

CS141: Intermediate Data Structures and Algorithms

Analysis of Algorithms

Amr Magdy

Analyzing Algorithms

1. Algorithm Correctness
 - a. Termination
 - b. Produces the correct output for all possible input.

2. Algorithm Performance
 - a. Either runtime analysis,
 - b. or storage (memory) space analysis
 - c. or both

Algorithm Correctness



- › Sorting problem
 - › Input: an array A of n numbers
 - › Output: the same array in ascending sorted order (smallest number in $A[1]$ and largest in $A[n]$)

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- › Insertion Sort

INSERTION-SORT(A, n)

for $j = 2$ **to** n

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 // Insert $A[j]$ into the sorted sequence $A[1 .. j - 1]$.

$i = j - 1$

while $i > 0$ and $A[i] > key$

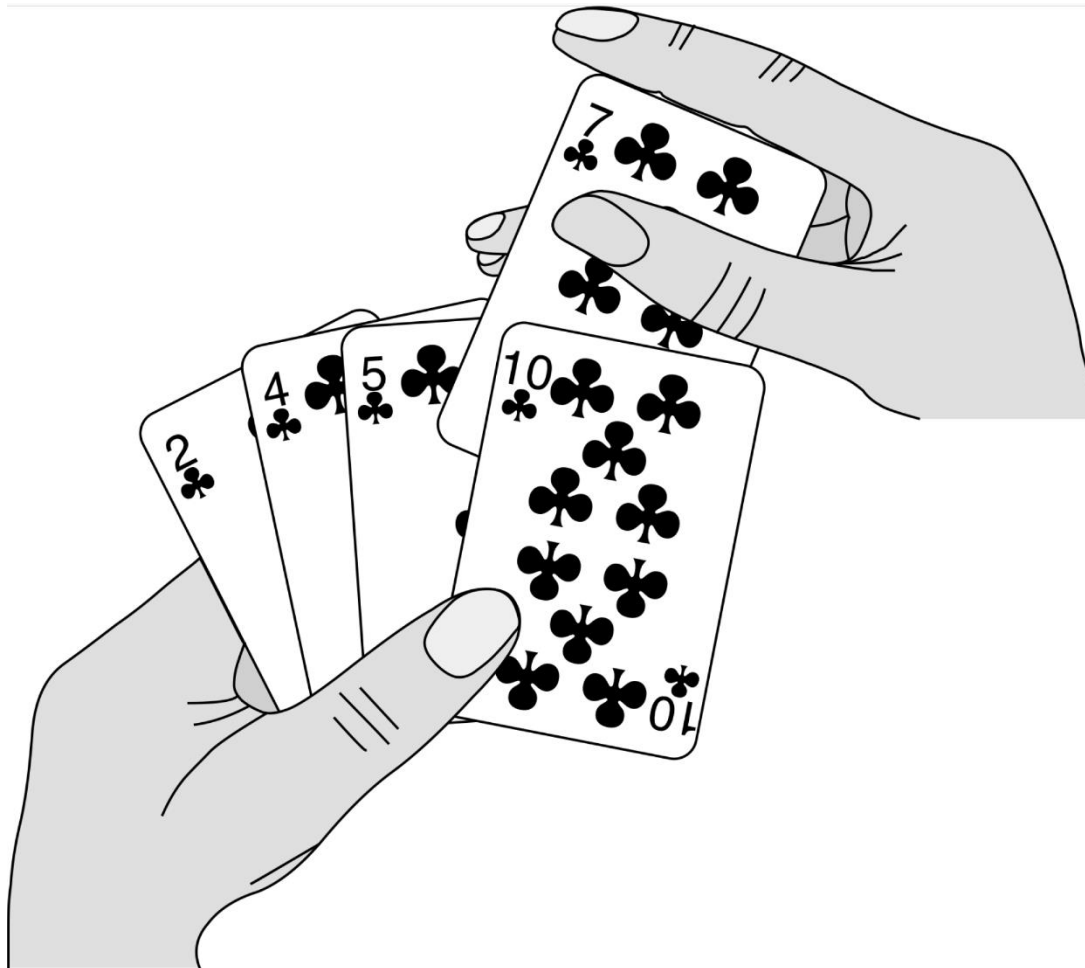
$A[i + 1] = A[i]$

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Algorithm Correctness

- › How does insertion sort work?



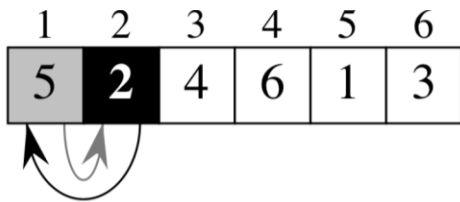
Algorithm Correctness



5	2	4	6	1	3
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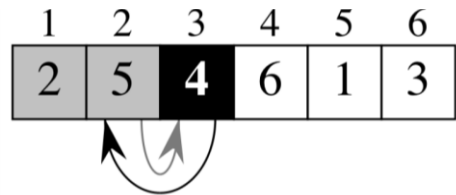
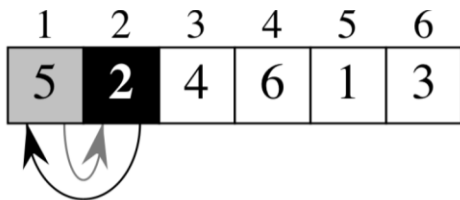
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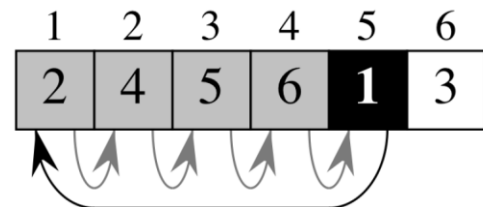
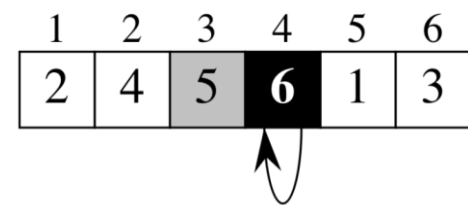
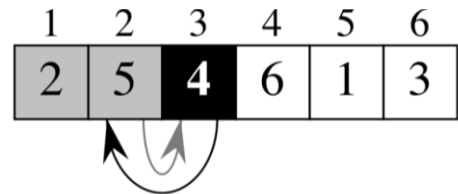
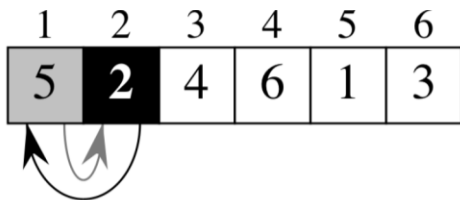
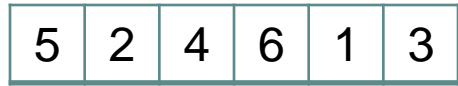
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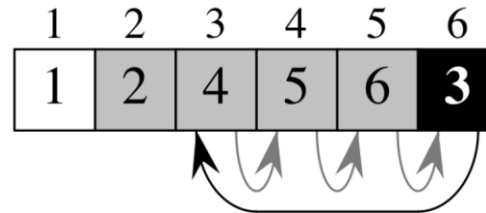
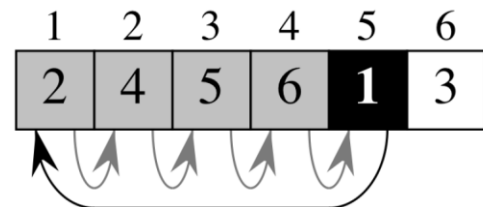
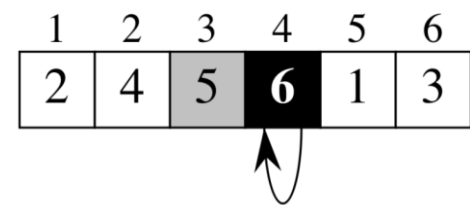
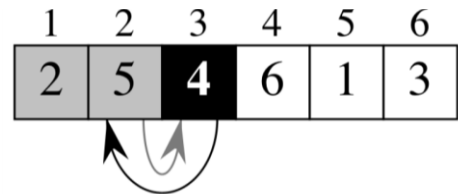
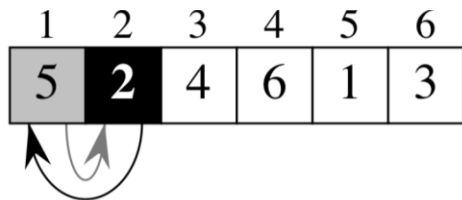
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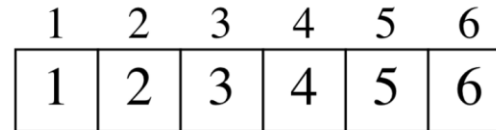
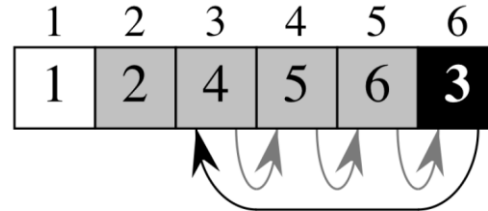
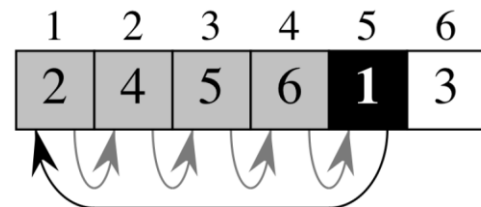
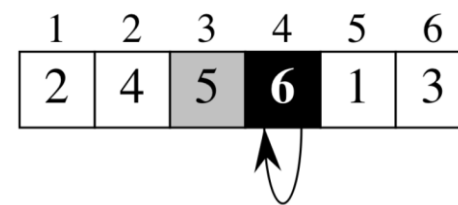
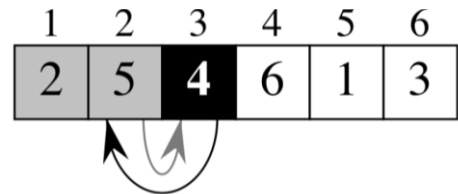
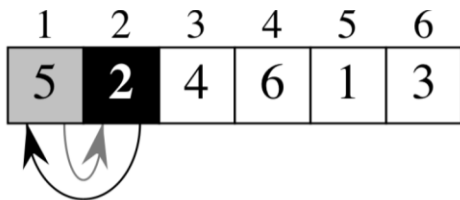
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 - › It is a property that is true before and after each loop iteration.

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- › Loop invariant:
 - › It is a property that is true before and after each loop iteration.
- › Insertion sort loop invariant (ISLI):
 - › The first $(j-1)$ array elements $A[1..j-1]$ are:
(a) the original $(j-1)$ elements, and (b) sorted.

INSERTION-SORT(A, n)

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- › Is insertion sort a correct algorithm?
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 - › Halts and produces the correct output

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- › Loop invariant (LI) correctness
 1. Initialization:
 - LI is true prior to the 1st iteration.
 2. Maintenance:
 - If LI true before the iteration, it remains true before the next iteration
 3. Termination:
 - After the loop terminates, the output is correct.

Algorithm Correctness

- ▶ ISLI: The first $(j-1)$ array elements $A[1..j-1]$ are:
(a) the original $(j-1)$ elements, and (b) sorted.

1. Initialization:

Prior to the 1st iteration, $j=2$, the first $(2-1)$ is sorted by definition.

2. Maintenance:

The $(j-1)$ th iteration inserts the j th element in a sorted order, so after the iteration, the first $(j-1)$ elements remains the same and sorted.

3. Termination:

The loop terminates after $(n-1)$ iterations, $j=n+1$, so the first n elements are sorted, then the output is correct.

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Correct

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Algorithms Performance Analysis



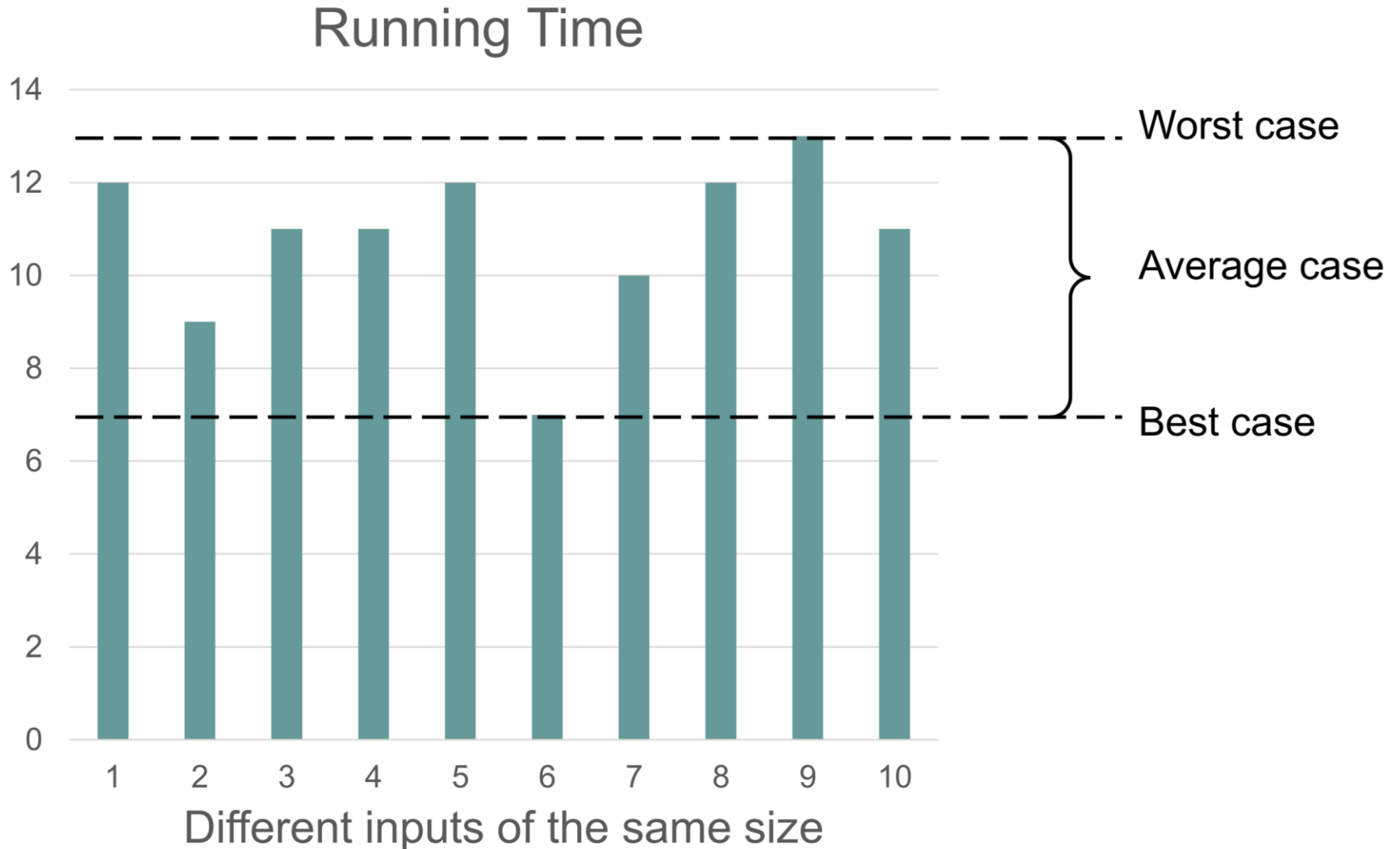
- › Which criteria should be taken into account?
- › Running time
- › Memory footprint
- › Disk IO
- › Network bandwidth
- › Power consumption
- › Lines of codes
- › ...

Algorithms Performance Analysis

› Which criteria should be taken into account?

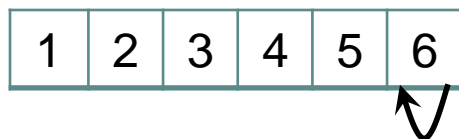
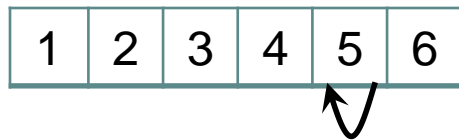
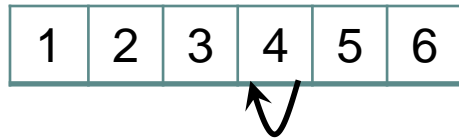
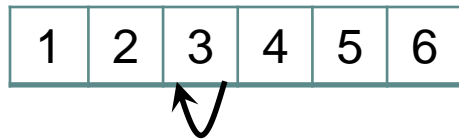
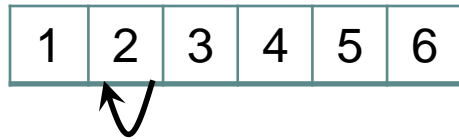
- › Running time
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Average Case vs. Worst Case



Insertion Sort Best Case

- › Input array is sorted



Insertion Sort Best Case

- Input array is sorted



INSERTION-SORT(A, n)

```

for  $j = 2$  to  $n$  ..... c1
     $key = A[j]$  ..... c2
    // Insert  $A[j]$  into the sorted sequence  $A[1 .. j - 1]$  ..... 0
     $i = j - 1$  ..... c3
    while  $i > 0$  and  $A[i] > key$  ..... c4
         $A[i + 1] = A[i]$ 
         $i = i - 1$ 
        } do not execute ..... 0 } 1
     $A[i + 1] = key$  ..... c5
    
```

(n-1)

Insertion Sort Best Case



- Input array is sorted



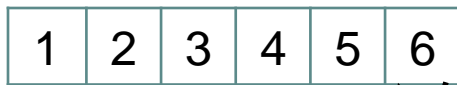
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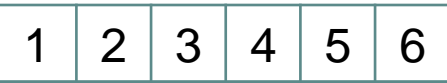
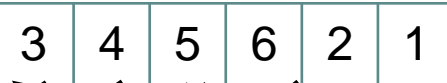
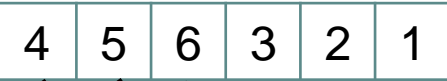
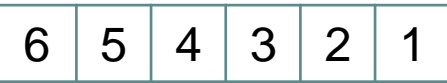


$$T(n) = (n-1) * (c1 + c2 + 0 + c3 + 1 * (c4 + 0) + c5)$$

$$T(n) = cn - c, \quad \text{const } c = c1 + c2 + c3 + c4 + c5$$

Insertion Sort Worst Case

- › Input array is reversed



Insertion Sort Worst Case

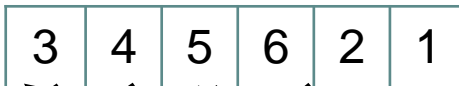
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INSERTION-SORT(A, n)

```

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    while  $i > 0$  and  $A[i] > key$  .....c4
         $A[i + 1] = A[i]$  .....c5
         $i = i - 1$  .....c6
     $A[i + 1] = key$  .....c7
  
```

} (n-1)
} i



Insertion Sort Worst Case

- Input array is reversed



INSERTION-SORT(A, n)

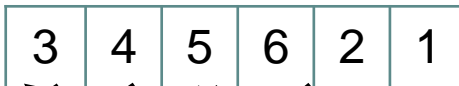


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for  $j = 2$  to  $n$  .....  $c_1$ 
     $key = A[j]$  .....  $c_2$ 
    // Insert  $A[j]$  into the sorted sequence  $A[1..j - 1]$  .....  $0$ 
     $i = j - 1$  .....  $c_3$ 
    while  $i > 0$  and  $A[i] > key$  .....  $c_4$ 
         $A[i + 1] = A[i]$  .....  $c_5$ 
         $i = i - 1$  .....  $c_6$ 
     $A[i + 1] = key$  .....  $c_7$ 
    
```

} (n-1)

} i



$$T(n) = (n-1) * (c_1 + c_2 + 0 + c_3 + i * (c_4 + c_5 + c_6) + c_7)$$

$$T(n) = (n-1) * (c_1 + c_2 + 0 + c_3 + c_7) + \sum i * (c_4 + c_5 + c_6), \text{ for all } 1 \leq i < n$$

$$T(n) = (cn - c) + \sum i * d, \text{ c \& d are constants}$$

$$\sum i * d = 1 * d + 2 * d + 3 * d + \dots + (n-1) * d = d * (1 + 2 + 3 + \dots + (n-1)) = d * n(n-1) / 2$$

$$T(n) = (cn - c) + dn^2 / 2 - dn / 2 = d * n^2 + c_{11} * n + c_{12}, \text{ c's \& d are consts}$$

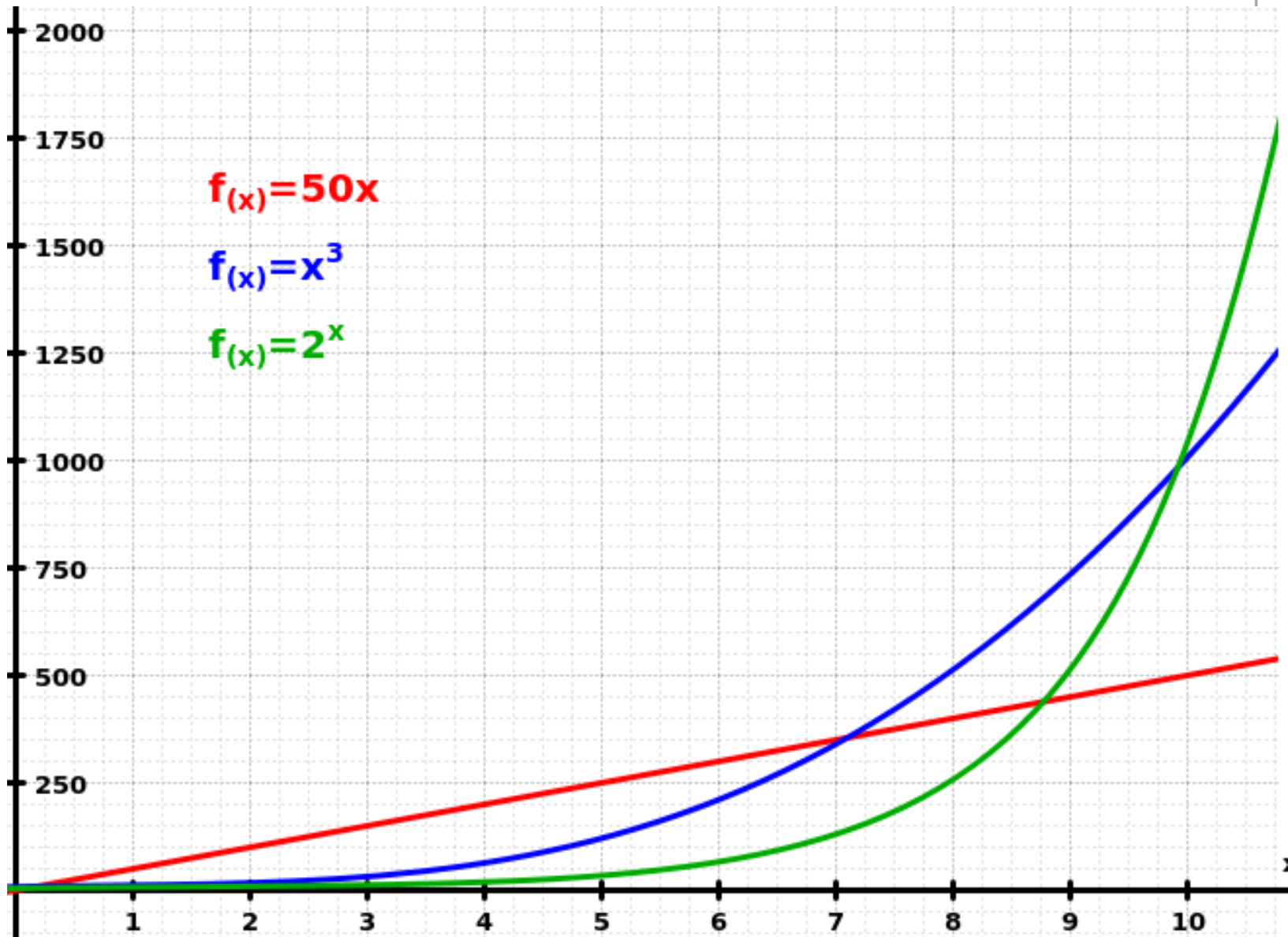
Insertion Sort Average Case

- › Average = (Best + Worst)/2
- › $T(n) = cn^2 + dn + e$, c, d, e are consts

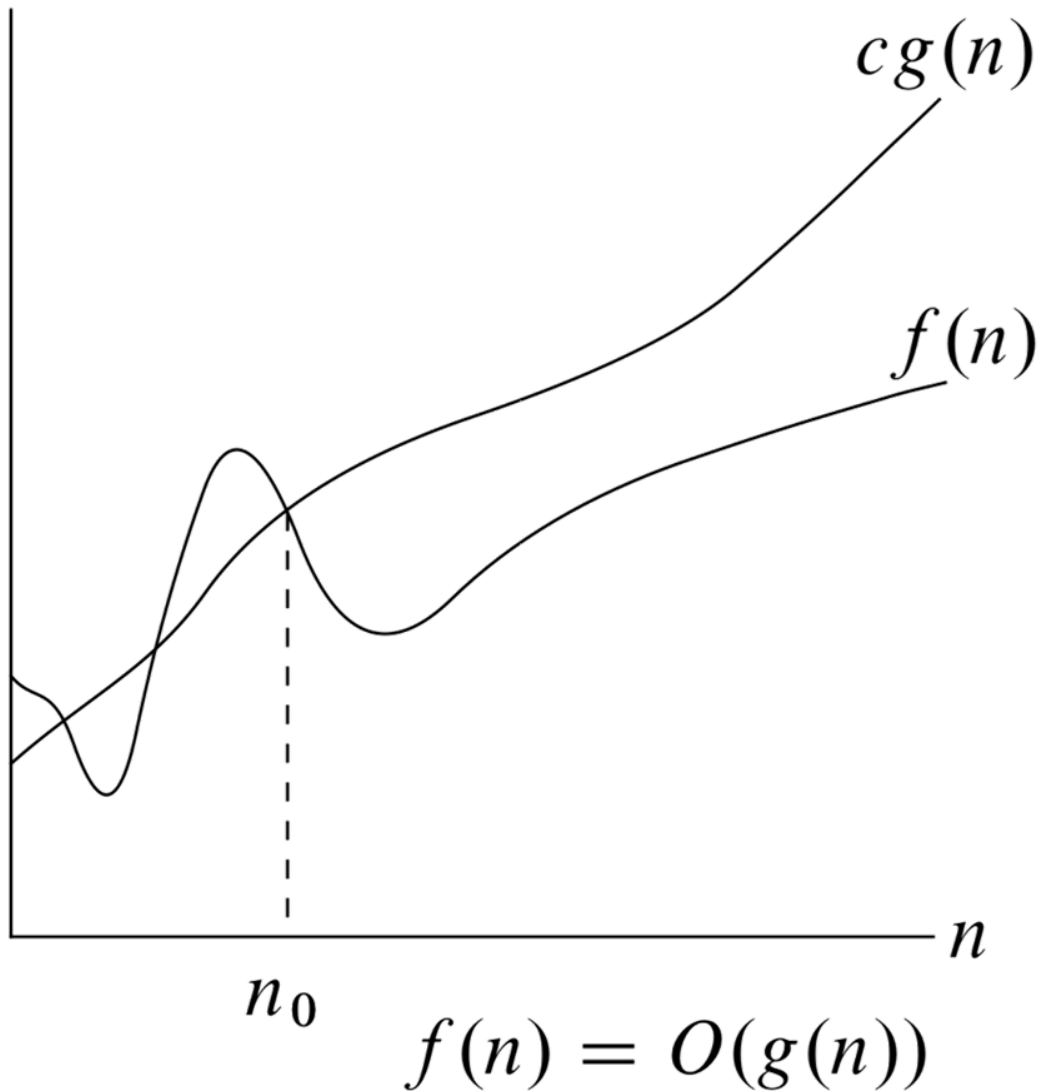
Growth of Functions

- › It is hard to compute the actual running time for more complex algorithms
- › The cost of the worst-case is a good measure
- › The growth of the cost function is what interests us (when input size is large)
- › We are more concerned with comparing two cost functions, i.e., two algorithms.

Growth of Functions



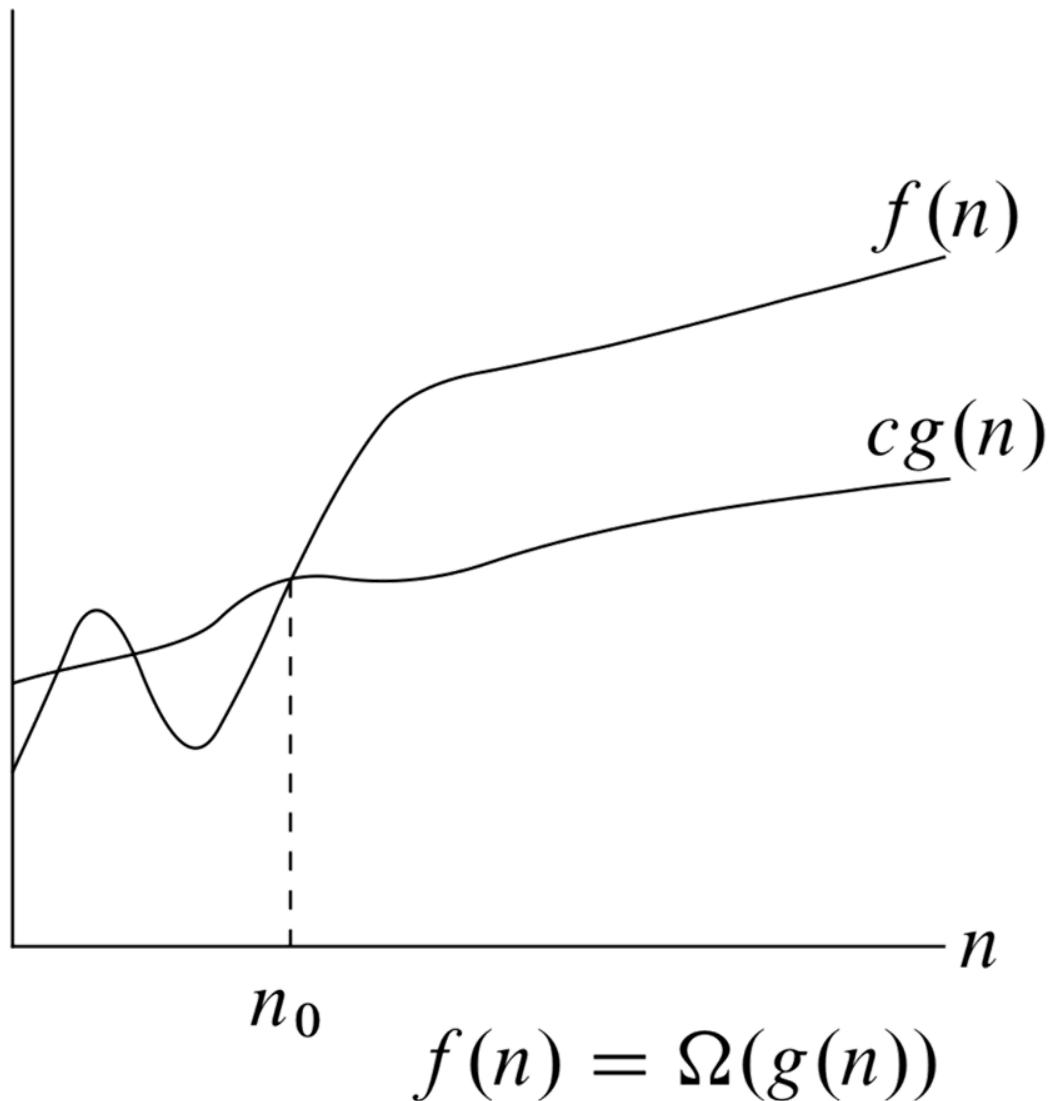
O-notation



$$\begin{aligned} \exists c > 0, n_0 > 0 \\ 0 \leq f(n) \leq cg(n) \\ n \geq n_0 \end{aligned}$$

$g(n)$ is an asymptotic **upper-bound** for $f(n)$

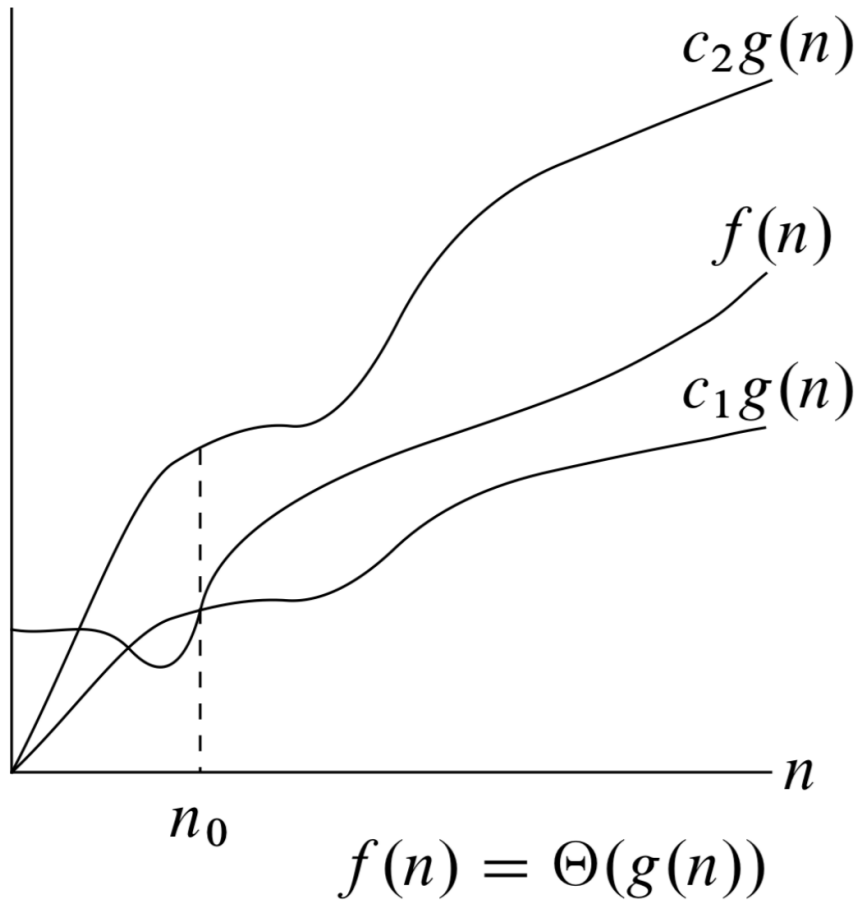
Ω -notation



$$\begin{aligned} \exists c > 0, n_0 > 0 \\ 0 \leq cg(n) \leq f(n) \\ n \geq n_0 \end{aligned}$$

$g(n)$ is an asymptotic **lower**-bound for $f(n)$

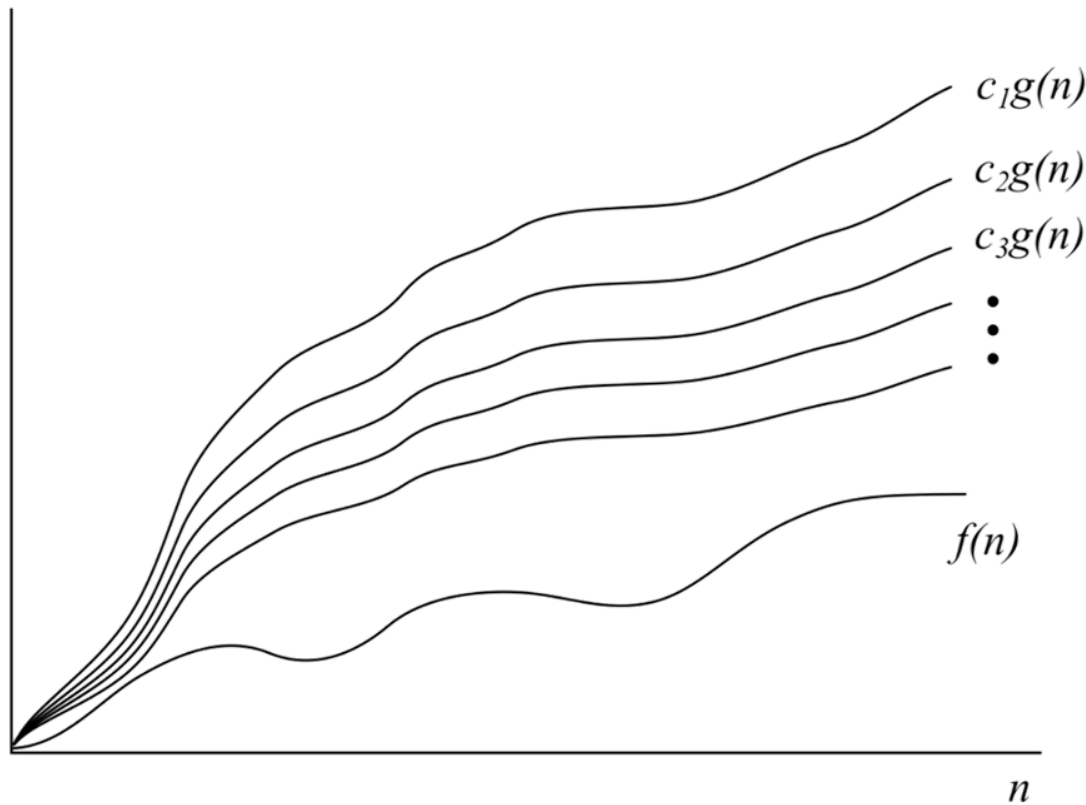
Θ -notation



$$\begin{aligned} &\exists c_1, c_2 > 0, n_0 > 0 \\ &0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \\ &n \geq n_0 \end{aligned}$$

$g(n)$ is an asymptotic **tight**-bound for $f(n)$

o-notation



$$f(n) = o(g(n))$$

$$\forall c > 0$$

$$\exists n_0 > 0$$

$$0 \leq f(n) \leq cg(n)$$

$$n \geq n_0$$

$g(n)$ is a **non-tight**
asymptotic **upper-**
bound for $f(n)$

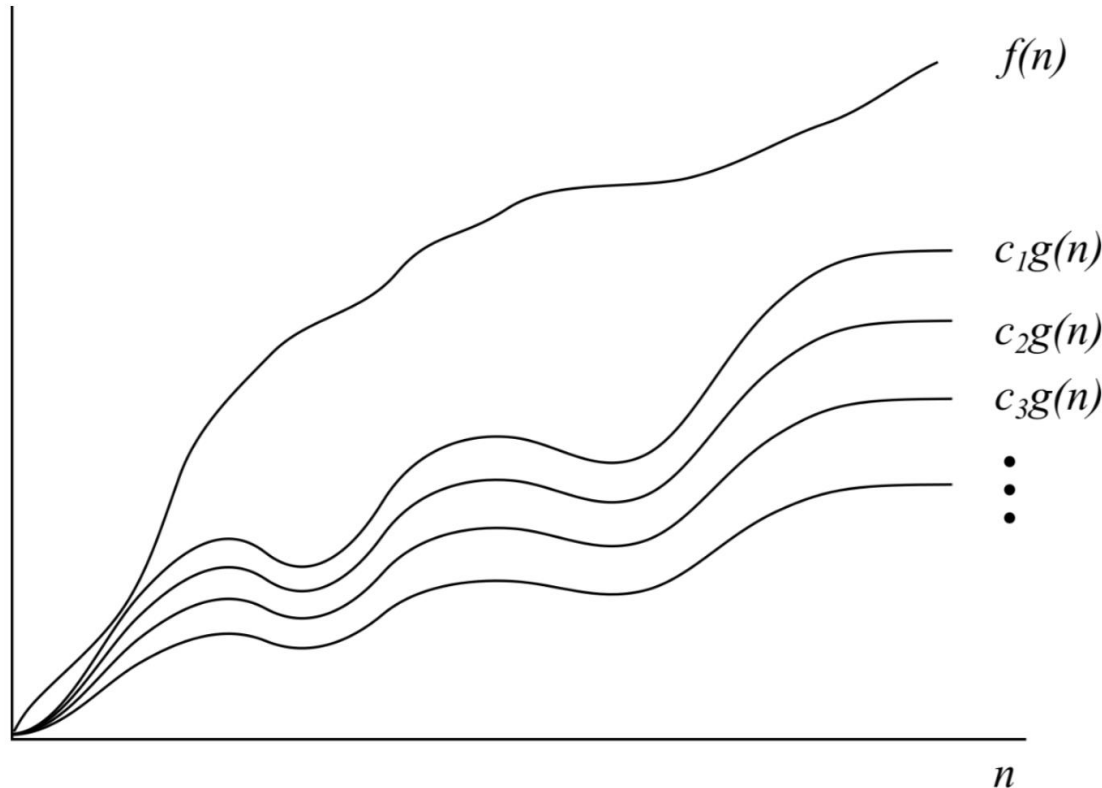
ω -notation

$$\forall c > 0$$

$$\exists n_0 > 0$$

$$0 \leq cgn(n) \leq f(n)$$

$$n \geq n_0$$



$$f(n) = \omega(g(n))$$

$g(n)$ is a **non-tight** asymptotic **lower-bound** for $f(n)$

Comparing Two Functions

- > $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$
- > 0: $f(n) = o(g(n))$
- > $c > 0$: $f(n) = \Theta(g(n))$
- > ∞ : $f(n) = \omega(g(n))$

Analogy to Real Numbers

Functions	Real numbers
$f(n) = O(g(n))$	$a \leq b$
$f(n) = \Omega(g(n))$	$a \geq b$
$f(n) = \Theta(g(n))$	$a = b$
$f(n) = o(g(n))$	$a < b$
$f(n) = \omega(g(n))$	$a > b$

Simple Rules

- › We can omit constants
- › We can omit lower order terms
- › $\Theta(an^2+bn+c)$ becomes $\Theta(n^2)$
- › $\Theta(c_1)$ and $\Theta(c_2)$ become $\Theta(1)$
- › $\Theta(\log_{k_1} n)$ and $\Theta(\log_{k_2} n)$ become $\Theta(\log n)$
- › $\Theta(\log(n^k))$ becomes $\Theta(\log n)$
- › $\log^{k_1}(n) = o(n^{k_2})$ for any positive constants k_1 and k_2

Popular Classes of Functions

Constant: $f(n) = \Theta(1)$

Logarithmic: $f(n) = \Theta(\lg(n))$

Sublinear: $f(n) = o(n)$

Linear: $f(n) = \Theta(n)$

Super-linear: $f(n) = \omega(n)$

Quadratic: $f(n) = \Theta(n^2)$

Polynomial: $f(n) = \Theta(n^k)$; k is a constant

Exponential: $f(n) = \Theta(k^n)$; k is a constant

Insertion Sort Worst Case (Revisit)

- Input array is reversed

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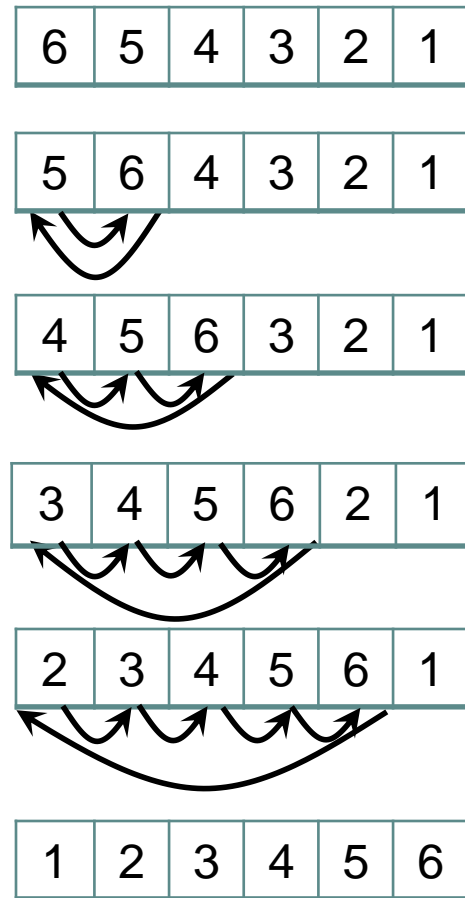
$i = i - 1$

$A[i + 1] = key$

max n

(n-1)

$$T(n) = (n-1) * n = O(n^2)$$



Comparing two algorithms

- › $T1(n) = 2n + 10000000$
- › $T2(n) = 200n + 1000$
- › Which is better? Why?
 - › In terms of order of growth?
 - › In terms of actual runtime?
- › What is the main usage of asymptotic notation analysis?

Analyzing Algorithms



```
> Algorithm 1
    for i = 1 to n
        j = 2*i
    for j = 1 to n/2
        print j
```

Analyzing Algorithms

- › Algorithm 2

```
    for i = 1 to n/2
```

```
        for j = 1 to n, step j = j*2
```

```
            print i*j
```

Analyzing Algorithms



› Algorithm 3

input x (+ve integer)

while $x > 0$

 print x

$x = \lfloor x/5 \rfloor$

Credits & Book Readings

- › Book Readings
 - › 2.1, 2.2, 3.1, 3.2
- › Credits
 - › Prof. Ahmed Eldawy notes
 - › <http://www.cs.ucr.edu/~eldawy/17WCS141/slides/CS141-1-09-17.pdf>
 - › Online websites
 - › <https://commons.wikimedia.org/wiki/File:Exponential.svg>