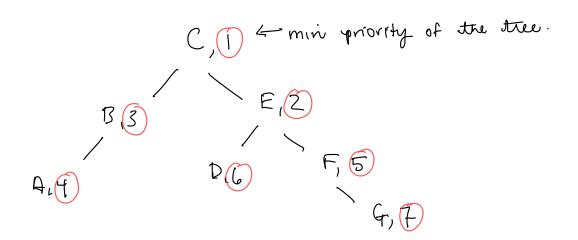
Random Trees

A Cartesian tree is a binary tree, introduced by Vuilleman (1980) for geometric range searching.

Prioritics = also called "time stamps".

Build a BST on the keys by inserting them in the order given by their priorities.



Note that the tree has min heap priority based on the Time stamps/priorities

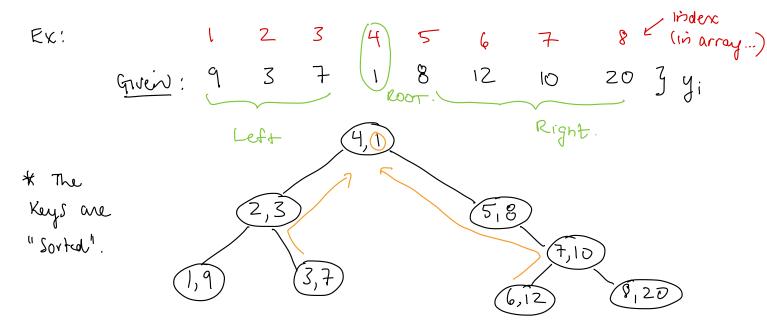
The *inorder* traversal of the tree results in the sorted keys. This is equivalent to BSTs.

**DEFINITION:** Cartesian tree

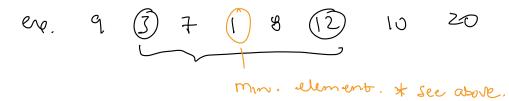
Given a set 
$$(x_1, y_1), \dots, (x_n, y_n)$$
  
 $f \in priority (time stamp).$   
Key  
A cartesian tree is a B.S.T. with respect to the Keys  
and a min. heap with respect to the priorities.

**Fact:** Given data  $(x_1, y_1, ) \dots (x_n, y_n)$ , the cartesian tree is unique.

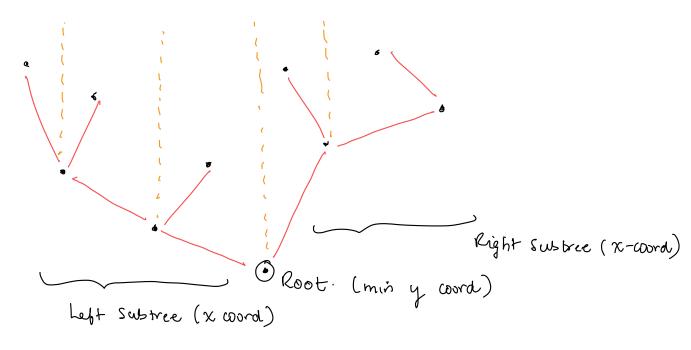
A Cartesian tree can be built on a sequence of items  $y_1, \ldots, y_n$  by using the y's as the priorities and the order of the items as the keys:



**Application:** Range Searching: Given 2 elemens in the above sequence/array, the minimum value in the sbuarray is the **lowest common ancestor** in the tree.



**Application:**Cartesean trees can be used to store 2-dim data: BST keys: x-coord Priorities: y-coord



Suppose we have keys  $x_1, \ldots, x_n$  for a **Binary search tree**. To build a Cartesian tree, we would need priorities. We could use the priorities  $1, \ldots, n$ .

$$X_1 - \dots + X_n$$
 Keys 7 The result  $\Rightarrow$  Keys get inserted into  
 $1 - \dots + n$  prioritics. BST in order  $X_1 \rightarrow X_n$ . Could be  
Unbalanced!!

In a **Random binary search tree** the keys are inserted in **Random** order. This could be achieved by **permuting** the data and the inserting them one by one into the Cartesean tree with priorities  $1, \ldots, n$ .

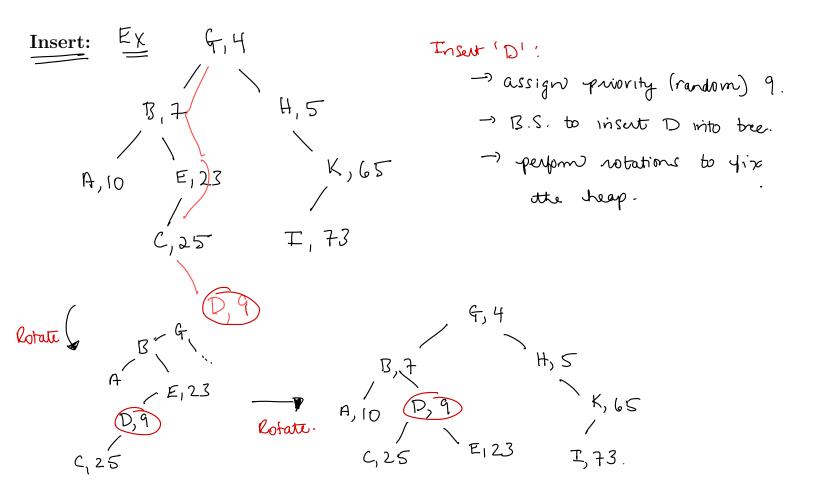
The above construction of a Cartesean tree is equivalent to a Random binary search tree. So it's depth is logarithmic with high probability!.

**Problem:** How could we maintain this randomized property as items are added/removed from the tree? This happens if we receive only 1 item at a time...

## TREAPS

First introduced by Aragon, Seidel (1989). It is a Cartesean tree where each node received a **Randomly chosen priority** 

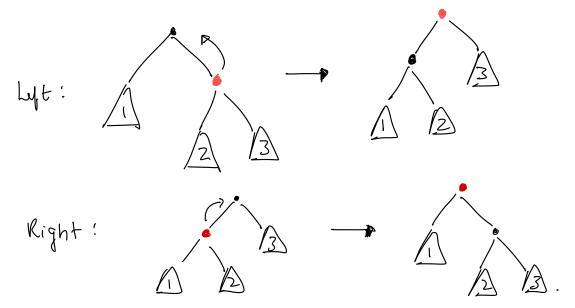
Operations: Goal is to "maintain" the random BST relationship ...

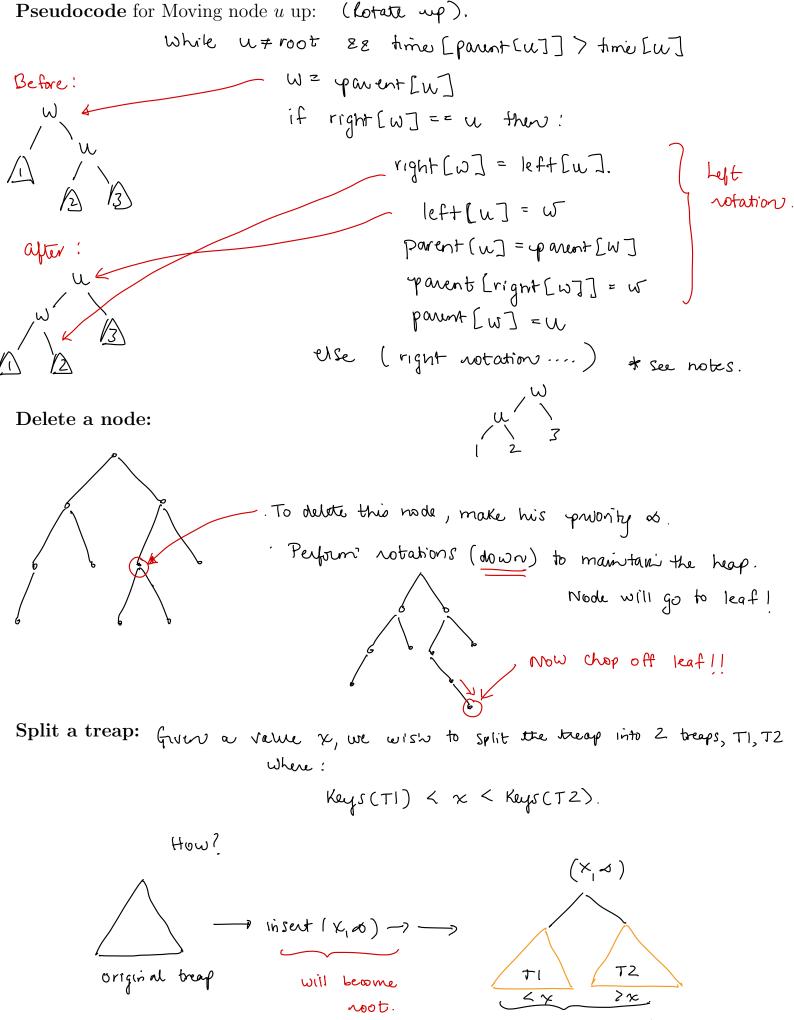


The insert operation:

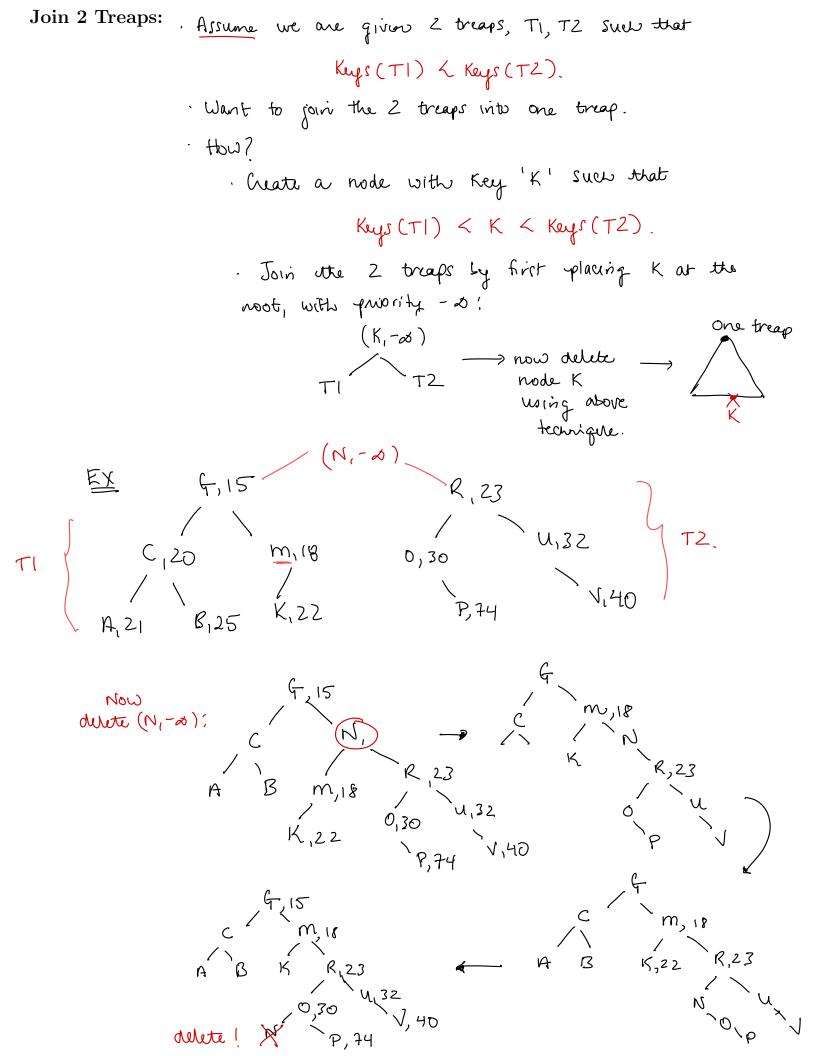
- assigns the new node a random priority
- searches for the position according to the BST
- performs a sequence of rotations to maintain the heap property

The rotate operation:

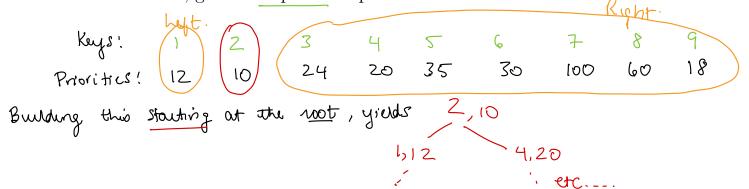




These are your 2 breaps.



Recall the Cartesian tree, given a sequence of priorities:

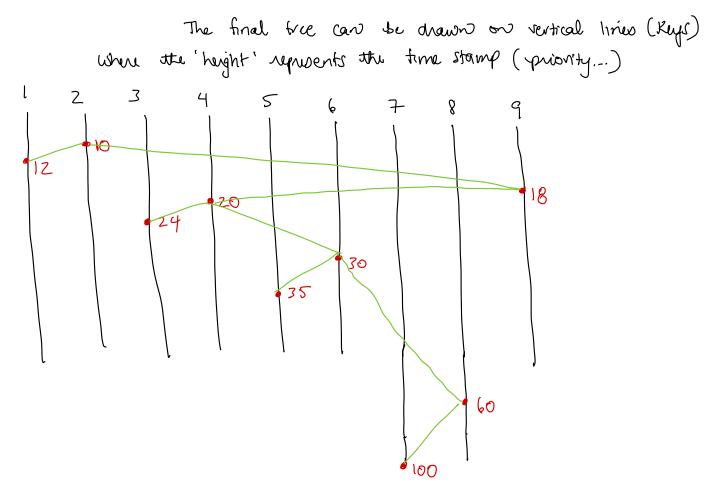


The tree can also be constructed **sequentally** in linear time, using the above insert algorithm:

· Insert the items one by ones maintaining a cartesian tree at each • The new Key will be inserted as the right child of the last element. Step... (1, 12) $(1,12) \xrightarrow{R} (2,10)$  (1,12) (1,12)(2,10) (1,12) (3,24) (4) (2, 10)(2,10)  $(1,12) (3,24) \xrightarrow{k} (1,12) (4,20) \\ (4,20) (3,24) \\ (4,20) (3,24) \\ (5,24)$ (Z, IO) $\begin{array}{c} (2,10) \\ (1,12) \\ (1,12) \\ (3,24) \\ (6,30) \\ (6,30) \\ (6,35) \\ (6,35) \\ (7,100) \\ (7,10) \\$ (8,60)

ETC...

A Cartesean tree on the sequence above can be pictured on an abacus...



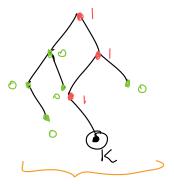
Towards the analysis of treap depth:

Recall that for a discrete Random Variable,

$$E(X) = \sum_{n} \chi_{n} P(X = \chi_{n})$$

Assume that the data of our treap is  $(1, T_1), \ldots, (n, T_n)$ . In other words, our keys are  $1 \ldots n$  for Let the depth of node k be  $D_k$ : Keys simplicity.

$$D_{K} = \sum_{j \neq K} X_{jK}$$
  
 $X_{jK} = \int I \quad if \quad j \quad is \quad ancestor \quad of \quad K$   
(O otherwise.



depths of K=3

1+1+1+0+0+0+0+0+0+0

To find the **Expected** depth we sum up the chance that each node in the tree is an ancestor of k:

$$E(D_{K}) = \sum_{j \neq K} P(j \text{ is an use for of } K)$$

$$= \sum_{j \leq K} P(j \text{ is an use for of } K) + \sum_{j \geq K} P(j \text{ is an use for } K)$$

$$= \sum_{j \leq K} P(T_{j} \text{ is smallust of } K) + \sum_{j \geq K} P(T_{j} \text{ is smallust} K)$$

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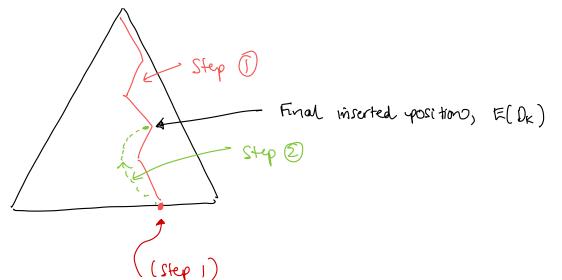
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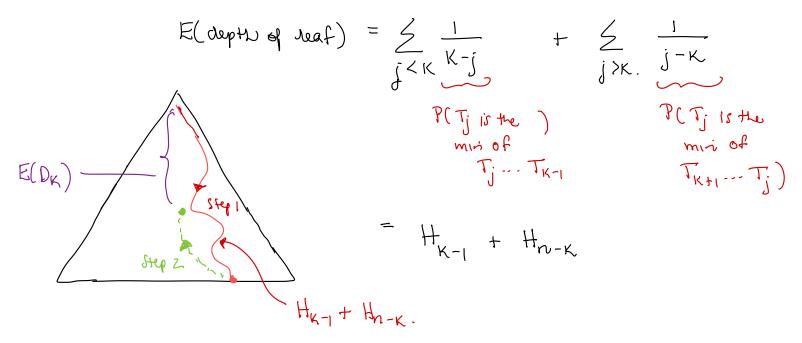
$$= \sum_{j < K} P(T_{j} \text{ is an use for } K) + \sum_{j \geq K} P(T_{j} \text{ is$$

The **insert** operation:

Recall it first searches through the BST to insert the key as a leaf, and the rotates up to maintain the heap property of the priorities.



How deep is the leaf expected to be? The expected depth of a leaf is computed in the same way, however we assume it's priority is  $\infty$  since it is at the bottom of the tree.



After the node is rotated up to its finial position, its depth is that of a typical random node:

$$E(D_{K}) = H_{K} + H_{n-K+1} - 2.$$

Thus the number of rotations is the difference between these 2 paths:

$$\operatorname{Greev} \coloneqq \left( \operatorname{H}_{\mathsf{K}-1} + \operatorname{H}_{\mathsf{n}-\mathsf{K}} \right) - \left( \operatorname{H}_{\mathsf{K}} + \operatorname{H}_{\mathsf{n}-\mathsf{K}+1} - Z \right) \leq 2$$

The insert time is bounded by the expected depth of the leaf.

 $\sim 2 \text{Nm2.}$  Similary, delete in a treap is  $O(\log n)$ .

Note that the expected height of the treap is  $O(\log n)$ , (not shown).

## QuickSort:

Note that the algorithm for quicksort is essentially the same as building a RBST.

The number of comparisons is thus the same as those made when building a BST:

$$\begin{aligned}
\# & \text{comp} = \sum_{i=1}^{n} D_{i} \\
E(\#_{\text{comp}}) &= \sum_{i=1}^{n} E(D_{i}) \\
&= \sum_{i=1}^{n} H_{i} + H_{n-i+1} - 2 \\
(H_{i} + H_{3}) + (H_{2} + H_{2}) + (H_{3} + H_{1}) \\
&= (2 \sum_{i=1}^{n} H_{i}) - 2n \\
&= 2 \sum_{i=1}^{n} \frac{1}{j} - 2n \\
&= 2 \sum_{i=1}^{n}$$

$$= \left(2nH_{nv} - 2Z(1) + 2Z'_{j}\right) - 2nz$$

$$= 2nvH_{nv} - 2nv + 2H_{nv} - 2nz$$

$$= 2nvH_{nv} + 2H_{nv} - 4nz$$

$$\stackrel{\sim}{=} 2nvH_{nv} + 2H_{nv} - 4nz$$

$$\stackrel{\sim}{=} 2nvInnv \stackrel{\sim}{=} 1.38nv\log_{2}nz$$

•