

$$F = G \frac{m_1 m_2}{d^2}$$

Design and Analysis of Algorithms

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$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}$$

$$\frac{df}{dt} = \lim_{h \rightarrow 0} \frac{f(t+h) - f(t)}{h}$$

$$\phi(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$dS \geq 0$$

$$E = mc^2$$

$$F - E + V = 2$$

Properties of $O(n)$

Properties of the O notation

- Constant factors may be ignored
 - $\forall k > 0, kf$ is $O(f)$
- Higher powers grow faster
 - n^r is $O(n^s)$ if $0 \leq r \leq s$
- Fastest growing term dominates a sum
 - If f is $O(g)$, then $f + g$ is $O(g)$
 - eg $an^4 + bn^3$ is $O(n^4)$
- Polynomial's growth rate is determined by leading term
 - If f is a polynomial of degree d , then f is $O(n^d)$

Properties of the O notation

- f is $O(g)$ is transitive
 - If f is $O(g)$ and g is $O(h)$ then f is $O(h)$
- Product of upper bounds is upper bound for the product
 - If f is $O(g)$ and h is $O(r)$ then fh is $O(gr)$
- Exponential functions grow faster than powers
 - n^k is $O(b^n) \forall b > 1$ and $k \geq 0$
e.g. n^{20} is $O(1.05^n)$

Important!

- Logarithms grow more slowly than powers
 - $\log_b n$ is $O(n^k) \forall b > 1$ and $k > 0$
e.g. $\log_2 n$ is $O(n^{0.5})$
- All logarithms grow at the same rate
 - $\log_b n$ is $O(\log_d n) \forall b, d > 1$

Order of growth

- We usually consider one algorithm to be more efficient than another if its worst case running time has a lower order of growth.
- Due to constant factors and lower order terms, an algorithm whose running time has a higher order of growth might take less time for small inputs than an algorithm whose running time has a lower order of growth. But for large enough inputs, $\Theta(n^2)$ algorithm, for example, will run more quickly in the worst case than $\Theta(n^3)$ algorithm



The Sorting Problem

- **Input:**

- A sequence of n numbers a_1, a_2, \dots, a_n

- **Output:**

- A permutation (reordering) a_1', a_2', \dots, a_n' of the input sequence such that $a_1' \leq a_2' \leq \dots \leq a_n'$

Why Study Sorting Algorithms?

- There are a variety of situations that we can encounter
 - Do we have randomly ordered keys?
 - Are all keys distinct?
 - How large is the set of keys to be ordered?
 - Need guaranteed performance?
- Various algorithms are better suited to some of these situations

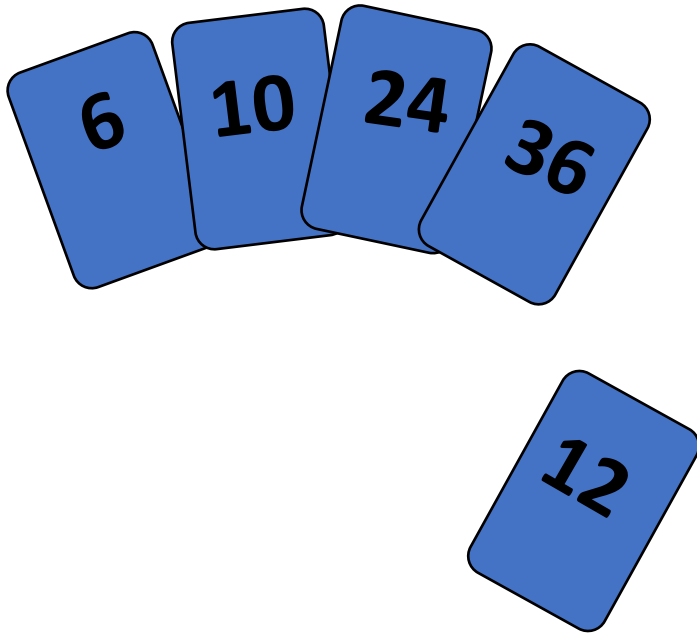
Insertion Sort

- Idea: like sorting a hand of playing cards
 - Start with an empty left hand and the cards facing down on the table.
 - Remove one card at a time from the table, and insert it into the correct position in the left hand
 - compare it with each of the cards already in the hand, from right to left
 - The cards held in the left hand are sorted
 - these cards were originally the top cards of the pile on the table

Insertion Sort

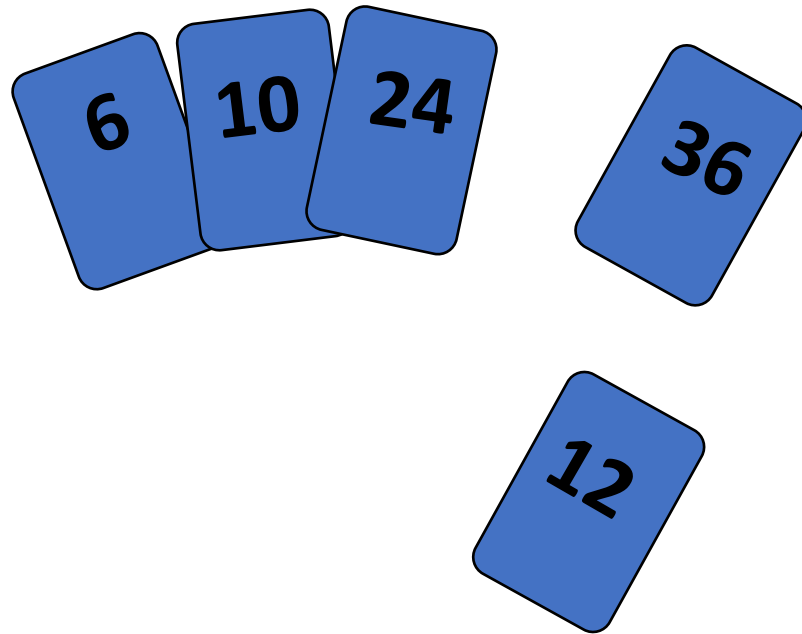
- The list is divided into two parts: sorted and unsorted
- In each pass, the following steps are performed
 - First element of the unsorted part (i.e., sub-list) is picked up
 - Transferred to the sorted sub-list
 - Inserted at the appropriate place
- A list of n elements will take at most $n-1$ passes to sort the data

Insertion Sort

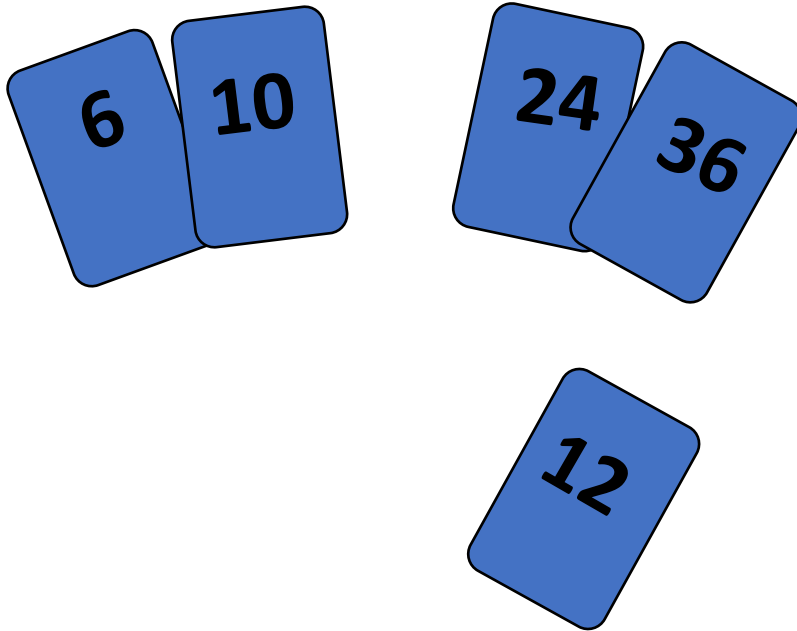


To insert 12, we need to make room for it by moving first 36 and then 24.

Insertion Sort



Insertion Sort



Example of Insertion Sort

8 2 4 9 3 6

Example of Insertion Sort



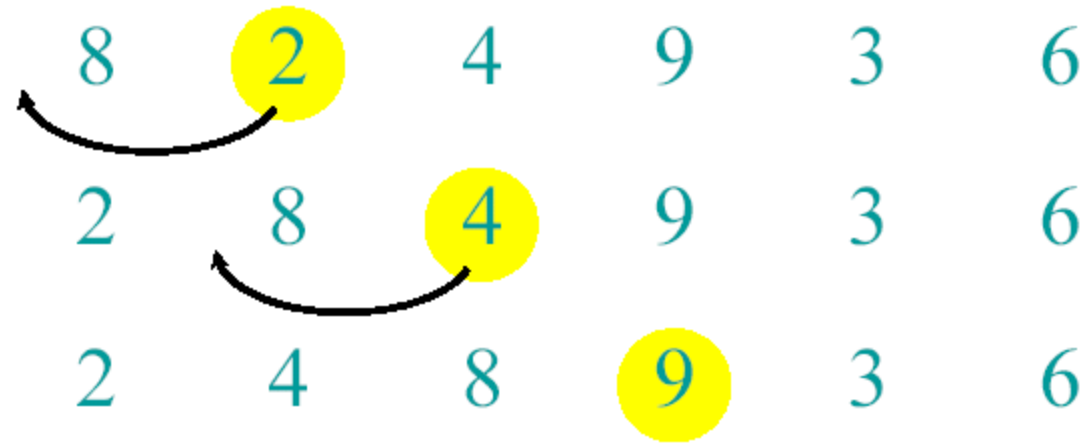
Example of Insertion Sort



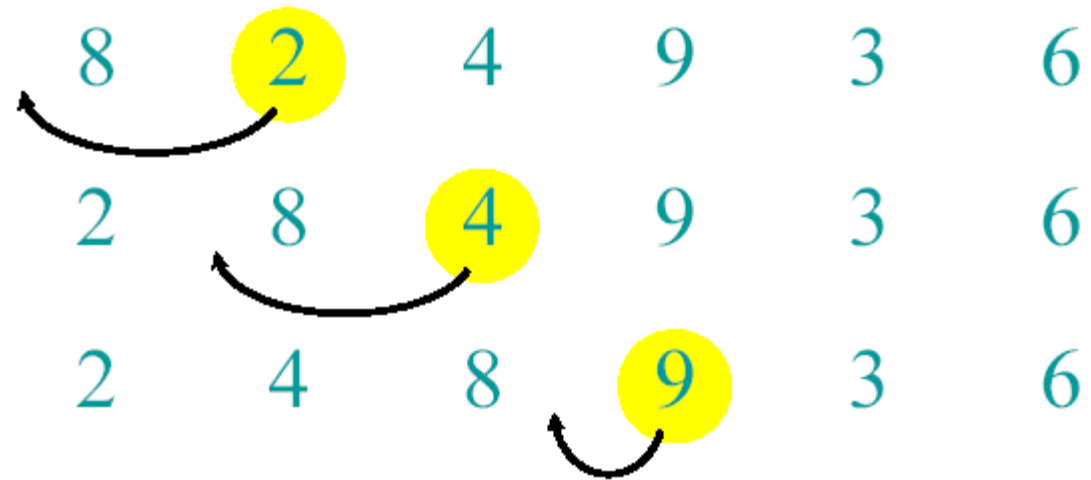
Example of Insertion Sort



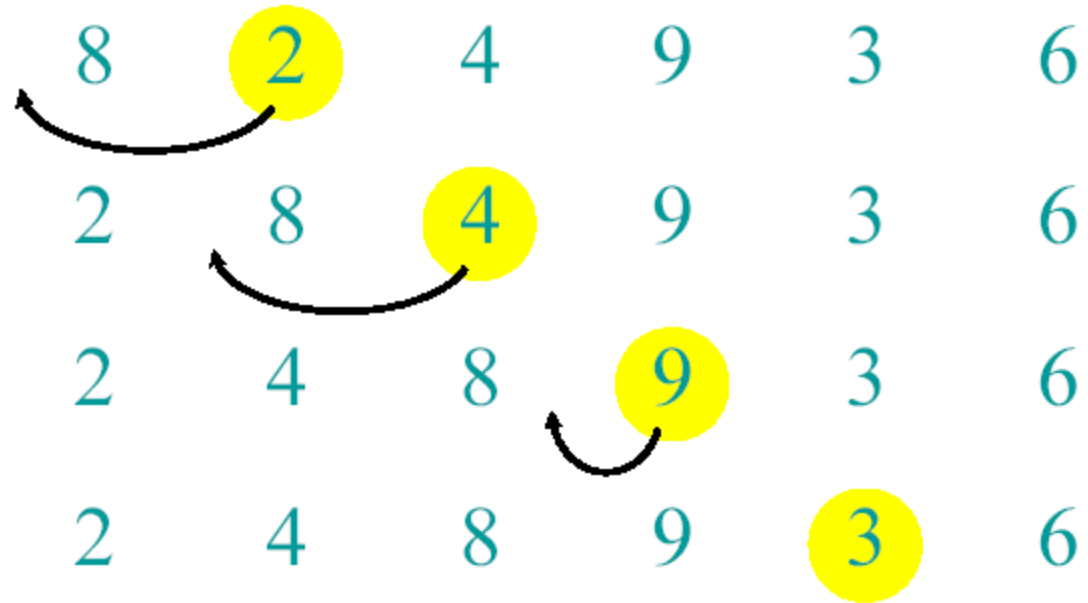
Example of Insertion Sort



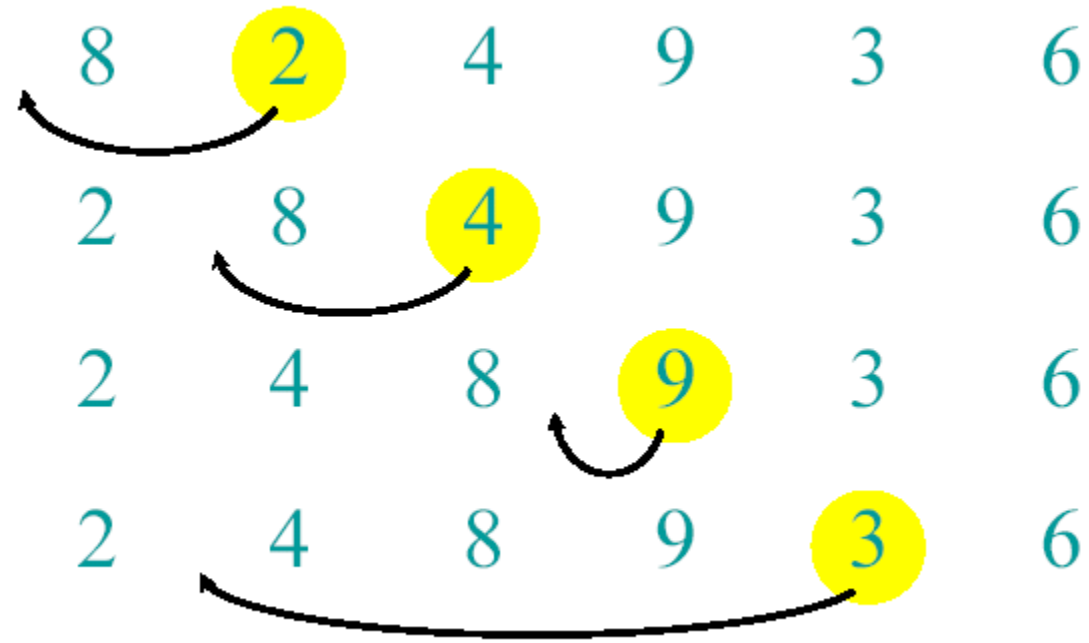
Example of Insertion Sort



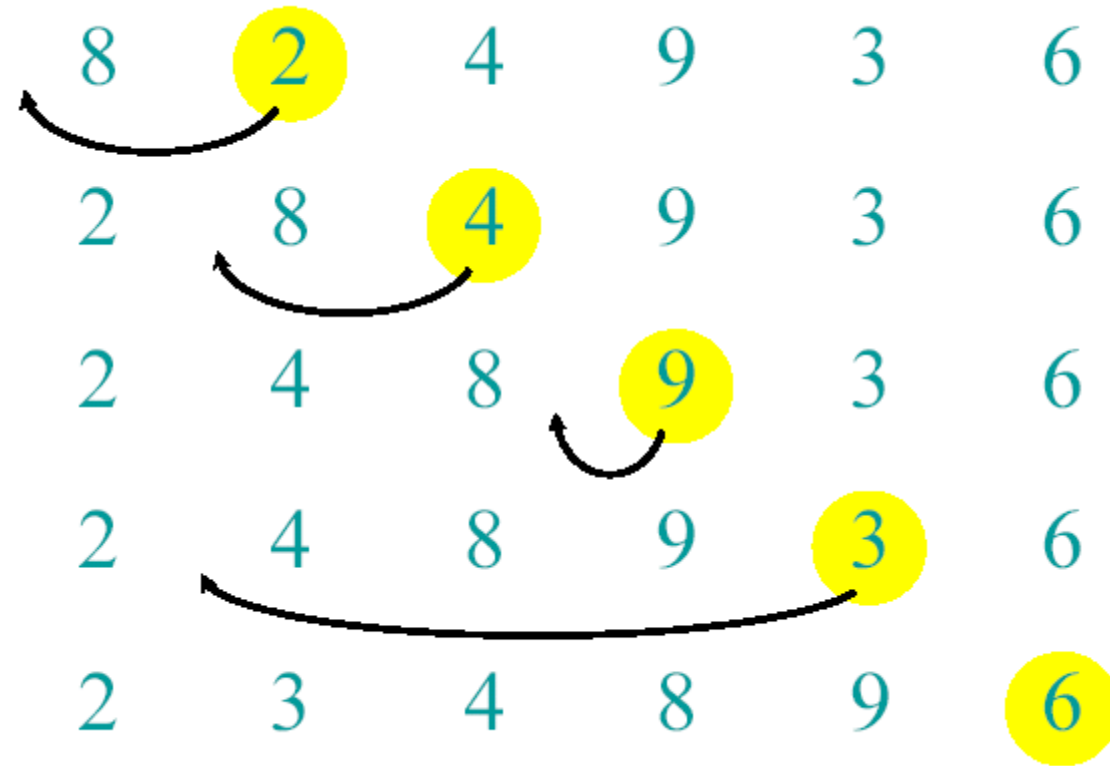
Example of Insertion Sort



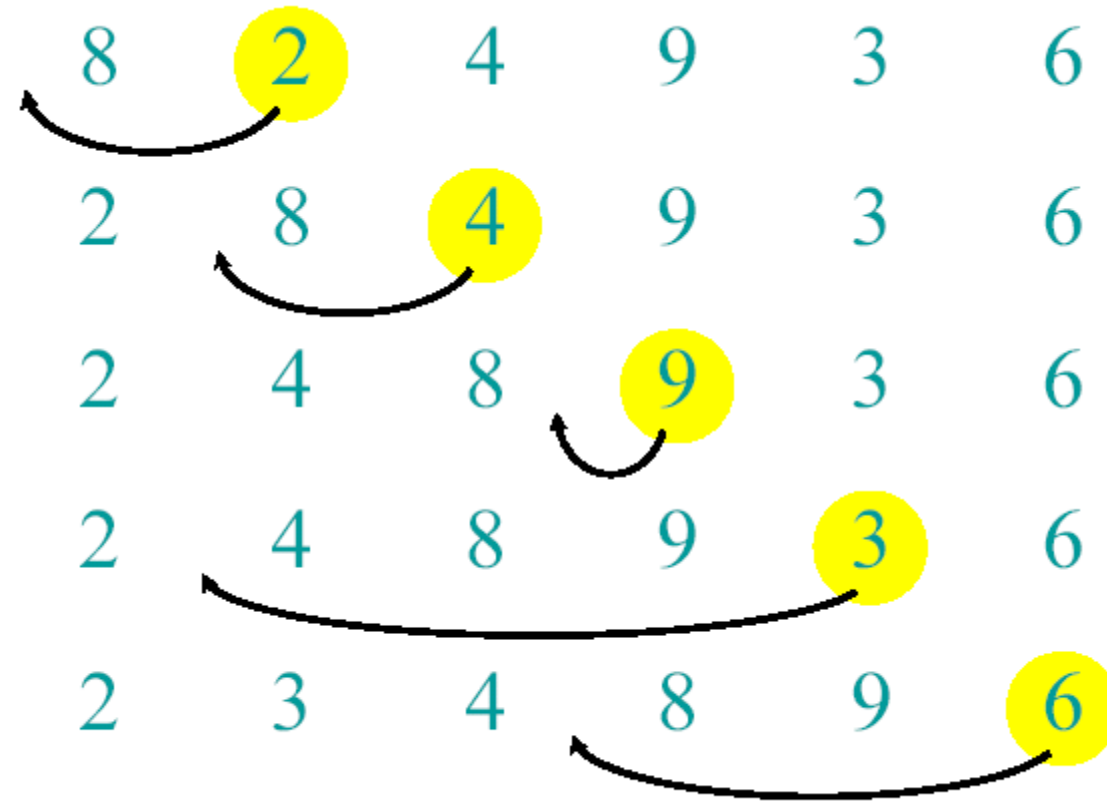
Example of Insertion Sort



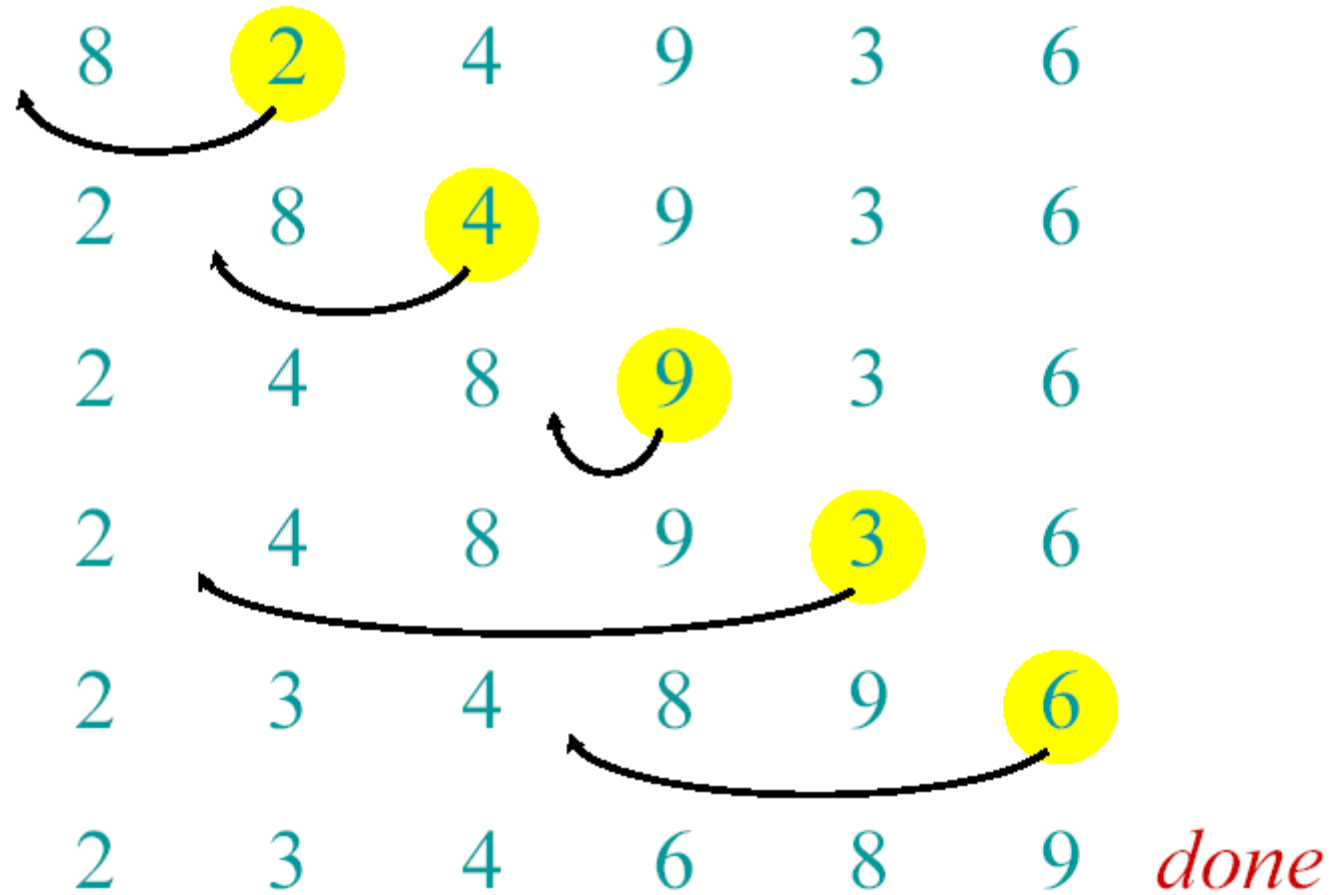
Example of Insertion Sort



Example of Insertion Sort



Example of Insertion Sort



Insertion Sort

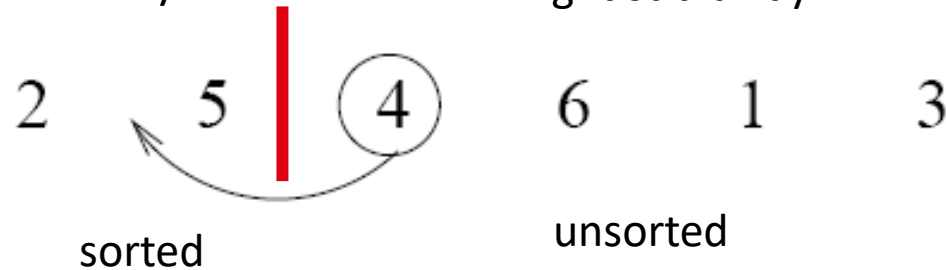
input array

5 2 4 6 1 3

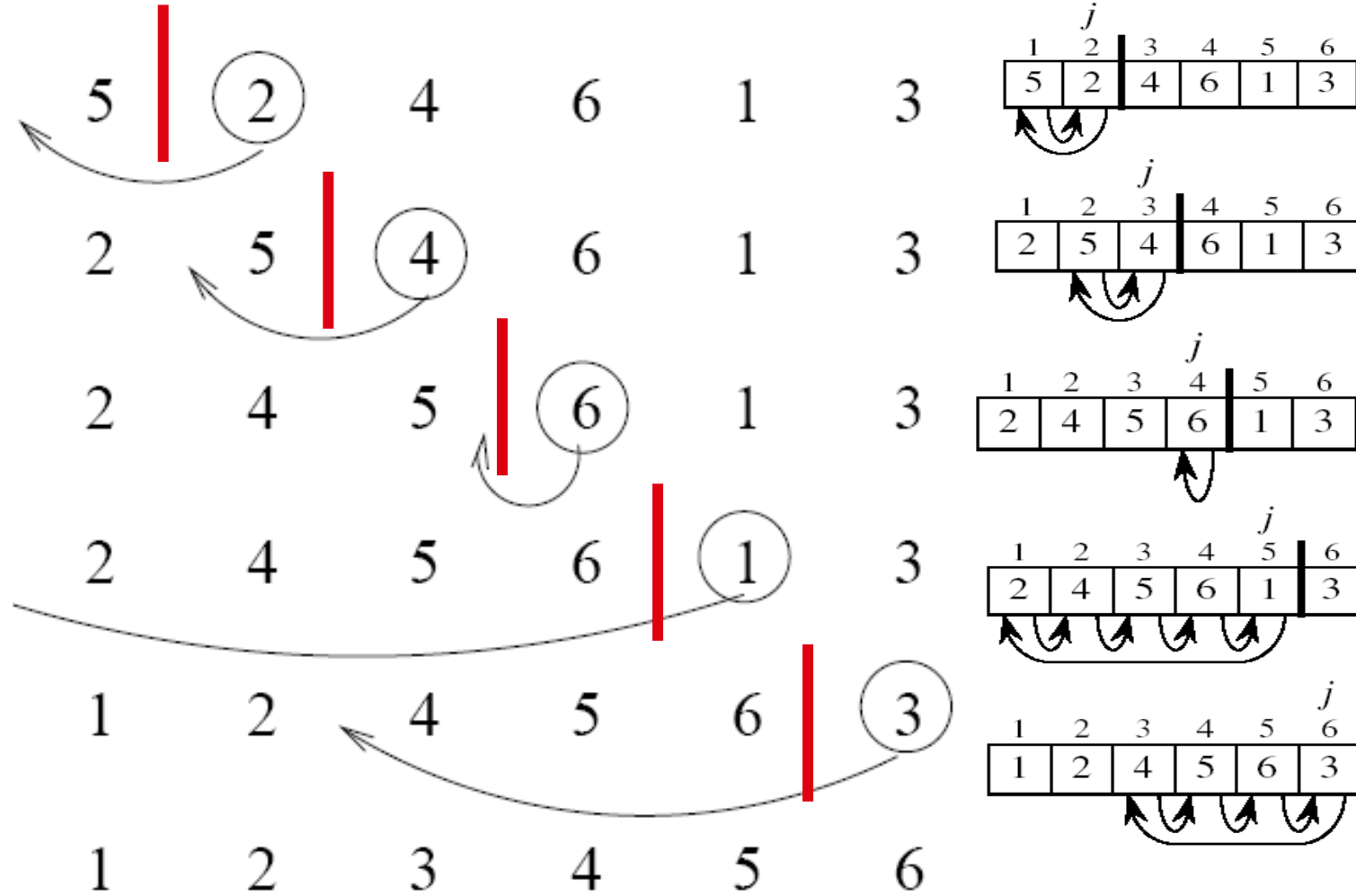
at each iteration, the array is divided in two sub-arrays:

left sub-array

right sub-array



Insertion Sort



Best Case

1

2

3

4

5

Worst Case

5	4	3	2	1
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Running time of Insertion Sort

- The running time depends on the input: an already sorted sequence is easier to sort.
- Parameterize the running time by the size of the input,
 - short sequences are easier to sort than long ones.
- Generally, we seek upper bounds on the running time, because everybody likes a guarantee.

Kinds of Analyses

- Worst-case: (Usually)
 - $T(n)$ = (maximum time of algorithm) on any input of size n .
- Average-case (Sometimes):
 - $T(n)$ = (expected time of algorithm) over all inputs of size n .
 - Need assumption of statistical distribution of inputs
- Best-case: (bogus)
 - Cheat with a slow algorithm that works fast on some input

INSERTION-SORT

Alg.: INSERTION-SORT(A)

for $j \leftarrow 2$ **to** n

do $key \leftarrow A[j]$

 Insert $A[j]$ into the sorted sequence $A[1 \dots j-1]$

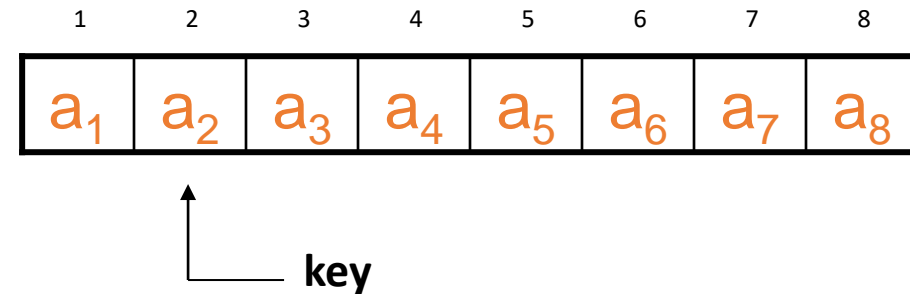
$i \leftarrow j - 1$

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

do $A[i + 1] \leftarrow A[i]$

$i \leftarrow i - 1$

$A[i + 1] \leftarrow key$



- Insertion sort – sorts the elements in place

Analysis of Insertion Sort

INSERTION-SORT(A)

for $j \leftarrow 2$ **to** n

do $\text{key} \leftarrow A[j]$

 ▷ Insert $A[j]$ into the sorted sequence $A[1 \dots j-1]$

$i \leftarrow j - 1$

while $i > 0$ and $A[i] > \text{key}$

do $A[i + 1] \leftarrow A[i]$

$i \leftarrow i - 1$

$A[i + 1] \leftarrow \text{key}$

cost times

c_1 n

c_2 $n-1$

0 $n-1$

c_4 $n-1$

c_5 $\sum_{j=2}^n t_j$

c_6 $\sum_{j=2}^n (t_j - 1)$

c_7 $\sum_{j=2}^n (t_j - 1)$

c_8 $n-1$

t_j : # of times the while statement is executed at iteration j

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8 (n-1)$$

Best Case Analysis

- The array is already sorted “while i > 0 and A[i] > key”
 - $A[i] \leq \text{key}$ upon the first time the **while** loop test is run
(when $i = j - 1$)
 - $t_j = 1$

$$\begin{aligned}
 T(n) &= c_1 \cdot n + (c_2 + c_4) \cdot (n-1) + c_5 \cdot (n-1) + c_8 \cdot (n-1) \\
 &= n \cdot (c_1 + c_2 + c_4 + c_5 + c_8) \\
 &\quad + (-c_2 - c_4 - c_5 - c_8) \\
 &= c_9 n + c_{10}
 \end{aligned}$$

$$T(n) = c_1 n + c_2(n-1) + c_4(n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n-1)$$

Comparisons and Exchanges in Insertion Sort

INSERTION-SORT(A)

for $j \leftarrow 2$ **to** n

cost times

c_1 n

do $\text{key} \leftarrow A[j]$

c_2 $n-1$

Insert $A[j]$ into the sorted sequence $A[1 \dots j-1]$

0 $n-1$

$i \leftarrow j - 1$

$\approx n^2/2$ comparisons

c_4 $n-1$

while $i > 0$ and $A[i] > \text{key}$

c_5 $\sum_{j=2}^n t_j$

do $A[i + 1] \leftarrow A[i]$

c_6 $\sum_{j=2}^n (t_j - 1)$

$i \leftarrow i - 1$

$\approx n^2/2$ exchanges

c_7 $\sum_{j=2}^n (t_j - 1)$

$A[i + 1] \leftarrow \text{key}$

c_8 $n-1$

Worst Case Analysis

- The array is in reverse sorted order “while i > 0 and A[i] > key”
 - Always $A[i] > \text{key}$ in **while** loop test
 - Have to compare **key** with all elements to the left of the j -th position
 \Rightarrow compare with $j-1$ elements $\Rightarrow t_j = j$

using $\sum_{j=1}^n j = \frac{n(n+1)}{2} \Rightarrow \sum_{j=2}^n j = \frac{n(n+1)}{2} - 1 \Rightarrow \sum_{j=2}^n (j-1) = \frac{n(n-1)}{2}$ we have:

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

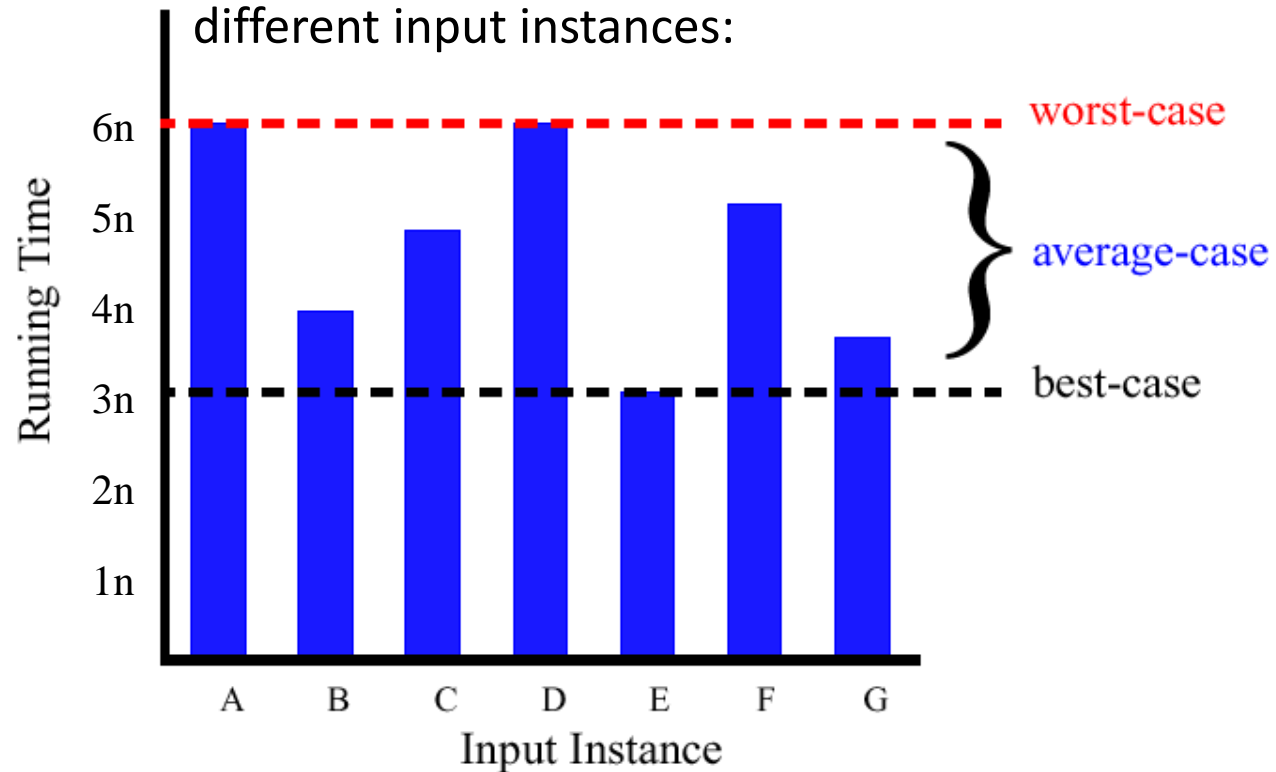
$$= an^2 + bn + c \quad \text{a quadratic function of } n$$

- $T(n) = \Theta(n^2)$ order of growth in n^2

$$T(n) = c_1 n + c_2(n-1) + c_4(n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n-1)$$

Best/Worst/Average Case (2)

- For a specific size of input n , investigate running times for different input instances:



Insertion Sort Analysis

- Is insertion sort a fast sorting algorithm?
 - Moderately so, for small n
 - Not at all, for large n
 - sorting "almost sorted" lists

Reference

- Introduction to Algorithms
- Chapter # 2
 - Thomas H. Cormen
 - 3rd Edition
- <https://visualgo.net/bn/sorting>