# CSCA48 TUTORIAL WEEK #11

**TUT 0006** 

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Welcome Objectives Insertion Sort Complexity Big-Oh Notation

# THIS WEEK

Sorting algorithm: insertion sort

Algorithm complexity analysis

Big-Oh notation

## INSERTION SORT

The algorithm – to the board!

[6, 5, 3, 1, 8, 7, 2, 4]

The implementation – to Wing!

# ALGORITHM COMPLEXITY ANALYSIS

Assuming every operation in a computer uses same amount of time

e.g. computing the value of an expression, querying a variable in the memory, assigning value to a variable, etc.

We call each of those operations a "step"

Let's see how many step does every line of our insertion sort function takes!

## ALGORITHM COMPLEXITY ANALYSIS

```
def insertion_sort(L):
for i in range(1,len(L)):
                                              # 5 steps
    current_val = L[i]
                                              # 3 steps
    i = i
                                              # 2 steps
    while(j > 0 and L[j-1] < current_val):
                                              # 8 steps
        L[j] = L[j-1]
                                              # 5 steps
        j = j - 1
                                              # 3 steps
    L[j] = current_val
                                              # 3 steps
```

Welcome Objectives Insertion Sort Complexity Big-Oh Notation

Big O notation is used to describe the performance or complexity of an algorithm

Big O specifically describes the worst-case scenario

Welcome Objectives Insertion Sort Complexity Big-Oh Notation

**O(1)** describes an algorithm that will always execute in the same time regardless of the size of the input data set.

```
def is_empty(L):
return len(L) == 0
```

**O(N)** describes an algorithm whose performance will grow linearly and in direct proportion to the size of the input data set.

```
def contains(L, element):
found = False
i = 0
while i < len(L) and not found:
    if L[i] == element:
        found = True
    i += 1
return found</pre>
```

O(N<sup>2</sup>) represents an algorithm whose performance is directly proportional to the square of the size of the input data set.

```
def insertion_sort(L):
for i in range(1,len(L)):
    current_val = L[i]
    i = i
    while(j > 0 and L[j-1] < current_val):</pre>
        L[j] = L[j-1]
         j = j - 1
    L[i] = current_val
```

 $O(2^N)$  denotes an algorithm whose growth doubles with each addition to the input data set. The growth curve of an  $O(2^N)$  function is exponential.

```
def fib(n):
if(n < 2):
    result = 1
else:
    result = fib(n - 1) + fib(n - 2)
return result</pre>
```

#### O(log N) represents binary algorithms.

```
def binary_search(L, s):
if(len(L) == 0):
    result = False
else:
    mid_index = len(L) // 2
    mid_element = L[mid_index]
    if(mid_element == s):
        result = True
    elif(mid_element < s):</pre>
        result = binary_search(L[mid_index + 1:], s)
    else:
        result = binary_search(L[: mid_index], s)
return result
```

Just for fun:  $O(\infty)$  denotes an algorithm which will never finish in the worst case scenario.

```
def bogo_sort(x):
while not inorder(x):
    shuffle(x)
return x
```

O(1)

O(N)

 $O(N^2)$ 

 $O(2^{N})$ 

O(log N)