

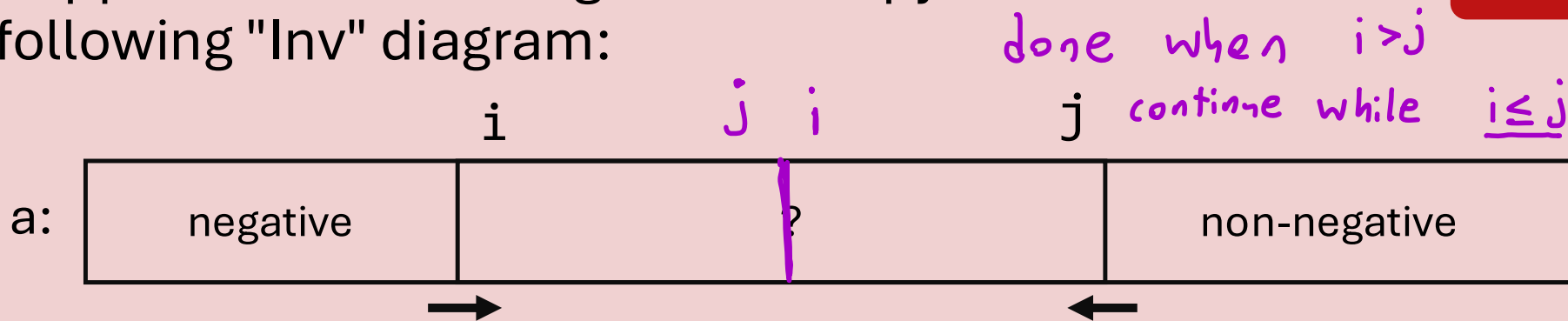
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Suppose we are writing some "loopy" code with the following "Inv" diagram:



What should be the guard on our loop?

$a[i] \geq 0$ (A)

$i < j$ (C)

$a[i] < a[j]$ (B)

$i \leq j$ (D)



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:

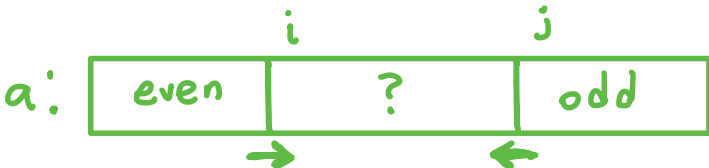
- Time user experience (lag), use more data, get better answers, time-sensitive domains (flight software)
- Memory Space embedded systems have space constraints
- Power
- CPU time, network bandwidth, cache space

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 illustrates the state of an array `a` during the execution of the `paritySplit` method. The array is represented as a horizontal rectangle divided into three sections. The first section on the left is labeled `even`. A green arrow points from the right boundary of this section to a green box labeled `i`. The middle section is labeled with a question mark `?`. A green arrow points from the left boundary of this section to a green box labeled `j`. The third section on the right is labeled `odd`. The entire diagram is labeled `a:` on the left.

```
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.

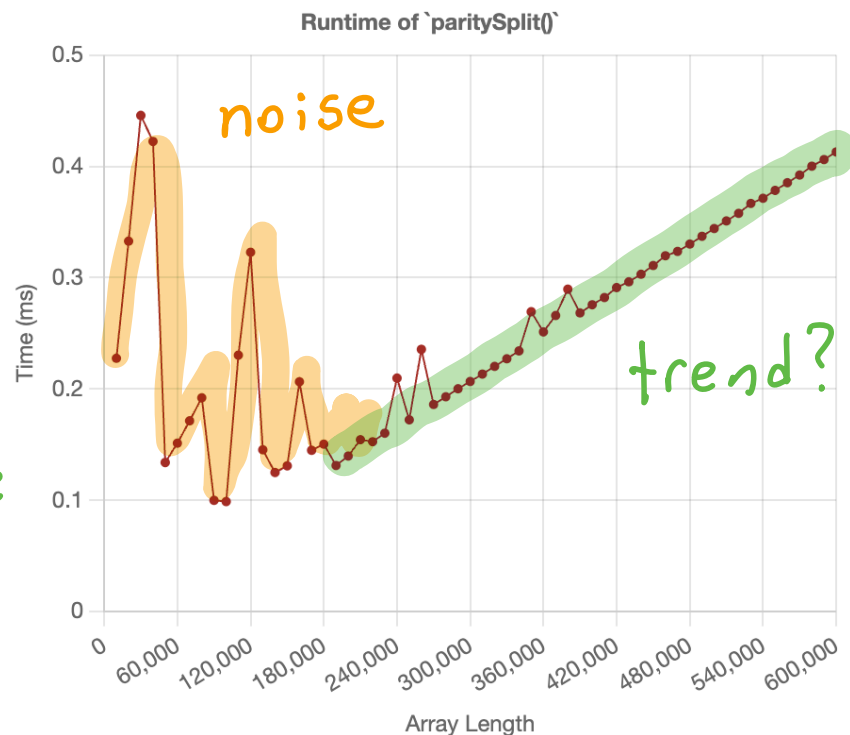
- Bigger inputs will take longer

- 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 = 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;  
}
```

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

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

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

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

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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; }
```

ops

$1 + 2 = 3$

3

4

3

$1 + 2 + 4 + 10$

3

1

times executed

1

$N + 1$

N

0

N

N

1

total # ops

3

$3N + 3$

$4N$

0

$17N$

$3N$

1

$T(N) = 27N + 7$

best-case
will hit
if branch

worst-case
will hit
else branch

Asymptotic (Big-O) Notation

$T(N) = 27N + 7$ is a bit too specific

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

– For large N , "low-order" term 7 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()
has Duplicates()
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--) {
```

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

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

```
                return true;
```

```
            }}
```

```
        return false;
```

```
    }
```

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

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

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

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

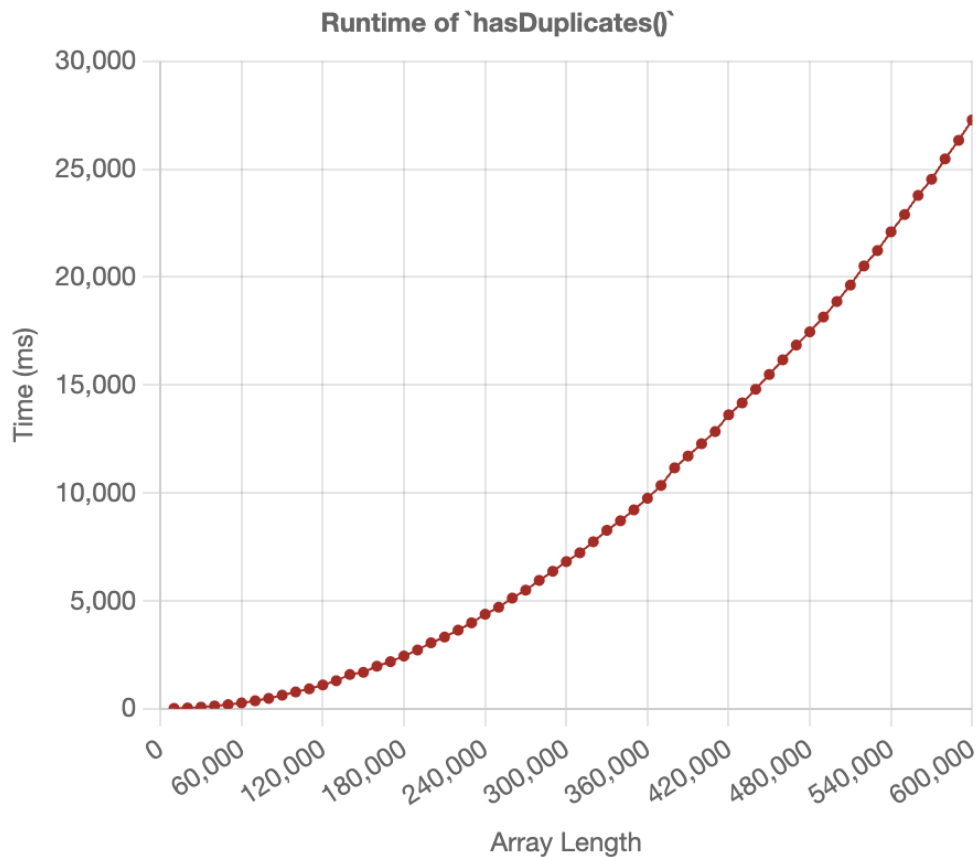
Asymptotic Analysis (Worst - Case)

```
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
$O(1)$	$O(N)$	$O(N)$
$O(1)$	$\sum_{i=0}^{N-1} (i+1) = O(N^2)$	$O(N^2)$
$O(1)$	$O(N^2)$	$O(N^2)$
$O(1)$	$O(1)$	$O(1)$
		<u>$O(N^2)$</u>

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

$\times 2$

(A)

40 seconds

quadruples runtime

$\times 2^2$

(B)

100 seconds

(C)

Linear Search

```
/** Returns the smallest index `i` such that `a[i] == v` or returns
 * `a.length` if `a` does not contain `v`. */
static int linearSearch(int[] a, int v) { ... }
```

Pre a: ?

Inv a:

v not here	?
------------	---

i

→

Post a:

v not here	v	?
------------	---	---

ⁱ (or

v not here

ⁱ)



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>$O(N)$</u>

Binary Search

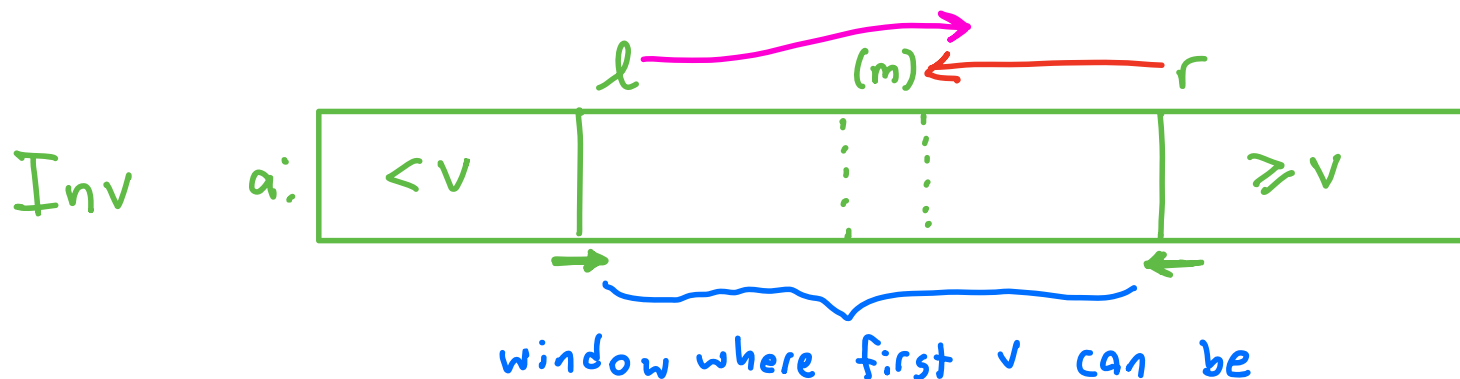
If array a is sorted, can we find the first index of v faster?

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

