

CSE 431/531: Algorithm Analysis and Design (Fall 2024)

# Divide-and-Conquer

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# Outline

- 1 Divide-and-Conquer
- 2 Counting Inversions
- 3 Quicksort and Selection
  - Quicksort
  - Lower Bound for Comparison-Based Sorting Algorithms
  - Selection Problem
- 4 Polynomial Multiplication
- 5 Solving Recurrences
- 6 Computing  $n$ -th Fibonacci Number

## Greedy Algorithm

- mainly for combinatorial optimization problems
- trivial algorithm runs in exponential time
- greedy algorithm gives an efficient algorithm
- main focus of analysis: correctness of algorithm

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## Divide-and-Conquer

- not necessarily for combinatorial optimization problems
- trivial algorithm already runs in polynomial time
- divide-and-conquer gives a more efficient algorithm
- main focus of analysis: running time

# Divide-and-Conquer

- **Divide:** Divide instance into many smaller instances
- **Conquer:** Solve each of smaller instances recursively and separately
- **Combine:** Combine solutions to small instances to obtain a solution for the original big instance

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## Running time analysis

- recursive programs: recurrence

## merge-sort( $A, n$ )

```
1: if  $n = 1$  then  
2:   return  $A$   
3: else  
4:    $B \leftarrow \text{merge-sort}(A[1..\lfloor n/2 \rfloor], \lfloor n/2 \rfloor)$   
5:    $C \leftarrow \text{merge-sort}(A[\lfloor n/2 \rfloor + 1..n], \lceil n/2 \rceil)$   
6:   return  $\text{merge}(B, C, \lfloor n/2 \rfloor, \lceil n/2 \rceil)$ 
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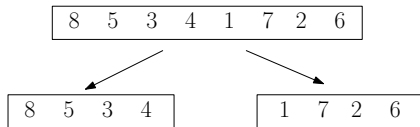
- Divide: trivial
- Conquer: 4, 5
- Combine: 6



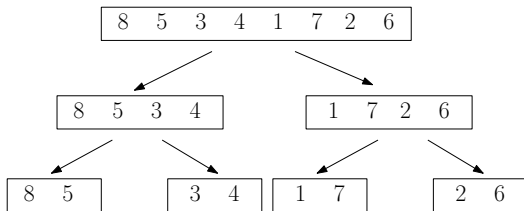
# merge-sort()

8	5	3	4	1	7	2	6
---	---	---	---	---	---	---	---

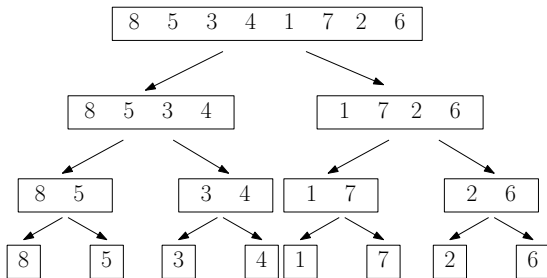
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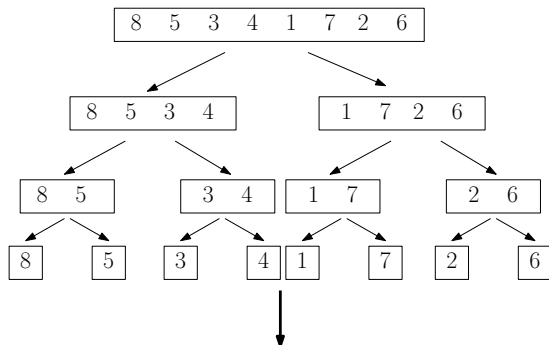
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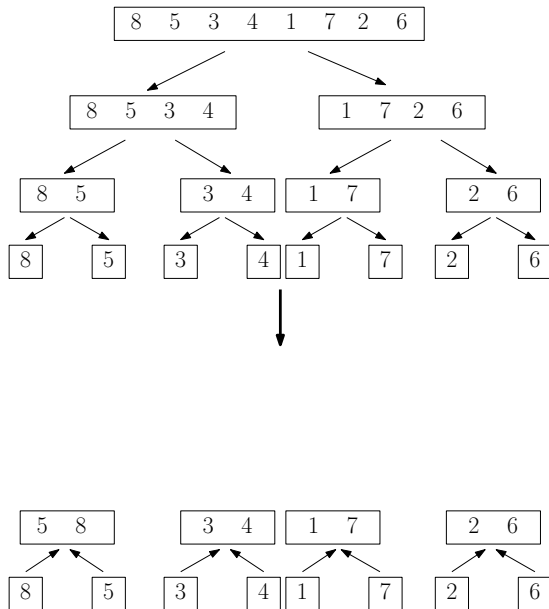


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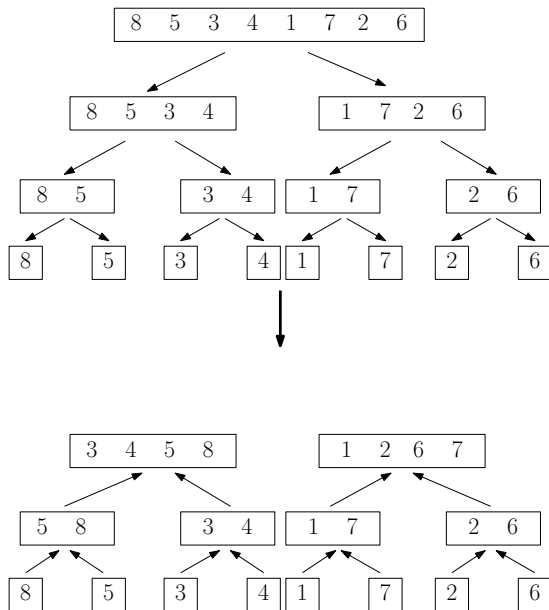


[8] [5] [3] [4] [1] [7] [2] [6]

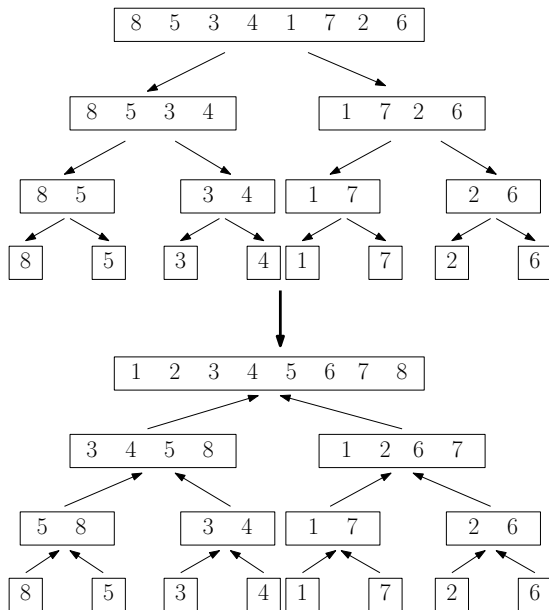
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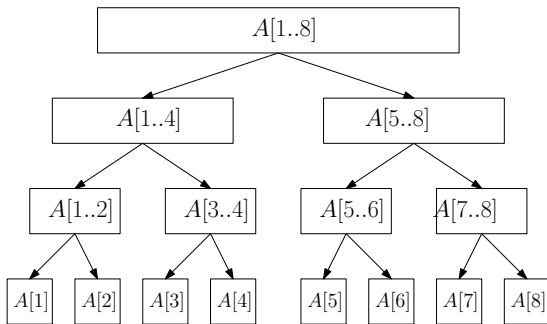


# merge-sort()





# Running Time for Merge-Sort



- Each level takes running time  $O(n)$
- There are  $O(\lg n)$  levels
- Running time =  $O(n \lg n)$
- Better than insertion sort

# Running Time for Merge-Sort

## Implementation

- Divide  $A[a, b]$  by  $q = \lfloor (a + b)/2 \rfloor$ :  $A[a, q]$  and  $A[q + 1, b]$ ; or  $A[a, q - 1]$  and  $A[q, b]$ ?

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## Stable sorting algorithm

- Stable sorting algorithm has the property that equal items will appear in the final sorted list in the same relative order that they appeared in the initial input.

# Running Time for Merge-Sort Using Recurrence

- $T(n)$  = running time for sorting  $n$  numbers, then

$$T(n) = \begin{cases} O(1) & \text{if } n = 1 \\ T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + O(n) & \text{if } n \geq 2 \end{cases}$$

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- Solving this recurrence, we have  $T(n) = O(n \lg n)$  (we shall show how later)