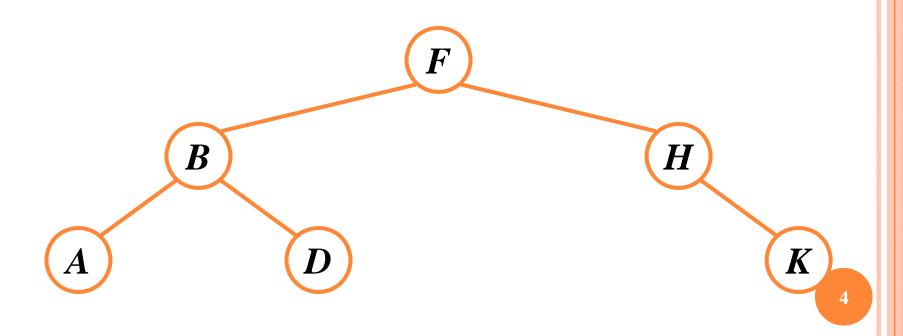
Binary Search Trees

BINARY SEARCH TREES

- Binary Search Trees (BSTs) are an important data structure for dynamic sets
- In addition to satellite data, elements have:
 - *key*: an identifying field inducing a total ordering
 - *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: $key[left(x)] \le key[x] \le key[right(x)]$
- Example:



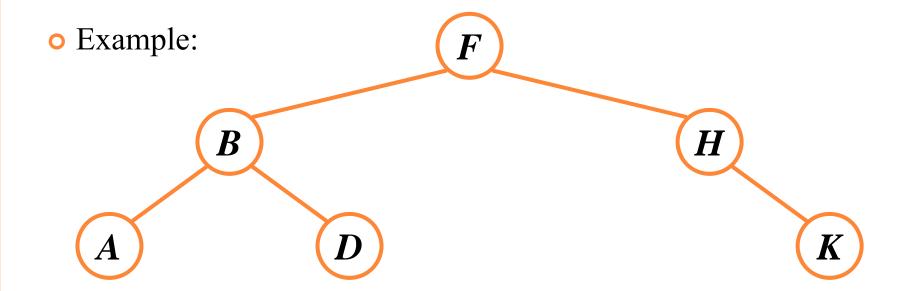
INORDER TREE WALK (TRAVERSAL)

• What does the following code do?

```
TreeWalk(x)
  TreeWalk(left[x]);
  print(x);
  TreeWalk(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



- Output: ABDFHK
- o How long will a tree walk take? □
 - *Theorem 12.1*

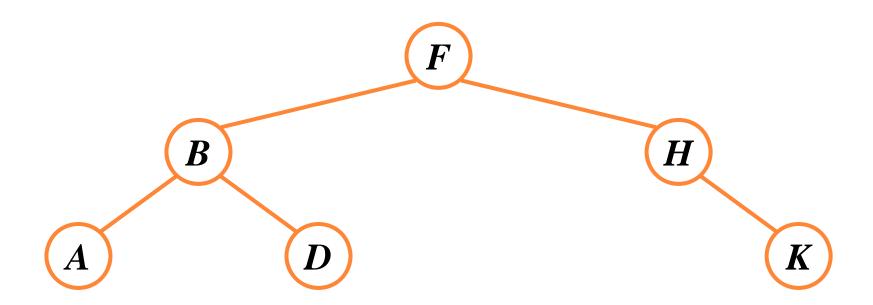
OPERATIONS ON BSTs: SEARCH

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

```
TreeSearch(x, k)
  if (x = NULL or k = key[x])
    return x;
  if (k < key[x])
    return TreeSearch(left[x], k);
  else
    return TreeSearch(right[x], k);</pre>
```

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 != NULL and k != key[x])

if (k < key[x])

x = left[x];

else

x = right[x];

return x;
```

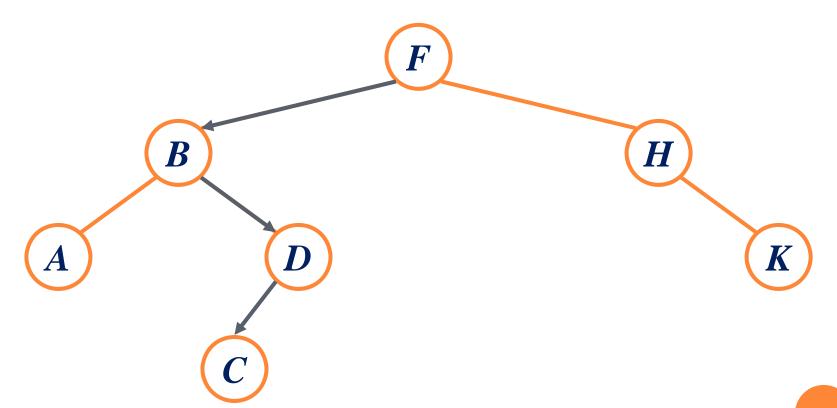
• Which of these two functions is more efficient?

OPERATIONS OF BSTs: INSERT

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

BST INSERT: EXAMPLE

• Example: Insert *C*



BST SEARCH/INSERT: RUNNING TIME

- The height of a binary search tree is h
- What is the running time of TreeSearch() or TreeInsert()?
 - O(h)
- What is the height of a binary search tree?
 - Worst case: h = O(n) when tree is just a linear string of left or right children

Minimum of BST

- TREE-Minimum(x)
- 1 while $left(x) \neq NIL$
- 2 do $x \leftarrow left[x]$
- 3 Return x

Maximum of BST

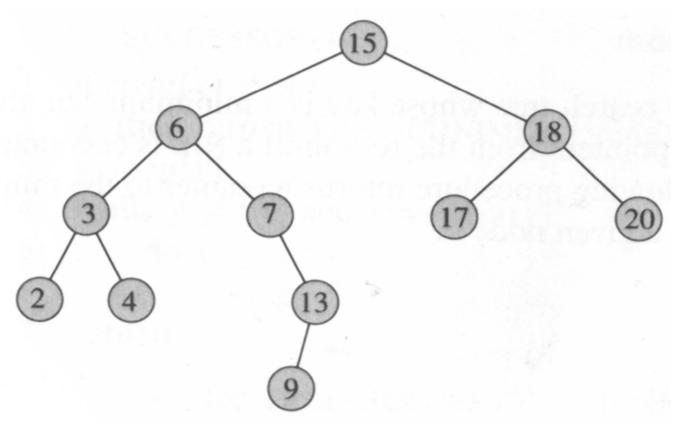
- TREE-Maximum(x)
- 1 while right(x) \neq NIL
- 2 do $x \leftarrow right[x]$
- 3 Return x

BST OPERATIONS: SUCCESSOR

- The successor of the current node is the one in the inorder tree walk.
- Two cases:
 - x has a right subtree: successor is minimum node in 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.

BST OPERATIONS: SUCCESSOR

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



• How about predecessor?

BST OPERATIONS: DELETE

Deletion is a bit tricky

o 3 cases:

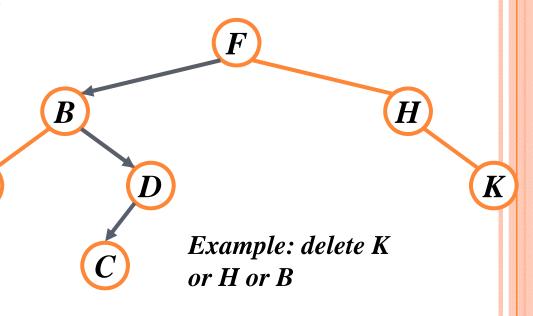
• x has no children:

•Remove x

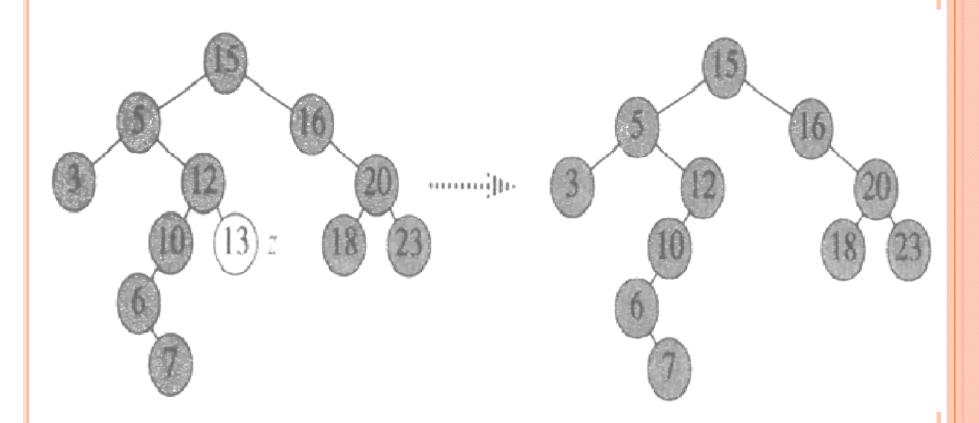
• x has one child:

•Splice out x

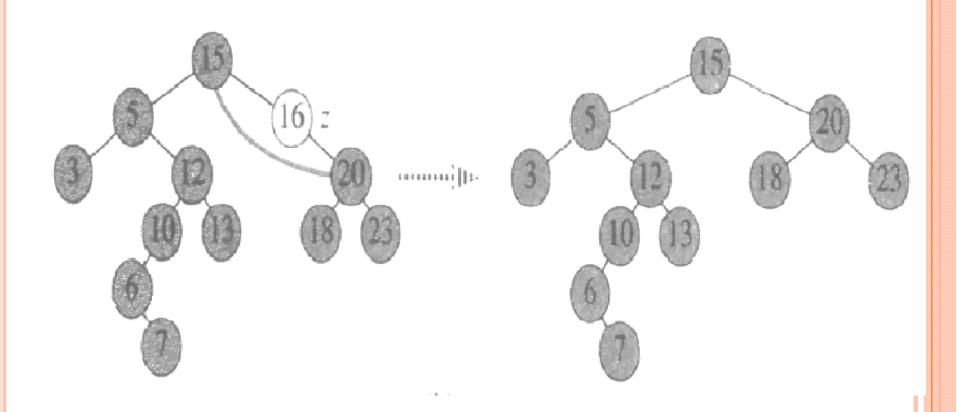
- x has two children:
 - Swap x with successor
 - oPerform case 1 or 2 to delete it



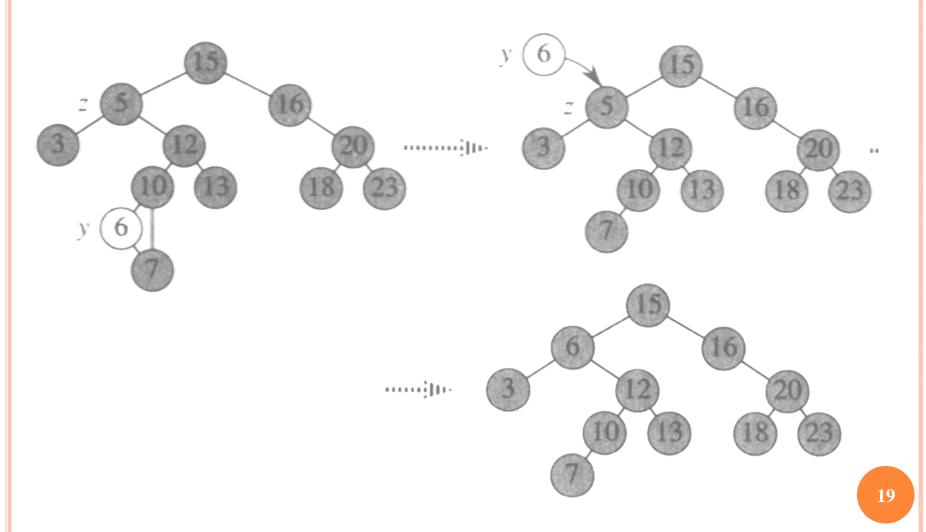
Z HAS NO CHILDREN



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.

SORTING WITH BINARY SEARCH TREES

• Can you come out an algorithm for sorting by BST?

3182675

SORTING WITH BINARY SEARCH TREES

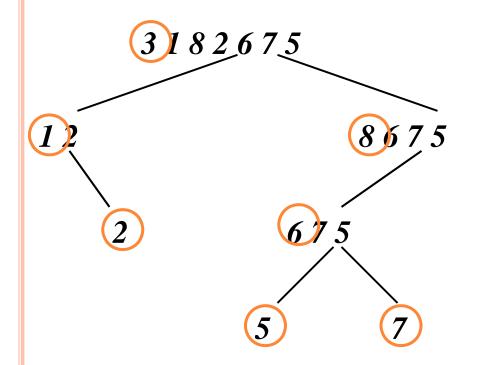
• Informal code for sorting array A of length *n*:

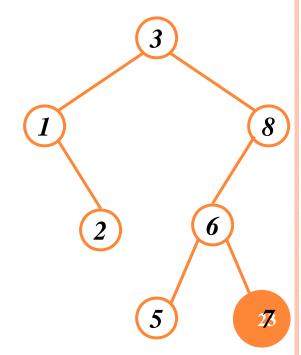
```
BSTSort(A)
for i=1 to n
TreeInsert(A[i]);
InorderTreeWalk(root);
```

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

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

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





- 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.

- 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?

- 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
 - 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 one that arises from inserting the keys in random order into an initially empty tree, where each of the *n*! permutations of the input keys is equally likely.

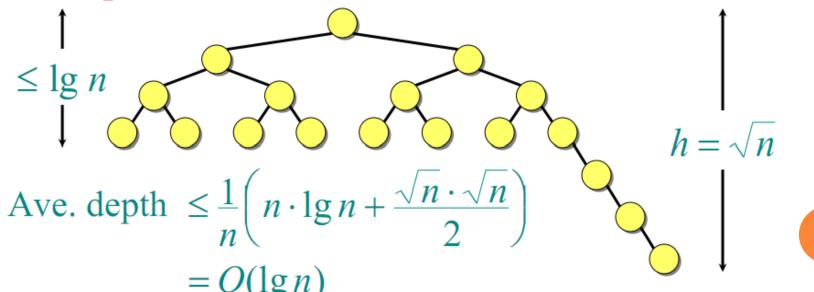
RANDOMLY BUILT BINARY SEARCH TREE

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

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.



HEIGHT OF A RANDOMLY BUILT BINARY SEARCH TREE

Outline of the analysis:

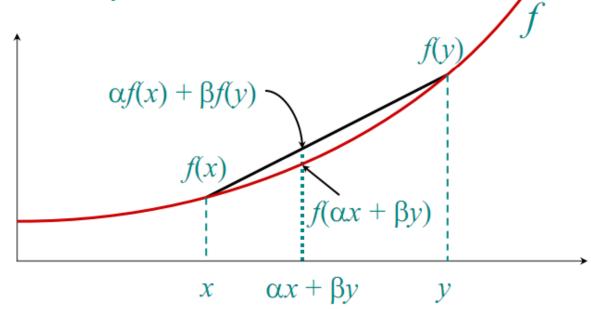
- Prove *Jensen's inequality*, which says that $f(E[X]) \le E[f(X)]$ for any convex function f and random variable X.
- Analyze the *exponential height* of a randomly built BST on n nodes, which is the random variable $Y_n = 2^{X_n}$, where X_n is the random variable denoting the height of the BST.
-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

CONVEX FUNCTIONS

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

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

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



CONVEXITY LEMMA

Lemma. Let $f: \mathbb{R} \to \mathbb{R}$ be a convex function, and let $\{\alpha_1, \alpha_2, ..., \alpha_n\}$ be a set of nonnegative constants such that $\sum_k \alpha_k = 1$. Then, for any set $\{x_1, x_2, ..., x_n\}$ 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 $f(\alpha_1 x_1) \le \alpha_1 f(x_1)$ trivially.

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:

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

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

Convexity.

PROOF (CONTINUED)

Inductive step:

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

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

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

Induction.

PROOF (CONTINUED)

Inductive step:

$$f\left(\sum_{k=1}^{n} \alpha_{k} x_{k}\right) = f\left(\alpha_{n} x_{n} + (1 - \alpha_{n}) \sum_{k=1}^{n-1} \frac{\alpha_{k}}{1 - \alpha_{n}} x_{k}\right)$$

$$\leq \alpha_{n} f(x_{n}) + (1 - \alpha_{n}) f\left(\sum_{k=1}^{n-1} \frac{\alpha_{k}}{1 - \alpha_{n}} x_{k}\right)$$

$$\leq \alpha_{n} f(x_{n}) + (1 - \alpha_{n}) \sum_{k=1}^{n-1} \frac{\alpha_{k}}{1 - \alpha_{n}} f(x_{k})$$

$$= \sum_{k=1}^{n} \alpha_{k} f(x_{k}). \quad \square \quad \text{Algebra.}$$

JENSEN'S INEQUALITY

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

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, $f(E[X]) \le E[f(X)]$.

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

$$\leq \sum_{k=-\infty}^{\infty} f(k) \cdot \Pr\{X = k\}$$

Convexity lemma (generalized).

JENSEN'S INEQUALITY

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

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

$$\leq \sum_{k=-\infty}^{\infty} f(k) \cdot \Pr\{X = k\}$$
$$= E[f(X)]. \square$$

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

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

ANALYSIS (CONTINUED)

Define the indicator random variable Z_{nk} as

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

Thus,
$$\Pr\{Z_{nk} = 1\} = \operatorname{E}[Z_{nk}] = 1/n$$
, and
$$Y_n = \sum_{k=1}^n Z_{nk} (2 \cdot \max\{Y_{k-1}, Y_{n-k}\}).$$

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

Take expectation of both sides.

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

Linearity of expectation.

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

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

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

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

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

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

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

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

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

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

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

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

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

$$= \frac{4}{n} \sum_{k=0}^{n-1} E[Y_{k}]$$
Each term appears twice, and reindex.

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

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

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

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

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

Substitution.

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

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

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

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

Integral method.

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

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

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

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

$$= \frac{4c}{n} \left(\frac{n^4}{4}\right)$$

Solve the integral.

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

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

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

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

$$= \frac{4c}{n} \left(\frac{n^4}{4}\right)$$

$$= cn^3. \text{ Algebra.}$$

Putting it all together, we have

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

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

Putting it all together, we have

$$2^{E[X_n]} \le E[2^{X_n}]$$
$$= E[Y_n]$$

Definition.

Putting it all together, we have

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

$$= E[Y_n]$$

$$\le cn^3.$$

What we just showed.

Putting it all together, we have

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

$$= E[Y_n]$$

$$\le cn^3.$$

Taking the lg of both sides yields

$$E[X_n] \le 3 \lg n + O(1).$$

HOMEWORK

o 12.1-5,12.2-5

Oue day:10.10