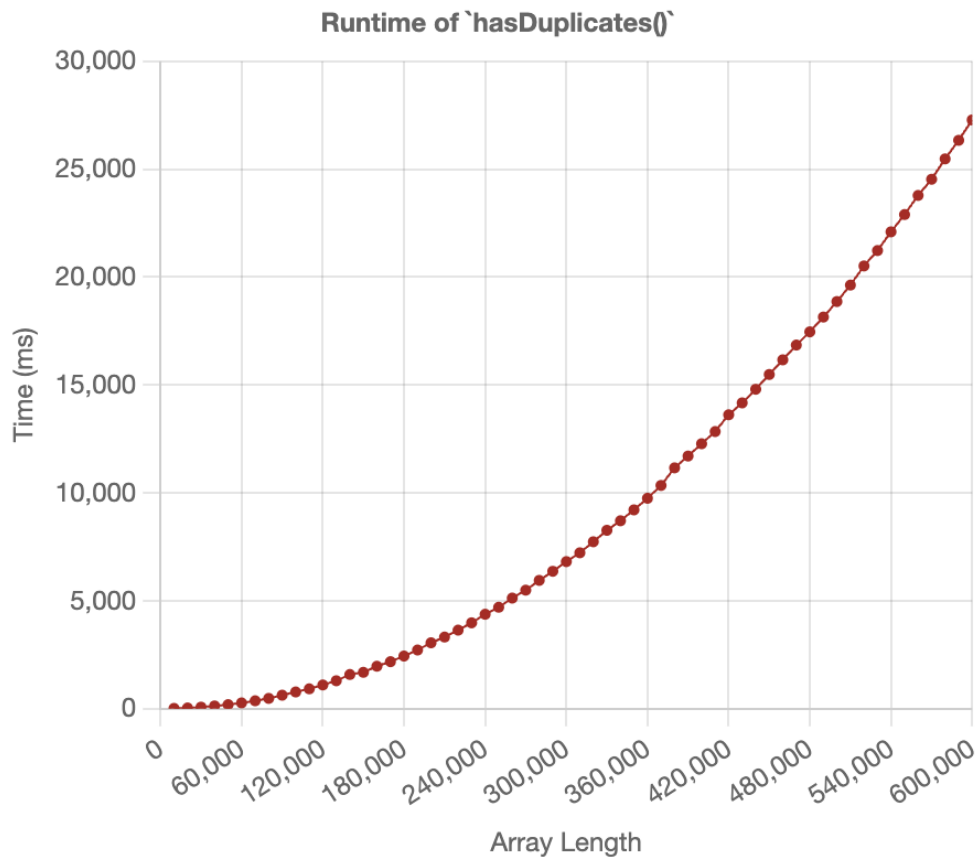
# Asymptotic Analysis (Worst - Case)

```
static boolean hasDuplicates(int[] a) {  
    for (int i = a.length - 1; i >= 0; i--) {  
        for (int j = 0; j < i; j++) {  
            if (a[i] == a[j]) {  
                return true;   
            }  
        }  
    }  
    return false;  
}
```

← not used  
by worst-  
case input

complexity	# executions	total complexity
$O(1)$	$O(N)$	$O(N)$
$O(1)$	$\sum_{i=0}^{N-1} (i+1) = O(N^2)$	$O(N^2)$
$O(1)$	$O(N^2)$	$O(N^2)$
$O(1)$	$O(1)$	$O(1)$
		<hr/> $O(N^2)$

# Visualizing hasDuplicates() Runtime



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hasDuplicates() ran in 10 seconds for an array with 360,000 elements.

How long should it take to run on an array with 720,000 elements?

*$O(N^2)$  time complexity means*

20 seconds

*doubling input size*

*x 2*

**(A)**

40 seconds

*quadruples runtime*

*x  $2^2$*

**(B)**

100 seconds

**(C)**





# Coding Demo: Linear Search



# Runtime Analysis

```
static int linearSearch(int[] a, int v) {  
    int i = 0;  
    /* Loop inv: `a[..i)` does not contain `v`. */  
    while (i < a.length) {  
        if (a[i] == v) {  
            return i; - early return not worst-case  
        }  
        i++;  
    }  
    return a.length;  
}
```

complexity	# executions	total complexity
$O(1)$	$O(1)$	$O(1)$
$O(1)$	$O(N)$	$O(N)$
$O(1)$	$O(N)$	$O(N)$
$O(1)$	$O(N)$	$O(N)$
$O(1)$	$O(1)$	$O(1)$
		<u><math>O(N)</math></u>

# Binary Search

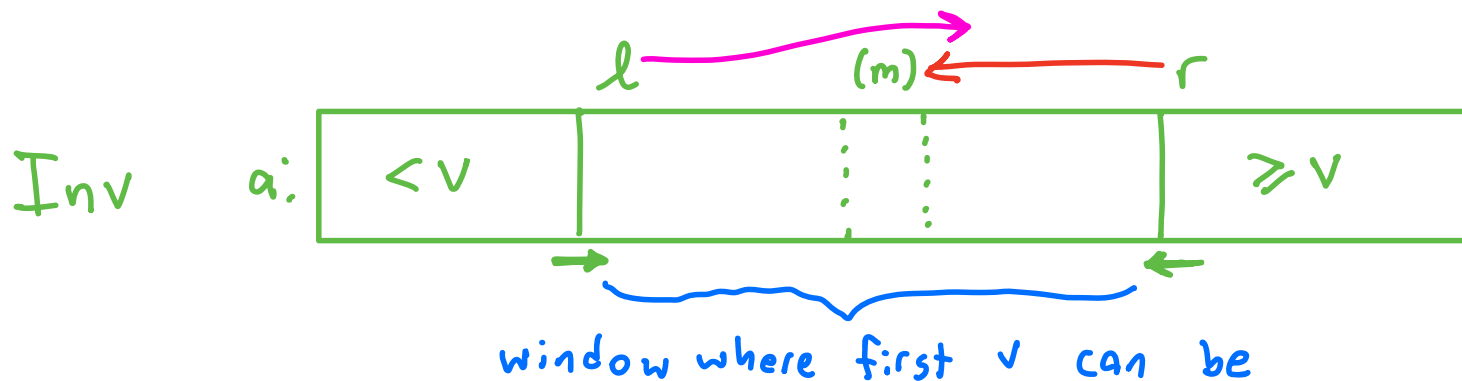
If array  $a$  is sorted, can we find the first index of  $v$  faster?

Idea: Inspect the middle element  $a[m]$

-  $a[m] \geq v$ : don't need to consider  $a[m..]$

-  $a[m] < v$ : don't need to consider  $a[..m]$

\* like searching in a (paper) dictionary





# Coding Demo: Binary Search

