

# Binary Search Trees

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# BINARY SEARCH TREES

- *Binary Search Trees* (BSTs) are an important data structure for dynamic sets (Operations).
- In addition to satellite data, elements have:
  - *key*: an identifying field inducing a total ordering (other satellite data)
  - *left*: pointer to a left child (may be NULL)
  - *right*: pointer to a right child (may be NULL)
  - *p*: pointer to a parent node (NULL for root)

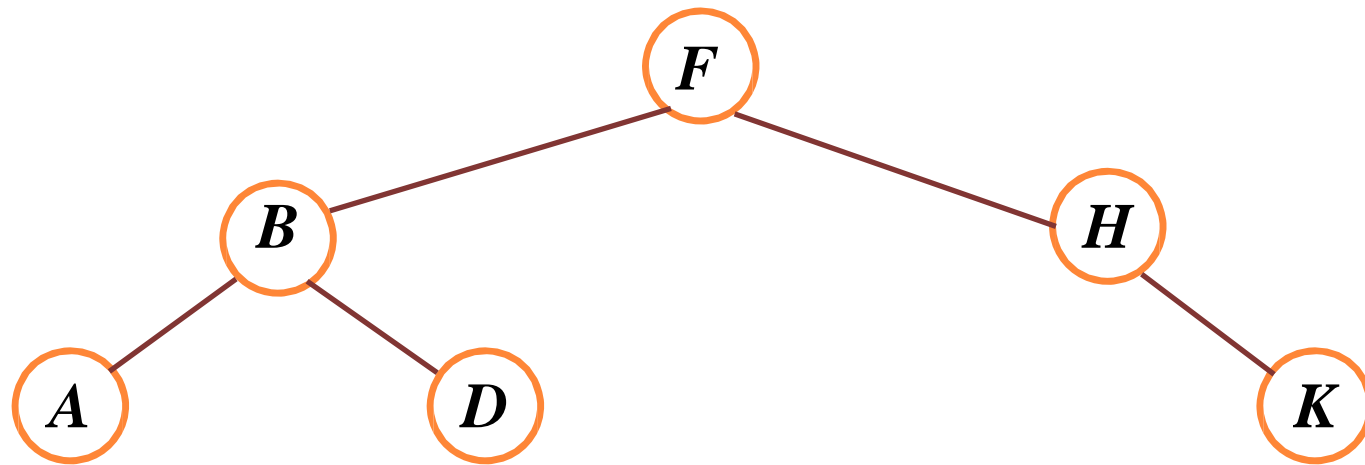


# BINARY SEARCH TREES

- BST property:

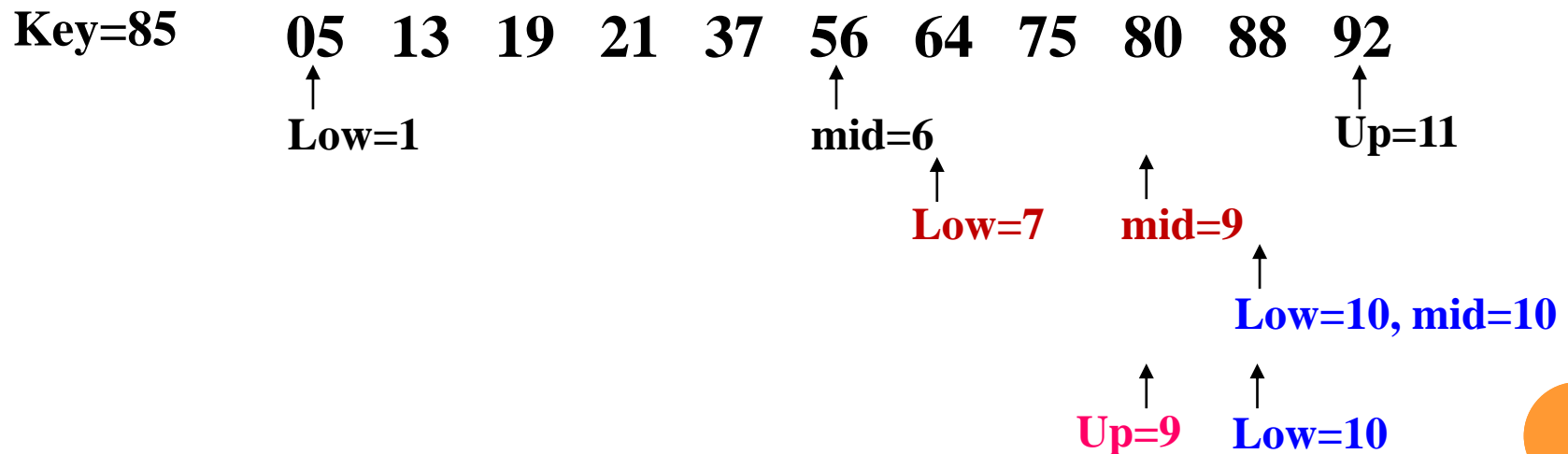
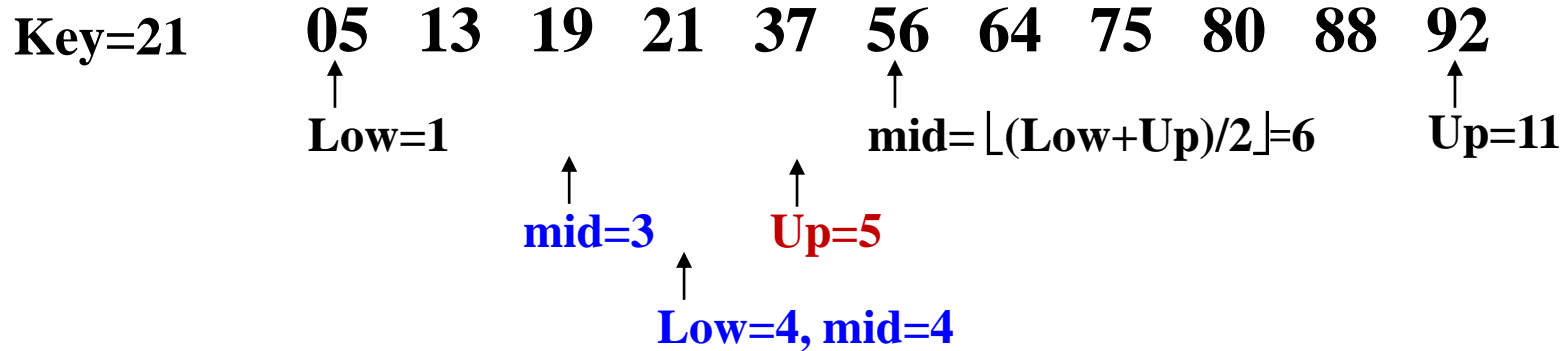
$$key[\text{left}(x)] \leq key[x] \leq key[\text{right}(x)]$$

- Example:



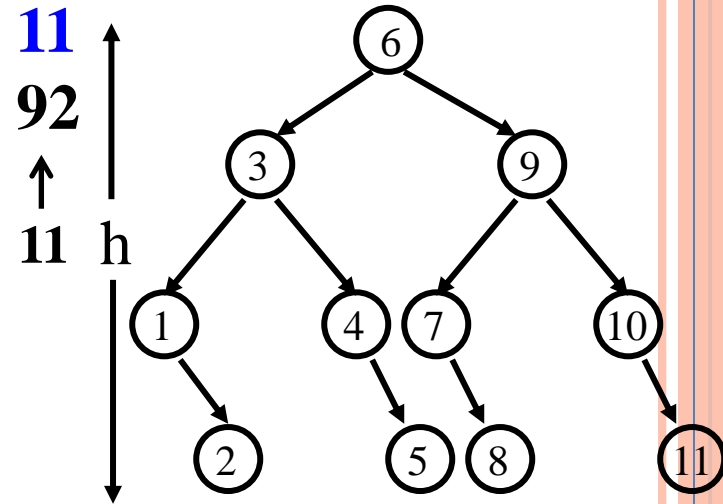
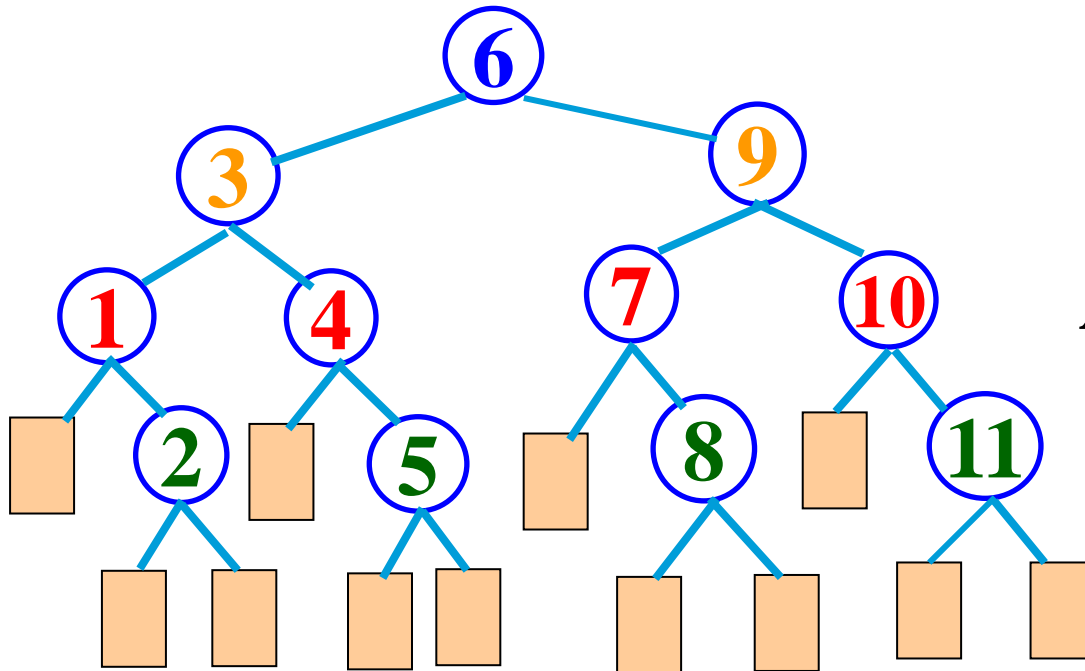
# BINARY SEARCH TREES

$A=[05,13,19,21,37,56,64,75,80,88,92]$



# BINARY SEARCH TREES

1	2	3	4	5	6	7	8	9	10	11
05	13	19	21	37	56	64	75	80	88	92
↑	←-- ↑ --→				↑		←-- ↑ --→			↑
1	3				6		9			11



$$\begin{aligned}
 ASL_{bs} &= \sum_{i=1}^n P_i \cdot C_i & P_i &= 1/n \\
 &= 1/n \cdot \sum_{j=1}^h j \cdot 2^{j-1} \\
 &= (n+1)/n \cdot \log_2(n+1) - 1 \\
 &\approx \log_2(n+1) - 1
 \end{aligned}$$



# INORDER TREE WALK (TRAVERSAL)

- *What does the following code do?*

**TreeWalk( $x$ )**

1. **if  $x \neq \text{NIL}$**
2.     **TreeWalk( $\text{left}[x]$ );**
3.     **print( $x$ );**
4.     **TreeWalk( $\text{right}[x]$ );**

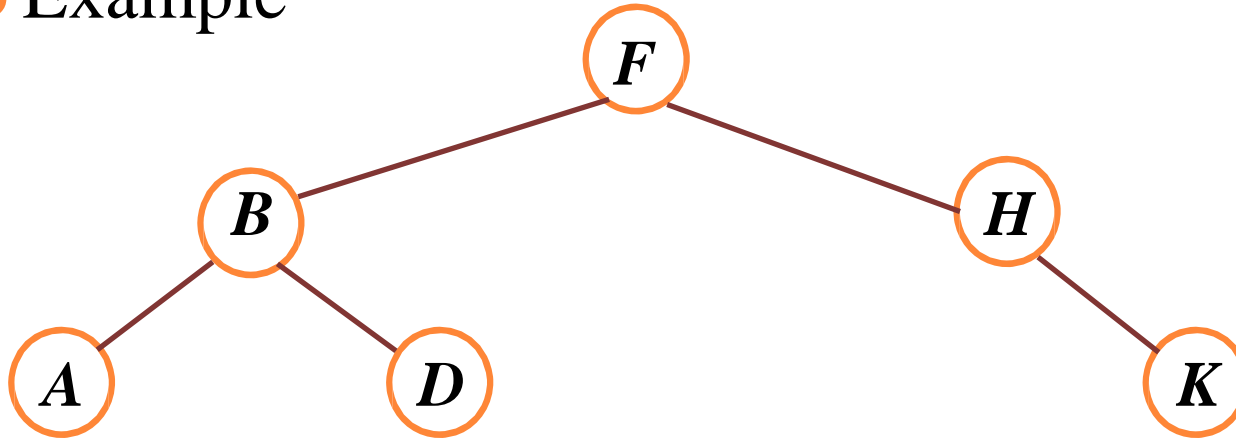
**A:** prints elements in sorted (increasing) order

- This is called an *Inorder Tree Walk*
  - *Preorder tree walk*: print root, then left, then right
  - *Postorder tree walk*: print left, then right, then root



# INORDER TREE WALK

- Example



- Output:  $A B D F H K$
- *How long will a tree walk take?*
  - Theorem 12.1

If  $x$  is the root of an  $n$ -node subtree, then the call INORDER-TREE-WALK( $x$ ) takes  $\Theta(n)$  time.

# OPERATIONS ON BSTs: SEARCH

- *Search*: Given a key and a pointer to a node, returns an element with that key or NULL:

**TreeSearch( $x, k$ )**

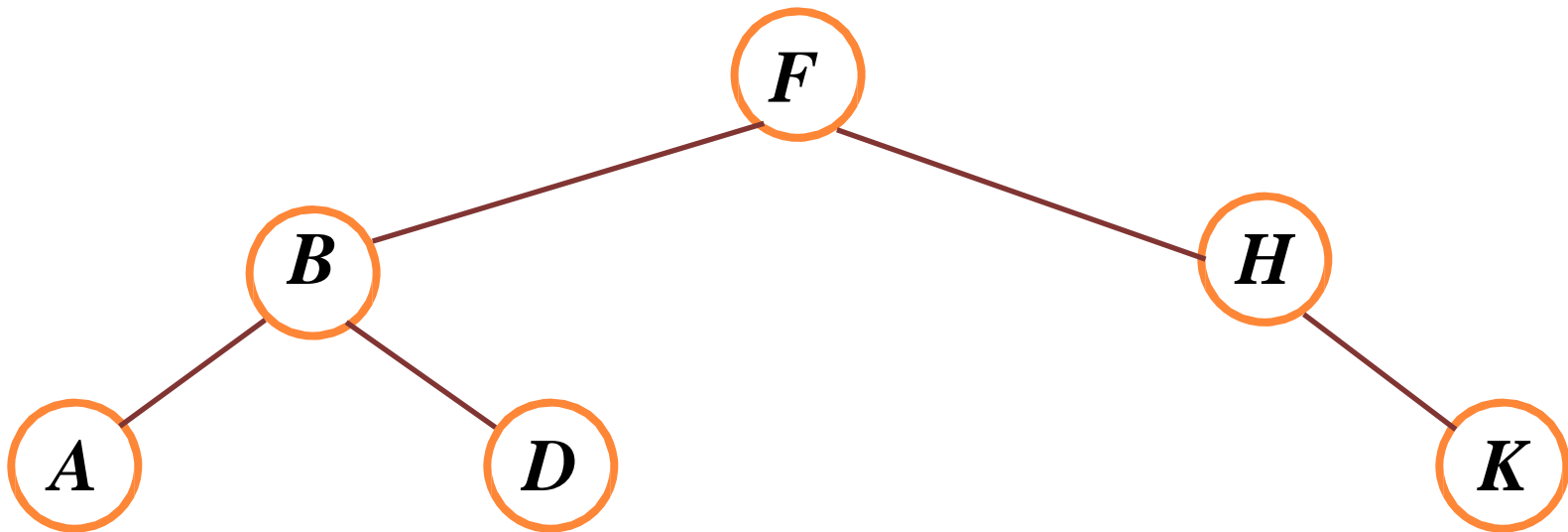
- 1. if ( $x = \text{NULL}$  or  $k = \text{key}[x]$ ) return  $x$ ;**
- 2. if ( $k < \text{key}[x]$ )**
- 3.   return TreeSearch( $\text{left}[x], k$ );**
- 4. else**
- 5.   return TreeSearch( $\text{right}[x], k$ );**





# BST SEARCH: EXAMPLE

- Search for *D* and *C*:



# OPERATIONS ON BSTs: SEARCH

- Here's another function that does the same:

**TreeSearch( $x, k$ )**

**while ( $x \neq \text{NULL}$  and  $k \neq \text{key}[x]$ )**

**if ( $k < \text{key}[x]$ )**

**$x = \text{left}[x];$**

**else**

**$x = \text{right}[x];$**

**return  $x$ ;**

**Minimum and  
maximum?**

- *Which of these two functions is more efficient?*



# OPERATIONS ON BSTs: MIN-MAX

## Minimum of BST

○ TREE-Minimum( $x$ )

1 while  $left(x) \neq \text{NIL}$

2 do  $x \leftarrow left[x]$

3 Return  $x$

## Maximum of BST

○ TREE-Maximum( $x$ )

1 while  $right(\underline{x}) \neq \text{NIL}$

2 do  $x \leftarrow right[x]$

3 Return  $x$



# OPERATIONS OF BSTs: INSERT

- Adds an element  $z$  to the tree so that the binary search tree property continues to hold
- The basic algorithm (straightforward)
  - Like the search procedure above
  - Insert  $x$  in place of NULL



# OPERATIONS OF BSTs: INSERT

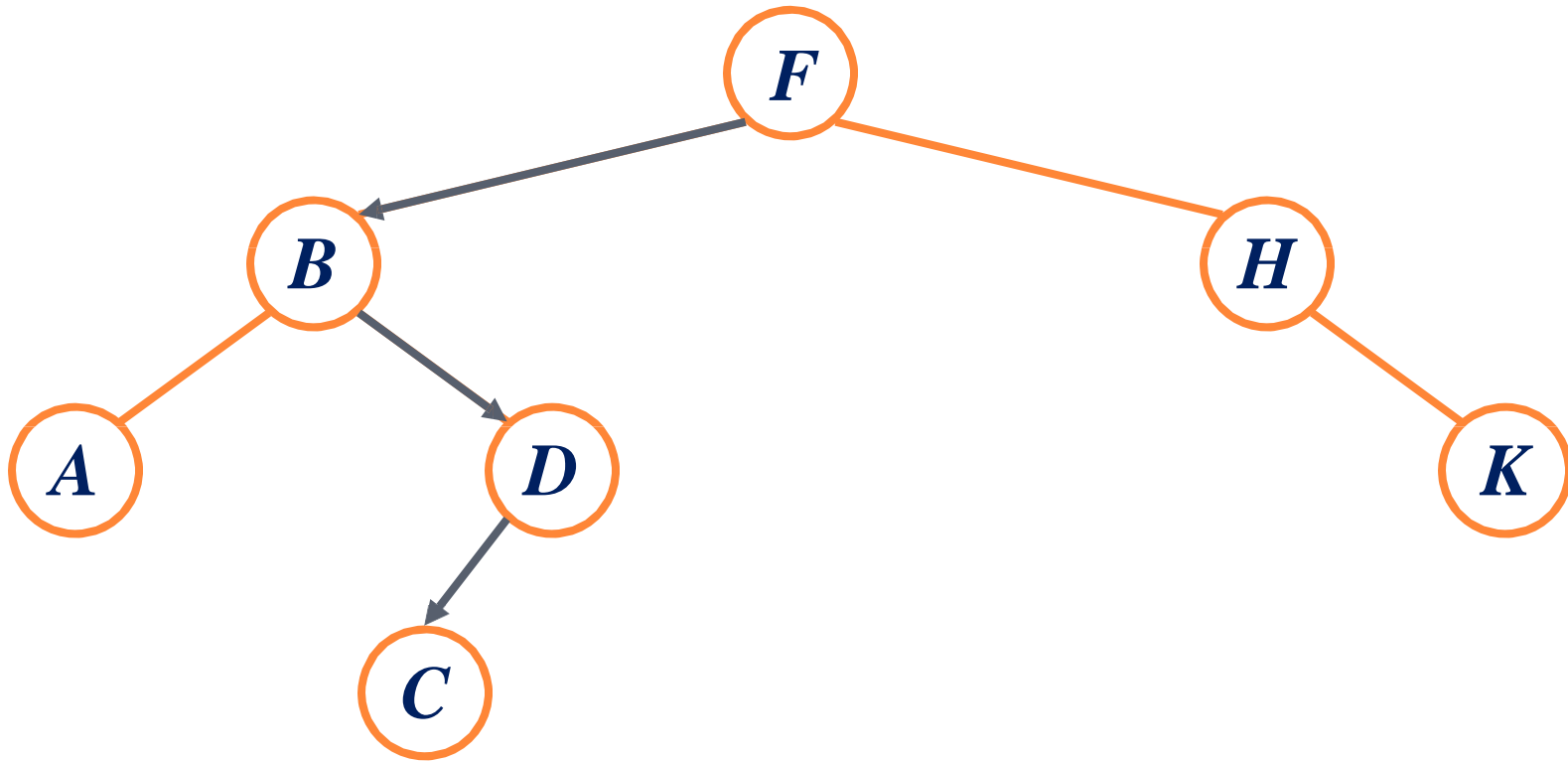
TREE-INSERT(  $T, z$  )

1.  $y = \text{NIL};$
2.  $x = T.\text{root};$
3. *while*  $x \neq \text{NIL}$
4.      $y = x;$
5.     *if*  $z.\text{key} < x.\text{key}$
6.          $x = x.\text{left};$
7.     *else*  $x = x.\text{right};$
8.  $z.p = y;$
9.   *if*  $y == \text{NIL}$
10.          $T.\text{root} = z;$   
              // empty tree T
11. *elseif*  $z.\text{key} < y.\text{key}$
12.          $y.\text{left} = z;$
13. *else*  $y.\text{right} = z;$



# BST INSERT: EXAMPLE

- Example: Insert *C*




# BST SEARCH/INSERT: RUNNING TIME

- TREE-INSERT begins at the root of the tree and the pointer  $x$  traces a simple path downward *looking for a NIL* to replace with the input item  $z$ .
- The height of a binary search tree is  $h$
- What is the running time of TreeSearch( ) or TreeInsert( )?
  - $O(h)$
- What determines the height of a binary search tree?
  - Worst case:  $h = O(n)$  when tree is just a linear string of left or right children



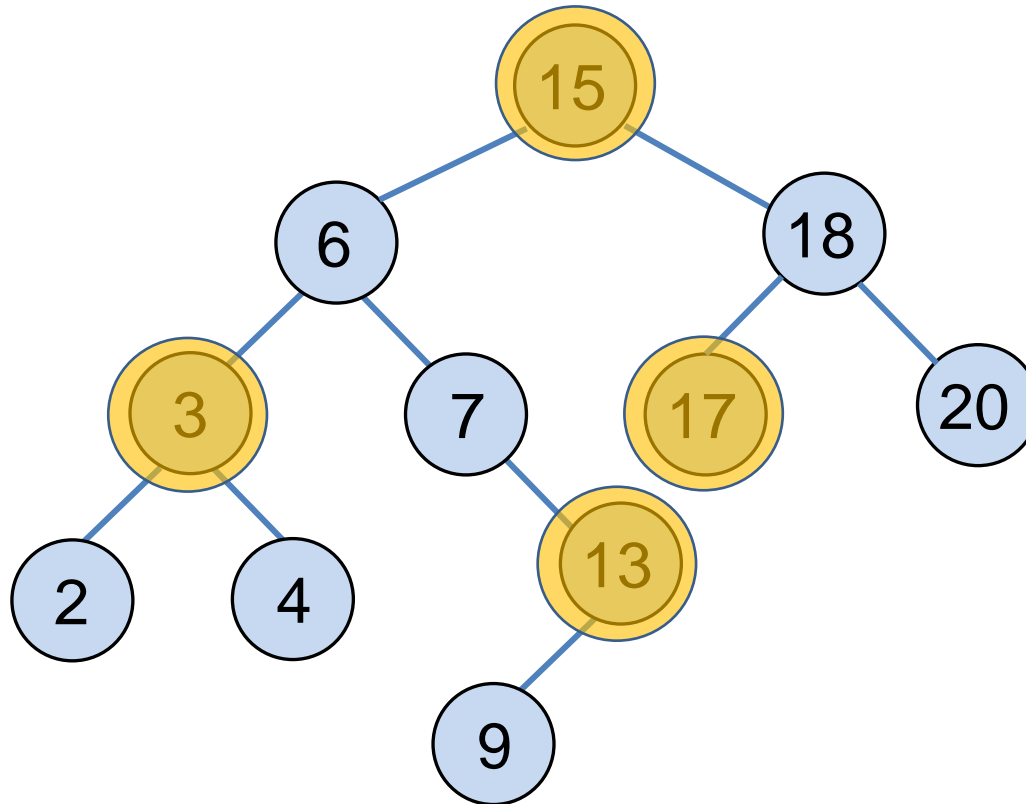
# BST OPERATIONS: SUCCESSOR

- The successor of the current node is the one in the in-order tree walk (distinct keys).
  - Two cases:
    - **$x$  has a right subtree:** successor is minimum node in right subtree (the leftmost node in  $x$ 's right subtree).
    - **$x$  has no right subtree:** successor is **lowest ancestor** of  $x$  whose **left child** is also one ancestor of  $x$  (every node is its own ancestor)
      - *Intuition:* As long as you move to the left up the tree, you're visiting smaller nodes.
      - To find  $y$ , we simply go up the tree from  $x$  until we encounter a node that is the *left child* of its parent.
- 



# BST OPERATIONS: SUCCESSOR

- What is the successor of node 3? 15? 13? 17?



- How about the Predecessor?



# BST OPERATIONS: SUCCESSOR

## ○ Theorem 12.2

We can implement the dynamic-set operations SEARCH, MINIMUM, MAXIMUM, SUCCESSOR, and PREDECESSOR so that each one runs in  $O(h)$  time on a binary search tree of height  $h$ .



# BST OPERATIONS: DELETE

- Deletion is a bit tricky.

- **Three** main cases:

- Case 1:  $z$  has **no** children

- Remove  $x$

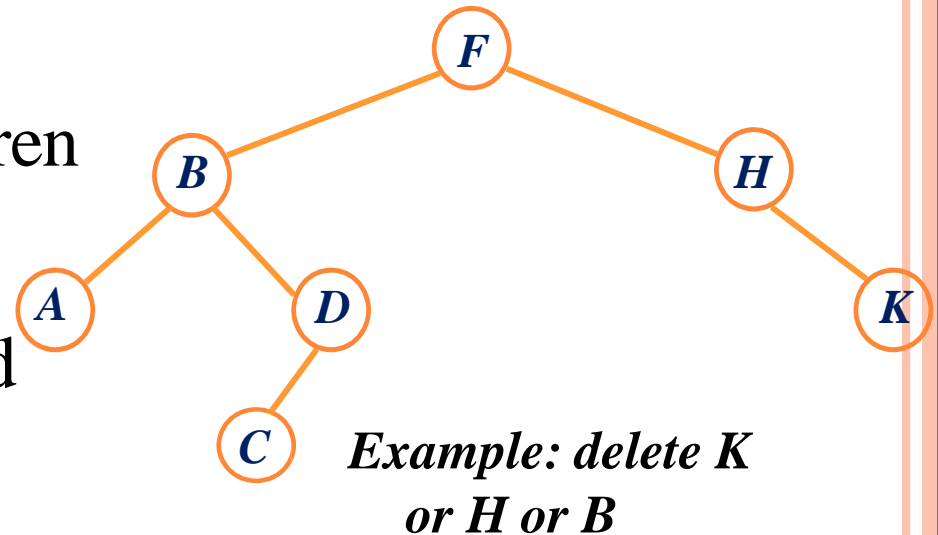
- Case 2:  $z$  has **one** child

- Splice out  $z$

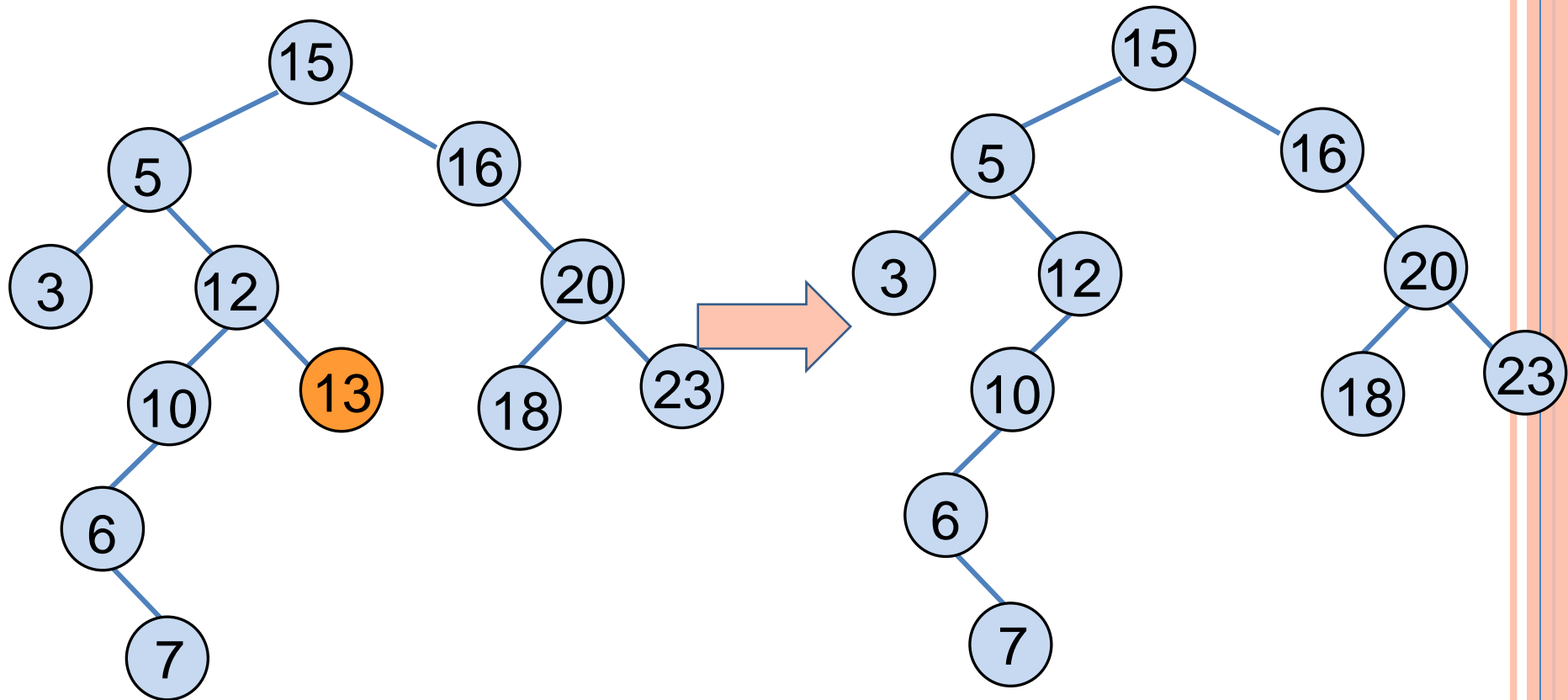
- Case 3:  $z$  has **two** children

- Swap  $z$  with successor

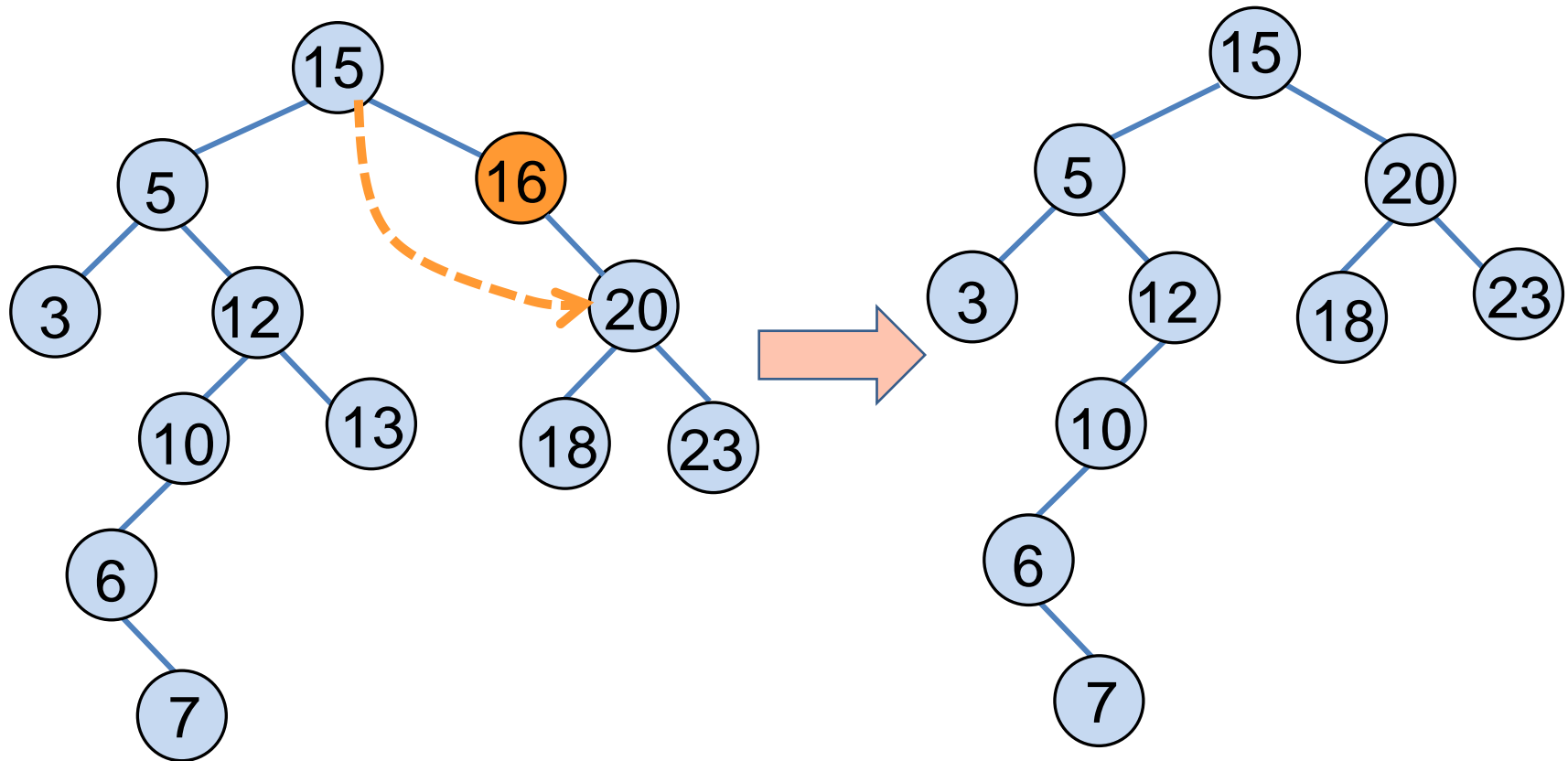
- Perform case 1 or 2 to delete it



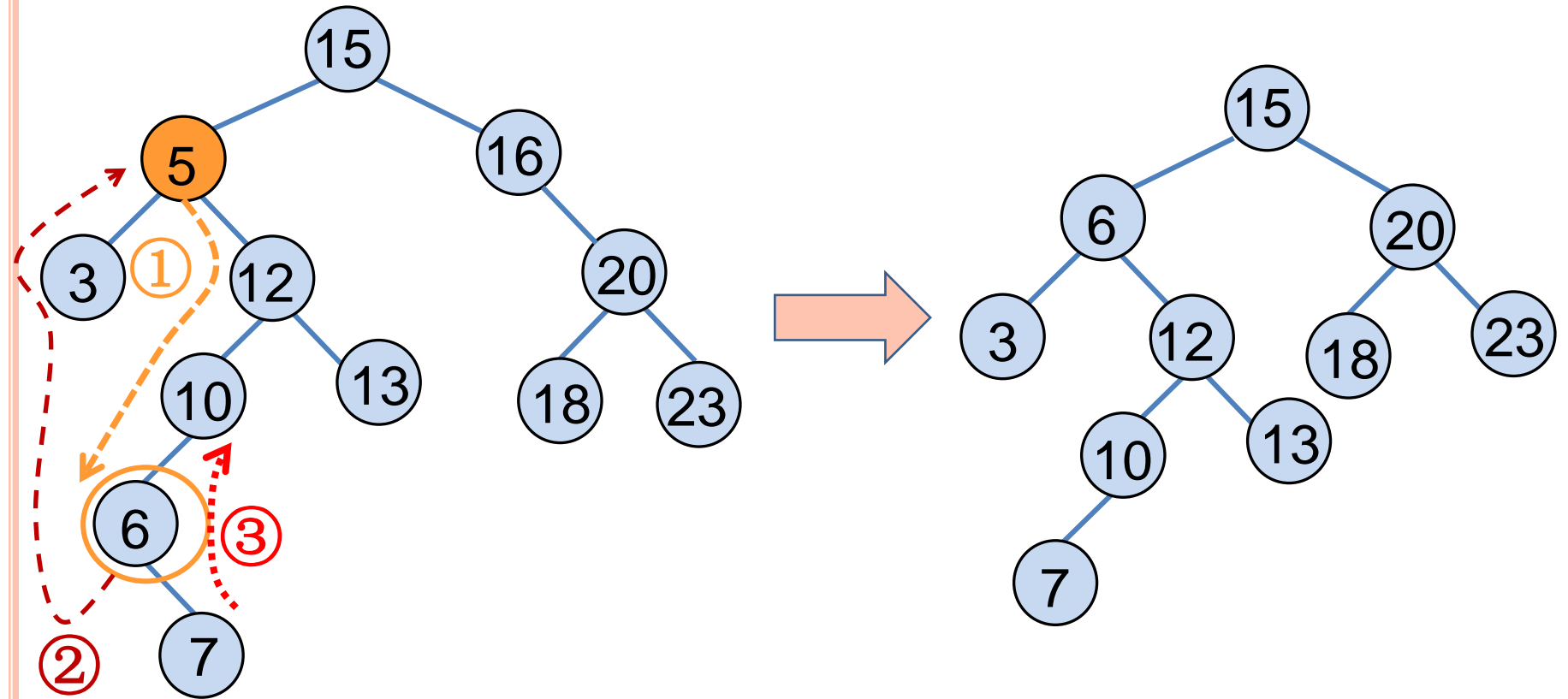
# Z HAS NO CHILDREN



# Z HAS ONLY ONE CHILD



# Z HAS TWO CHILDREN

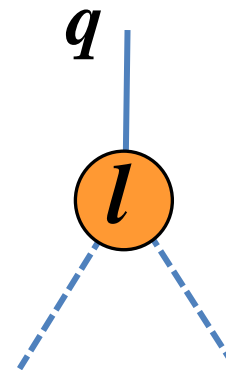
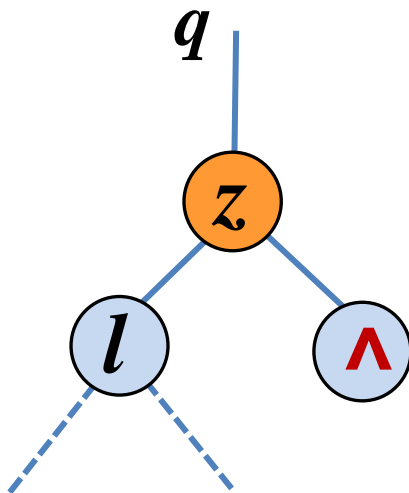
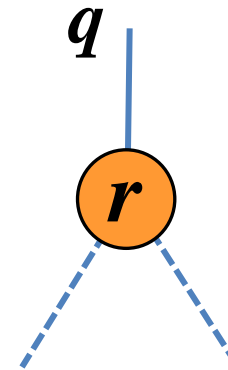
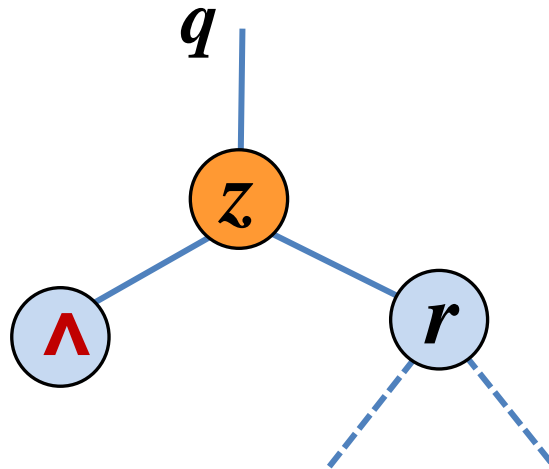


# BST OPERATIONS: DELETE

- *Why will case 2 always go to case 0 or case 1?*
  - When  $x$  has 2 children, its successor is the minimum in its right subtree.
- *Could we swap  $x$  with predecessor instead of successor?*
  - Yes.

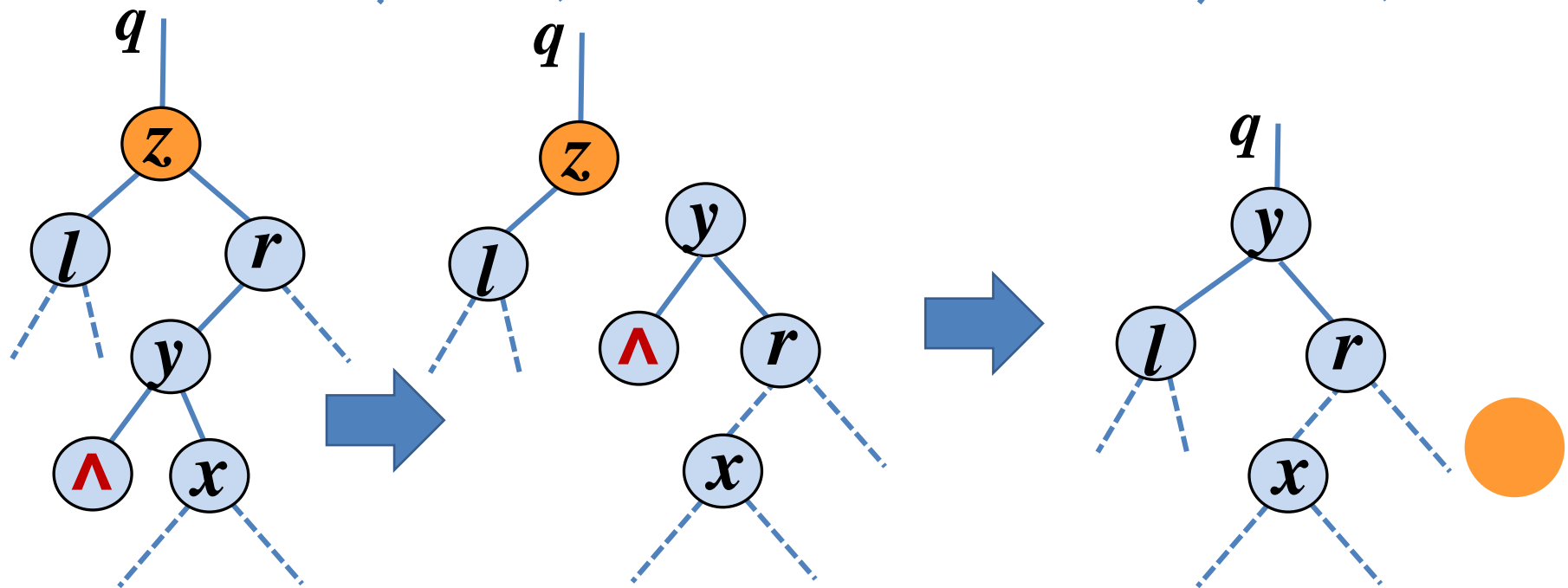
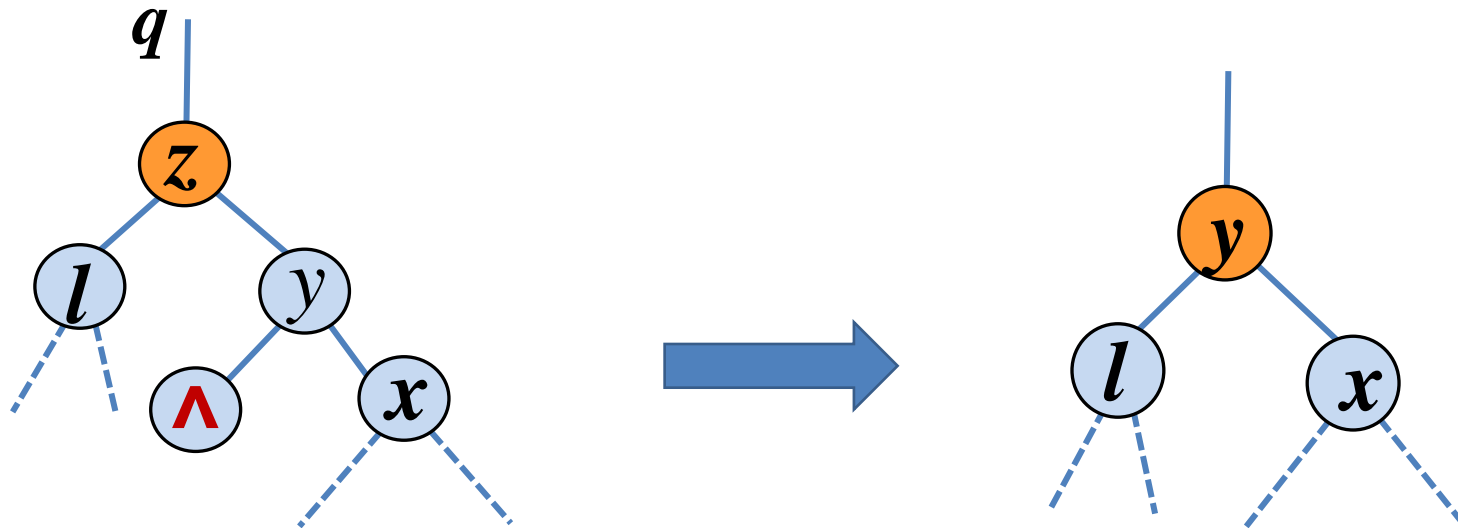


# BST OPERATIONS: DELETE (MORE)





# BST OPERATIONS: DELETE (MORE)



# SORTING WITH BINARY SEARCH TREES

- Can you come out an algorithm for **sorting** by BST?

**3 1 8 2 6 7 5**

1. By inserting nodes to build a BST
2. Inorder tree walk



# SORTING WITH BINARY SEARCH TREES

- Informal code for sorting array  $A$  of length  $n$ :

**BSTSort( $A$ )**

**for  $i = 1$  to  $n$**

**TREEINSERT( $A[i]$ );**

**InorderTreeWalk(root);**

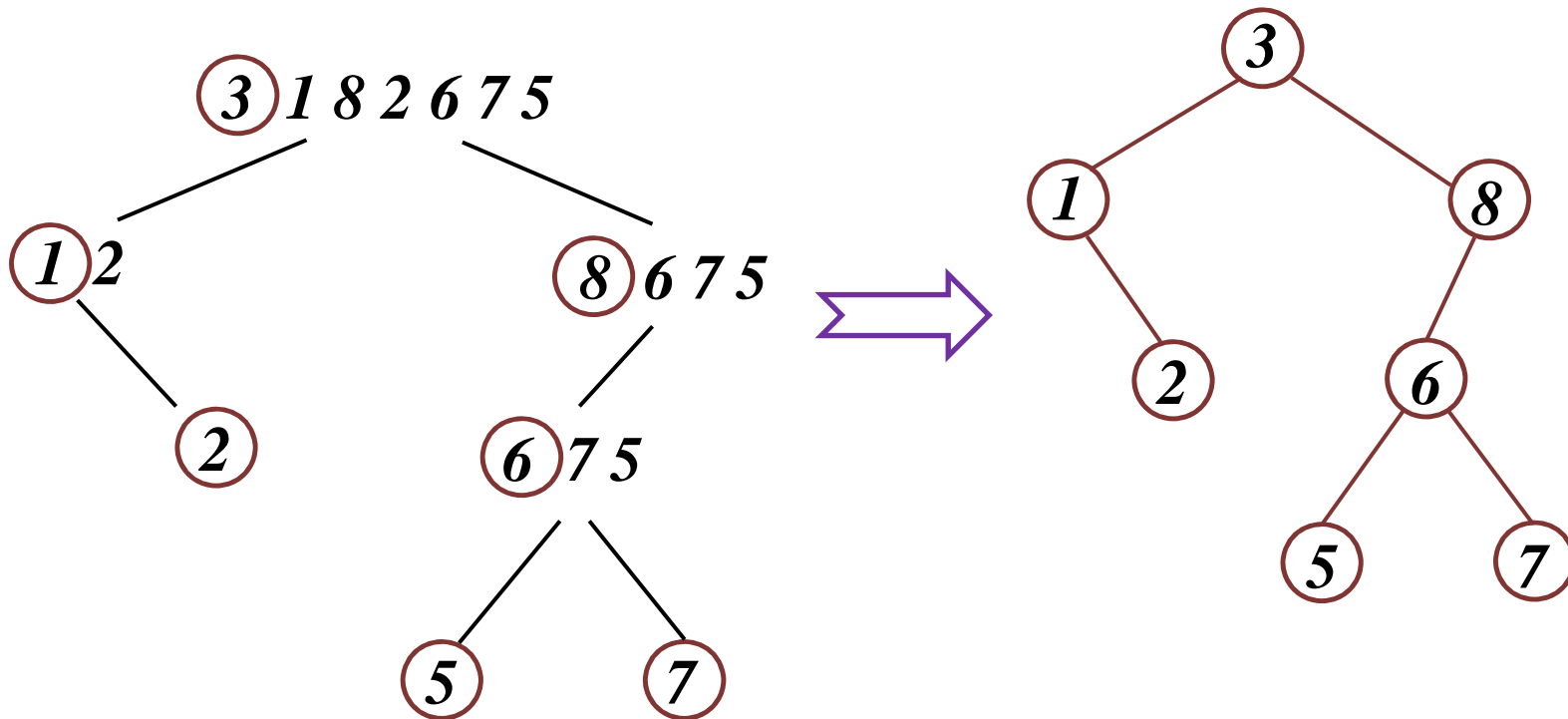
- *What will be the running time in the*
  - *Worst case?*
  - *Best case?*
  - *Average case?*



# SORTING WITH BSTs

- Average case analysis
  - It's a form of **quicksort**!

```
for  $i = 1$  to  $n$   
    TreeInsert( $A[i]$ );  
InorderTreeWalk( $root$ );
```



# SORTING WITH BSTs

- Inserted nodes are similar to partition pivot used in quicksort, but in a different order.
- BST does not partition immediately after picking the inserted node.



# SORTING WITH BSTs

- Since run time is proportional to the number of comparisons, same time as quicksort:  $O(n \lg n)$
- Which do you think is better, quicksort or BSTSort? Why?
  - **Quicksort**
  - **Sorts in place (no extra space)**
  - **Doesn't need to build data structure**

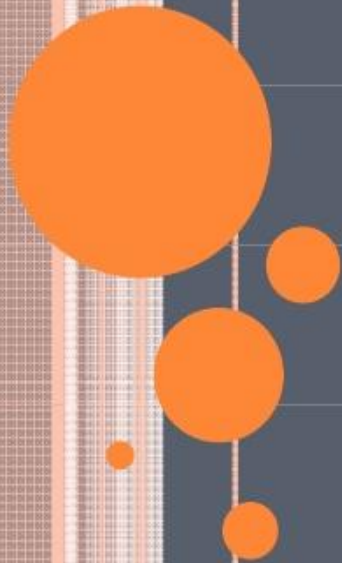


# MORE BST OPERATIONS

- BSTs are good for more than sorting. For example, can implement a priority queue.
- *What operations must a priority queue have?*
  - Insert
  - Minimum



# Randomly Built Binary Search Tree





# DEFINITION

- A *randomly built binary search tree* on  $n$  keys as the one that arises from inserting the keys in random order into an initially empty tree.
- Each of the  $n!$  permutations of the input keys is equally likely.



# RANDOMLY BUILT BINARY SEARCH TREE

- **Theorem:** The average height of a randomly-built binary search tree of  $n$  distinct keys is  $O(\lg n)$ .
- **Corollary:** The dynamic operations like *Successor*, *Predecessor*, *Search*, *Min*, *Max*, *Insert*, and *Delete* all have  $O(\lg n)$  average complexity on randomly-built binary search trees.



# NODE DEPTH

- The depth of a node = the number of comparisons made during TREE-INSERT. Assuming all input permutations are equally likely, we have
- **Average node depth**

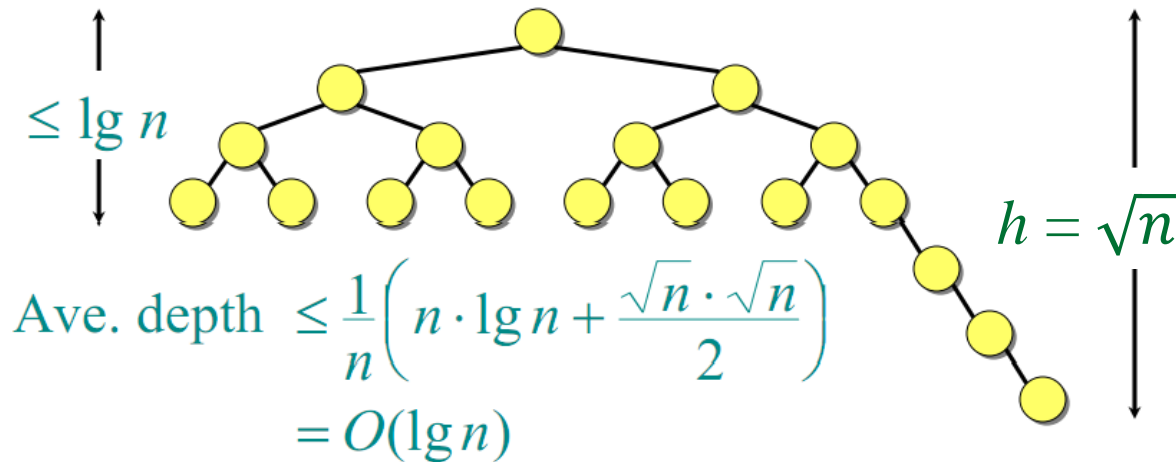
$$\begin{aligned} &= \frac{1}{n} \left[ \sum_{i=1}^n (\# \text{ comparisons to insert node } i) \right] \\ &= E \left( \sum_{i=1}^n d(x_i) \right) = \sum_{i=1}^n E(d(x_i)) \\ &= \sum_{i=1}^n \frac{1}{n} d(x_i) = \frac{1}{n} \sum_{i=1}^{\log(n+1)} i * 2^{i-1} < \log(n+1) \quad (\text{quicksort analysis}) \\ &= O(\log n) \end{aligned}$$



# EXPECTED TREE HEIGHT

- Average node depth of a randomly built BST =  $O(\lg n)$  does not necessarily mean that its expected height is also  $O(\lg n)$  (although it is).

**Example.**



$$\begin{aligned} & \sum_{i=1}^{\lg n} i \cdot 2^{i-1} - \sum_{i=1}^{\lg n} i + \sum_{i=1}^{\sqrt{n}} i \\ & \leq n \lg n + 1 - n + \frac{\sqrt{n}(\sqrt{n} + 1)}{2} \leq n(\lg n + \frac{n}{2}) \end{aligned}$$



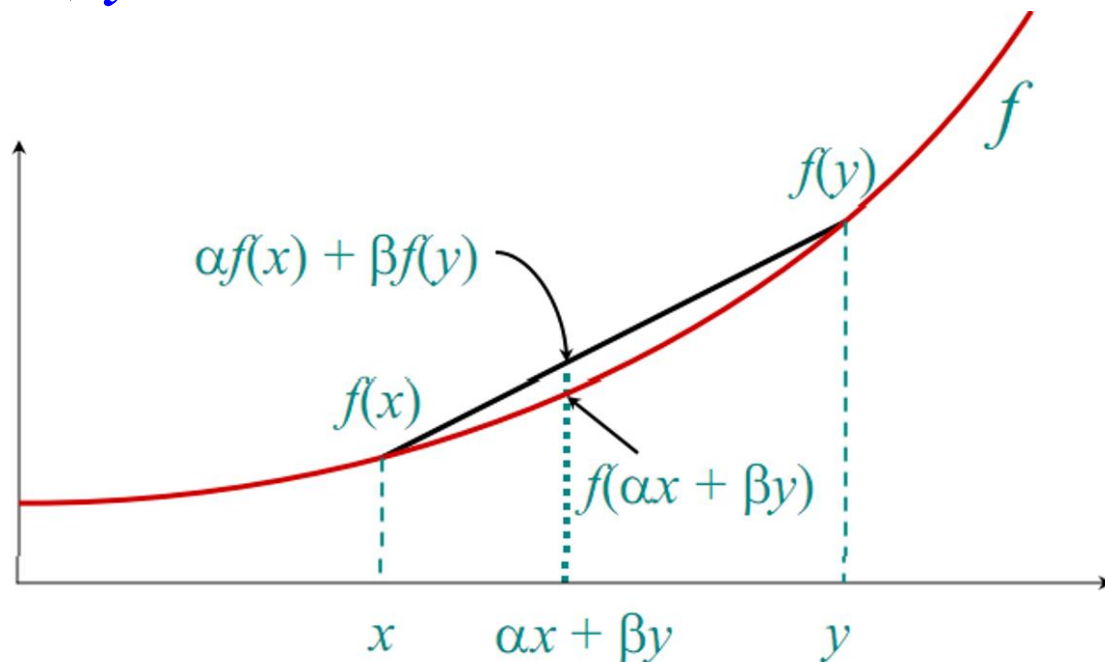
# CONVEX FUNCTIONS

## Jensen's Inequality:

- A function  $f: \mathbb{R} \rightarrow \mathbb{R}$  is *convex*, if for all  $\alpha, \beta \geq 0$  such that  $\alpha + \beta = 1$ , we have

$$f(\alpha x + \beta y) \leq \alpha f(x) + \beta f(y)$$

for all  $x, y \in \mathbb{R}$ .



# EXPECTED TREE HEIGHT OF A RANDOMLY BUILT BST

## Outline of the analysis:

- Based on the **Jensen's inequality**, we can say that  $f(E[X]) \leq E[f(X)]$  for any **convex** function  $f$  and random variable  $X$ .

$$\begin{aligned} f(E[X]) &= f\left(\sum_{k=-\infty}^{+\infty} k * P(x = k)\right) \\ &\leq \sum_{k=-\infty}^{+\infty} f(k) * P(x = k) \\ &= E[f(X)] \end{aligned}$$

$$\begin{aligned} f(\alpha x + \beta y) &\leq \alpha f(x) + \beta f(y) \\ \text{s.t. } \alpha + \beta &= 1 \end{aligned}$$



# EXPECTED TREE HEIGHT OF A RANDOMLY BUILT BST

## Outline of the analysis (Three main steps):

- Based on the **Jensen's inequality**, we can say that  $f(E[X]) \leq E[f(X)]$  for any **convex** function  $f$  and random variable  $X$ .
- Analyze the *exponential height* of a randomly built BST on  $n$  nodes, which is the random variable  $Y_n = 2^{X_n}$ , where  $X_n$  is the random variable denoting the height of the BST.
- Prove that  $2^{E[X_n]} \leq E[2^{X_n}] = E[Y_n] = O(n^3)$ , and hence that  $E[X_n] = O(\lg n)$ .



# CONVEXITY LEMMA

**Lemma.** Let  $f: \mathbb{R} \rightarrow \mathbb{R}$  be a convex function  $f$ , and let  $\{\alpha_1, \alpha_2, \dots, \alpha_n\}$  be a set of nonnegative constants ( $[0,1]$ ) such that  $\sum_k \alpha_k = 1$ . Then, for any set  $\{x_1, x_2, \dots, x_n\}$  of real numbers, we have

$$f\left(\sum_{k=1}^n \alpha_k x_k\right) \leq \sum_{k=1}^n \alpha_k f(x_k)$$

**Proof.** By induction on  $n$ . For  $n=1$ , we have  $\alpha_1=1$ , and hence  $f(\alpha_1 x_1) \leq \alpha_1 f(x_1)$  trivial.

$$\begin{aligned} f(\alpha x + \beta y) &\leq \alpha f(x) + \beta f(y) \\ \text{s.t. } \alpha + \beta &= 1 \end{aligned}$$





# PROOF (CONTINUED)

Inductive step:

$$f\left(\sum_{k=1}^n \alpha_k x_k\right) = f\left(\alpha_n x_n + (1 - \alpha_n) \sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right)$$

Algebra.



# PROOF (CONTINUED)

Inductive step:

$$\begin{aligned} f\left(\sum_{k=1}^n \alpha_k x_k\right) &= f\left(\alpha_n x_n + (1 - \alpha_n) \sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right) \\ &\leq \alpha_n f(x_n) + (1 - \alpha_n) f\left(\sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right) \end{aligned}$$

Convexity.



# PROOF (CONTINUED)

Inductive step:

$$\begin{aligned} f\left(\sum_{k=1}^n \alpha_k x_k\right) &= f\left(\alpha_n x_n + (1 - \alpha_n) \sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right) \\ &\leq \alpha_n f(x_n) + (1 - \alpha_n) f\left(\sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right) \\ &\leq \alpha_n f(x_n) + (1 - \alpha_n) \sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} f(x_k) \end{aligned}$$

Induction.



# PROOF (CONTINUED)

Inductive step:

$$\begin{aligned} f\left(\sum_{k=1}^n \alpha_k x_k\right) &= f\left(\alpha_n x_n + (1 - \alpha_n) \sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right) \\ &\leq \alpha_n f(x_n) + (1 - \alpha_n) f\left(\sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} x_k\right) \\ &\leq \alpha_n f(x_n) + (1 - \alpha_n) \sum_{k=1}^{n-1} \frac{\alpha_k}{1 - \alpha_n} f(x_k) \\ &= \sum_{k=1}^n \alpha_k f(x_k). \quad \square \end{aligned}$$

Algebra.



# JENSEN'S INEQUALITY

**Lemma.** Let  $f$  be a convex function and let  $X$  be a random variable. Then, that  $f(E[X]) \leq E[f(X)]$ .

*Proof.*

$$f(E[X]) = f\left(\sum_{k=-\infty}^{\infty} k \cdot \Pr\{X = k\}\right)$$

Definition of expectation.



# JENSEN'S INEQUALITY

**Lemma.** Let  $f$  be a convex function and let  $X$  be a random variable. Then, that  $f(E[X]) \leq E[f(X)]$ .

*Proof.*

$$\begin{aligned} f(E[X]) &= f\left(\sum_{k=-\infty}^{\infty} k \cdot \Pr\{X = k\}\right) \\ &\leq \sum_{k=-\infty}^{\infty} f(k) \cdot \Pr\{X = k\} \end{aligned}$$

Convex lemma (generalized)



# JENSEN'S INEQUALITY

**Lemma.** Let  $f$  be a convex function and let  $X$  be a random variable. Then, that  $f(E[X]) \leq E[f(X)]$ .

*Proof.*

$$\begin{aligned} f(E[X]) &= f\left(\sum_{k=-\infty}^{\infty} k \cdot \Pr\{X = k\}\right) \\ &\leq \sum_{k=-\infty}^{\infty} f(k) \cdot \Pr\{X = k\} \\ &= E[f(X)] \end{aligned}$$

Tricky step, but true –think about it.



# ANALYSIS OF BST HEIGHT

- Let  $X_n$  be the random variable denoting the height of a randomly built binary search tree on  $n$  nodes, and let  $Y_n = 2^{X_n}$  be its exponential height.
- If the root of the tree has rank  $k$ , then

$$X_n = 1 + \max \{X_{k-1}, X_{n-k}\}$$

Since each of the left and right subtrees of the root are randomly built.

Hence, we have

$$Y_n = 2 * \max \{Y_{k-1}, Y_{n-k}\}.$$





# ANALYSIS OF BST HEIGHT (CONTINUED)

- Define the indicator random variable  $Z_{nk}$  as

$$Z_{nk} = \begin{cases} 1 & \text{if the root has rank } k, \\ 0, & \text{otherwise.} \end{cases}$$

$$Z_{nk} = I(X_n = k)$$

Thus,  $Pr\{Z_{nk} = 1\} = E[Z_{nk}] = 1/n$ , and

$$Y_n = \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\})$$



# EXPONENTIAL HEIGHT RECURRENCE

$$E(Y_n) = E \left[ \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right]$$

Take expectation of both sides



# EXPONENTIAL HEIGHT RECURRENCE

$$\begin{aligned} E(Y_n) &= E \left[ \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right] \\ &= \sum_{k=1}^n E \left[ Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right] \end{aligned}$$

Linearity of Expectation.



# EXPONENTIAL HEIGHT RECURRENCE

$$\begin{aligned} E(Y_n) &= E \left[ \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right] \\ &= \sum_{k=1}^n E \left[ Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right] \\ &= 2 \sum_{k=1}^n E[Z_{nk}] E[\max\{Y_{k-1}, Y_{n-k}\}] \end{aligned}$$

Independence of the rank of the root  
from the ranks of subtree roots.



# EXPONENTIAL HEIGHT RECURRENCE

$$\begin{aligned} E(Y_n) &= E \left[ \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right] \\ &= \sum_{k=1}^n E[Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\})] \\ &= 2 \sum_{k=1}^n E[Z_{nk}] E[\max\{Y_{k-1}, Y_{n-k}\}] \\ &\leq \frac{2}{n} \sum_{k=1}^n E[Y_{k-1} + Y_{n-k}] \end{aligned}$$

The max of two nonnegative numbers is at most their sum, and  $E[Z_{nk}] = 1/n$ .



# EXPONENTIAL HEIGHT RECURRENCE

$$E(Y_n) = E \left[ \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}) \right]$$

$$= \sum_{k=1}^n E[Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\})]$$

$$= 2 \sum_{k=1}^n E[Z_{nk}] E[\max\{Y_{k-1}, Y_{n-k}\}]$$

$$\leq \frac{2}{n} \sum_{k=1}^n E[Y_{k-1} + Y_{n-k}]$$

$$= \frac{4}{n} \sum_{k=0}^{n-1} E[Y_k]$$

Each term appears twice  
and reindex.



# SOLVING THE RECURRENCE

- Use substitution to show the  $E[Y_n] \leq cn^3$  for some positive constant  $c$ , which we can pick sufficiently large to handle the initial conditions.

$$E(Y_n) \leq \frac{4}{n} \sum_{k=0}^{n-1} E[Y_k]$$



# SOLVING THE RECURRENCE

- Use substitution to show the  $E[Y_n] \leq cn^3$  for some positive constant  $c$ , which we can pick sufficiently large to handle the initial conditions (**inductive**).

$$E(Y_n) \leq \frac{4}{n} \sum_{k=0}^{n-1} E[Y_k]$$

$$\leq \frac{4}{n} \sum_{k=0}^{n-1} ck^3$$

Substitution.





# SOLVING THE RECURRENCE

- Use substitution to show the  $E[Y_n] \leq cn^3$  for some positive constant  $c$ , which we can pick sufficiently large to handle the initial conditions.

$$E(Y_n) \leq \frac{4}{n} \sum_{k=0}^{n-1} E[Y_k]$$

$$\leq \frac{4}{n} \sum_{k=0}^{n-1} ck^3$$

$$\leq \frac{4c}{n} \int_0^n x^3 dx$$

Integral method.



# SOLVING THE RECURRENCE

- Use substitution to show the  $E[Y_n] \leq cn^3$  for some positive constant  $c$ , which we can pick sufficiently large to handle the initial conditions.

$$\begin{aligned} E(Y_n) &\leq \frac{4}{n} \sum_{k=0}^{n-1} E[Y_k] \\ &\leq \frac{4}{n} \sum_{k=0}^{n-1} ck^3 \\ &\leq \frac{4c}{n} \int_0^n x^3 dx \\ &= \frac{4c}{n} \left( \frac{n^4}{4} \right) \end{aligned}$$

Solve the Integral.



# SOLVING THE RECURRENCE

- Use substitution to show the  $E[Y_n] \leq cn^3$  for some positive constant  $c$ , which we can pick sufficiently large to handle the initial conditions.

$$\begin{aligned} E(Y_n) &\leq \frac{4}{n} \sum_{k=0}^{n-1} E[Y_k] \\ &\leq \frac{4}{n} \sum_{k=0}^{n-1} ck^3 \\ &\leq \frac{4c}{n} \int_0^n x^3 dx \\ &= \frac{4c}{n} \left( \frac{n^4}{4} \right) = cn^3 \end{aligned}$$

Algebra.



# THE GRAND FINALE

- Putting it all together, and we have

$$2^{E[X_n]} \leq E[2^{X_n}]$$

Jensen's Inequality, since  
 $f(x) = 2^x$  is convex.



# THE GRAND FINALE

- Putting it all together, and we have

$$\begin{aligned} 2^{E[X_n]} &\leq E[2^{X_n}] \\ &= E[Y_n] \end{aligned}$$

Definition.



# THE GRAND FINALE

- Putting it all together, and we have

$$\begin{aligned} 2^{E[X_n]} &\leq E[2^{X_n}] \\ &= E[Y_n] \\ &\leq cn^3 \end{aligned}$$

What we just showed.

- Taking the  $\lg$  of both sides yields

$$E[Y_n] \leq 3 \lg n + O(1).$$

