

Algorithm Analysis

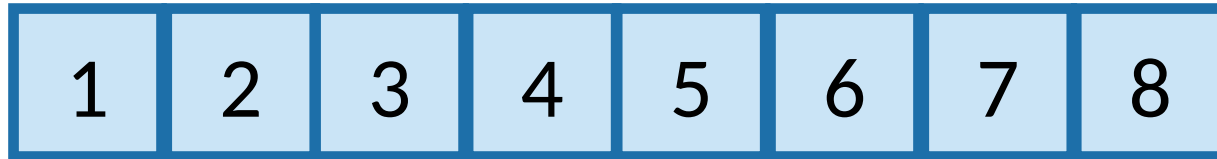
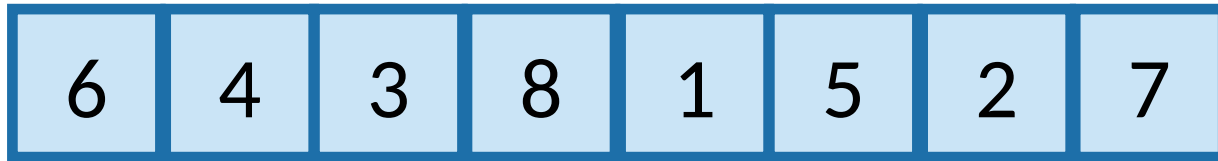
Insertion Sort

Sorting

- Arrange an unordered list of elements in some order.
- Some common algorithms
 - Bubble Sort
 - Insertion Sort
 - Merge Sort
 - Quick Sort

Sorting

- Important primitive
- For today, we'll pretend all elements are distinct.



Length of the list is n

Insertion Sort

INSERTION-SORT(A, n)

1 **for** $i = 2$ **to** n

2 $key = A[i]$

3 *// Insert $A[i]$ into the sorted subarray $A[1 : i - 1]$.*

4 $j = i - 1$

5 **while** $j > 0$ and $A[j] > key$

6 $A[j + 1] = A[j]$

7 $j = j - 1$

8 $A[j + 1] = key$

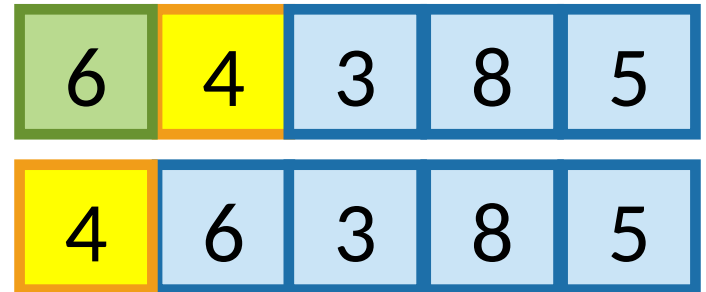
Insertion Sort

example

```
Insertion-Sort(A, n)
  for i = 1 to n - 1
    key = A[i]
    j = i - 1
    while j >= 0 and A[j] > key
      A[j + 1] = A[j]
      j = j - 1
    A[j + 1] = key
```



Start by moving A[1] toward the beginning of the list until you find something smaller (or can't go any further):



Insertion Sort

example

```
Insertion-Sort(A, n)
  for i = 1 to n - 1
    key = A[i]
    j = i - 1
    while j >= 0 and A[j] > key
      A[j + 1] = A[j]
      j = j - 1
    A[j + 1] = key
```



Then move A[2]:

key = 3



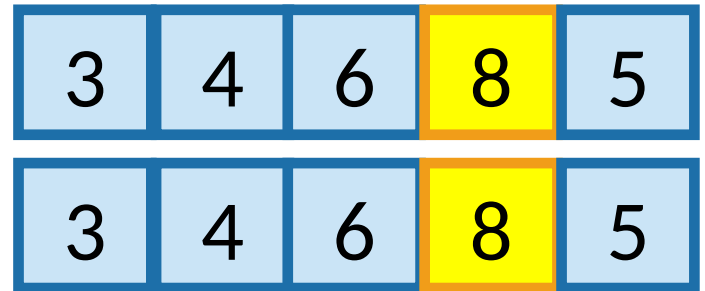
Insertion Sort

example

```
Insertion-Sort(A, n)
  for i = 1 to n - 1
    key = A[i]
    j = i - 1
    while j >= 0 and A[j] > key
      A[j + 1] = A[j]
      j = j - 1
    A[j + 1] = key
```



Then move A[3]:
key = 8



Insertion Sort

example

```
Insertion-Sort(A, n)
  for i = 1 to n - 1
    key = A[i]
    j = i - 1
    while j >= 0 and A[j] > key
      A[j + 1] = A[j]
      j = j - 1
    A[j + 1] = key
```



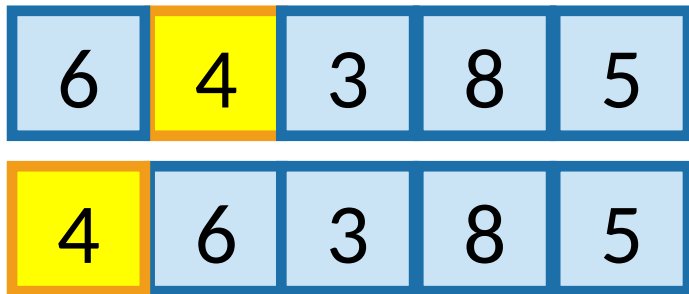
Then move A[4]:
key = 5



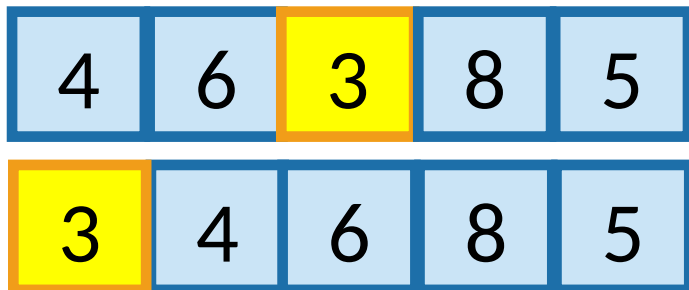
Insertion Sort

example

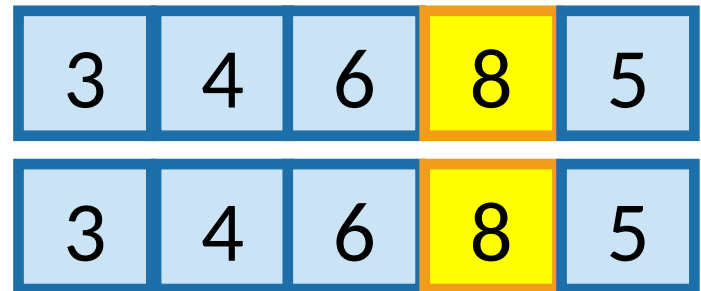
Start by moving $A[1]$ toward the beginning of the list until you find something smaller (or can't go any further):



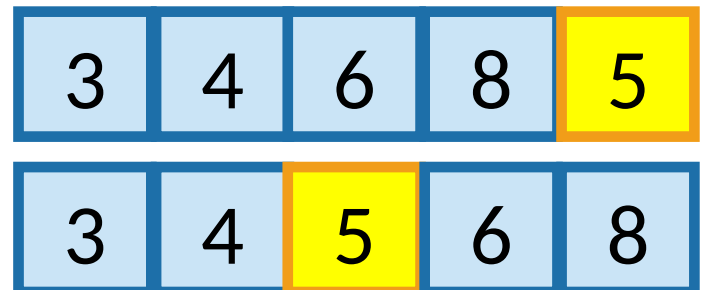
Then move $A[2]$:



Then move $A[3]$:



Then move $A[4]$:



Then we are done!

Why does this work?

- Say you have a sorted list,

3	4	6	8
---	---	---	---

, and another element

5

.
- Insert

5

 right after the largest thing that's still smaller than

5

. (Aka, right after

4

).
- Then you get a sorted list:

3	4	5	6	8
---	---	---	---	---

This sounds like a job for...

**Proof By
Induction!**

Outline of a proof by induction

Let A be a list of length n

- **Base case:**
 - $A[:1]$ is sorted at the end of the 0'th iteration. ✓
- **Inductive Hypothesis:**
 - $A[:i+1]$ is sorted at the end of the i^{th} iteration (of the outer loop).
- **Inductive step:**
 - For any $0 < k < n$, if the inductive hypothesis holds for $i=k-1$, then it holds for $i=k$.
 - Aka, if $A[:k]$ is sorted at step $k-1$, then $A[:k+1]$ is sorted at step k (previous slide)
- **Conclusion:**
 - The inductive hypothesis holds for $i = 0, 1, \dots, n-1$.
 - In particular, it holds for $i=n-1$.
 - At the end of the $n-1^{\text{st}}$ iteration (aka, at the end of the algorithm), $A[:n] = A$ is sorted.
 - That's what we wanted! ✓

Algorithm Analysis

- Estimate the resources required by an algorithm
 - Memory
 - Communication Bandwidth
 - Energy Consumption
 - Computational Time
- Helps identify the most efficient one

Algorithm Analysis (Insertion Sort)

- Timing the run of insertion sort on our computer
- We will get runtime estimates for,
 - A particular computer
 - A particular input
 - A particular implementation
 - A particular compiler/interpreter
 - Particular libraries and background tasks
- What about others?

Algorithm Analysis

- Assumptions
 - One-processor
 - Random-access machine (RAM) model of computation

Random-access Machine

- Random-access machine (RAM)
 - Instructions execute one after another
 - No concurrent operations
 - Each instructions takes the same amount of time
 - Each data access takes the same amount of time
 - Contains commonly found instructions
 - Arithmetic (add subtract, multiply, divide, remainder, floor, ceiling)
 - Data movement (load, store, copy) and
 - Control (branching, call and return)
 - Includes common data types
 - Doesn't model memory hierarchy

Algorithm Analysis (Insertion Sort)

- Runtime depends on inputs
 - Sort an array of 10000 items vs Sort an array of 3 items
- Input Size
 - Problem specific
 - For sorting, the number of items in the input
 - For multiplication, the total number of bits needed for representation
 - For graph, the number of nodes and edges

Algorithm Analysis (Insertion Sort)

- Running time of an algorithm
 - For a given input
 - The number of instructions and data access executed
- Assumption
 - Constant time taken by each line of the pseudocode

Algorithm Analysis (Insertion Sort)

INSERTION-SORT(A, n)		<i>cost</i>	<i>times</i>
1	for $i = 2$ to n	c_1	n
2	$key = A[i]$	c_2	$n - 1$
3	// Insert $A[i]$ into the sorted subarray $A[1 : i - 1]$.	0	$n - 1$
4	$j = i - 1$	c_4	$n - 1$
5	while $j > 0$ and $A[j] > key$	c_5	$\sum_{i=2}^n t_i$
6	$A[j + 1] = A[j]$	c_6	$\sum_{i=2}^n (t_i - 1)$
7	$j = j - 1$	c_7	$\sum_{i=2}^n (t_i - 1)$
8	$A[j + 1] = key$	c_8	$n - 1$

t_i denotes the number of times the **while** loop test in line 5 is executed for that value of i

Algorithm Analysis (Insertion Sort)

- $T(n)$ denote the running time of an algorithm on an input of size of n

INSERTION-SORT(A, n)		<i>cost</i>	<i>times</i>
1	for $i = 2$ to n	c_1	n
2	$key = A[i]$	c_2	$n - 1$
3	// Insert $A[i]$ into the sorted subarray $A[1 : i - 1]$.	0	$n - 1$
4	$j = i - 1$	c_4	$n - 1$
5	while $j > 0$ and $A[j] > key$	c_5	$\sum_{i=2}^n t_i$
6	$A[j + 1] = A[j]$	c_6	$\sum_{i=2}^n (t_i - 1)$
7	$j = j - 1$	c_7	$\sum_{i=2}^n (t_i - 1)$
8	$A[j + 1] = key$	c_8	$n - 1$

Algorithm Analysis (Insertion Sort)

$$T(n) = c_1n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{i=2}^n t_i + c_6 \sum_{i=2}^n (t_i - 1) \\ + c_7 \sum_{i=2}^n (t_i - 1) + c_8(n - 1) .$$

INSERTION-SORT(*A*, *n*)

	<i>cost</i>	<i>times</i>
1 for <i>i</i> = 2 to <i>n</i>	c_1	n
2 $key = A[i]$	c_2	$n - 1$
3 // Insert $A[i]$ into the sorted subarray $A[1 : i - 1]$.	0	$n - 1$
4 $j = i - 1$	c_4	$n - 1$
5 while $j > 0$ and $A[j] > key$	c_5	$\sum_{i=2}^n t_i$
6 $A[j + 1] = A[j]$	c_6	$\sum_{i=2}^n (t_i - 1)$
7 $j = j - 1$	c_7	$\sum_{i=2}^n (t_i - 1)$
8 $A[j + 1] = key$	c_8	$n - 1$

Algorithm Analysis (Insertion Sort)

$$T(n) = c_1n + c_2(n-1) + c_4(n-1) + c_5 \sum_{i=2}^n t_i + c_6 \sum_{i=2}^n (t_i - 1) \\ + c_7 \sum_{i=2}^n (t_i - 1) + c_8(n-1) .$$

- Best Case
 - The array is sorted
 - $t_i = 1$ for all $i = 2, 3, \dots, n$

$$T(n) = c_1n + c_2(n-1) + c_4(n-1) + c_5(n-1) + c_8(n-1) \\ = (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8) .$$

Algorithm Analysis (Insertion Sort)

$$T(n) = c_1n + c_2(n-1) + c_4(n-1) + c_5 \sum_{i=2}^n t_i + c_6 \sum_{i=2}^n (t_i - 1) \\ + c_7 \sum_{i=2}^n (t_i - 1) + c_8(n-1).$$

- Best Case

- The array is sorted

- $t_i = 1$ for all $i = 2, 3, \dots, n$

A linear function of n

$$T(n) = an + b$$

where $a = c_1 + c_2 + c_4 + c_5 + c_8$

and $b = -(c_2 + c_4 + c_5 + c_8)$

Algorithm Analysis (Insertion Sort)

$$T(n) = c_1n + c_2(n-1) + c_4(n-1) + c_5 \sum_{i=2}^n t_i + c_6 \sum_{i=2}^n (t_i - 1) \\ + c_7 \sum_{i=2}^n (t_i - 1) + c_8(n-1).$$

- Worst Case
 - The array is sorted in reverse order
 - $t_i = i$ for all $i = 2, 3, \dots, n$

$$T(n) = c_1n + c_2(n-1) + c_4(n-1) + c_5 \left(\frac{n(n+1)}{2} - 1 \right) \\ + c_6 \left(\frac{n(n-1)}{2} \right) + c_7 \left(\frac{n(n-1)}{2} \right) + c_8(n-1) \\ = \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2} \right) n^2 + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8 \right) n \\ - (c_2 + c_4 + c_5 + c_8).$$


Algorithm Analysis (Insertion Sort)

$$T(n) = c_1n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{i=2}^n t_i + c_6 \sum_{i=2}^n (t_i - 1) \\ + c_7 \sum_{i=2}^n (t_i - 1) + c_8(n - 1) .$$

- Best Case

- The array is sorted
- $t_i = i$ for all $i = 2, 3, \dots, n$

A quadratic function of n


$$T(n) = an^2 + bn + c$$

where a, b and c are constants

Algorithm Analysis (Insertion Sort)

- Instead of exact function, we estimate the **rate of growth** or the **order of growth**
- Best Case
 - Most significant term an
 - Linear
- Worst Case
 - Most significant term an^2
 - Quadratic

Reference

- CLRS Chapter 2
 - Sections 2.1, 2.2