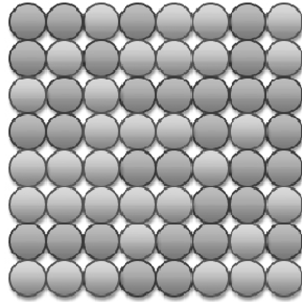

UNIT 4C
Iteration: Scalability & Big O

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

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
  - ...

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
```

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
```


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.

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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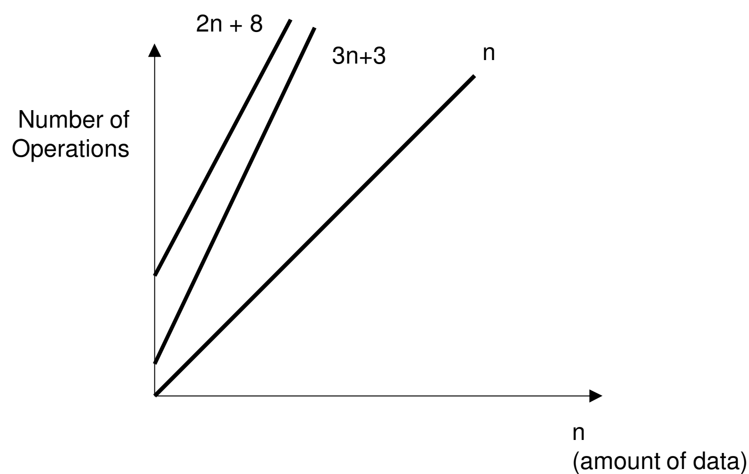
Order of Complexity

- 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:

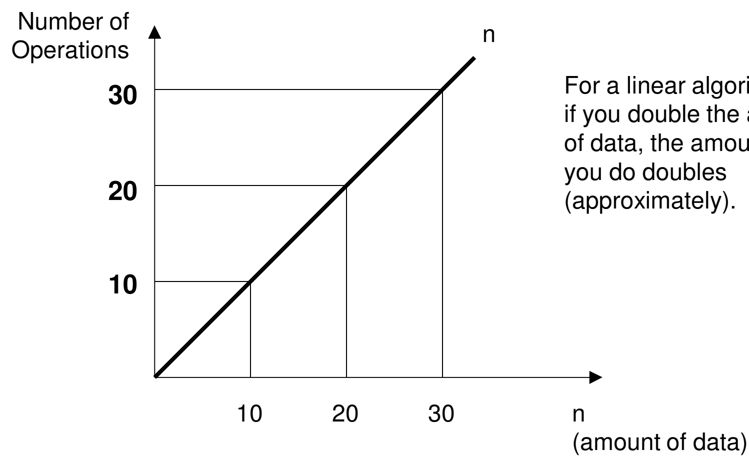
Number of operations	Order of Complexity
n	O(n)
3n+3	O(n)
2n+8	O(n)

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

O(n) ("Linear")



O(n)



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

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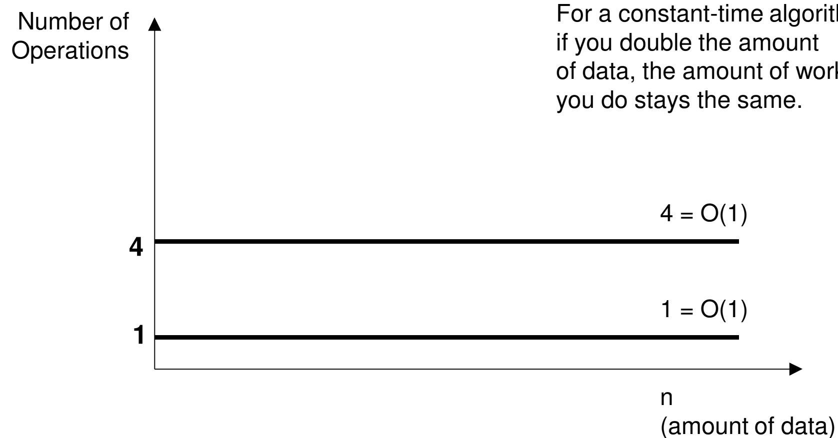
O(1) (“Constant-Time”)

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



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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

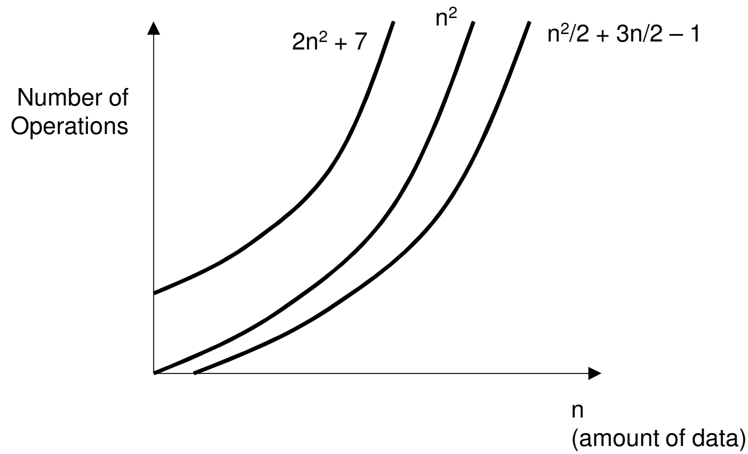
- So the total number of operations is
  n + (n-1 move_left's)
- But each move_left performs i+1 operations, where i varies from 1 to n-1:
- n-1 move_left's = 2 + 3 + 4 + ... + n
- Since 1 + 2 + ... + n = n(n+1)/2,
  n-1 move_left's = n(n+1)/2 – 1
- The total number of operations is:
  $n + n(n+1)/2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

Order of Complexity

Number of operations	Order of Complexity
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")



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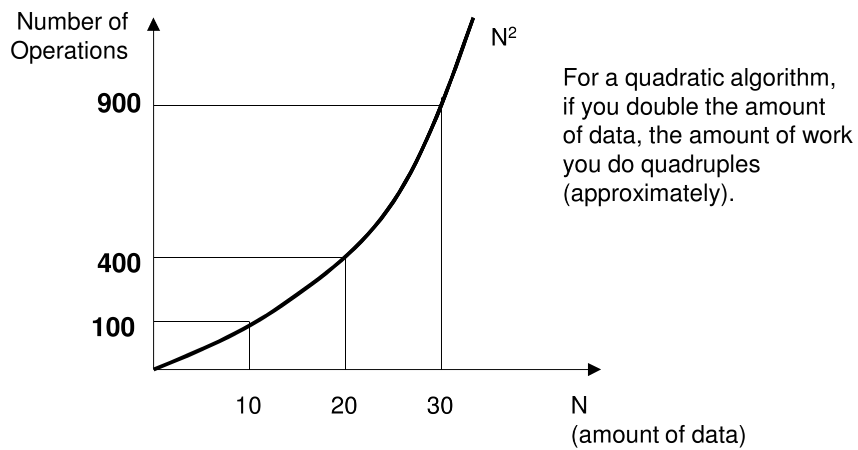
$O(n^2)$



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

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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.