## VE281

Data Structures and Algorithms

Basic Sorting Algorithm; Merge Sort

## Outline

- Basic Sorting Algorithms
- Merge Sort

#### **Insertion Sort**

- A[0] alone is a sorted array.
- For **i=1** to **N-1** 
  - Insert A[i] into the appropriate location in the sorted array A[0], ..., A[i-1], so that A[0], ..., A[i] is sorted.
  - To do so, save **A**[i] in a temporary variable t, shift sorted elements greater than t right, and then insert t in the gap.
- Time comlexity?  $O(N^2)$
- In place? Yes. O(1) additional memory.
- Stable?
  - Yes, because elements are visited in order and equal elements are inserted after its equals.

#### **Insertion Sort**

Best Case Time Complexity

- For **i=1** to **N-1** 
  - Insert A[i] into the appropriate location in the sorted array A[0], ..., A[i-1], so that A[0], ..., A[i] is sorted.
- The **best case** time complexity is O(N).
  - It happens when the array is already sorted.
  - For other sorting algorithms we will talk, their best case time complexity is  $\Omega(N \log N)$ .

#### Selection Sort

- For **i=0** to **N-2** 
  - Find the smallest item in the array A[i], ..., A[N-1]. Then, swap that item with A[i].
- Finding the smallest item requires linear search.
- Time complexity?
  - $O(N^2)$  best case?
- In place?
  - Yes. O(1) additional memory.
- Stable?
  - No. (3, e), (3, b), (2, a) (2, a), (3, b), (3, e)

#### **Bubble Sort**

```
For i=N-2 downto 0
For j=0 to i
If A[j]>A[j+1] swap A[j] and A[j+1]
```

- Compares two adjacent items and swap them to keep them in ascending order.
  - From the beginning to the end. The last item will be the largest.
- Time complexity?  $O(N^2)$
- In place? Yes.
- Stable?
  - Yes, because equal elements will not be swapped.

# Two Problems with Simple Sorts

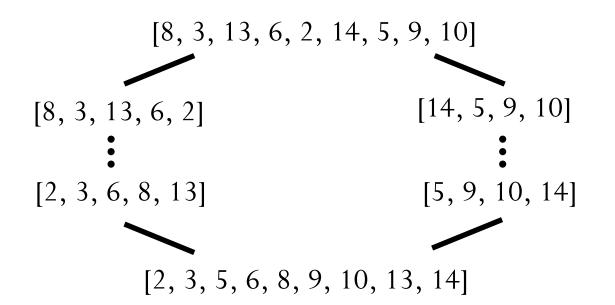
- They learn only one piece of information per comparison and hence might compare every pair of elements.
  - Contrast with binary search: learns N/2 pieces of information with first comparison.
- They often move elements one place at a time (bubble sort and insertion sort), even if the element is "far" from its **final** place.
  - Contrast with selection sort, which moves each element exactly to its final place.
- Fast sorts attack these two problems.
  - Two famous ones: merge sort and quick sort.

## Outline

- Basic Sorting Algorithms
- Merge Sort

#### Algorithm

- Spilt array into two (roughly) equal subarrays.
- Merge sort each subarray recursively.
  - The two subarrays will be sorted.
- Merge the two sorted subarrays into a sorted array.



Pseudo-code

```
void mergesort(int *a, int left, int
  right) {
    if (left >= right) return;
    int mid = (left+right)/2;
    mergesort(a, left, mid);
    mergesort(a, mid+1, right);
    merge(a, left, mid, right);
}
```

# Merge Two Sorted Arrays

- For example, merge A = (2, 5, 6) and B = (1, 3, 8, 9, 10).
- Compare the smallest element in the two arrays A and B and move the smaller one to an additional array C.
- Repeat until one of the arrays becomes empty.
- Then append the other array at the end of array C.

## Merge Two Sorted Arrays

#### **Implementation**

- We actually do not "remove" element from arrays A and B.
  - We just keep a pointer indicating the smallest element in each array.
  - We "remove" element by incrementing that pointer.

Time complexity is O(sizeA + sizeB)

Time Complexity

```
void mergesort(int *a, int left, int
  right) {
    if (left >= right) return;
    int mid = (left+right)/2;
    mergesort(a, left, mid);  T(N/2)
    mergesort(a, mid+1, right);  T(N/2)
    merge(a, left, mid, right);  O(N)
}
```

- Let T(N) be the time required to merge sort N elements.
- Merge two sorted arrays with total size N takes O(N).

```
Recursive relation: T(N) = 2T(N/2) + O(N)
```

How to solve the recurrence?

#### Solve Recurrence: Master Method

- A "black box" for solving recurrence.
- However, there is an important assumption: all sub-problems have roughly equal sizes.
  - E.g., merge sort
  - Not apply to unbalanced division.

### Solve Recurrence: Master Method

- Recurrence:  $T(n) \le aT\left(\frac{n}{h}\right) + O(n^d)$ 
  - Base case:  $T(n) \leq constant$  for all sufficiently small n.
  - $a = \text{number of recursive calls (integer } \ge 1)$
  - b = input size shrinkage factor (integer > 1)
  - $O(n^d)$ : the runtime of merging solutions. d is real value  $\geq 0$ .
  - a, b, d are independent of n.

• Claim:

base doesn't matter

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

base matters!

# Example of Merge Sort

Recurrence: 
$$T(n) \le aT\left(\frac{n}{h}\right) + O(n^d)$$

Claim: 
$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

- $a = 2, b = 2, d = 1 \implies b^d = a$
- $T(n) = O(n \log n)$

# Another Example: Binary Search

Recurrence: 
$$T(n) \le aT\left(\frac{n}{h}\right) + O(n^d)$$

Claim: 
$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

• Exercise: What is a, b, d?

#### Characteristics

- Not in-place
  - For efficient merging two sorted arrays, we need an auxiliary O(N) space.
  - Recursion needs up to  $O(\log N)$  stack space.
- Stable if **merge()** maintains the relative order of equal keys.

# Divide-and-Conquer Approach

- Merge sort uses the divide-and-conquer approach.
- Recursively **breaking** down a problem into two or more sub-problems of the same (or related) type, until these become simple enough to be solved directly.
  - For merge sort, split an array into two and sort them respectively.
- The solutions to the sub-problems are then **combined** to give a solution to the original problem.
  - For merge sort, merge two sorted arrays.