

# Design and Analysis of Algorithms

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Lecture 1: Getting Started

# About this lecture

- Study a few simple algorithms for sorting
  - Insertion Sort
  - Selection Sort
  - Merge Sort
- Show why these algorithms are correct
- Try to analyze the efficiency of these algorithms (how fast they run)

# The Sorting Problem

Input: A list of  $n$  numbers

Output: Arrange the numbers in increasing order

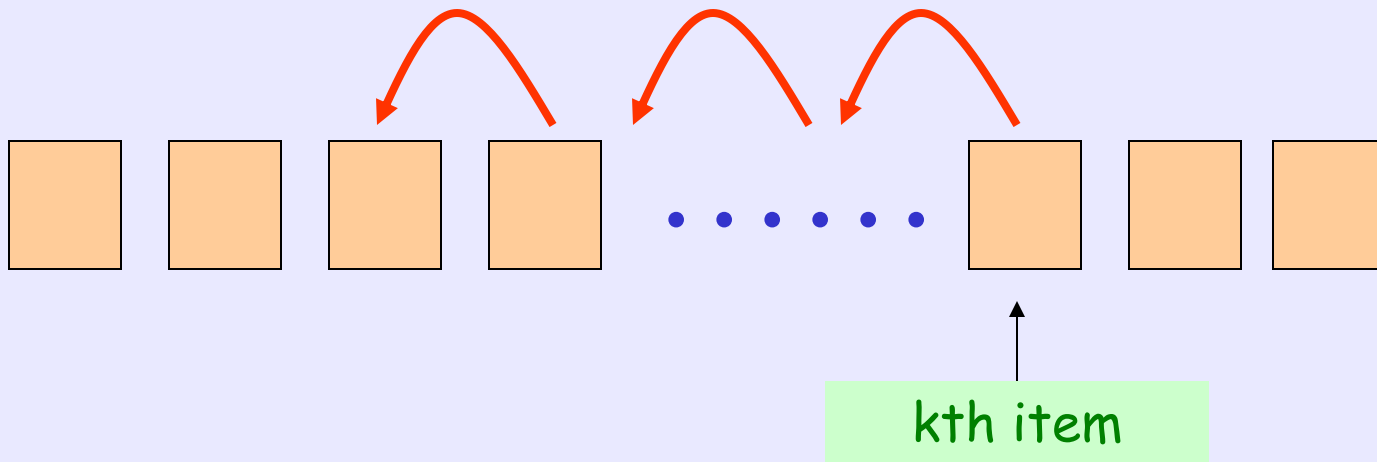
**Remark:** Sorting has many applications.

E.g., if the list is already sorted, we can search a number in the list faster

# Insertion Sort

- Operates in  $n$  rounds
- At the  $k^{\text{th}}$  round,

Swap towards left side ;  
Stop until seeing an item  
with a smaller value.



Question: Why is this algorithm correct?

# Selection Sort

- Operates in  $n$  rounds
- At the  $k^{\text{th}}$  round,
  - Find minimum item after  $(k-1)^{\text{th}}$  position
  - Let's call this minimum item  $X$
  - Insert  $X$  at  $k^{\text{th}}$  position in the list

Question: Why is this algorithm correct?

# Divide and Conquer

- Divide a big problem into smaller problems
  - ➔ solve smaller problems separately
  - ➔ combine the results to solve original one
- This idea is called **Divide-and-Conquer**
- Smart idea to solve complex problems (why?)
- Can we apply this idea for sorting ?

# Divide-and-Conquer for Sorting

- What is a smaller problem ?
  - ➔ E.g., sorting fewer numbers
  - ➔ Let's divide the list to two shorter lists
- Next, solve smaller problems (how?)
- Finally, combine the results
  - ➔ "merging" two sorted lists into a single sorted list (how?)

# Merge Sort

- The previous algorithm, using divide-and-conquer approach, is called **Merge Sort**
- The key steps are summarized as follows:
  - Step 1. Divide list to two halves, **A** and **B**
  - Step 2. Sort **A** using Merge Sort
  - Step 3. Sort **B** using Merge Sort
  - Step 4. Merge sorted lists of **A** and **B**

**Question:** Why is this algorithm correct?



# Analyzing the Running Times

- Which of previous algorithms is the best?
- Compare their running time on a computer
  - But there are many kinds of computers !!!

Standard assumption: Our computer is a RAM (Random Access Machine), so that

- each arithmetic (such as  $+$ ,  $-$ ,  $\times$ ,  $\div$ ), memory access, and control (such as conditional jump, subroutine call, return) takes constant amount of time

# Analyzing the Running Times

- Suppose that our algorithms are now described in terms of RAM operations
  - we can count # of each operation used
  - we can measure the running time !
- Running time is usually measured as a function of the input size
  - E.g.,  $n$  in our sorting problem

# Insertion Sort (Running Time)

The following is a pseudo-code for Insertion Sort.  
Each line requires constant RAM operations.

INSERTION-SORT( <i>A</i> )		<i>cost</i>	<i>times</i>
1	<b>for</b> $j \leftarrow 2$ <b>to</b> $\text{length}[A]$	$c_1$	$n$
2	<b>do</b> $\text{key} \leftarrow A[j]$	$c_2$	$n - 1$
3	▷ Insert $A[j]$ into the sorted sequence $A[1 \dots j - 1]$ .	0	$n - 1$
4	$i \leftarrow j - 1$	$c_4$	$n - 1$
5	<b>while</b> $i > 0$ and $A[i] > \text{key}$	$c_5$	$\sum_{j=2}^n t_j$
6	<b>do</b> $A[i + 1] \leftarrow A[i]$	$c_6$	$\sum_{j=2}^n (t_j - 1)$
7	$i \leftarrow i - 1$	$c_7$	$\sum_{j=2}^n (t_j - 1)$
8	$A[i + 1] \leftarrow \text{key}$	$c_8$	$n - 1$

$t_j = \#$  of times *key* is compared at round  $j$

# Insertion Sort (Running Time)

- Let  $T(n)$  denote the running time of insertion sort, on an input of size  $n$
- By combining terms, we have

$$T(n) = c_1n + (c_2+c_4+c_8)(n-1) + c_5\sum t_j + (c_6+c_7) \sum (t_j - 1)$$

- The values of  $t_j$  are dependent on the input (not the input size)

# Insertion Sort (Running Time)

- **Best Case:**

The input list is sorted, so that all  $t_j = 1$

$$\begin{aligned}\text{Then, } T(n) &= c_1n + (c_2 + c_4 + c_5 + c_8)(n-1) \\ &= Kn + c \rightarrow \text{linear function of } n\end{aligned}$$

- **Worst Case:**

The input list is sorted in **decreasing** order, so that all  $t_j = j-1$

$$\begin{aligned}\text{Then, } T(n) &= K_1n^2 + K_2n + K_3 \\ &\rightarrow \text{quadratic function of } n\end{aligned}$$

# Worst-Case Running Time

- In our course (and in most CS research), we concentrate on worst-case time
- Some reasons for this:
  1. Gives an upper bound of running time
  2. Worst case occurs fairly often

**Remark:** Some people also study **average-case** running time (they assume input is drawn **randomly**)

# Try this at home

- Revisit pseudo-code for Insertion Sort
  - make sure you understand what's going on
- Write pseudo-code for Selection Sort

# Merge Sort (Running Time)

The following is a partial pseudo-code for Merge Sort.

```
MERGE-SORT( $A, p, r$ )  
1  if  $p < r$   
2      then  $q \leftarrow \lfloor (p + r)/2 \rfloor$   
3          MERGE-SORT( $A, p, q$ )  
4          MERGE-SORT( $A, q + 1, r$ )  
5          MERGE( $A, p, q, r$ )
```

The subroutine  $\text{MERGE}(A, p, q, r)$  is missing.

Can you complete it?

Hint: Create a temp array for merging



# Merge Sort (Running Time)

- Let  $T(n)$  denote the running time of merge sort, on an input of size  $n$
- Suppose we know that Merge( ) of two lists of total size  $n$  runs in  $c_1 n$  time
- Then, we can write  $T(n)$  as:
$$T(n) = 2T(n/2) + c_1 n + c_2 \quad \text{when } n > 1$$
$$T(n) = c_3 \quad \text{when } n = 1$$
- Solving the recurrence, we have
- $T(n) = K_1 n \log n + K_2 n + K_3$

# Which Algorithm is Faster?

- Unfortunately, we still cannot tell
  - since constants in running times are unknown
- But we **do** know that if **n** is VERY large, worst-case time of Merge Sort must be smaller than that of Insertion Sort
- Merge Sort is **asymptotically** faster than Insertion Sort