

UNIT 4C

Iteration: Scalability & Big O

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Efficiency

- A computer program should be totally correct, but it should also
 - execute as quickly as possible (time-efficiency)
 - use memory wisely (storage-efficiency)
- How do we compare programs (or algorithms in general) with respect to execution time?
 - various computers run at different speeds due to different processors
 - compilers optimize code before execution
 - the same algorithm can be written differently depending on the programming paradigm

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Counting Operations

- We measure time efficiency by counting the number of operations performed by the algorithm.
- But what is an operation?
 - assignment statements
 - comparisons
 - return statements
 - ...

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Linear Search: Worst Case

```
# let n = the length of list.
def search(list, key)
  index = 0                                1
  while index < list.length do            n+1
    if list[index] == key then            n
      return index
    end
    index = index + 1                      n
  end
  return nil                                1
end                                         Total: 3n+3
```

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Linear Search: Best Case

```
# let n = the length of list.
def search(list, key)
  index = 0                                1
  while index < list.length do            1
    if list[index] == key then            1
      return index                        1
    end
    index = index + 1
  end
  return nil
end                                         Total: 4
```

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Counting Operations

- How do we know that each operation we count takes the same amount of time? (We don't.)
- So generally, we look at the process more abstractly and count whatever operation depends on the amount or size of the data we're processing.
- For linear search, we would count the number of times we compare elements in the array to the key.

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Linear Search: Worst Case Simplified

```
# let n = the length of list.
def search(list, key)
  index = 0
  while index < list.length do
    if list[index] == key then          n
      return index
    end
    index = index + 1
  end
  return nil
end                                     Total:    n
```

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Linear Search: Best Case Simplified

```
# let n = the length of list.
def search(list, key)
  index = 0
  while index < list.length do
    if list[index] == key then        1
      return index
    end
    index = index + 1
  end
  return nil
end                                     Total:    1
```

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Order of Complexity

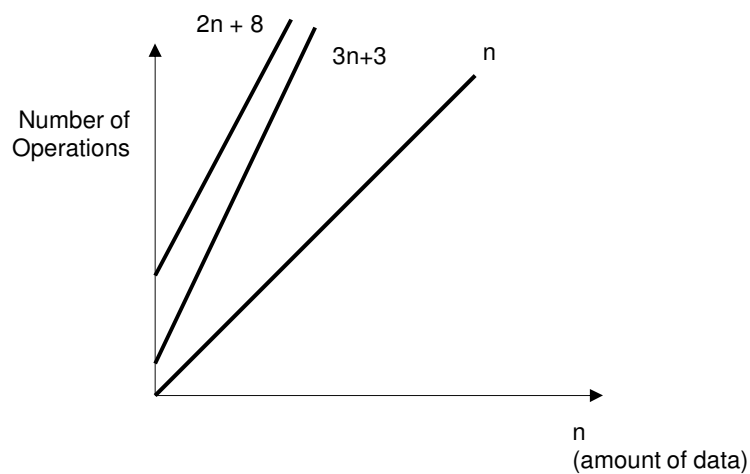
- For very large n , we express the number of operations as the (time) order of complexity.
- Order of complexity is often expressed using Big-O notation:

<u>Number of operations</u>	<u>Order of Complexity</u>	
n	$O(n)$	Usually doesn't matter what the constants are... we are only concerned about the highest power of n.
$3n+3$	$O(n)$	
$2n+8$	$O(n)$	

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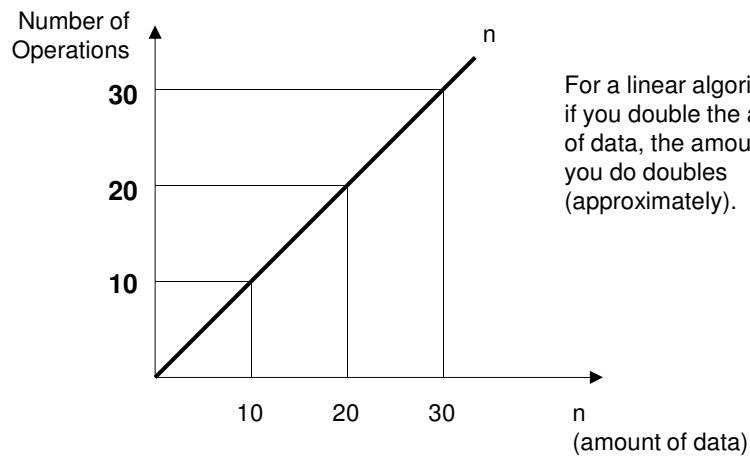
$O(n)$ ("Linear")



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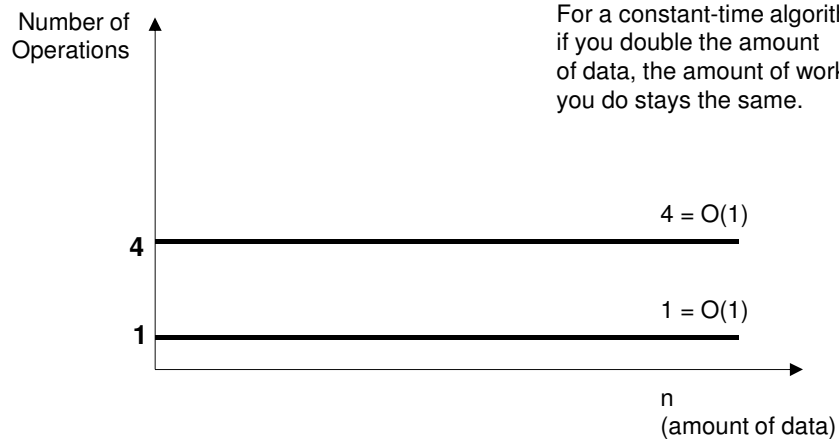
10

$O(n)$



For a linear algorithm, if you double the amount of data, the amount of work you do doubles (approximately).

$O(1)$ ("Constant-Time")



For a constant-time algorithm, if you double the amount of data, the amount of work you do stays the same.

Linear Search

- Worst Case: $O(n)$
- Best Case: $O(1)$
- Average Case: ?

Insertion Sort: Worst Case

```
# let n = the length of list.
def isort(list)
    a = list.clone           n
    i = 1
    while i != a.length do
        move_left(a, i)     n-1
        i = i + 1
    end
    return a
end
```

Insertion Sort: Worst Case

```
# let n = the length of list.
def move_left(a, i)
  x = a.slice!(i)
  j = i-1
  while j >= 0 && a[j] > x do      i+1
    j = j - 1
  end
  a.insert(j+1, x)
end
```

but how long do `slice!` and `insert` take?

move_left (alternate version)

```
# let n = the length of list.
def move_left(a, i)
  x = a[i]
  j = i-1
  while j >= 0 && a[j] > x do      i+1
    a[j+1] = a[j]
    j = j - 1
  end
  a[j+1] = x
end
```


Insertion Sort: Worst Case

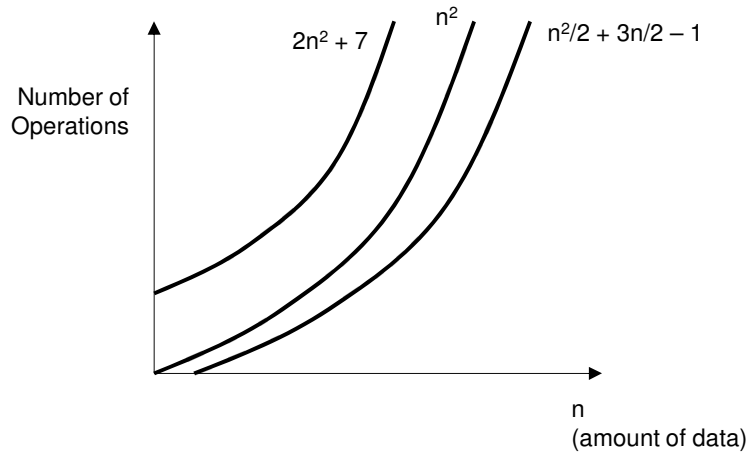
- So the total number of operations is $n + (n-1 \text{ move_left's})$
- But each `move_left` performs $i+1$ operations, where i varies from 1 to $n-1$:
- $n-1 \text{ move_left's} = 2 + 3 + 4 + \dots + n$
- Since $1 + 2 + \dots + n = n(n+1)/2$,
 $n-1 \text{ move_left's} = n(n+1)/2 - 1$
- The total number of operations is:
 $n + n(n+1)/2 - 1 = n + n^2/2 + n/2 - 1 = n^2/2 + 3n/2 - 1$

Order of Complexity

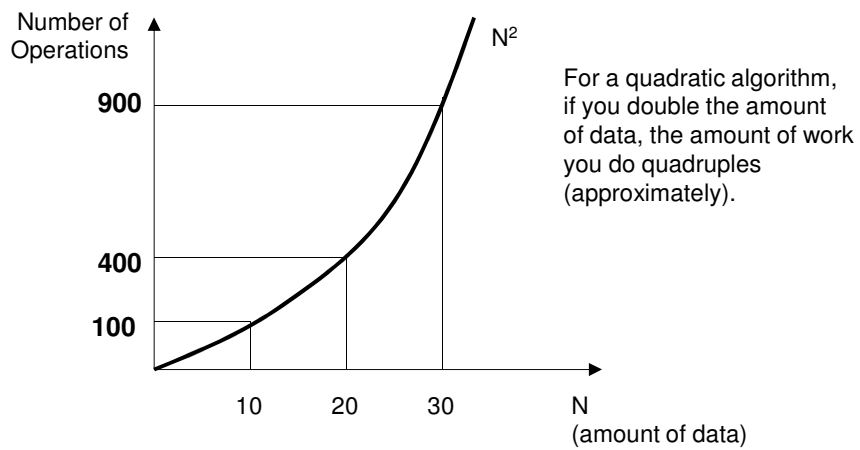
<u>Number of operations</u>	<u>Order of Complexity</u>
n^2	$O(n^2)$
$n^2/2 + 3n/2 - 1$	$O(n^2)$
$2n^2 + 7$	$O(n^2)$

Usually doesn't matter what the constants are... we are only concerned about the highest power of n .

$O(n^2)$ ("Quadratic")



$O(n^2)$



Insertion Sort

- Worst Case: $O(n^2)$
- Best Case: ?

We'll compare these algorithms with others soon to see how scalable they really are based on their order of complexities.