Efficient Algorithms

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Lecture 5: Searching and Sorting

Binary Search Tree: Properties

Properties:

- (1) Each node x in a binary search tree (BST) has a key x. key.
- (2) Nodes x other than the root have parents x. parent.
- (3) Nodes x may have a left child x. left or a right child x. right or both. (Branching Factor = 2)

NB: These nodes are **pointers** unlike in a heap (from Lecture 3).

Binary Search Tree: Invariant

Invariant:

For any node x, for all nodes y in the *left* subtree of node x, y. $key \le x$. key and for all nodes z in the *right* subtree of node x, z. $key \ge x$. key.

Binary Search Tree: Search

The running time of the **search operation** is O(h), where h is the height of the tree.

```
1: procedure BST-Search(root, key)
      if root == NULL then
 2:
         return NULL
 3:
      else
 4:
         if key < root.key then
 5:
             return BST-Search(root.left, key)
 6:
         else
7:
             if key > root.key then
8:
                return BST-Search(root.right, key)
9:
             else
10:
                return root
11:
12:
```

Binary Search Tree: Min and Max

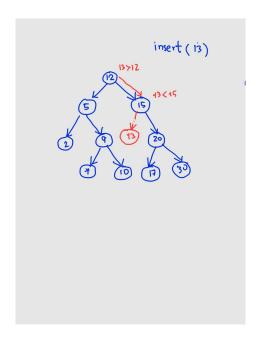
```
To find min, keep going left.
```

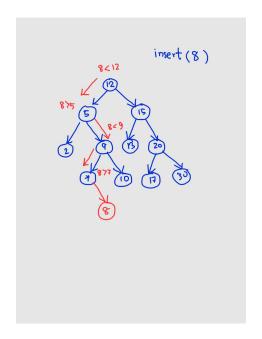
To find **max**, keep going **right**.

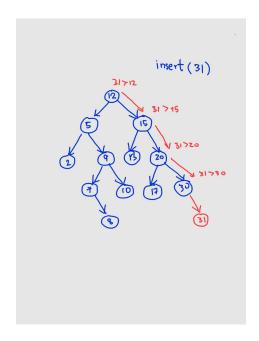
The running time of the min and max is the length of the downward path from root to the leftmost leaf and rightmost leaf, respectively.

Binary Search Tree: Insert

Follow the left and right pointers until the right position for the key being inserted is found.







Binary Search Tree: Delete

There are **three** cases.

Case I: The node x is a leaf.

- Follow the path until node x is reached.
- Remove x by modifying its parent to replace x with NULL as its child.

Binary Search Tree: Delete

There are **three** cases.

Case II: The node x has one child.

- Follow the path until x is reached.
- Remove x by elevating its child to take x's position and modifying x's parent to point to x's child.

Binary Search Tree: Delete

There are **three** cases.

Case III: The node x has two children.

- Locate the leftmost leaf node y in the right subtree of x
- Replace x with y
- Remove the now duplicated leaf node y

The Sorting Problem

The sorting problem can be stated as follows:

<u>Input</u>: a sequence of n numbers $< a_1$, a_2 , ..., $a_n >$,

Output: a permutation $< a_1', a_2', ..., a_n' >$ of the input sequence $a_1' \le a_2' \le \cdots \le a_n'$.

The numbers that we are sorting are also known as *keys*.

Insertion Sort

Idea:

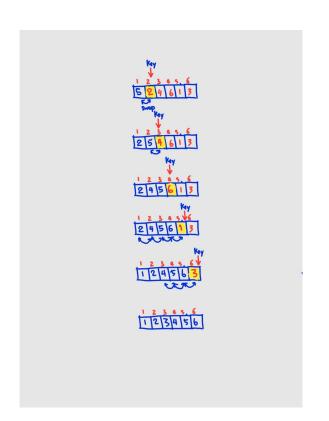
Imagine the way we sort a hand of playing cards.

- We start with an empty hand with the cards facing down on the table
- We pick up one card at a time from the table
- We insert the new card into the correct position in the left hand
 - To find the correct position, we compare the new card with the existing cards in the hand, from left to right.

Illustration taken

from CLRS

Insertion Sort



```
1: procedure Insertion-Sort(A)
2: for j = 2 \to A.length do
3: key = A[i]
4: i = j - 1
5: while i > 0 \land A[i] > key do
6: A[i + 1] = A[i]
7: i = i - 1
8: A[i + 1] = key
```

Insertion Sort

The index j indicates the *current card* being inserted into the left hand.

Observation:

At the beginning of each iteration of the for loop,

- the subarray A[1...j-1] constitutes the currently sorted hand
- the subarray A[j+1,...,n] corresponds to the pile of cards still on the table
- the elements A[1...j-1] are the elements originally in positions 1 through j-1, but now in sorted order.

Insertion Sort: Correctness Proof

Based on the *observation*, we propose the following *loop invariant*:

At the start of each iteration of the for loop, the subarray A[1...j-1] consists of the elements **originally** in A[1...j-1] but in **sorted** order.

We will use this *loop invariant* to prove the *correctness* of insertion sort.

Loop Invariant

<u>Initialization</u>: The invariant is true prior to the first iteration of the loop.

<u>Maintenance</u>: If the invariant is true before an iteration of the loop, it remains true before the next iteration.

<u>Termination</u>: When the loop terminates, the invariant provides a useful property that can be used to show that the algorithm is correct.

Loop Invariant

When the Initialization & Maintenance properties are true, the loop invariant is true prior to every iteration.

This is the concept of *mathematical induction*:

- *Initialization* corresponds to the *base case*
- Maintenance corresponds to the inductive step
 - Assume true for an iteration i and show true for the next iteration i'
 - This is to show that the loop is inductive

The *Termination* property differs from how we use *mathematical induction* where we apply the inductive step *infinitely*.

At termination, we stop the induction and use the invariant to show that the algorithm is correct.

Insertion Sort: Loop Invariant

<u>Initialization</u>: Before the first iteration, j = 2.

The subarray $A[1 ... 1] \equiv A[1]$ is **trivially sorted** and is also the original element in A[1] so the invariant holds prior to the first iteration.

Maintenance: Informally, the body of the for loop works by moving A[j-1], A[j-2], A[j-3] and so on by one position to the right until it finds the proper position for A[j], at which point A[j] is inserted.

The subarray A[1 ... j] consists of the elements originally in A[1 ... j], but in sorted order.

When the loop counter j is incremented by one, the loop invariant still holds.

Insertion Sort: Termination

Termination: Finally, we check what happens when the loop terminates.

Because each iteration increments j by one, we have j = n + 1 on termination.

Substituting j = n + 1 into the wording of the loop invariant, we have that: at the start of each iteration of the for loop, the subarray A[1 ... n], which is now the **entire array**, consists of the elements **originally** in A[1 ... n] but in **sorted** order.

Insertion Sort: Running Time Analysis

Barometer Instruction:

We count the total number of comparisons

• This corresponds to *Line 5*

```
1: procedure Insertion-Sort(A)
2: for j = 2 \to A.length do
3: key = A[i]
4: i = j - 1
5: while i > 0 \land A[i] > key do
6: A[i+1] = A[i]
7: i = i - 1
8: A[i+1] = key
```

Insertion Sort: Worst-Case Analysis

At each iteration i of the inner while loop, the elements in A[1 ... i] are all moved one position to the right.

- During each iteration, i = j 1 (Line 4)
- There can be at most $\sum_{j=2}^{n} (j-1)$ comparisons. (Line 5)

Summing up the number of comparisons:

$$T(n) = \sum_{j=2}^{n} (j-1)$$
$$= \frac{n^2}{2} - \frac{n}{2} \blacksquare$$

```
1: procedure Insertion-Sort(A)
2: for j = 2 \rightarrow A.length do
3: key = A[i]
4: i = j - 1
5: while i > 0 \land A[i] > key do
6: A[i+1] = A[i]
7: i = i - 1
8: A[i+1] = key
```

Insertion Sort: Average-Case Analysis

For a given iteration i, there are i possible cases as follows:

Case	A[i] is	#Comparisons
1	largest	1
2	second largest	2
3	third largest	3
i-1	second smallest	<i>i</i> − 1
i	smallest	i-1

Insertion Sort: Average-Case Analysis

Let N_i be a random variable representing the number of comparisons during iteration i.

$$E(N_i) = \sum_{k=1}^{i} n_k \Pr\{Case \ k \ happens\}$$

$$E(N_i) = 1 \cdot \frac{1}{i} + 2 \cdot \frac{1}{i} + \dots + (i-1) \cdot \frac{1}{i} + (i-1) \cdot \frac{1}{i}$$

$$E(N_i) = \frac{i-1}{2} + \frac{i-1}{i}$$

Insertion Sort: Average-Case Analysis

The **expected** total number N of comparisons, where $N = \sum_{i=2}^{n} N_i$, is

$$E(N) = \sum_{i=2}^{n} \left\{ \frac{i-1}{2} + \frac{i-1}{i} \right\} \le \sum_{i=2}^{n} \frac{i-1}{2} + (n-1) \qquad \left[\lim_{i \to \infty} \frac{i-1}{i} = 1 \right]$$

$$=\frac{n^2+3n-4}{4}$$

Therefore, $E(N) = \Theta(n^2)$

Insertion Sort: in-place algorithm

It is worth noting that insertion sort is an *in-place* algorithm in which O(1) *auxiliary storage* is required to store *intermediate results* during the execution of the sorting algorithm.

Line 3 indicates that the variable key is the auxiliary space used by

insertion sort.

```
1: procedure Insertion-Sort(A)
2: for j = 2 \to A.length do
3: key = A[i]
4: i = j - 1
5: while i > 0 \land A[i] > key do
6: A[i + 1] = A[i]
7: i = i - 1
8: A[i + 1] = key
```

Merge Sort

<u>Idea</u>: Merge sort is a *divide-and-conquer* algorithm

Divide the array into **two subarrays** of (approximately the same size) if the array size is larger than one \rightarrow Divide

Sort each subarray (*recursively*) → Conquer

Merge the *sorted* two subarrays → Combine

```
1: procedure MERGE-SORT(A, p, r)

2: if p < r then

3: q = \lfloor \frac{p+r}{2} \rfloor

4: MERGE-SORT(A, p, q)

5: MERGE-SORT(A, q + 1, r)

6: MERGE(A, p, q, r)
```

Merge Sort

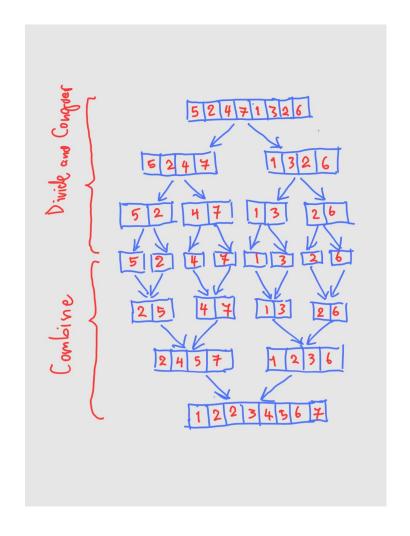
Merge sort has several interesting properties as follows:

- The *divide* and *conquer* parts are very simple
- The *combine* part is (computationally) complex

Merge Sort: Example

The *divide* and *conquer* parts are very simple

The *combine* part is (computationally) complex



Merge Sort: Merge

The *MERGE* routine is the *core part* of the *merge sort* algorithm as it is where most of the computational power is spent.

The MERGE assumes that the two subarrays $A[p \dots q]$ and $A[q+1 \dots r]$ are already sorted.

Let n_1 and n_2 denote the number of elements in the *left-half* and the *right-half* subarrays of the array A, respectively. (*Lines 2-3*)

New arrays L and R are created to hold the left-half and the right -half subarrays, respectively. (*Lines* 4-9)

The so called **two-finger** algorithm is used to compare and merge

the two subarrays. (*Lines 14-20*)

- * The $\it left\ finger$ points to the smallest element in the $\it left\ subarray\ L$ that has not been copied back to $\it A$
- ${}^{\bullet}$ $\,$ The $\it right\, finger\, points$ to the smallest element in the $\it right\, subarray\, R$ that has not been copied back to A

```
1: procedure MERGE(A, p, q, r)
       n_1 = q - p + 1
      n_2 = r - q
       CREATE L[1...n_1 + 1]
       CREATE R[1...n_2 + 1]
       for i=1 \rightarrow n_1 do
 6:
           L[i] = A[p+i-1]
       for i = 1 \rightarrow n_2 do
           R[i] = A[q+j]
       L[n_1+1]=\infty
10:
       R[n_2+1]=\infty
11:
       i=1
12:
       i=1
13:
       for k = p \rightarrow r do
14:
           if L[i] \leq R[j] then
15:
              A[k] = L[i]
16:
               i = i + 1
17:
           else
18:
               A[k] = R[j]
19:
               j = j + 1
20:
```

Since *MERGE* is the core part, we will focus on the correctness of the *MERGE* routine by considering its *loop invariant*.

Once we show the correctness of *MERGE*, it is easy to see the correctness of *MergeSort*.

Loop Invariant:

- (1) At the start of each iteration of the for loop of Lines 14-20, the subarray $A[p \dots k-1]$ contains the k-p smallest elements of $L[1 \dots n_1+1]$ and $R[1 \dots n_2+1]$ in sorted order.
- (2) L[i] and R[j] are the smallest elements of the two arrays that have not been copied back to A.

Initialization:

Prior to the first iteration, k = p.

Therefore, the subarray A[p ... p - 1] is **empty** and contains k - p = k - k = 0 smallest elements.

Moreover, i = j = 1. L[1] and R[1] contain the smallest elements of the two subarrays that have not been copied back into A.

Maintenance:

Consider the case $L[i] \leq R[j]$.

By the loop invariant and the fact that $L[i] \leq R[j]$, L[i] is the smallest element not copied back to A.

By the loop invariant, A[p ... k - 1] contains the smallest k - p elements.

In **Line 16**, A[k] = L[i].

Therefore, A[p ... k] contains the smallest k - p + 1 elements.

Moreover, L[i+1] and R[j] contain the smallest elements of the two subarrays that have not been copied back to A.

After *Line 17*, i is incremented and k is incremented.

This reestablishes the loop invariant at the start of the next iteration.

***The maintenance property for the other case L[i] > R[j] can be shown in a similar way.

Termination:

When the for loop terminates, k = r + 1.

By the definition of the loop invariant,

$$A[p \dots k-1] = A[p \dots (r+1)-1] = A[p \dots r] \text{ contains the smallest } k-p = (r+1)-p = r-p+1 \text{ elements in sorted order.} \blacksquare$$

Merge Sort: Time Complexity

The running time T(n) for the base case (n = 1) is O(1).

The running time T(n) for the recursive case (n > 1) is split into the following components:

the *divide* part costs O(1)the *conquer* part costs $2T(\frac{n}{2})$ the *combine* part costs O(n)

Therefore,
$$T(n) = O(1) + 2T\left(\frac{n}{2}\right) + O(n)$$

$$= 2T\left(\frac{n}{2}\right) + O(n) \text{ for } n > 1 \blacksquare$$

Merge Sort: Time Complexity

Solving the recurrence relation, we get

$$T(n) = \Theta(n \log n)$$

NB: you solved this using the **recursion tree method** in PS2.4 as homework.

Merge Sort: out-of-place

Merge sort needs *auxiliary space* $\Theta(n)$ to store *intermediate results* during the execution of the sorting algorithm.

Therefore, it is **not** an in-place algorithm.

Summary

We have covered the following topics:

- Binary Search Tree and its basic operations
- Insertion Sort
- Merge Sort
- Correctness Proof using Loop Invariants

Next time, we will cover *divide and conquer*.