

# Algorithms Analysis

## Algorithm Analysis with Kat

- A Brief introduction
- Sorting algorithms
  - Big-O notation
- Learning can be fun
  - Q & A

# Algorithms Analysis

## Introduction

**We are learning to create algorithms as junior developers**

**Algorithms produce the result we want.**

**Common algorithms we have created are**

- **Storing algorithms**
- **Searching algorithms**
- **Sorting algorithms**

**When two programs solve the same problem but look different, which one is better?**

# Sorting Algorithms

## Sorting Algorithms

- **Bubble Sort**
- **Selection Sort**
- **Insertion Sort**
- **Sorted() - comparison**

# Sorting Algorithms

# BubbleSort

86	25	24	52	10	99	68	44	67	5
25	86	24	52	10	99	68	44	67	5
25	24	86	52	10	99	68	44	67	5
25	24	52	86	10	99	68	44	67	5
25	24	52	10	86	99	68	44	67	5
25	24	52	10	86	99	68	44	67	5
25	24	52	10	86	99	68	44	67	5
25	24	52	10	86	68	99	44	67	5
25	24	52	10	86	68	99	44	67	5
25	24	52	10	86	68	44	99	67	5
25	24	52	10	86	68	44	67	99	5
25	24	52	10	86	68	44	67	5	99

# Sorting Algorithms

## Selection Sort

86	25	24	52	10	99	68	44	67	5
86	25	24	52	10	5	68	44	67	99
67	25	24	52	10	5	68	44	86	99
67	25	24	52	10	5	44	68	86	99
44	25	24	52	10	5	67	68	86	99
44	25	24	5	10	52	67	68	86	99
10	25	24	5	44	52	67	68	86	99
10	5	24	25	44	52	67	68	86	99
10	5	24	25	44	52	67	68	86	99
5	10	24	25	44	52	67	68	86	99

# Sorting Algorithms

## Insertion Sort

86	25	24	52	10	99	68	44	67	5
25	86	24	52	10	99	68	44	67	5
24	25	86	52	10	99	68	44	67	5
24	25	52	86	10	99	68	44	67	5
10	24	25	52	86	99	68	44	67	5
10	24	25	52	86	99	68	44	67	5
10	24	25	52	68	86	99	44	67	5
10	24	25	44	52	68	86	99	67	5
10	24	25	44	52	67	68	86	99	5
5	10	24	25	44	52	67	68	86	99

# Algorithms Analysis

## Python

- I am going to use Python to demonstrate some sorting algorithms.
- I am using Python because it is the programming language I am most familiar with.
- The same logic can be applied to any algorithm.

# Algorithm Efficiency

## Big-O notation

**When trying to characterise an algorithms efficiency in terms of execution, it is important to quantify the number of steps an algorithm will take.**

**An algorithms execution time changes with respect to the size of the problem.**

**A basic unit of computation for computing is to count the number of steps that are performed**

$$T(n) = 1 + n$$

"n" is the size of the problem

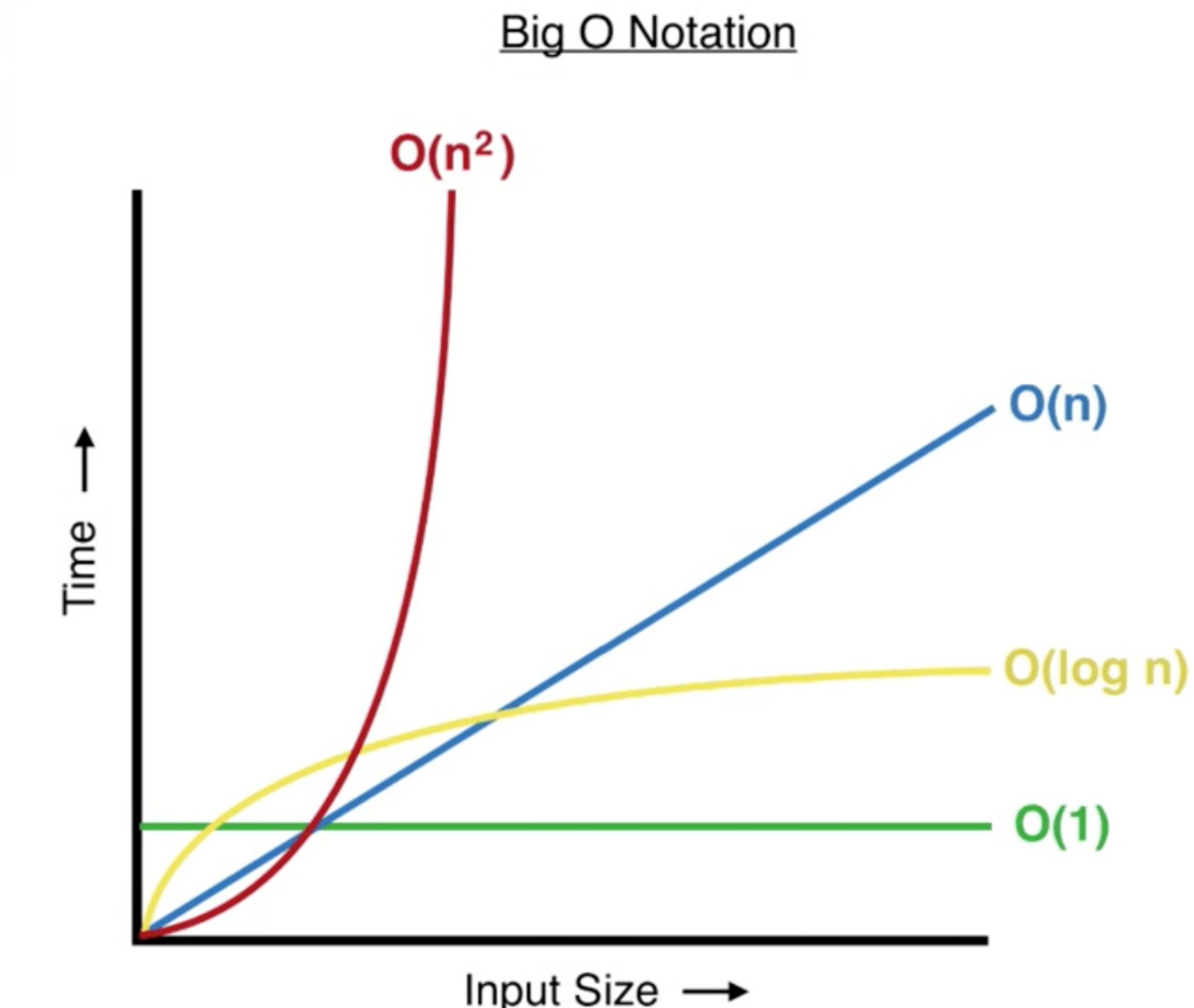
T(n) is the time taken to solve the problem

# Algorithm Efficiency

## Big-O notation

Order of magnitude functions " $O(f(n))$ ", often called Big-O notation ("O" for order)

$f(n)$	Name
1	Constant
$\log n$	Logarithmic
$n$	Linear
$n \log n$	Log Linear
$n^2$	Quadratic
$n^3$	Cubic
$2^n$	Exponential



# Big-O notation

## Big-O Bubble Sort

```
1 def bubbleSort(aList):
2     for passNumber in range (len(aList)-1,0,-1): # n(
3         for i in range(passNumber): # n(
4             if aList[i] > aList [i + 1]: # 1 +
5                 temp = aList[i] # 1 +
6                 aList[i] = aList[i +1] # 1 +
7                 aList[i +1] = temp # 1)
8
9     return aList
10
11 # T(n) = n(n(1 + 1 + 1 + 1))
12 # T(n) = n(n(4))
13 # T(n) = 4n^2
14 # O(n) = n^2
```

```
def bubbleSort(aList):
    for passNumber in range (len(aList)-1,0,-1): # n(
        for i in range(passNumber): # n(
            if aList[i] > aList [i + 1]: # 1 +
                temp = aList[i] # 1 +
                aList[i] = aList[i +1] # 1 +
                aList[i +1] = temp # 1)
    return aList
= n(n(1 + 1 + 1+ 1)
= n(n)
= n^2
```

Oder of Magnitude Big-O Highest Values wins:  $O(n^2)$   
It all depends on the value of n

# Big-O notation

## Selection Sort Bubble Sort

```
1 def selectionSort(aList):
2     for endSlot in range (len(aList) -1, 0, -1):           # n(
3         positionOfMax = 0                                     # 1 +
4         for location in range(1, endSlot + 1):               # n(
5             if aList[location] > aList[positionOfMax]:          # 1 +
6                 positionOfMax = location                      # 1 +
7             temp = aList[endSlot]                            # 1 +
8             aList[endSlot] = aList[positionOfMax]              # 1 +
9             aList[positionOfMax] = temp                      # 1 ))
10    return aList
11
12 # T(n) = n(1 + n( 1 + 1 + 1 + 1 + 1))
13 # T(n) = n(1 + n(5))
14 # T(n) = n(n(5))
15 # T(n) = 5n^2
16 # O(n) = n^2
```

```
def selectionSort(aList):
    for endSlot in range (len(aList) -1, 0, -1):           n(
        positionOfMax = 0                                     1 +
        for location in range(1, endSlot + 1):               n(
            if aList[location] > aList[positionOfMax]:          1 +
                positionOfMax = location                      1 +
            temp = aList[endSlot]                            1 +
            aList[endSlot] = aList[positionOfMax]              1 +
            aList[positionOfMax] = temp                      1 ))
    return aList
= n(1 + n(1 + 1 + 1+ 1+ 1))
= n(n)
= n^2
```

Oder of Magnitude Big-O Highest Values wins:  $O(n^2)$   
It all depends on the value of n

# Big-O notation

## Insertion Sort Bubble Sort

```
1 def insertionSort(aList):
2     for index in range(1, len(aList)):
3         currentValue = aList[index]
4         position = index
5         while position > 0 and aList[position] > currentValue:
6             aList[position] = aList[position - 1]
7             position = position - 1
8             aList[position] = currentValue
9     return aList
10
11 # T(n) = n(1 + 1 + n(1 + 1 + 1))
12 # T(n) = n(2 + n(3))
13 # T(n) = n(2 + 3n)
14 # T(n) = 2n + 3n^2
15 # O(n) = n^2
```

```
def insertionSort(aList):
    for index in range(1, len(aList)):
        currentValue = aList[index]
        position = index
        while position > 0 and aList[position] > currentValue:
            aList[position] = aList[position - 1]
            position = position - 1
            aList[position] = currentValue
    return aList
```

$$T(n) = n(1 + 1 + n(1 + 1 + 1))$$

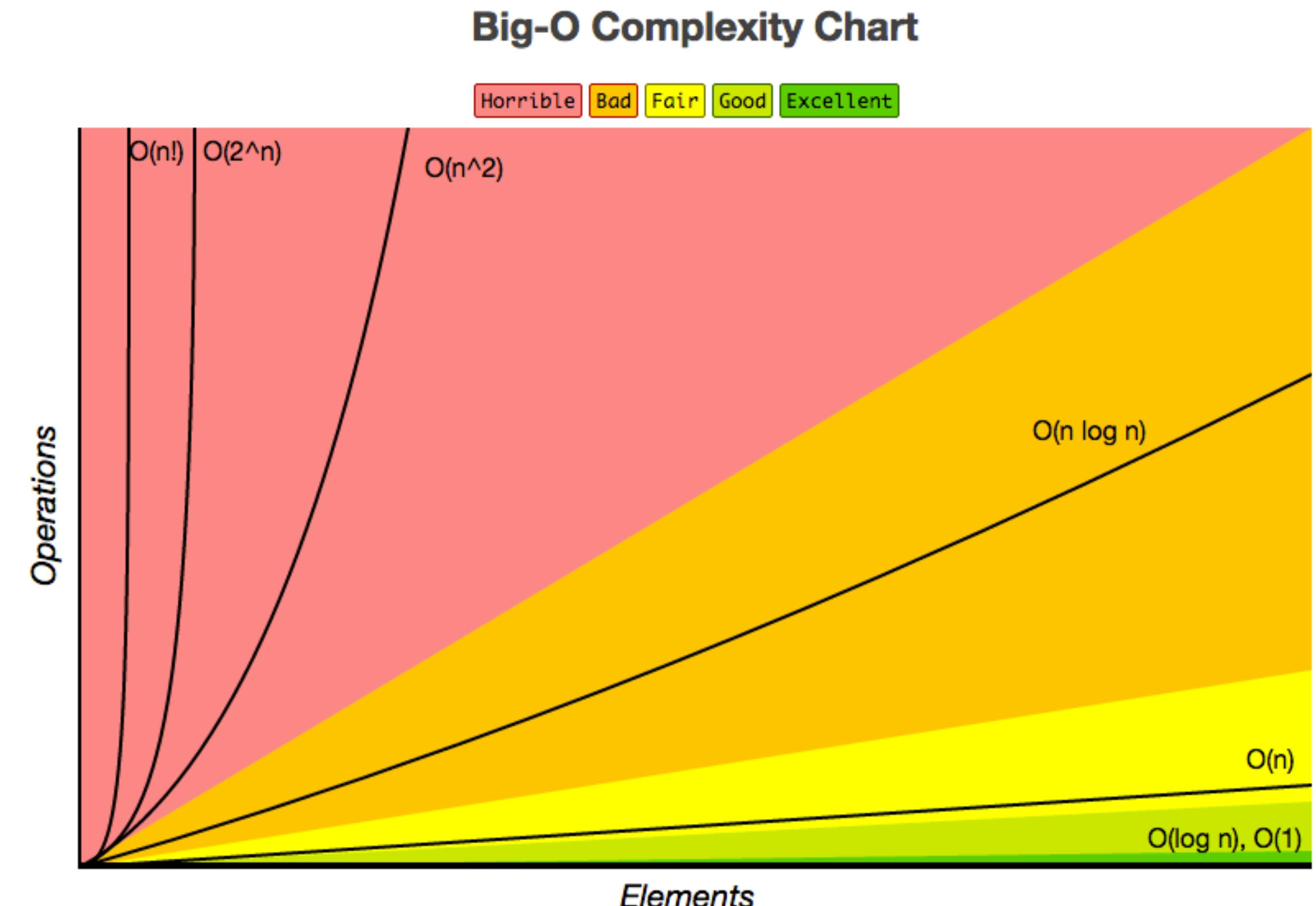
$$T(n) = n(n)$$

$$T(n) = n^2$$

Oder of Magnitude Big-O Highest Values wins:  $O(n^2)$   
It all depends on the value of n

# Algorithm Complexity

## Big-O Complexity



# Algorithm Complexity

## Best Case / Worst Case

Algorithm	Time Complexity			Space Complexity
	Best	Average	Worst	Worst
Quicksort	$\Omega(n \log(n))$	$\Theta(n \log(n))$	$O(n^2)$	$O(\log(n))$
Mergesort	$\Omega(n \log(n))$	$\Theta(n \log(n))$	$O(n \log(n))$	$O(n)$
Timsort	$\Omega(n)$	$\Theta(n \log(n))$	$O(n \log(n))$	$O(n)$
Heapsort	$\Omega(n \log(n))$	$\Theta(n \log(n))$	$O(n \log(n))$	$O(1)$
Bubble Sort	$\Omega(n)$	$\Theta(n^2)$	$O(n^2)$	$O(1)$
Insertion Sort	$\Omega(n)$	$\Theta(n^2)$	$O(n^2)$	$O(1)$
Selection Sort	$\Omega(n^2)$	$\Theta(n^2)$	$O(n^2)$	$O(1)$
Tree Sort	$\Omega(n \log(n))$	$\Theta(n \log(n))$	$O(n^2)$	$O(n)$
Shell Sort	$\Omega(n \log(n))$	$\Theta(n(\log(n))^2)$	$O(n(\log(n))^2)$	$O(1)$
Bucket Sort	$\Omega(n+k)$	$\Theta(n+k)$	$O(n^2)$	$O(n)$
Radix Sort	$\Omega(nk)$	$\Theta(nk)$	$O(nk)$	$O(n+k)$
Counting Sort	$\Omega(n+k)$	$\Theta(n+k)$	$O(n+k)$	$O(k)$
Cubesort	$\Omega(n)$	$\Theta(n \log(n))$	$O(n \log(n))$	$O(n)$

# Algorithm Complexity

## Common Data Structures

Data structures are the key in some of the most used algorithms and a good data structure is sometimes the way to achieve a very efficient and elegant solution.

# Common Data Structure

# Algorithm Complexity

## Sorting

Very important and useful type of algorithms for many common problems nowadays, Finder, iTunes, Excel and many more programs use some kind of sorting algorithm.

Algorithm	Time			Auxiliary Space
	Best	Average	Worst	
Bubble sort	$O(n)$	$O(n^2)$	$O(n^2)$	$O(1)$
Insertion sort	$O(n)$	$O(n^2)$	$O(n^2)$	$O(1)$
Selection sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$
Mergesort	$O(n \log(n))$	$O(n \log(n))$	$O(n \log(n))$	$O(n)$
Quicksort	$O(n \log(n))$	$O(n \log(n))$	$O(n^2)$	$O(\log(n))$
Heapsort	$O(n \log(n))$	$O(n \log(n))$	$O(n \log(n))$	$O(1)$
Shell Sort $(h=2^{k-1})$	$O(n \log(n))$	$O(n^{1.25})$ conjectured	$O(n^{3/2})$	$O(1)$
Bucket sort	$O(n+k)$	$O(n+k)$	$O(n^2)$	$O(nk)$
Radix sort	$O(nk)$	$O(nk)$	$O(nk)$	$O(n+k)$

## Sorting

# Algorithm Complexity

## Searching

### Searching

Searching efficiently is crucial for many operations in different parts of the software we use on a daily basis.

Algorithm	Data Structure	Time		Space
		Average	Worst	
Depth First Search (DFS)	Graph of $ V $ vertices and $ E $ edges	-	$O( E  +  V )$	$O( V )$
Breadth First Search (BFS)	Graph of $ V $ vertices and $ E $ edges	-	$O( E  +  V )$	$O( V )$
Binary Search	Sorted array of $n$ elements	$O(\log(n))$	$O(\log(n))$	$O(1)$
Linear (Brute force)	Array of $n$ elements	$O(n)$	$O(n)$	$O(1)$
Shortest path by Dijkstra, Min-heap as priority queue	Graph of $ V $ vertices and $ E $ edges	$O(( V + E ) \log V )$	$O(( V + E ) \log V )$	$O( V )$
Shortest path by Dijkstra, unsorted array as priority queue	Graph of $ V $ vertices and $ E $ edges	$O( V ^2)$	$O( V ^2)$	$O( V )$
Shortest path by Bellman-Ford	Graph of $ V $ vertices and $ E $ edges	$O( V  E )$	$O( V  E )$	$O( V )$

# Algorithm Complexity

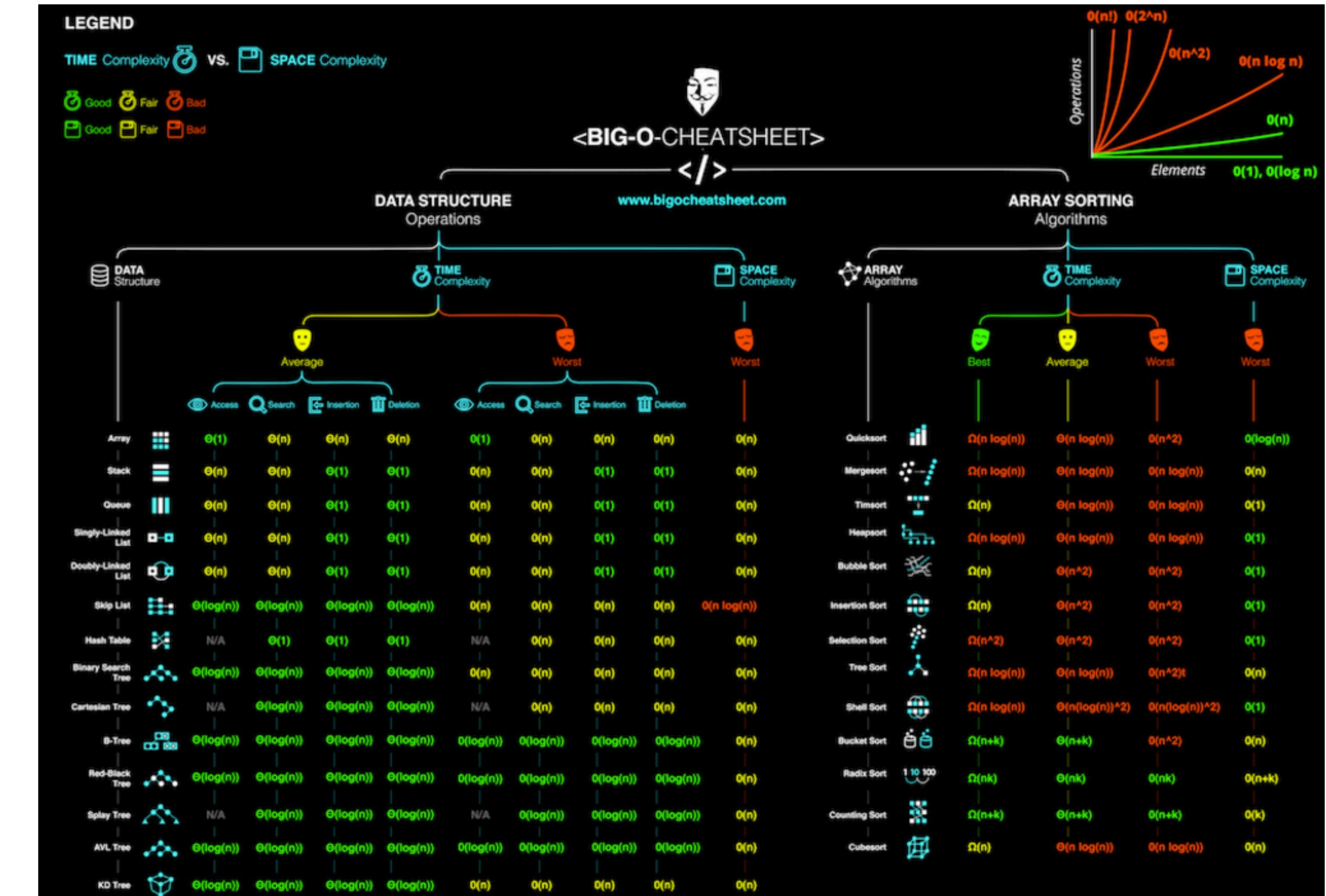
## Heaps

### Heaps

Heaps	Time						
	Heapify	Find Max	Extract Max	Increase Key	Insert	Delete	Merge
Linked List	-	O(1)	O(1)	O(n)	O(n)	O(1)	O(m+n)
Linked List (unsorted)	-	O(n)	O(n)	O(1)	O(1)	O(1)	O(1)
Binary Heap	O(n)	O(1)	O(n log(n))	O(n log(n))	O(n log(n))	O(n log(n))	O(m+n)
Binomial Heap	-	O(n log(n))	O(n log(n))	O(n log(n))	O(n log(n))	O(n log(n))	O(n log(n))
Fibonacci Heap	-	O(1)	O(log(n))*	O(1)*	O(1)	O(log(n))*	O(1)

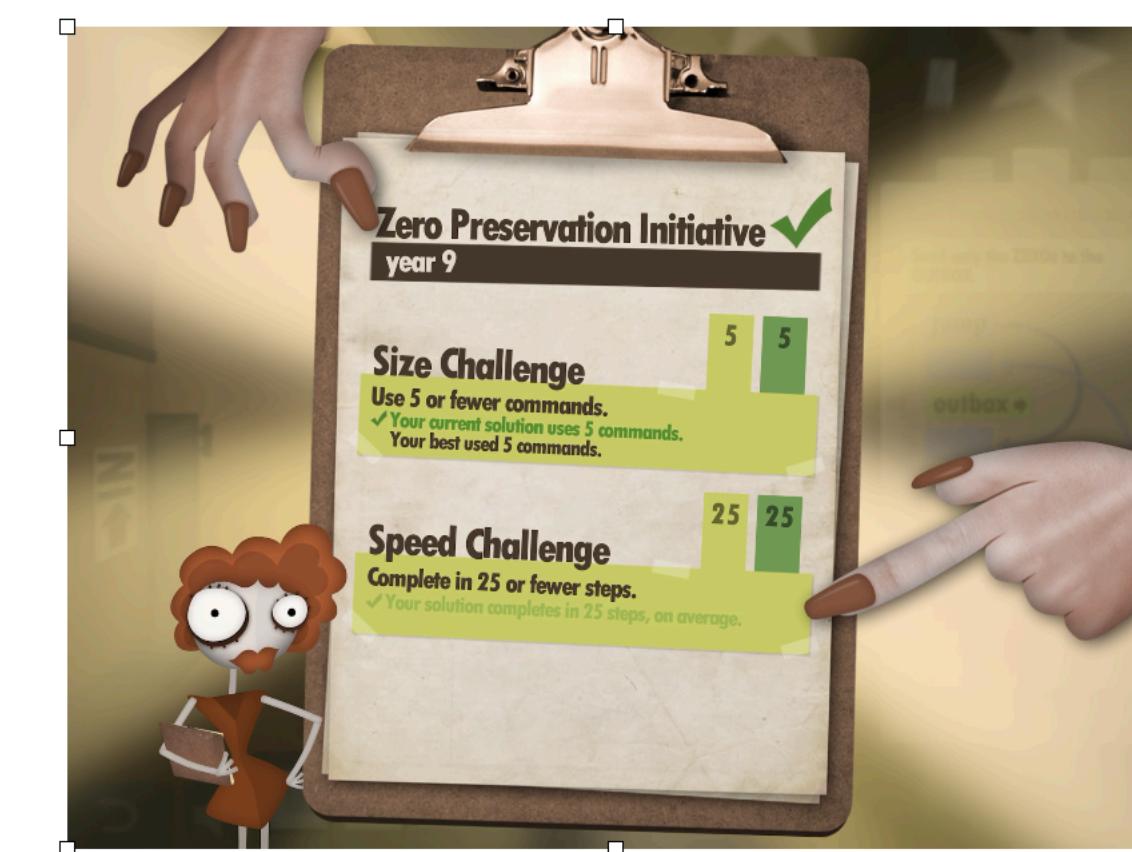
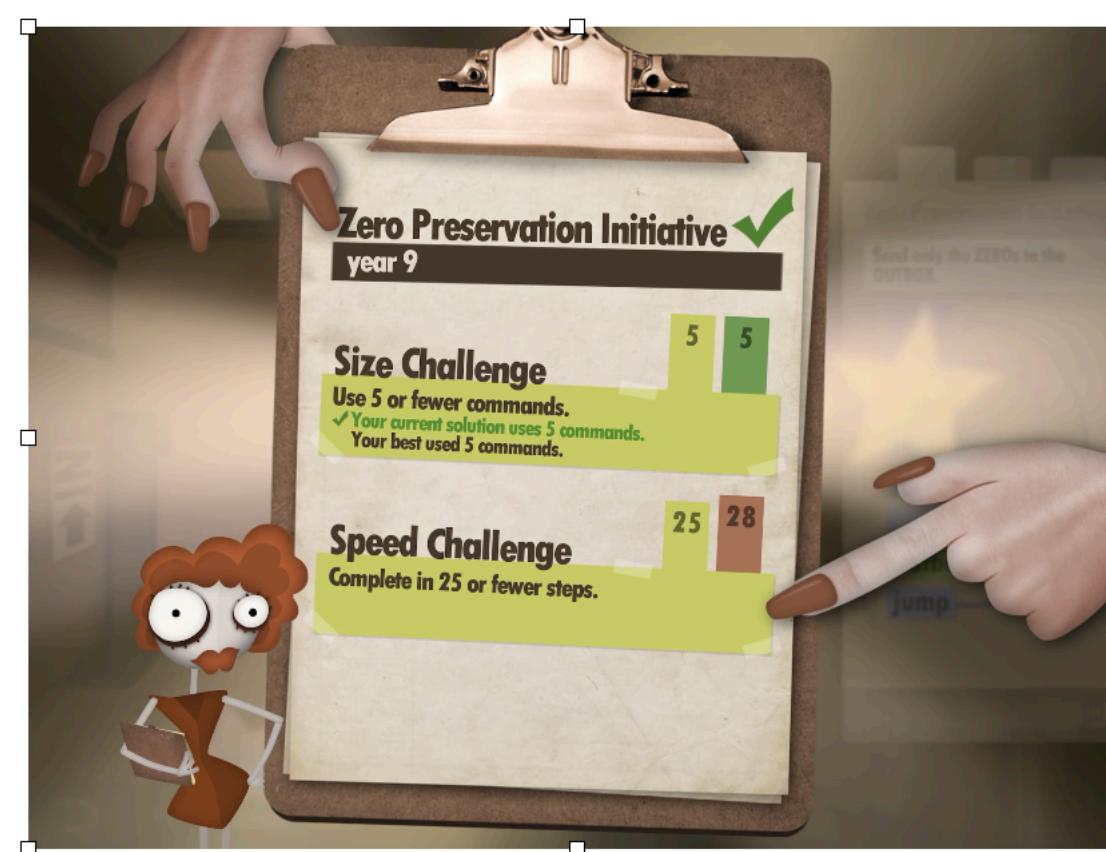
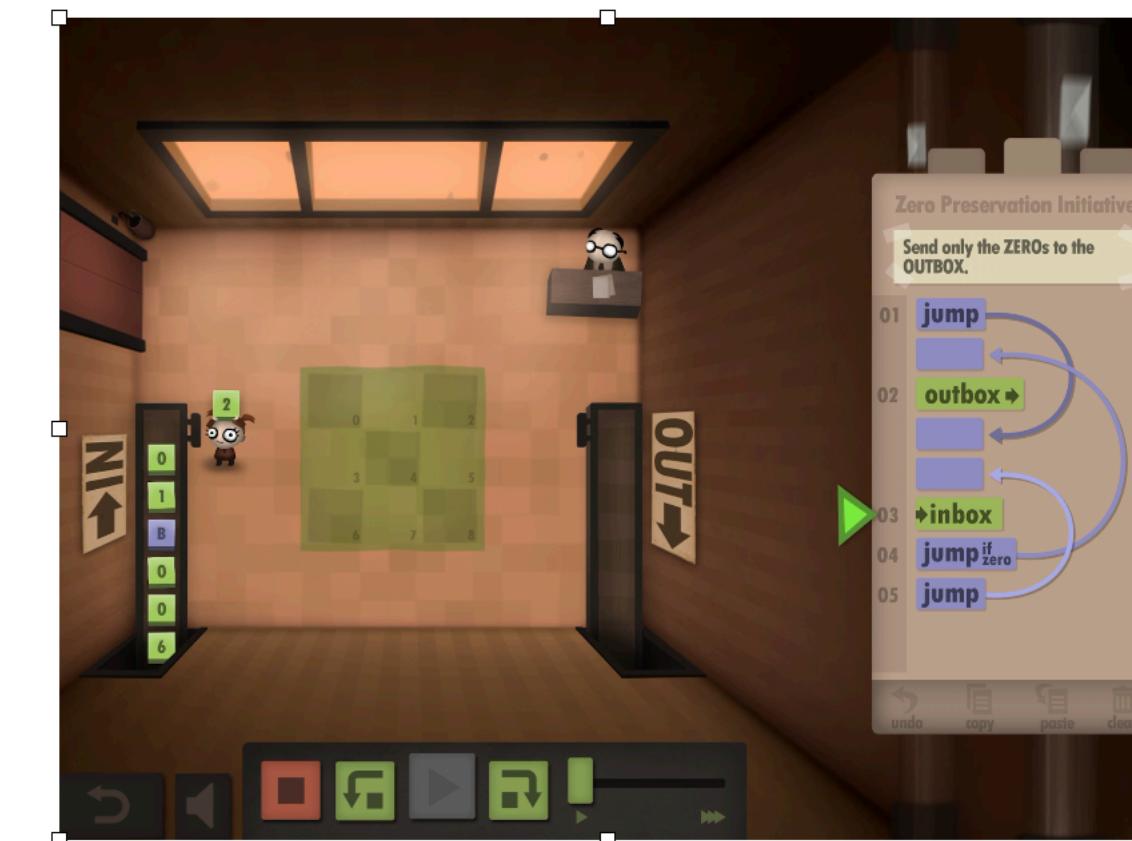
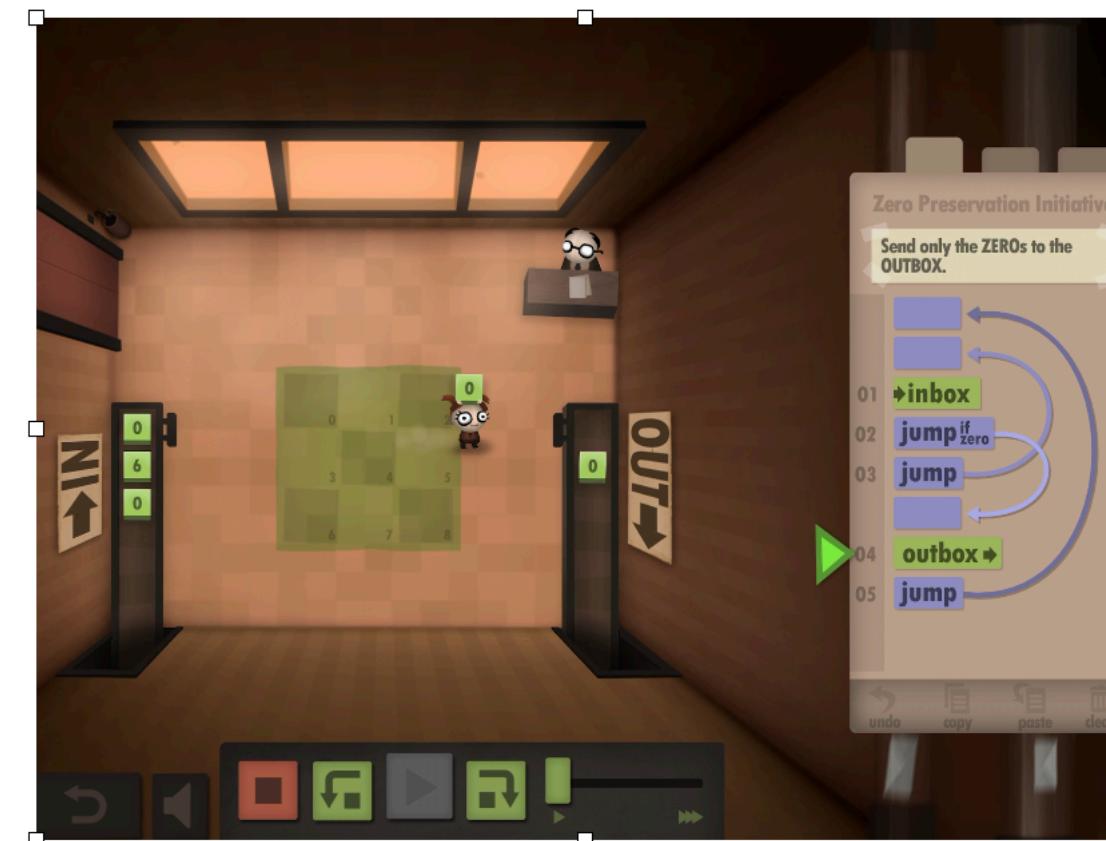
# Algorithm Complexity

## Big-O Cheat Sheet



# Algorithm Game

Fun!



# Algorithm Game

## Human Resource Machine



# Algorithm Game

## 7 Billion Humans



The End

**Thank you for watching!**

**Any Questions? ...**