

CSE 143

Computer Programming II

Efficiency



```
public void run() {  
    //for (int i = 0; i < 1000000; i++) {  
        //doLongCalculation();  
        //anotherAnalysis();  
        //solvePNP();  
    //}  
    System.out.println("Done!");  
}
```

What does it mean to have an “efficient program”?

```
1 System.out.println("hello");      vs.      1 System.out.print("h");
                                           2 System.out.print("e");
                                           3 System.out.print("l");
                                           4 System.out.print("l");
                                           5 System.out.println("o");
```

OUTPUT

```
>> left average run time is 1000 ns.
>> right average run time is 5000 ns.
```

We're measuring in NANoseconds!

Both of these run **very very** quickly. The first is definitely better style, but it's not “more efficient.”

hasDuplicate

Given a **sorted int array**, determine if the array has a duplicate.

```
public boolean hasDuplicate1(int[] array) {
    for (int i=0; i < array.length; i++) {
        for (int j=0; j < array.length; j++) {
            if (i != j && array[i] == array[j]) {
                return true;
            }
        }
    }
    return false;
}
```

```
public boolean hasDuplicate2(int[] array) {
    for (int i=0; i < array.length - 1; i++) {
        if (array[i] == array[i+1]) {
            return true;
        }
    }
    return false;
}
```

OUTPUT

```
>> hasDuplicate1 average run time is 5254712 ns.
>> hasDuplicate2 average run time is 2384 ns.
```

Timing programs is prone to error:

- We can't compare between computers
- We get noise (what if the computer is busy?)

Let's **count** the number of steps instead:

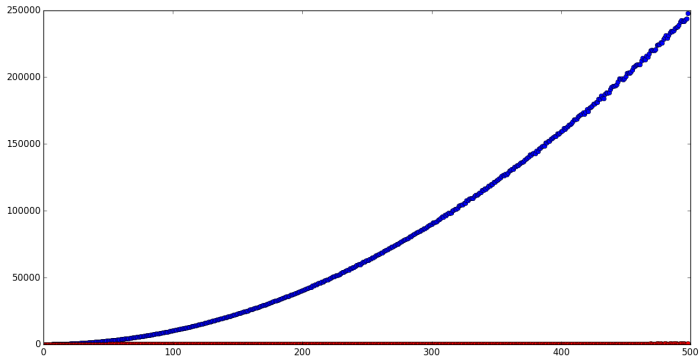
```
public int stepsHasDuplicate1(int[] array) {
    int steps = 0;
    for (int i=0; i < array.length; i++) {
        for (int j=0; j < array.length; j++) {
            steps++; // The if statement is a step
            if (i != j && array[i] == array[j]) {
                return steps;
            }
        }
    }
    return steps;
}
```

OUTPUT

```
>> hasDuplicate1 average number of steps is 9758172 steps.
>> hasDuplicate2 average number of steps is 170 steps.
```

This **still** isn't good enough! We're only trying **a single** array!

Instead, let's try running on arrays of size 1, 2, 3, ..., 1000000, and plot:



Runtime Efficiency

We've made the following observations:

- All “simple” statements (`println(“hello”), 3 + 7, etc.`) take **one** step to run.
- We should look at the “number of steps” a program takes to run.
- We should compare the **growth** of the runtime (not just one input).

```
1 statement1; }
2 statement2; } 3
3 statement3; }
4
5 for (int i = 0; i < N; i++) { } N
6     statement4;
7 }
8
9
10 for (int i = 0; i < N; i++) { } 4N
11     statement5;
12     statement6;
13     statement7;
14     statement8;
15 }
```

$5N + 3$

We measure **algorithmic complexity** by looking at the **growth rate** of the steps vs. the size of the input.

The algorithm on the previous slide ran in $5N + 3$ steps. As N gets very large, the “5” and the “3” become irrelevant.

We say that algorithm is $\mathcal{O}(N)$ (“Big-Oh-of- N ”) which means the number of steps it takes is **linear** in the input.

Some Common Complexities

$\mathcal{O}(1)$	Constant	The number of steps doesn't depend on n
$\mathcal{O}(n)$	Linear	If you double n , the number of steps doubles
$\mathcal{O}(n^2)$	Quadratic	If you double n , the number of steps quadruples
$\mathcal{O}(2^n)$	Exponential	The number of steps gets infeasible at $n < 100$


```
1 statement1; }
2 statement2; } 3
3 statement3; }
4
5 for (int i = 0; i < N; i++) {
6     statement4;
7     for (int j=0; j < N/2; j++) { } N/2
8         statement5;
9     }
10 }
11
12
13 for (int i = 0; i < N; i++) {
14     statement6;
15     statement7;
16     statement8;
17     statement9;
18 }
```

$N + N(N/2)$

$4N$

$0.5N^2 + 5N + 3$

So, the entire thing is $\mathcal{O}(N^2)$, because the quadratic term overtakes all the others.

<code>add(val)</code>	$\mathcal{O}(1)$
<code>add(idx, val)</code>	$\mathcal{O}(n)$
<code>get(idx)</code>	$\mathcal{O}(1)$
<code>set(idx, val)</code>	$\mathcal{O}(1)$
<code>remove(idx)</code>	$\mathcal{O}(n)$
<code>size()</code>	$\mathcal{O}(1)$

What are the time complexities of these functions?

```
1 public static void numbers1(int max) {  
2     ArrayList<Integer> list = new ArrayList<Integer>(); //O(1)  
3     for (int i = 1; i < max; i++) {  
4         list.add(i); //O(1)  
5     }  
6 }
```

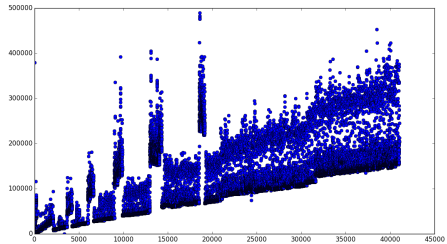
} $O(n)$

vs.

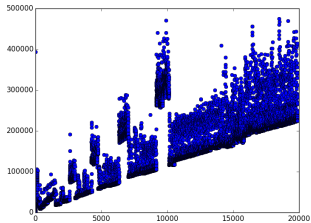
```
1 public static void numbers2(int max) {  
2     ArrayList<Integer> list = new ArrayList<Integer>(); //O(1)  
3     for (int i = 1; i < max; i++) {  
4         list.add(i); //O(1)  
5         list.add(i); //O(1)  
6     }  
7 }
```

} $O(n)$

numbers1



numbers2



```
1 public boolean is10(int number) {
2     return number == 10;
3 } } O(1)
4
5 public boolean two10s(int num1, int num2, int num3) {
6     return (is10(num1) && is10(num2) && !is10(num3)) ||
7            (is10(num1) && !is10(num2) && is10(num3)) ||
8            (!is10(num1) && is10(num2) && is10(num3));
9 } } O(1)
10
11 public void loops(int N) {
12     for (int i = 0; i < N; i++) {
13         for (int j = 0; j < N; j++) {
14             System.out.println(i + " " + j);
15         }
16     } } O(n2)
17
18
19     for (int i = 0; i < N; i++) {
20         System.out.println(N - i);
21     } } O(n)
22 }
```

```
1 public static int has5(int[] array) {  
2     for (int i = 0; i < array.length; i++) {  
3         System.out.println(array[i]); //O(1)  
4         if (array[i] == 5) { //O(1)  
5             return true; //O(1) } O(1)  
6     }  
7 }  
8 return false; //O(1)  
9 }
```

$O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$

$O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$ $O(1)$

$O(n)$ $O(n)$

Sometimes, these will finish in fewer than `array.length` steps, but **in the worse case**, we have to go through the whole array. This makes both of them $O(n)$.