

An Introduction to the Contraction Method

Ralph Neininger
Institute for Mathematics
Johann Wolfgang Goethe-Universität
Frankfurt am Main

Bucket Selection

Given: $U_1, \dots, U_n$ indep, $\text{unif}[0, 1]$ distributed.

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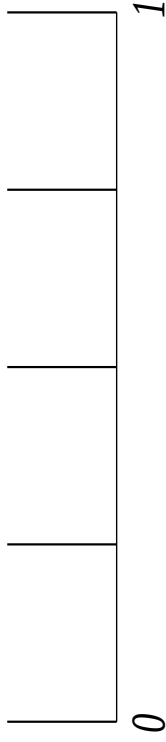
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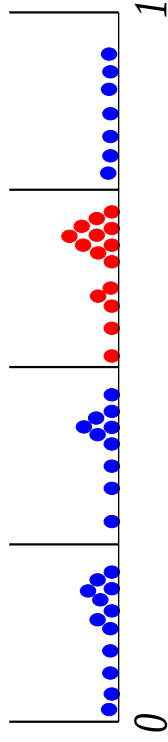
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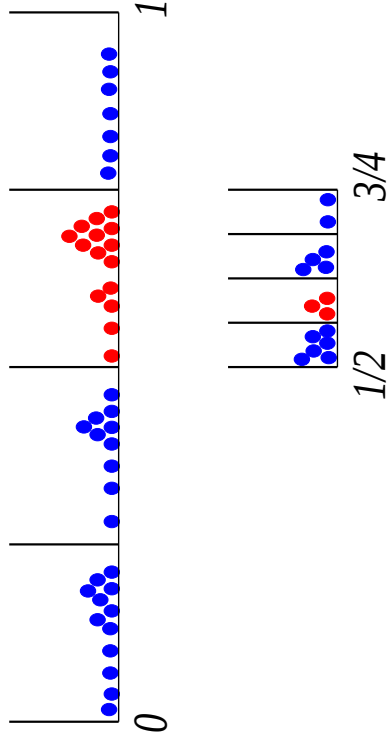
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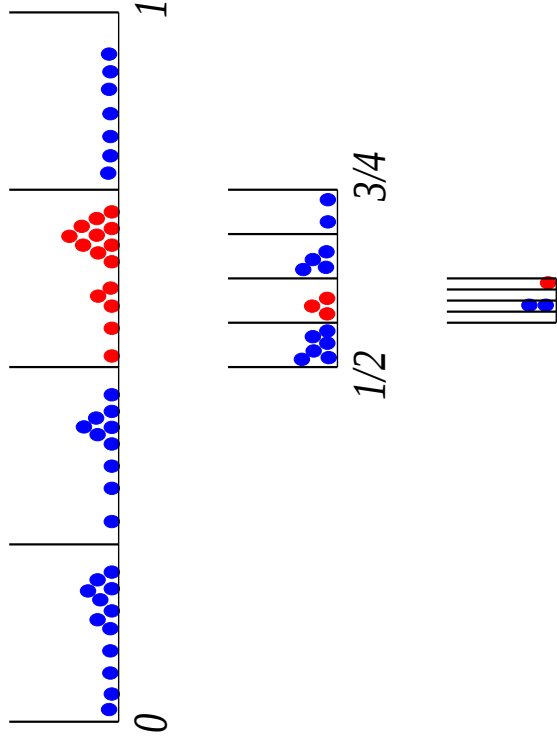
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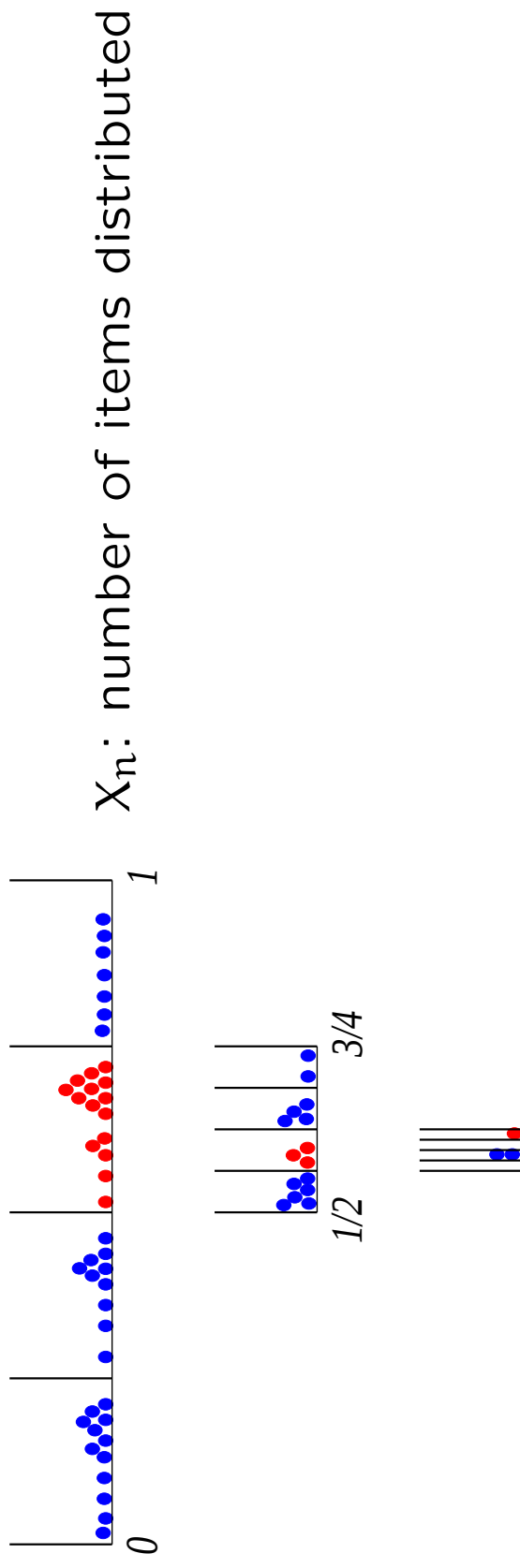
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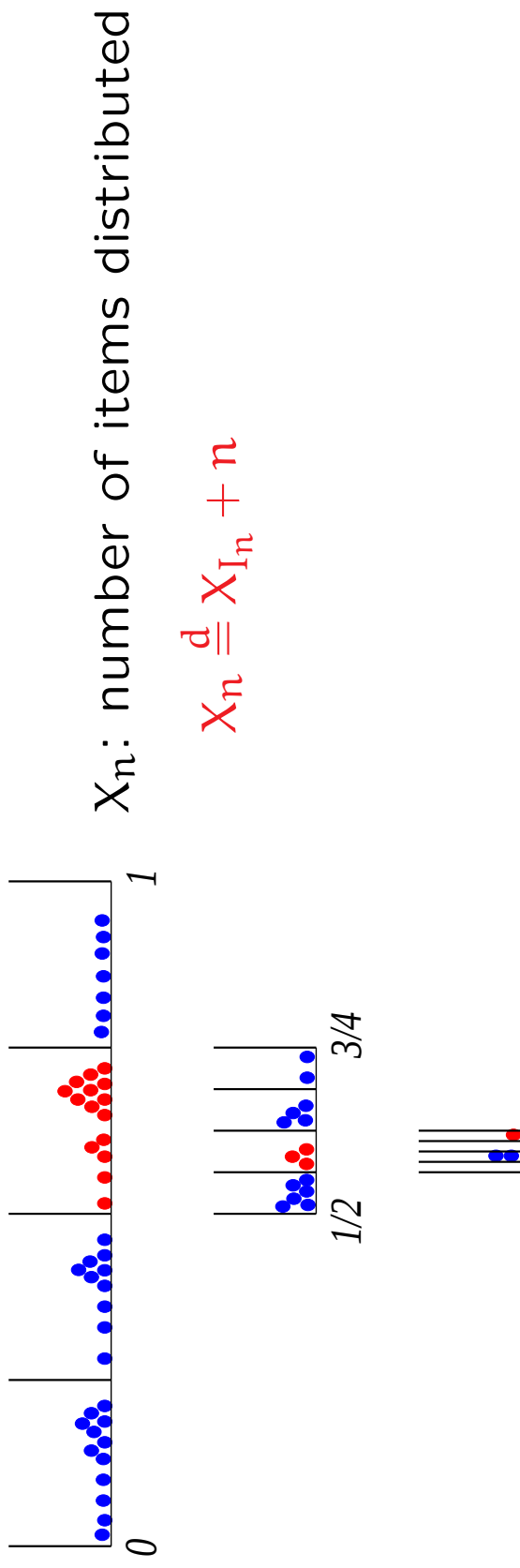
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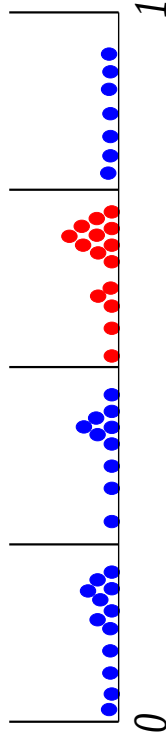
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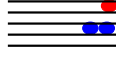
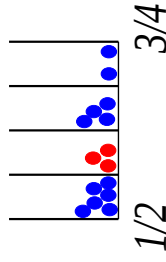
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X_n : number of items distributed

$$X_n \stackrel{d}{=} X_{I_n} + n$$

$I_n = \#$ items in relevant bucket



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Given: $x_1, \dots, x_n \in \mathbb{R}$.

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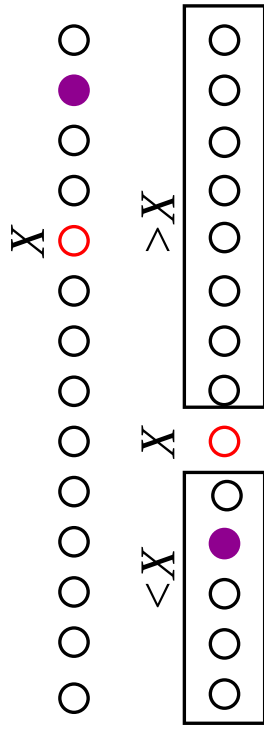
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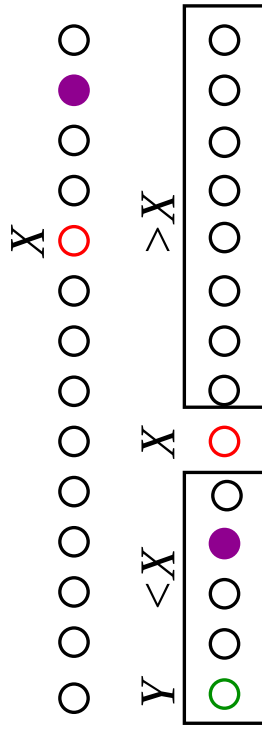
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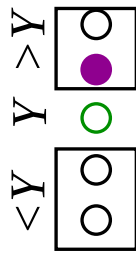
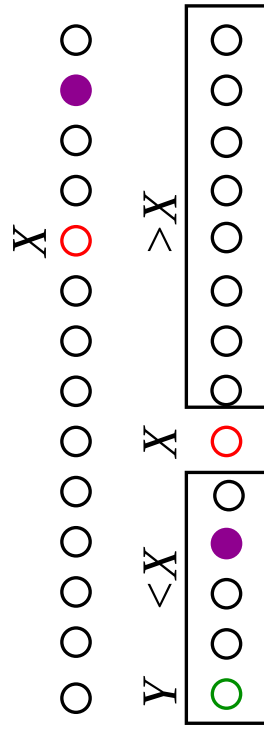
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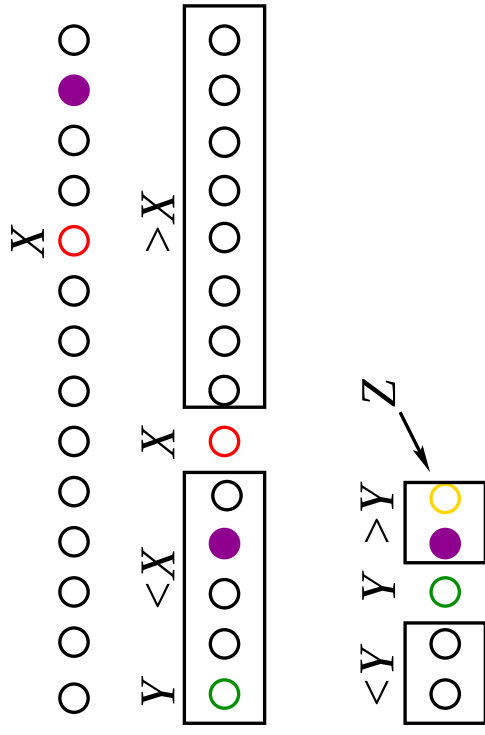
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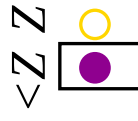
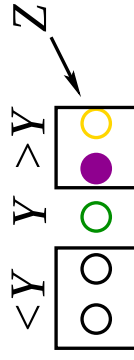
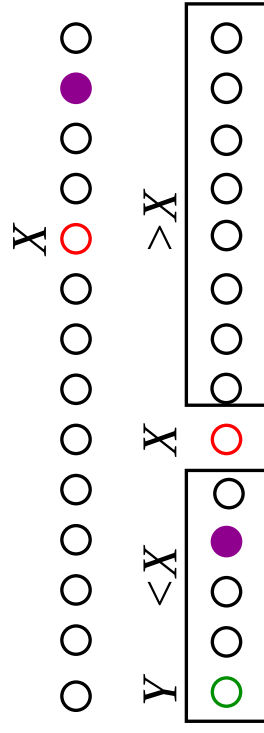
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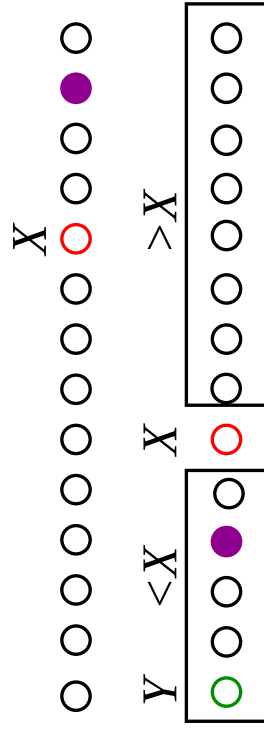
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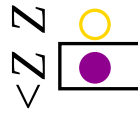
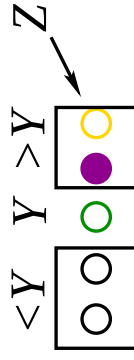
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X_n : number of key comparisons

$$X_n \stackrel{d}{=} X_{I_n} + n - 1, \quad n \geq 2.$$

$\overset{<Y}{\boxed{\circ \circ}} \overset{Y > Y}{>} \boxed{\circ \circ \circ \circ \circ} \overset{Z}{\swarrow}$

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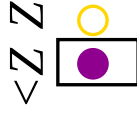
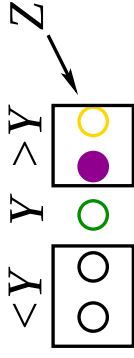
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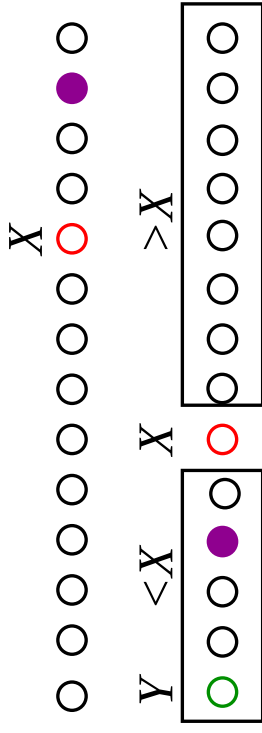
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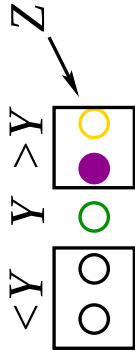
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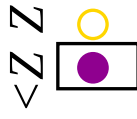
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For $k = 1$: $I_n \stackrel{d}{=} \text{unif}\{0, \dots, n - 1\}$



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Hence, this suggests

$$\mathbb{E} X_n = \mathbb{E}[nY_n] \sim n\mathbb{E} Y = 2n,$$

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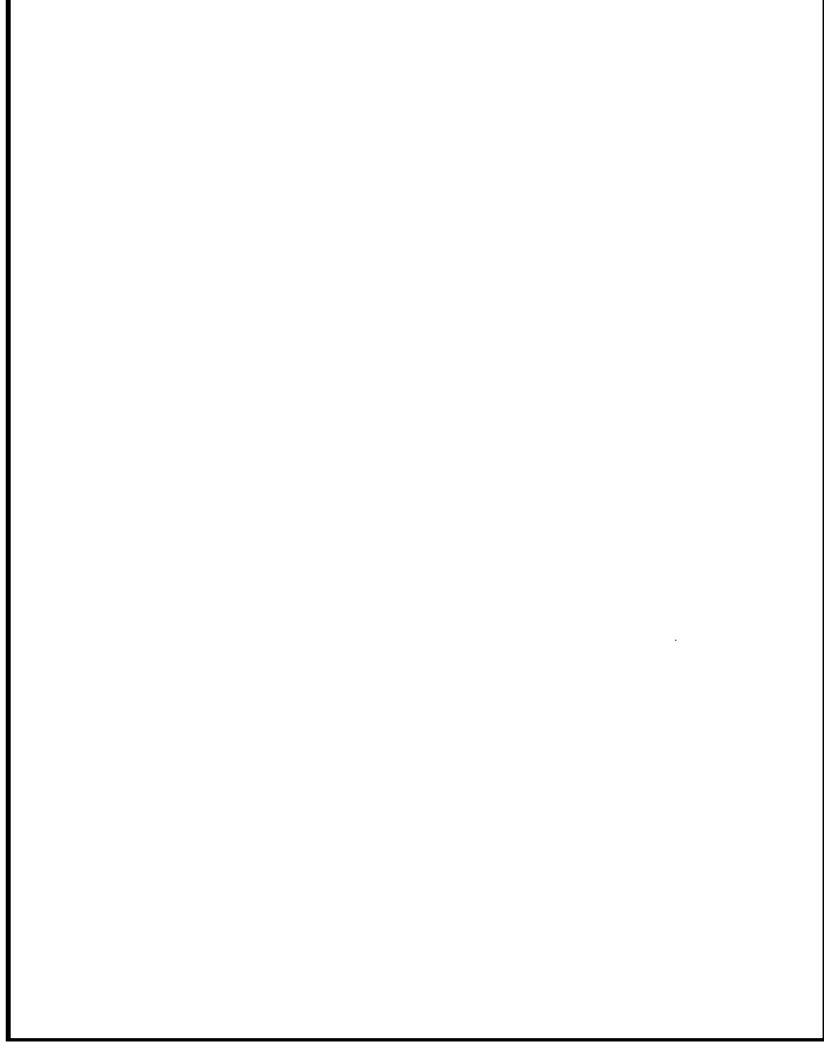
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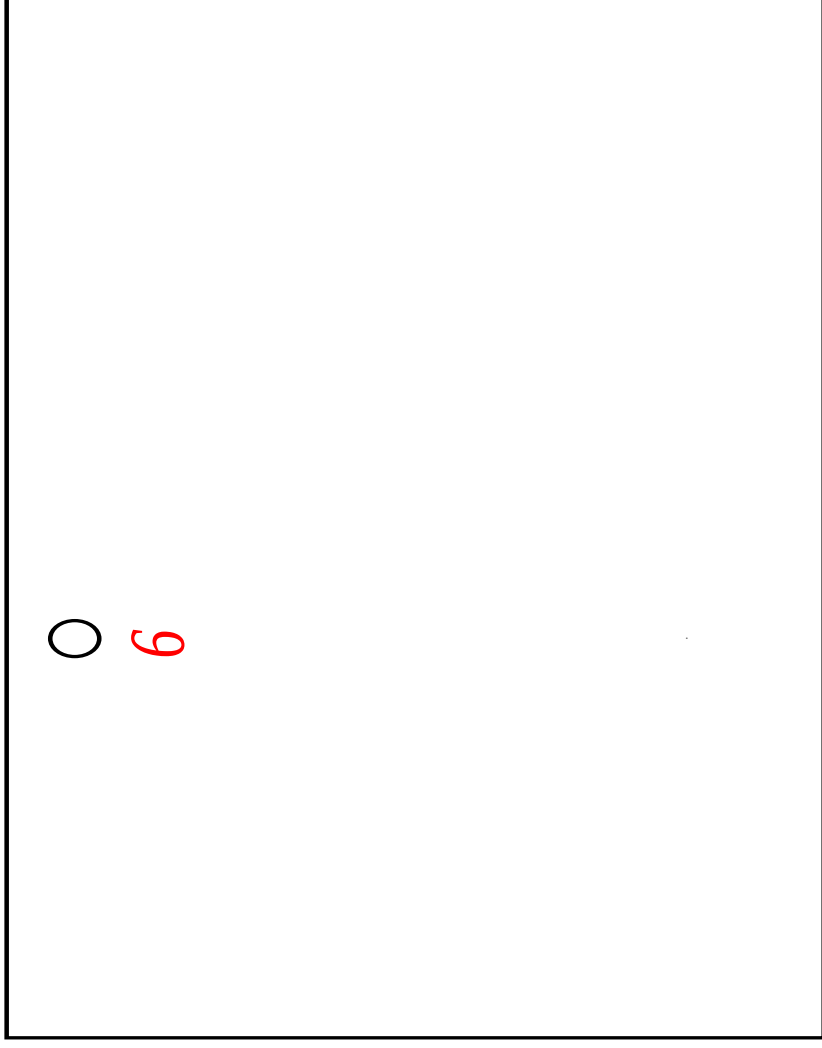
Binary search tree

Given numbers: 6, 1, 8, 7, 5, 3, 10, 2, 11, 4, 9.



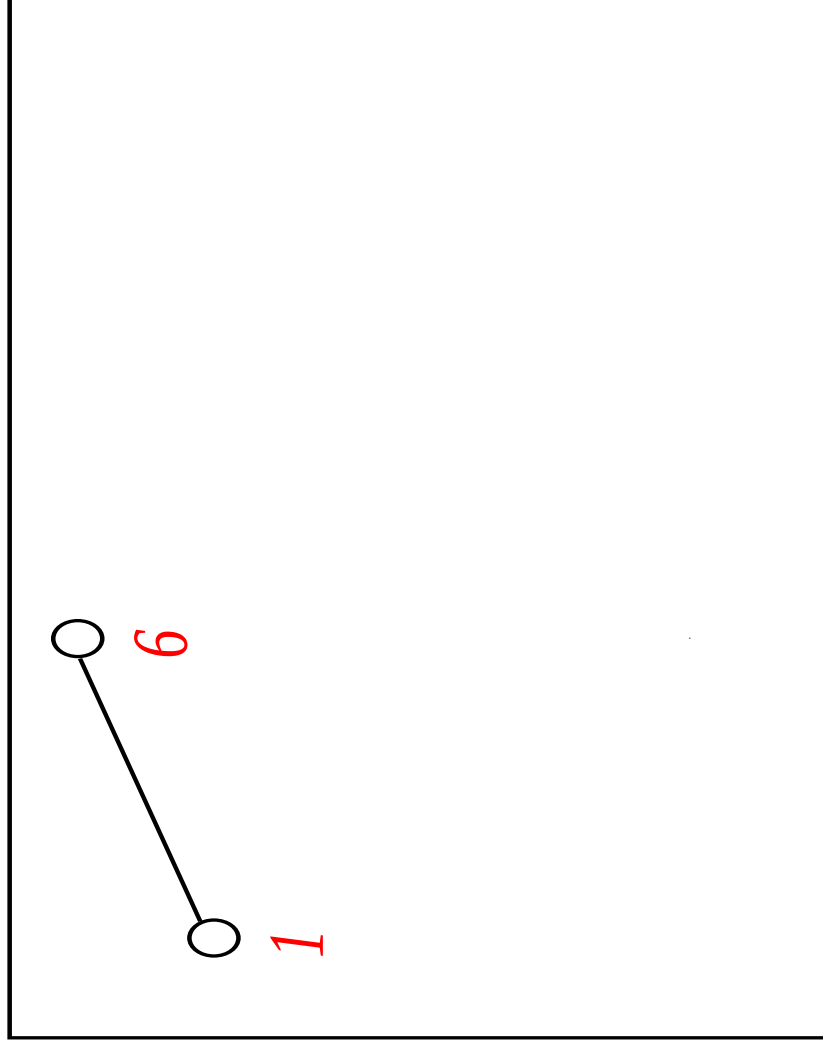
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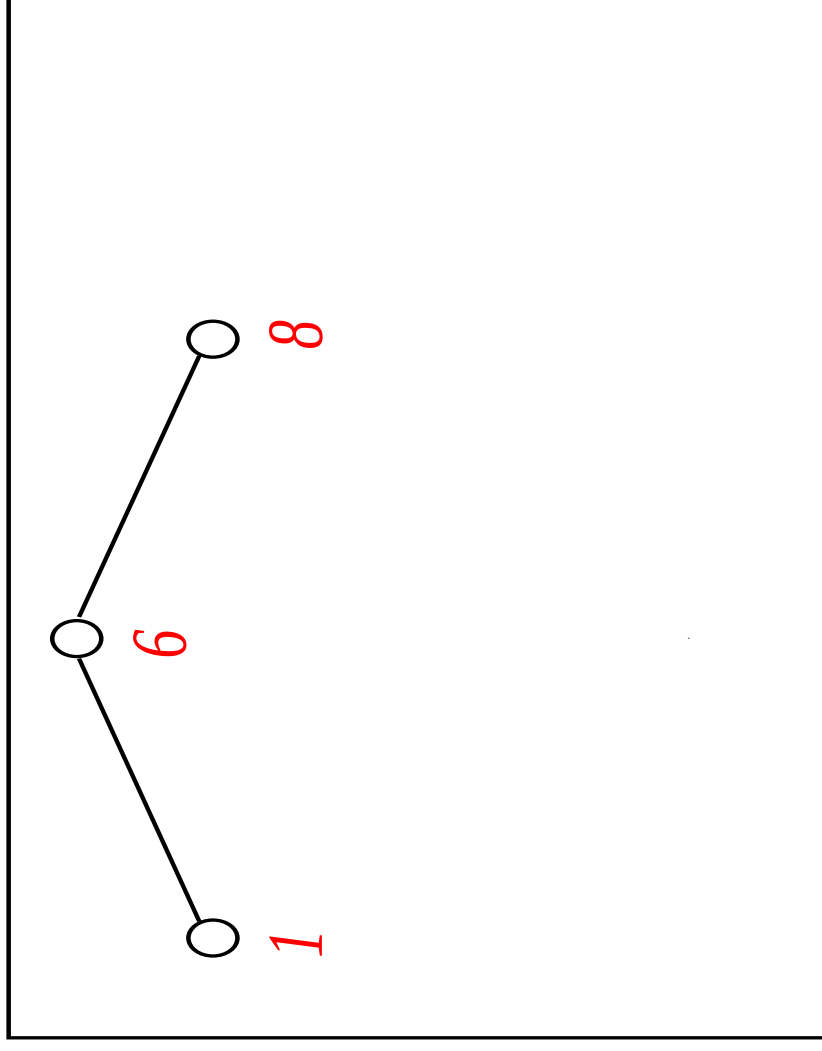
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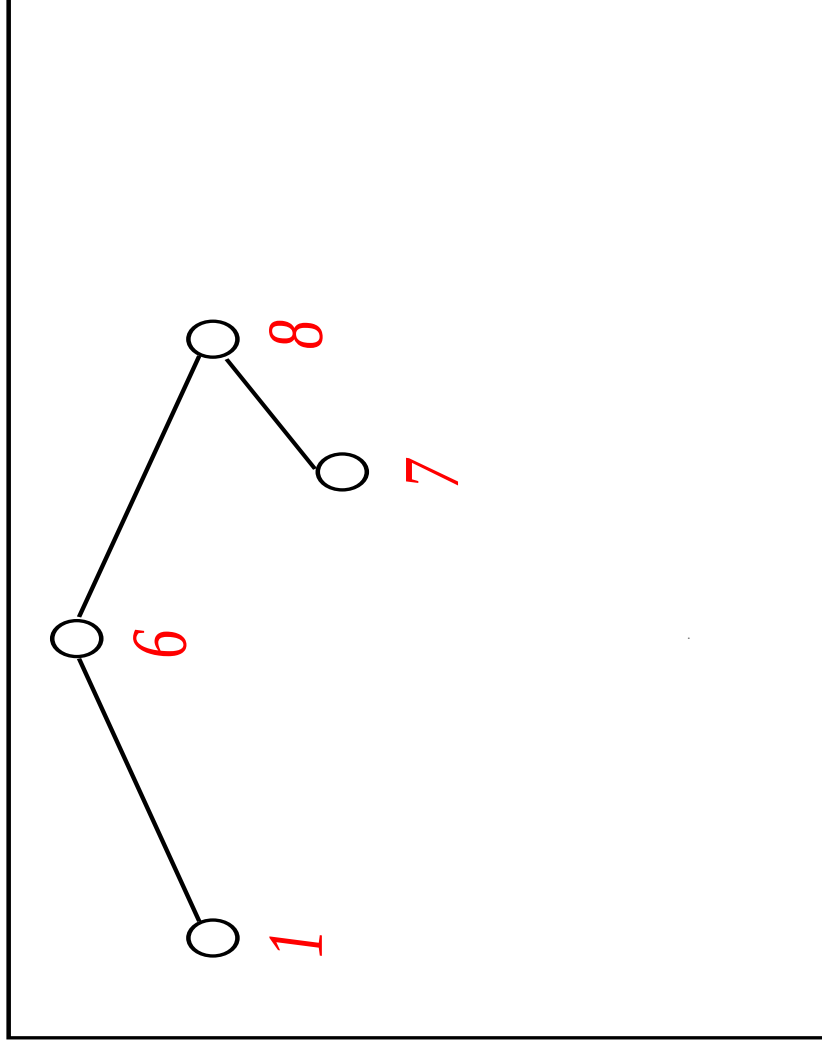
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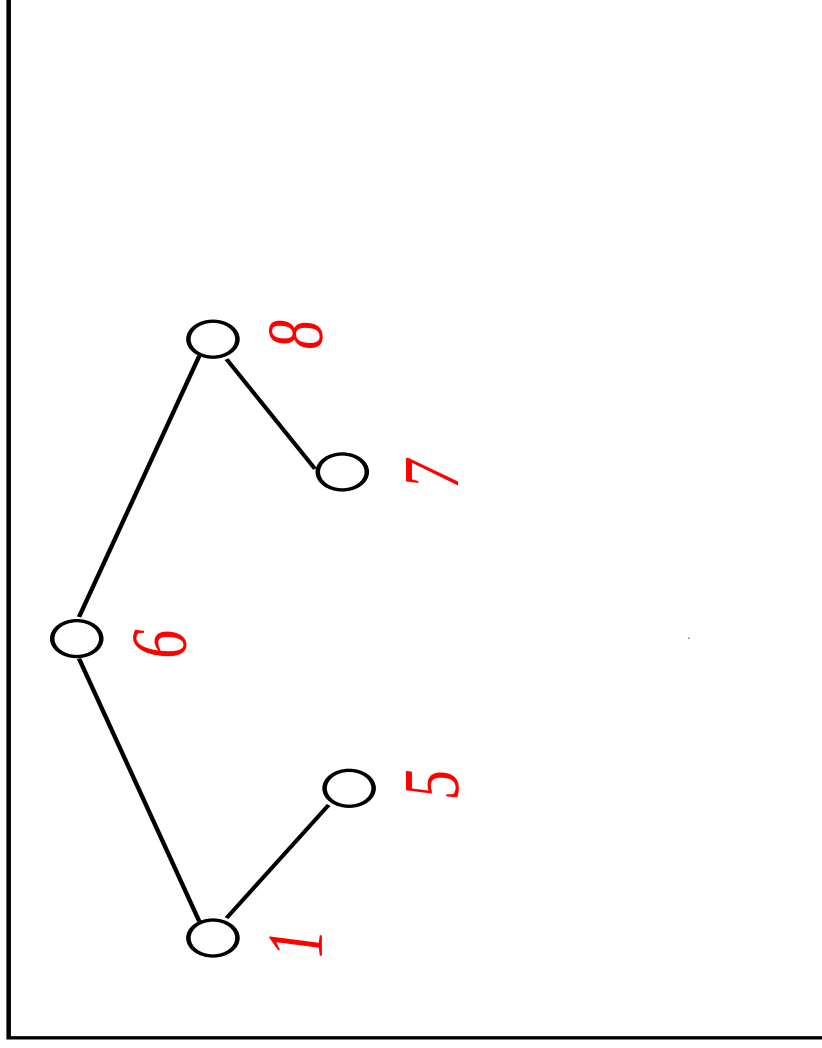
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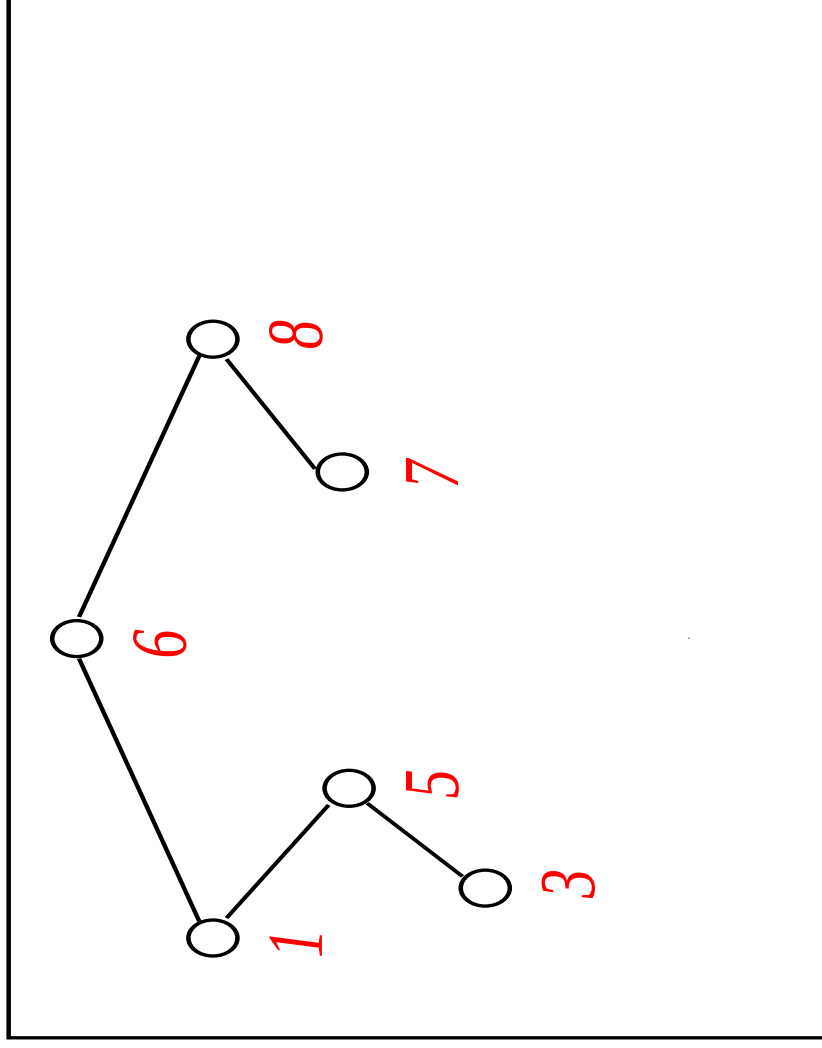
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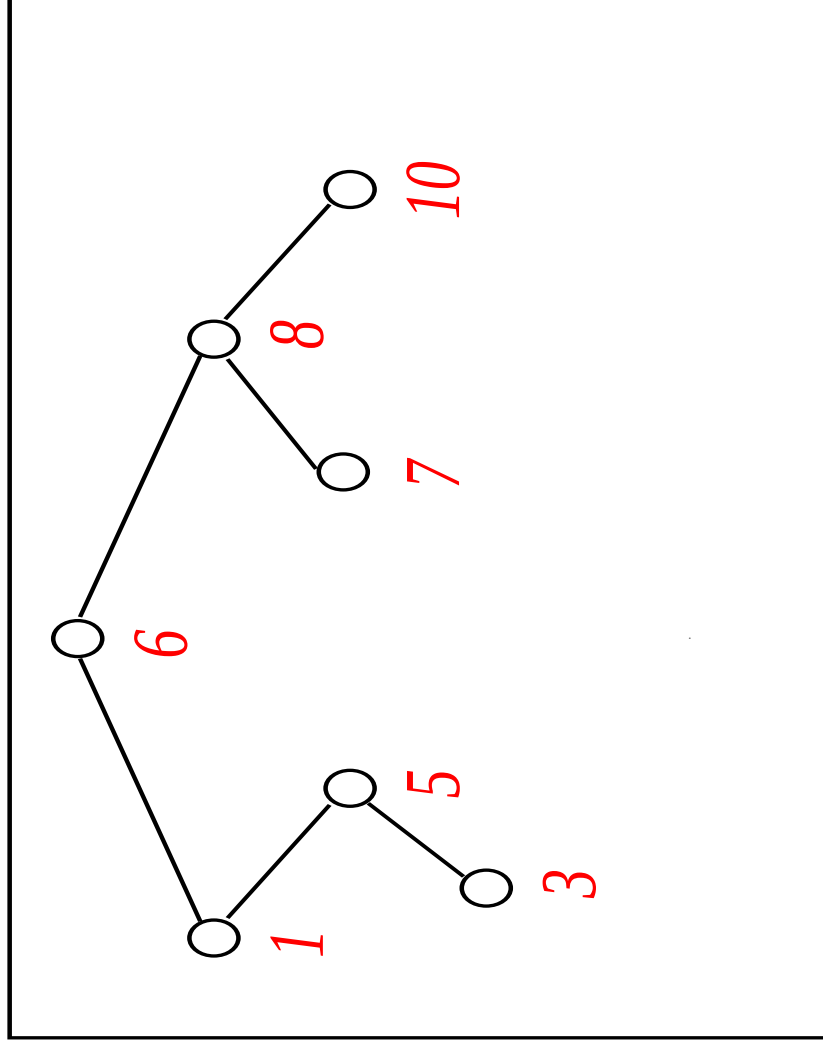
Binary search tree

Given numbers: 6, 1, 8, 7, 5, 3, 10, 2, 11, 4, 9.



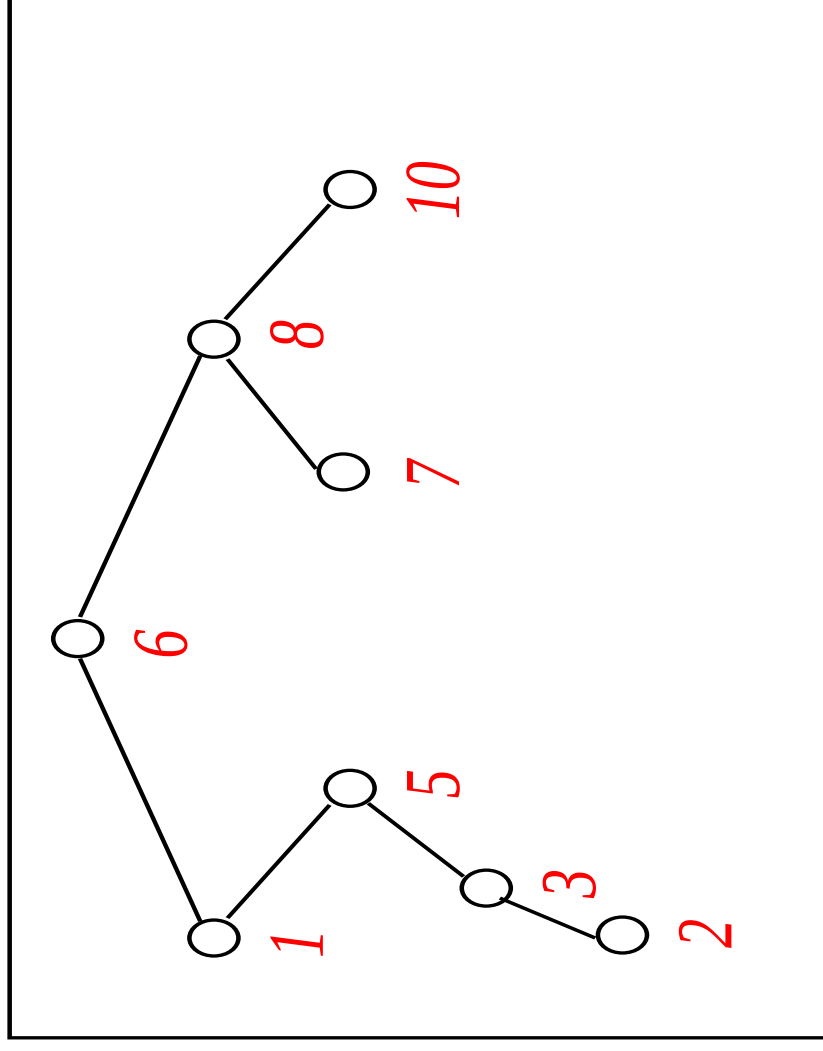
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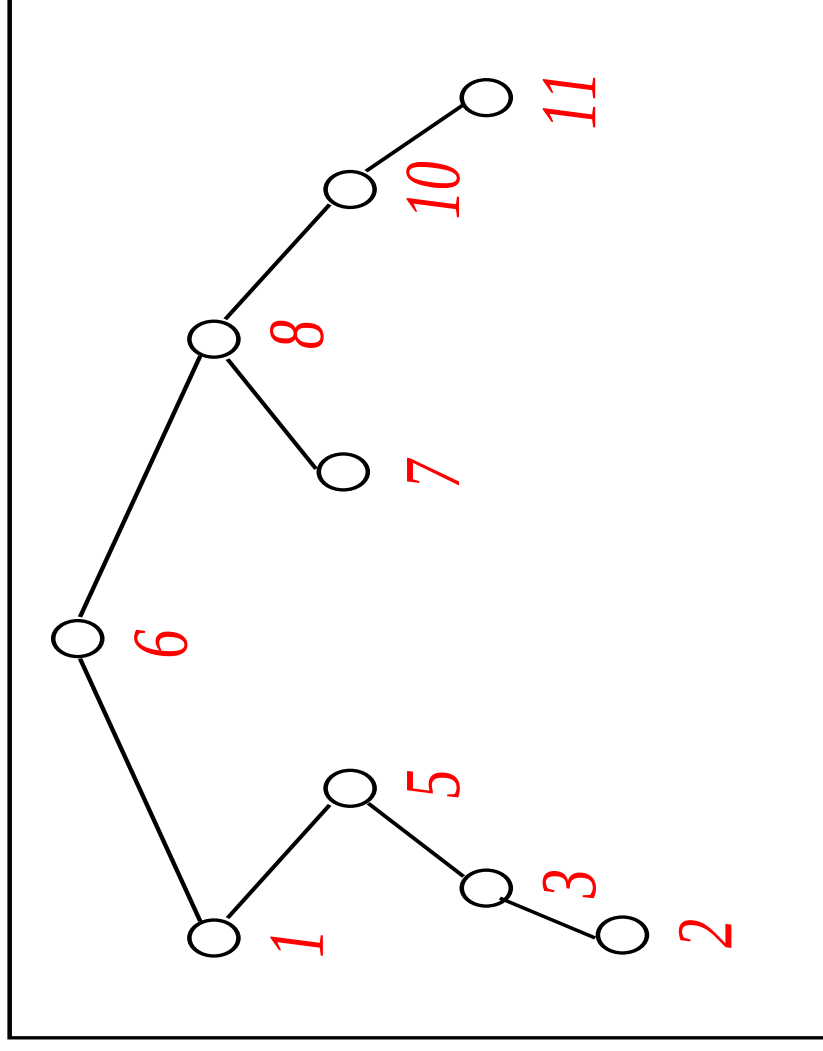
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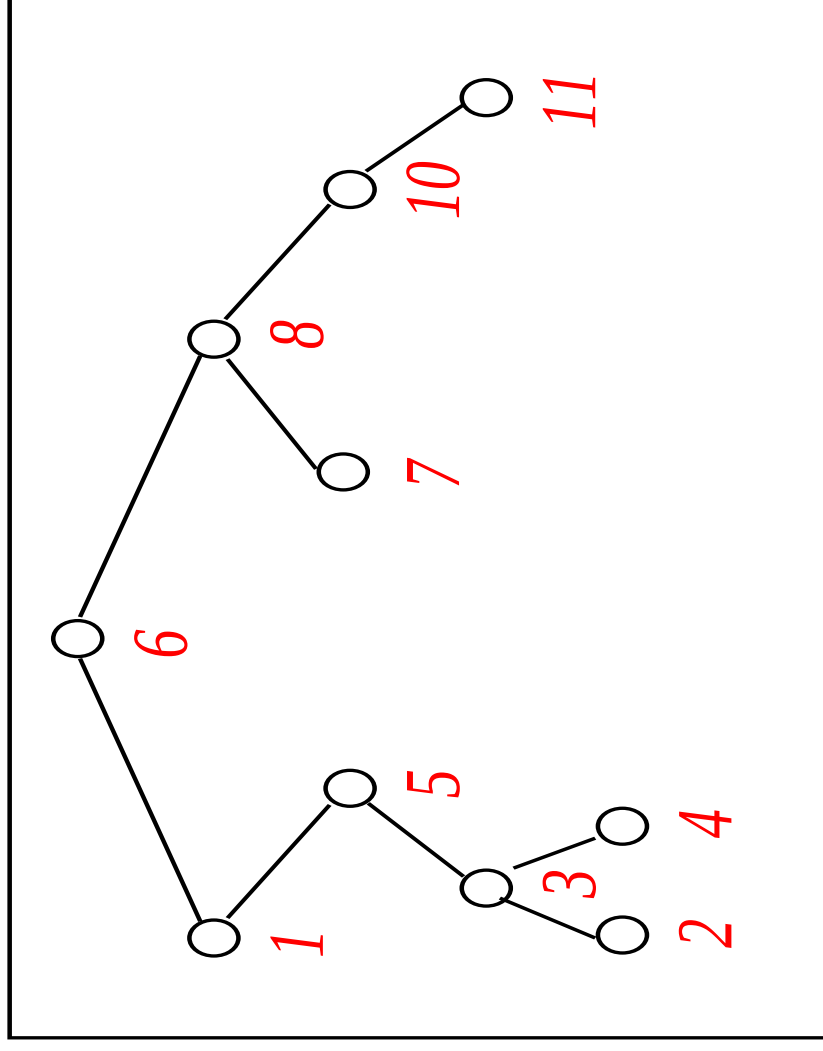
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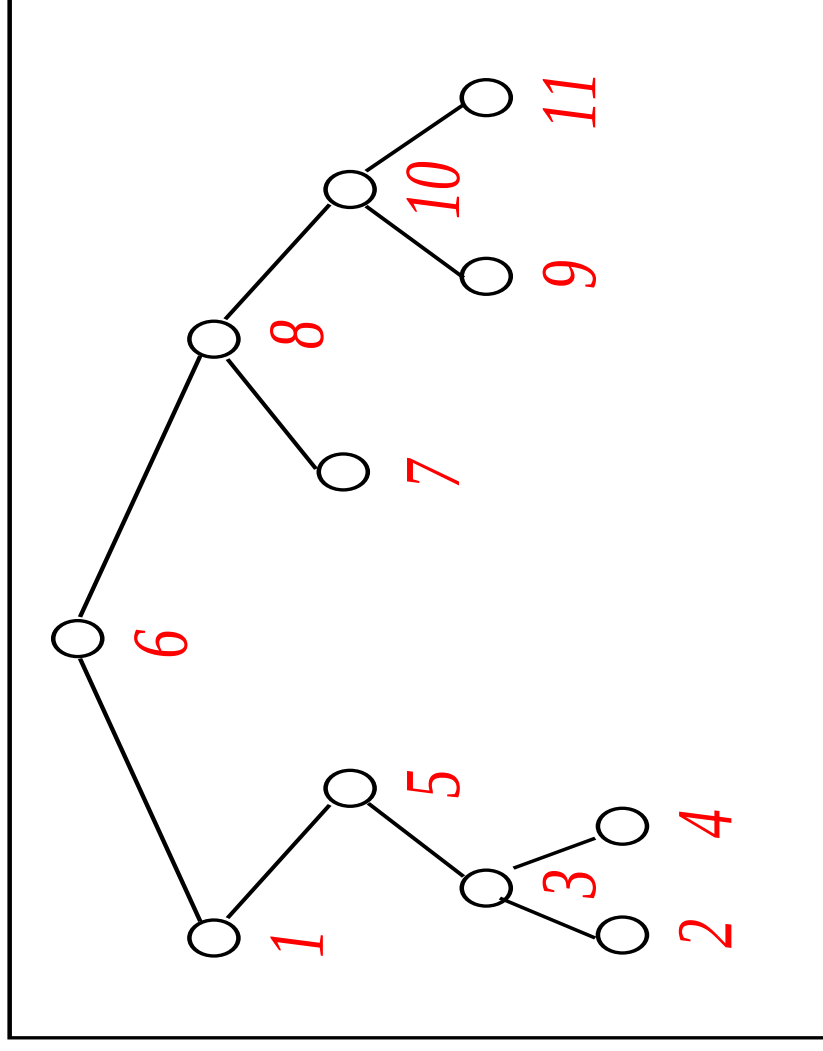
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Given numbers: 6, 1, 8, 7, 5, 3, 10, 2, 11, 4, 9.

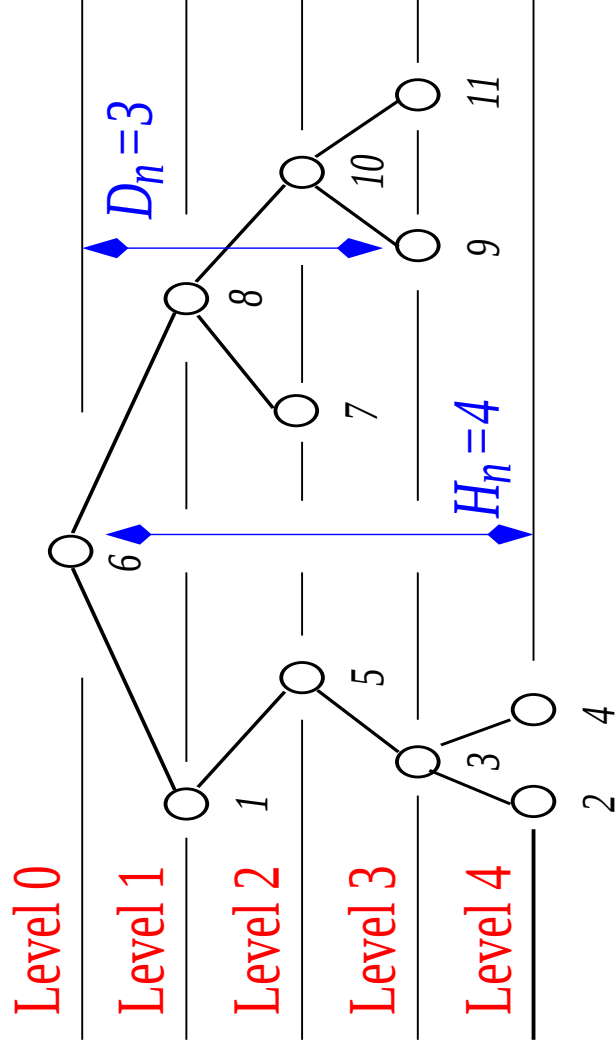


Binary search tree

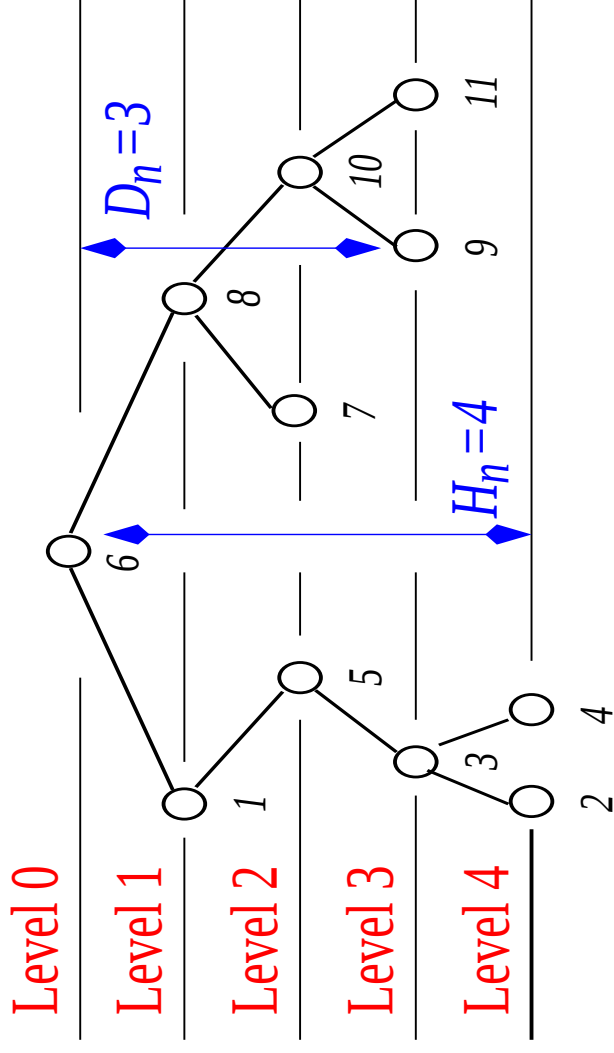
Given numbers: 6, 1, 8, 7, 5, 3, 10, 2, 11, 4, 9.



Quantities in BST

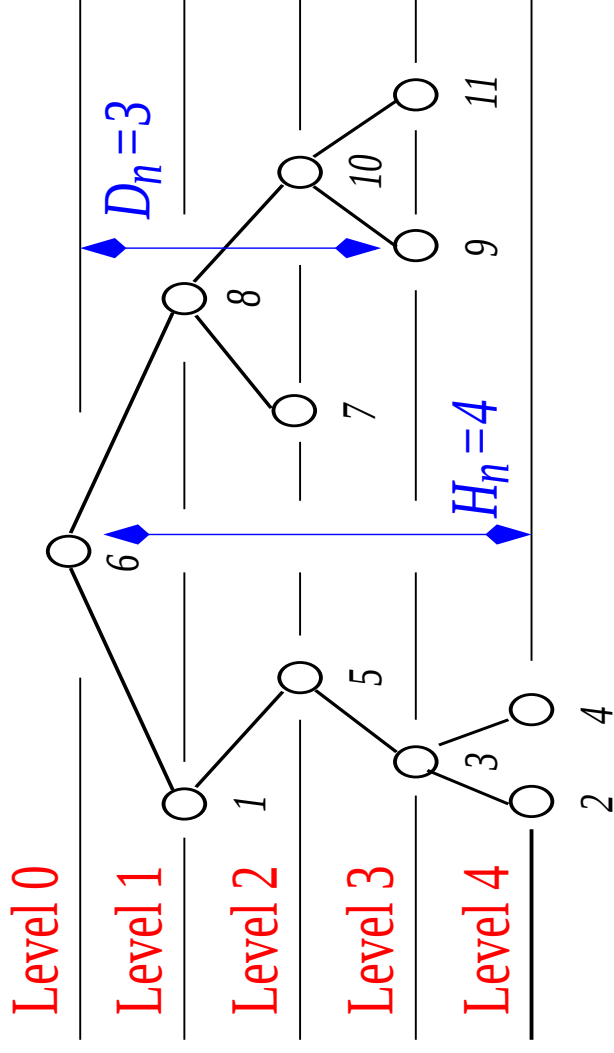


Quantities in BST



D_n — depth = distance root to n -th inserted node

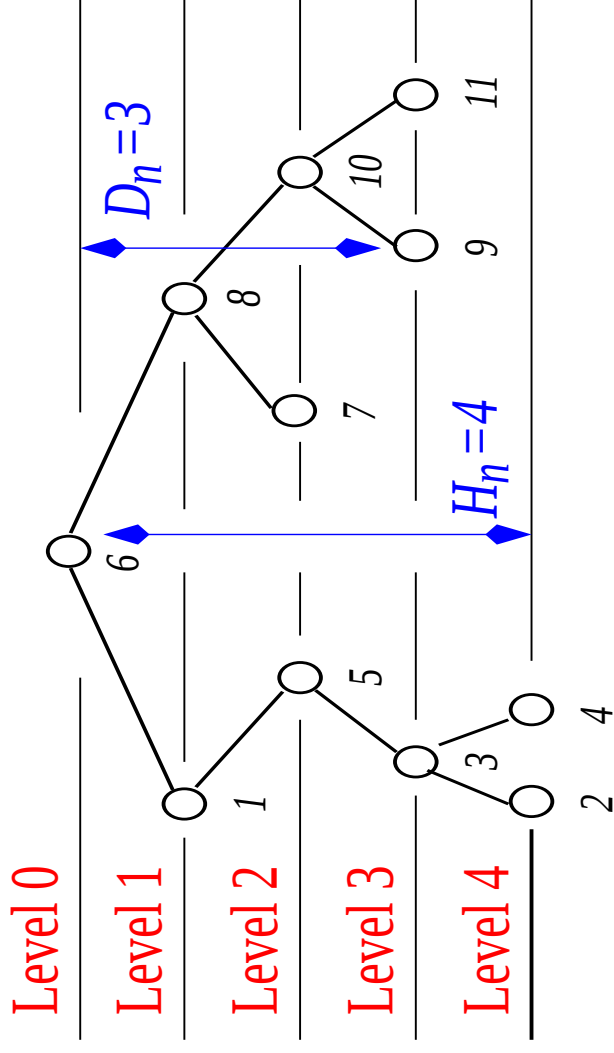
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Quantities in BST



D_n — depth = distance root to n -th inserted node

$H_n = \max_{1 \leq j \leq n} D_j$ — height

$Q_n = \sum_{1 \leq j \leq n} D_j$ — internal path length

Random binary search tree

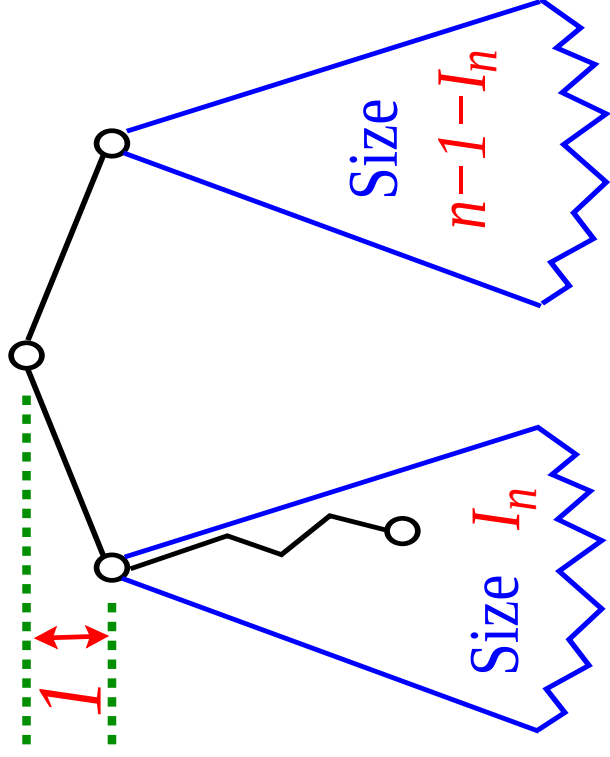
Model of randomness:

All permutations of $1, \dots, n$ equally likely.

Equivalent: Use $U_1, \dots, U_n$ i.i.d. $\text{unif}[0, 1]$.

Internal path length

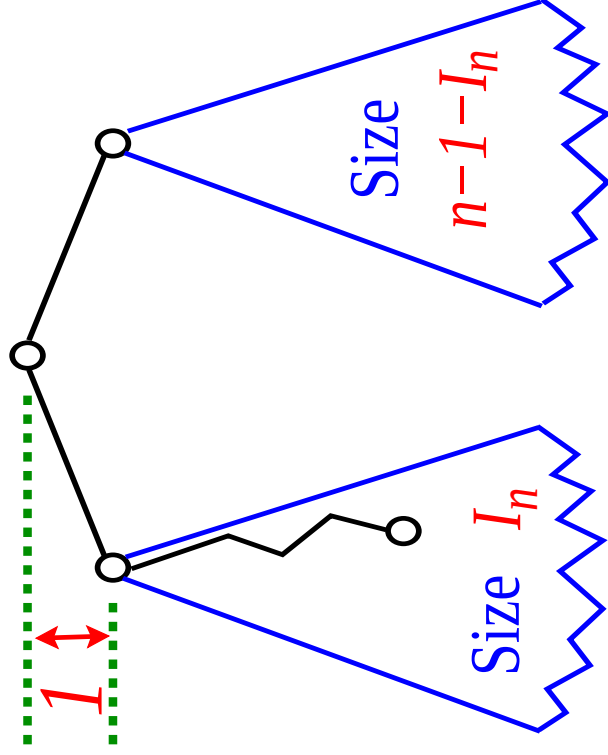
Internal path length Q_n :



Internal path length

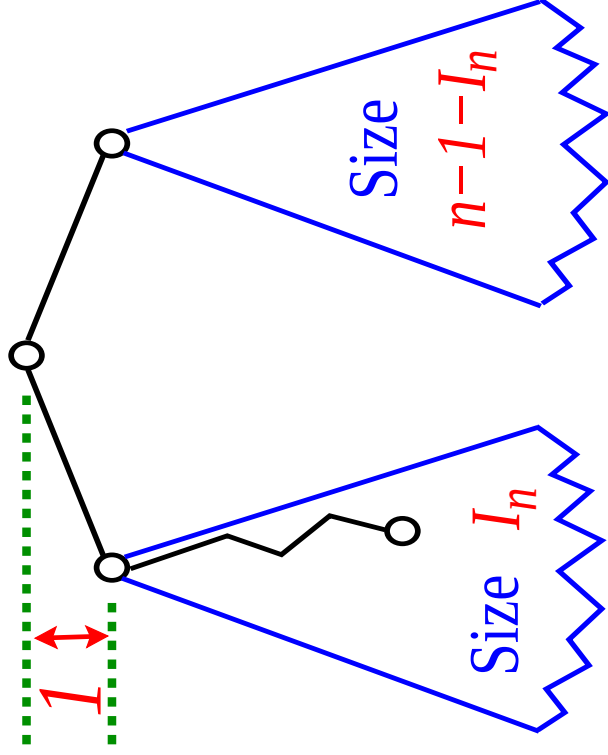
Internal path length Q_n :

$$Q_n \stackrel{d}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1$$



.

Internal path length



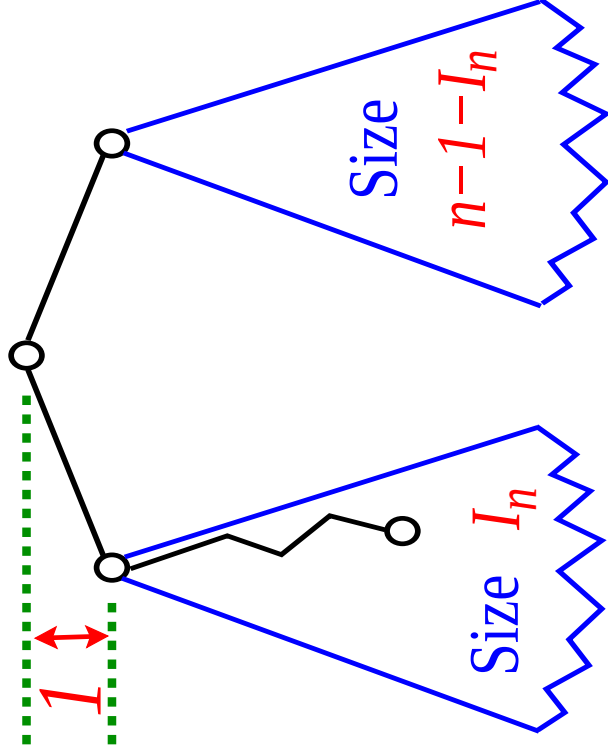
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$Q_0^*, \dots, Q_{n-1}^*, Q_0^{**}, \dots, Q_{n-1}^{**}, I_n$
indep.,

.

Internal path length



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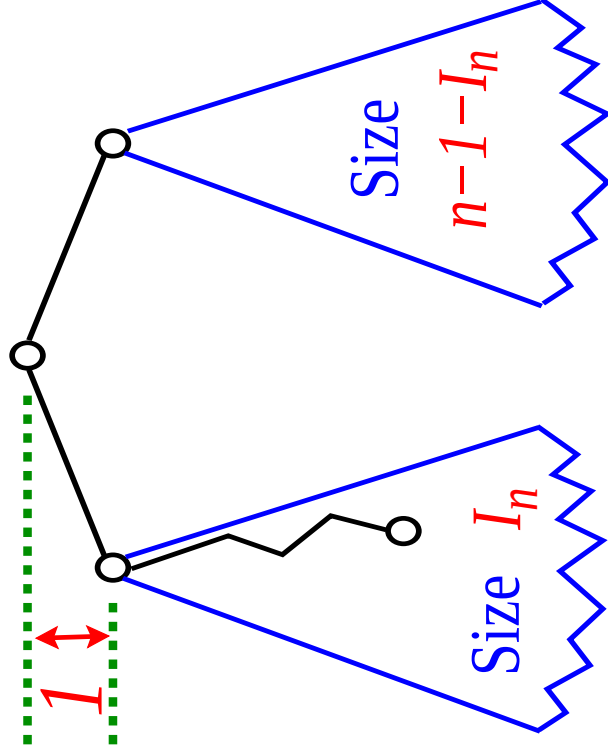
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.

Internal path length



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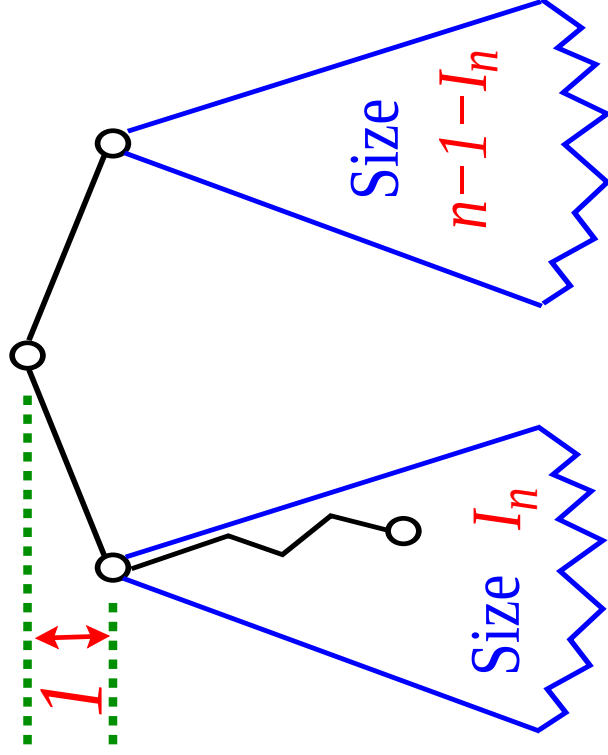
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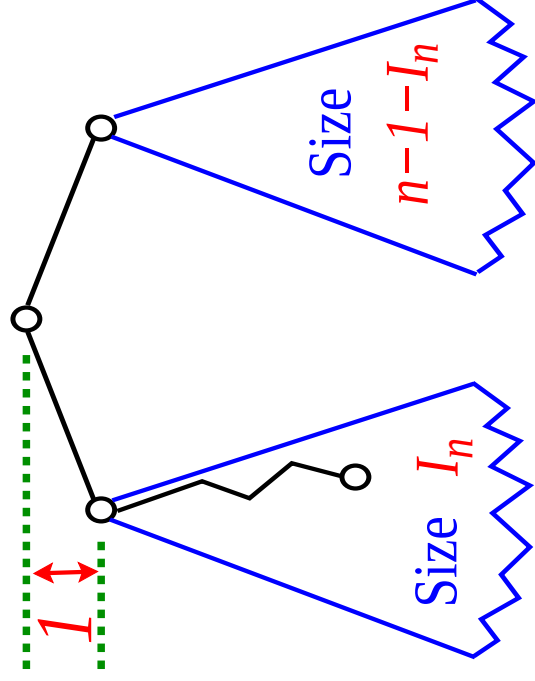
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This needs a proof!

Internal path length Q_n



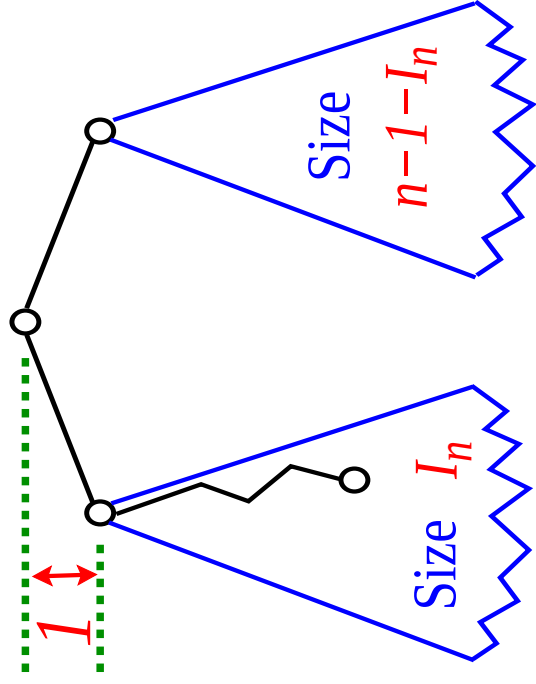
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$$Q_0^*, \dots, Q_{n-1}^*, Q_0^{**}, \dots, Q_{n-1}^{**}, \text{indep.},$$

$$\begin{array}{c} Q^j \\ \parallel \\ P \end{array} \quad \begin{array}{c} Q^{j,*} \\ \parallel \\ P \end{array} \quad \begin{array}{c} Q^{j,*} \end{array}$$

$$I_n \stackrel{d}{=} \text{unif}\{0, \dots, n-1\}.$$

Internal path length Q_n



$$Q_n \stackrel{d}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1 =: Z_n,$$

$$Q_0^*, \dots, Q_{n-1}^*, Q_0^{**}, \dots, Q_{n-1}^{**}, I_n$$

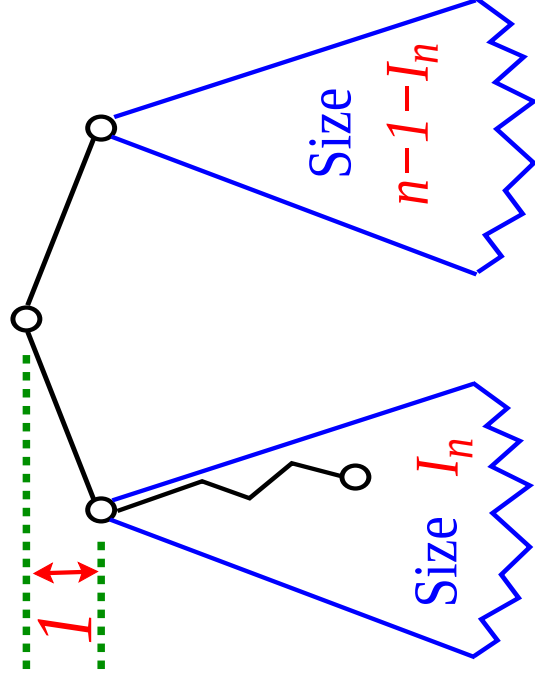
indep.,

$$Q_j^* \stackrel{d}{=} Q_j^{**} \stackrel{d}{=} Q_j,$$

$$I_n \stackrel{d}{=} \text{unif}\{0, \dots, n-1\}.$$

Sufficient: $\mathbb{P}(Q_n = j) = \mathbb{P}(Z_n = j)$ for all $j \in \mathbb{N}$.

Internal path length Q_n



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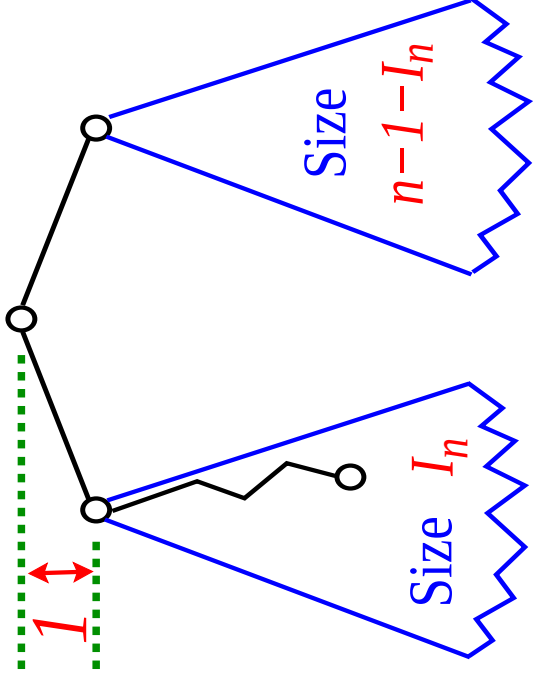
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Show: For all $j \in \mathbb{N}$, $k \in \{0, \dots, n-1\}$:

$$\mathbb{P}(Q_n = j \mid I_n = k) = \mathbb{P}(Z_n = j \mid I_n = k).$$

Internal path length Q_n



$$Q_n^d \equiv Q_{I_n}^* + Q_{n-1-I_n}^{*} + n - 1 \equiv: Z_n,$$

$$Q_0^*, \dots, Q_{n-1}^*, Q_0^*, \dots, Q_{n-1}^*, \text{In} \\ \text{indep.},$$

$$\begin{array}{c} Q^j_i \\ \parallel \\ Q^{*j}_i \\ \parallel \\ Q^{*j}_i \end{array}$$

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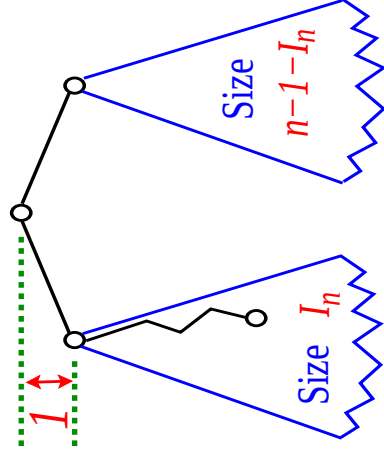
Show: For all $j \in \mathbb{N}$, $k \in \{0, \dots, n-1\}$:

$$\mathbb{P}(Q_n = j | I_n = k) = \mathbb{P}(Z_n = j | I_n = k).$$

[Total probability theorem yields:

$$\begin{aligned}\mathbb{P}(Q_n = j) &= \sum_k \mathbb{P}(I_n = k) \mathbb{P}(Q_n = j \mid I_n = k) \\ &= \sum_k \mathbb{P}(I_n = k) \mathbb{P}(Z_n = j \mid I_n = k) = \mathbb{P}(Z_n = j).\end{aligned}$$

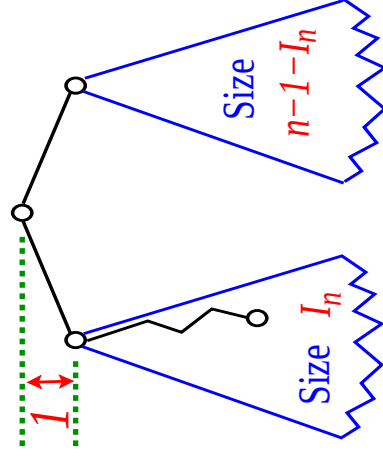
Proof of the recurrence



To prove:

$$\begin{aligned} & \mathbb{P}(Q_n = j \mid I_n = k) \\ &= \mathbb{P}(Q_k^* + Q_{n-1-k}^{* *} + n - 1 = j). \end{aligned}$$

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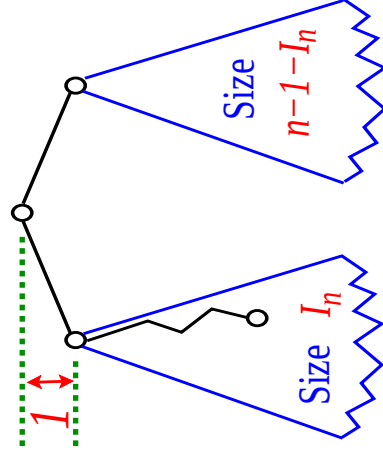
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We show: Given $\{I_n = k\}$ we have:

- a) $\mathcal{T}_1$ and $\mathcal{T}_2$ independent, b) $\mathcal{T}_1$ and $\mathcal{T}_2$ random BSB.

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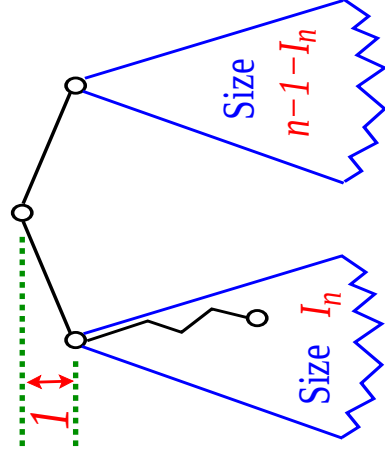
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$\pi = (5, 7, 3, 2, 6, 8, 1, 4)$ π equiprobable permutation in S_n .

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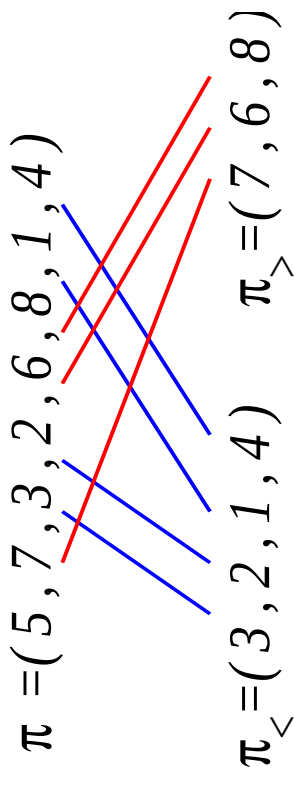
$\pi = (5, 7, 3, 2, 6, 8, 1, 4)$

$\pi_{<} = (3, 2, 1, 4)$ $\pi_{>} = (7, 6, 8)$

Construct $\pi_{<}$ and $\pi_{>}$

Proof of the recurrence II

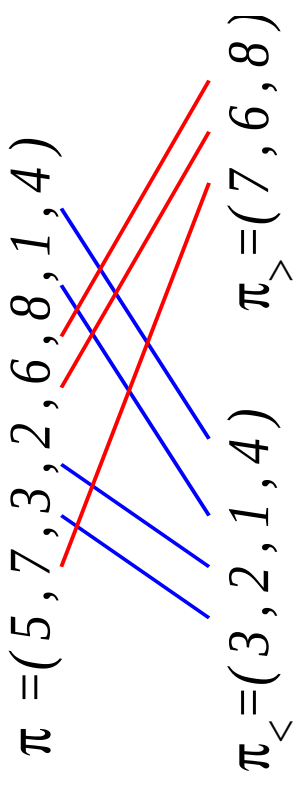
π equiprobable in S_n .



Proof of the recurrence II

π equiprobable in S_n .

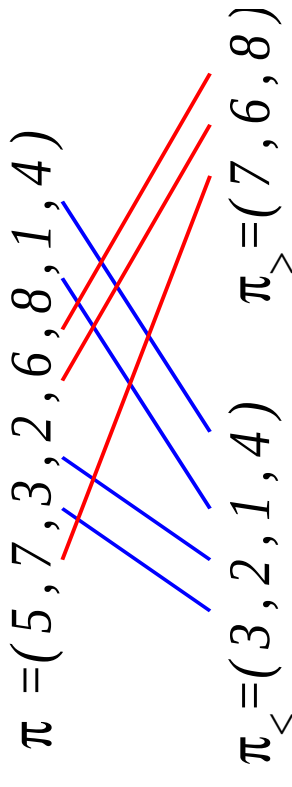
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$\pi_{<}$ and $\pi_{>}$ independent and equiprobable on S_k and S_{n-1-k} resp.

Proof of the recurrence II

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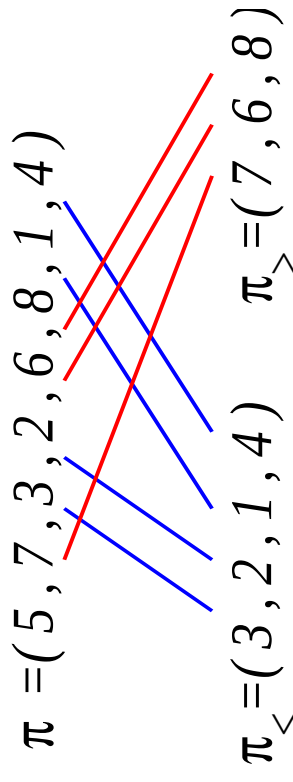
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For arbitrary $\sigma \in S_k$:

$$\mathbb{P}(\pi_{<} = \sigma \mid \pi_1 = k + 1) = \frac{1}{1/n} \frac{\binom{n-1}{k} (n-1-k)!}{n!} = \frac{1}{k!}$$

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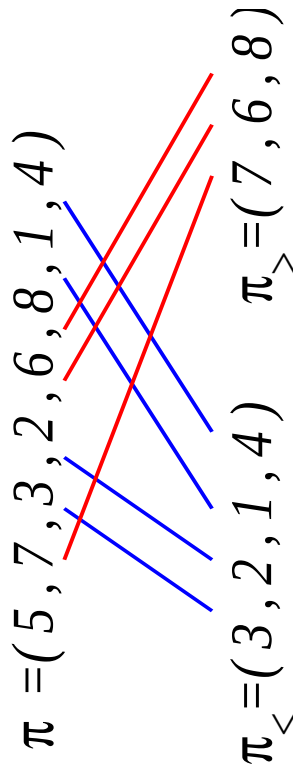
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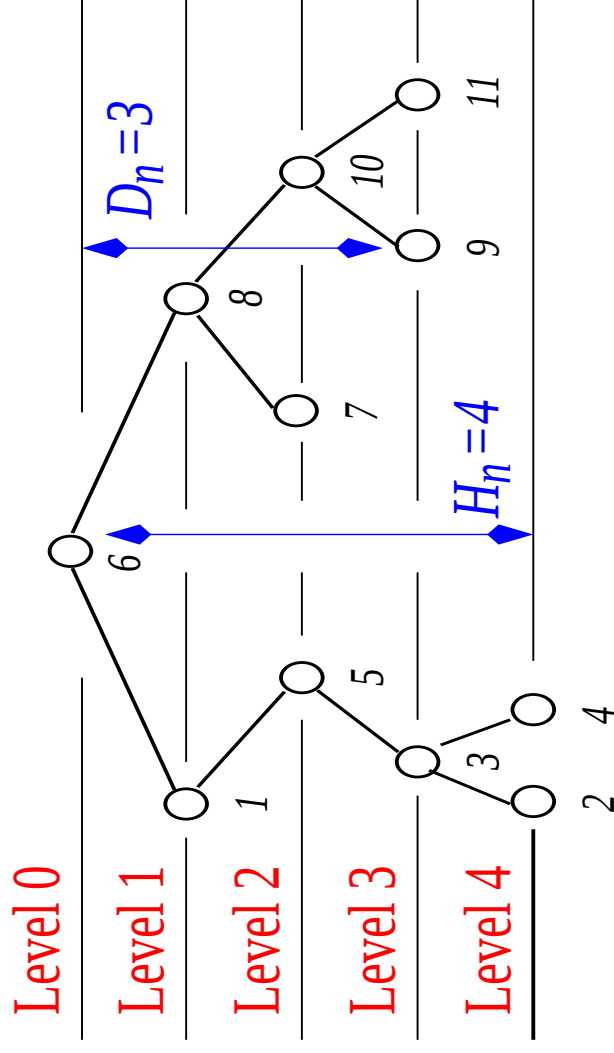
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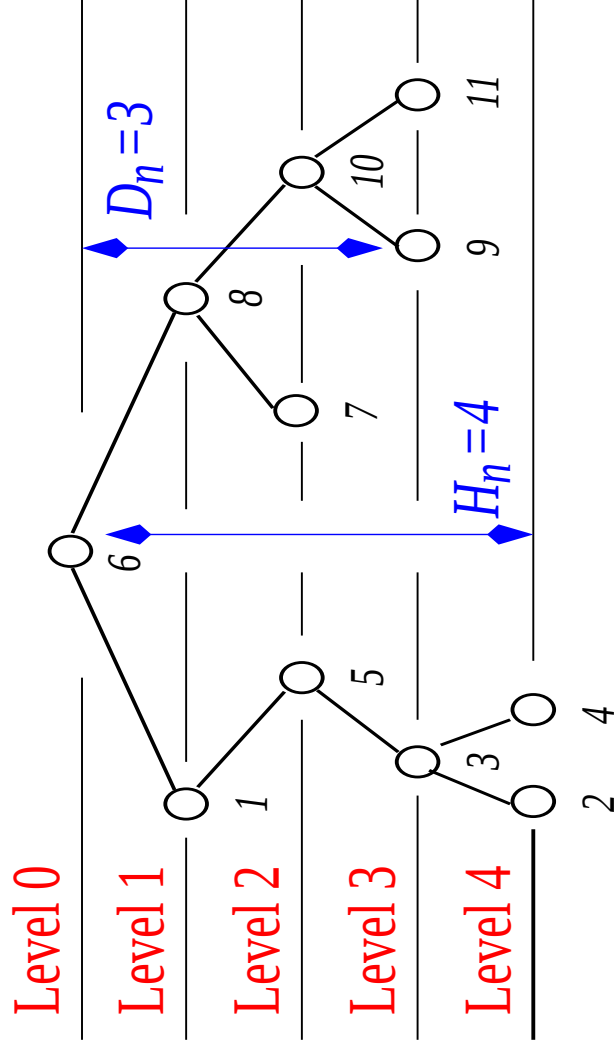
Second assertion similar.

Other BST recurrences



$$Q_n \stackrel{d}{=} Q_{I_n}^{(1)} + Q_{n-1-I_n}^{(2)} + n - 1$$

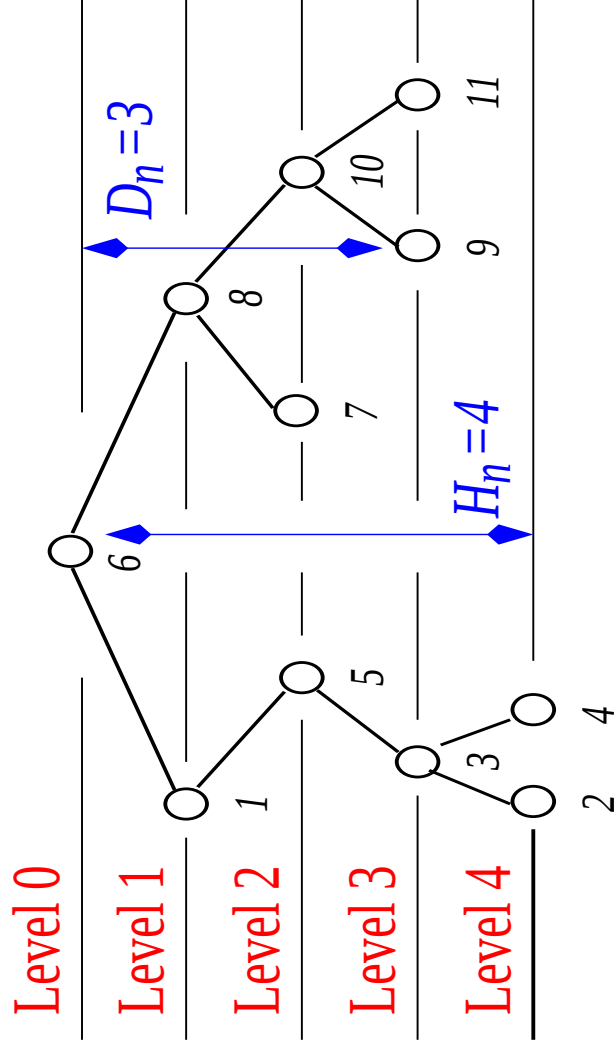
Other BST recurrences



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Other BST recurrences



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$$\mathbf{D}_n \stackrel{\text{d}}{=} \mathbf{1}_{A_n} \mathbf{D}_{I_n} + \mathbf{1}_{A_n^c} \mathbf{D}_{n-1-I_n} + \mathbf{1}$$

Expected internal path length

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$$Q_n \stackrel{d}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1, \quad I_n \stackrel{d}{=} \text{unif}\{0, \dots, n-1\}.$$

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$$q_n := \mathbb{E} Q_n = \mathbb{E} \left[Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1 \right]$$

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Expected internal path length

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Solves easily:

$$q_n = 2(n+1)\mathcal{H}_n - 4n = 2n \log(n) + (2\gamma - 4)n + o(n).$$

$$[\mathcal{H}_n := \sum_{i=1}^n 1/i \text{ harmonic numbers.}]$$

Rescaling

$$Q_n \stackrel{d}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1, \quad q_n = 2n \log(n) + cn + o(n).$$

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Scaling

$$Y_n := \frac{Q_n - q_n}{n}.$$

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Then

$$Y_n \stackrel{d}{=} \frac{1}{n} \left(Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1 - q_n \right)$$

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Rescaling

$$Q_n \stackrel{d}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1, \quad q_n = 2n \log(n) + cn + o(n).$$

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$$\begin{aligned} Y_n &\stackrel{d}{=} \frac{1}{n} \left(Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1 - q_n \right) \\ &= \frac{1}{n} \left(I_n \frac{Q_{I_n}^* \pm q_{I_n}}{I_n} + (n - 1 - I_n) \frac{Q_{n-1-I_n}^{**} \pm q_{n-1-I_n}}{n - 1 - I_n} + n - 1 - q_n \right) \\ &= \underbrace{\frac{I_n}{n} Y_{I_n}^* + \frac{n - 1 - I_n}{n} Y_{n-1-I_n}^{**}}_{\text{}} + b^{(n)} \end{aligned}$$

Rescaling

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$$Y_n := \frac{Q_n - q_n}{n}.$$

Then

$$\begin{aligned} Y_n &\stackrel{d}{=} \frac{1}{n} \left(Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1 - q_n \right) \\ &= \frac{1}{n} \left(I_n \frac{Q_{I_n}^* \pm q_{I_n}}{I_n} + (n - 1 - I_n) \frac{Q_{n-1-I_n}^{**} \pm q_{n-1-I_n}}{n - 1 - I_n} + n - 1 - q_n \right) \\ &= \underbrace{\frac{I_n}{n} Y_{I_n}^* + \frac{n - 1 - I_n}{n} Y_{n-1-I_n}^{**}}_{Y_n} + b^{(n)} \end{aligned}$$

with

$$b^{(n)} = \frac{1}{n} \left(q_{I_n} + q_{n-1-I_n} - q_n + n - 1 \right).$$

Rescaling

$$Q_n \stackrel{\text{d}}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1, \quad q_n = 2n \log(n) + cn + o(n).$$

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$$q_n = 2n \log(n) + cn + o(n).$$

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$$= 2 \frac{I_n}{n} \log \left(\frac{I_n}{n} \right) + 2 \frac{n-1-I_n}{n} \log \left(\frac{n-1-I_n}{n} \right) + 1 + o(1)$$

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 &\rightarrow 2u \log(u) + 2(1-u) \log(1-u) + 1 =: g(u)
 \end{aligned}$$

Rescaling: Summary

$$Q_n \stackrel{\text{d}}{=} Q_{I_n}^* + Q_{n-1-I_n}^{**} + n - 1, \quad q_n = 2n \log(n) + cn + o(n).$$

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Hence, this suggests

$$Y_n \rightarrow Y$$

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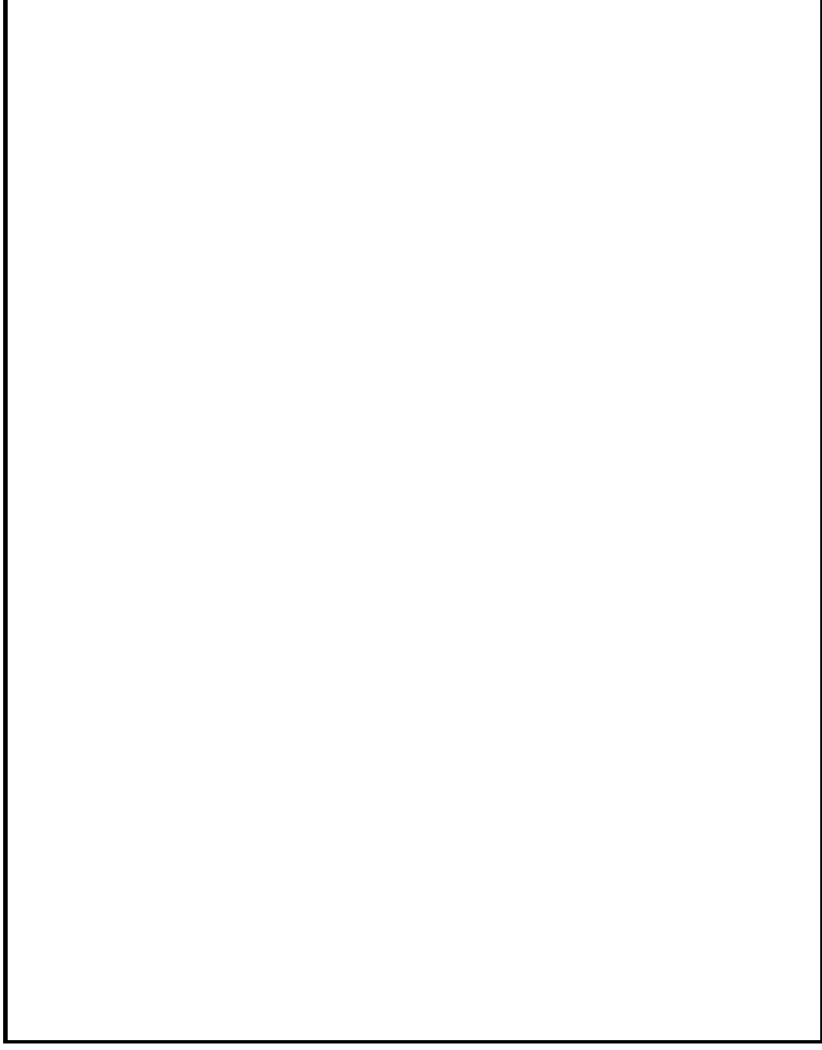
$$Y_n \rightarrow Y \stackrel{d}{=} uY^* + (1-u)Y^{**} + g(u),$$

with Y^*, Y^{**}, u independent, $Y \stackrel{d}{=} Y^* \stackrel{d}{=} Y^{**}$.

m-ary search trees

Example: $m = 4$

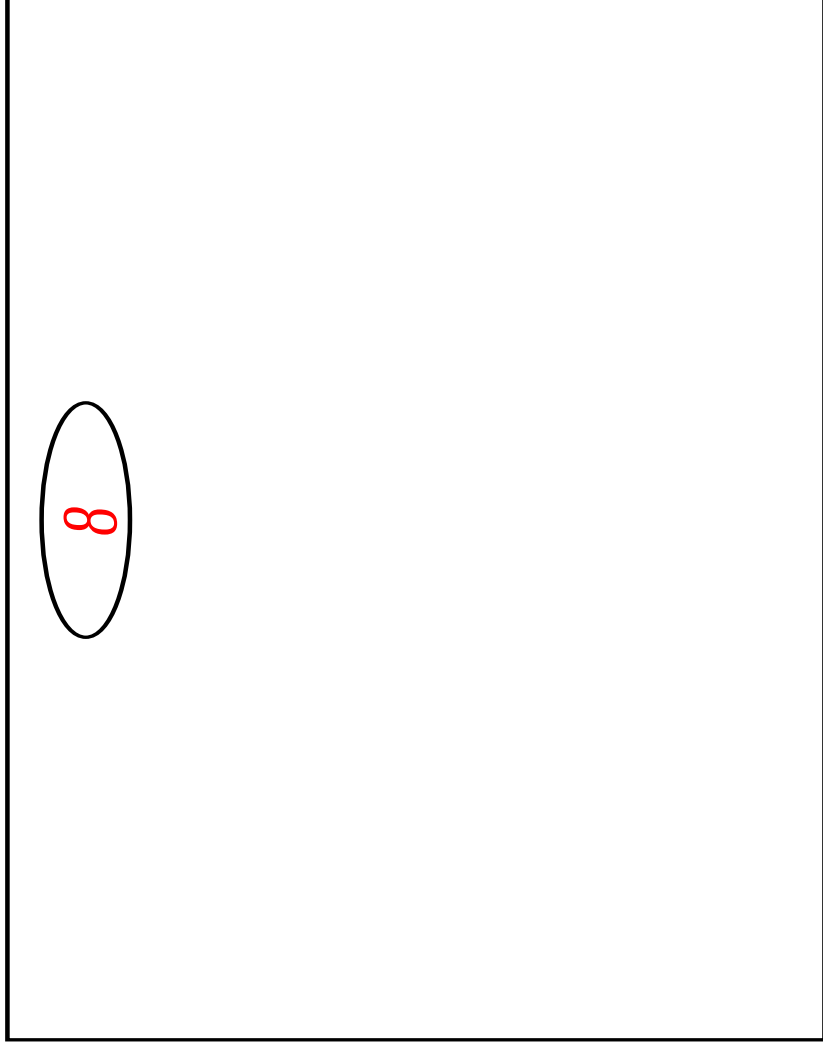
List of data: 8, 3, 9, 6, 2, 1, 11, 7, 10, 4, 5.



m-ary search trees

Example: $m = 4$

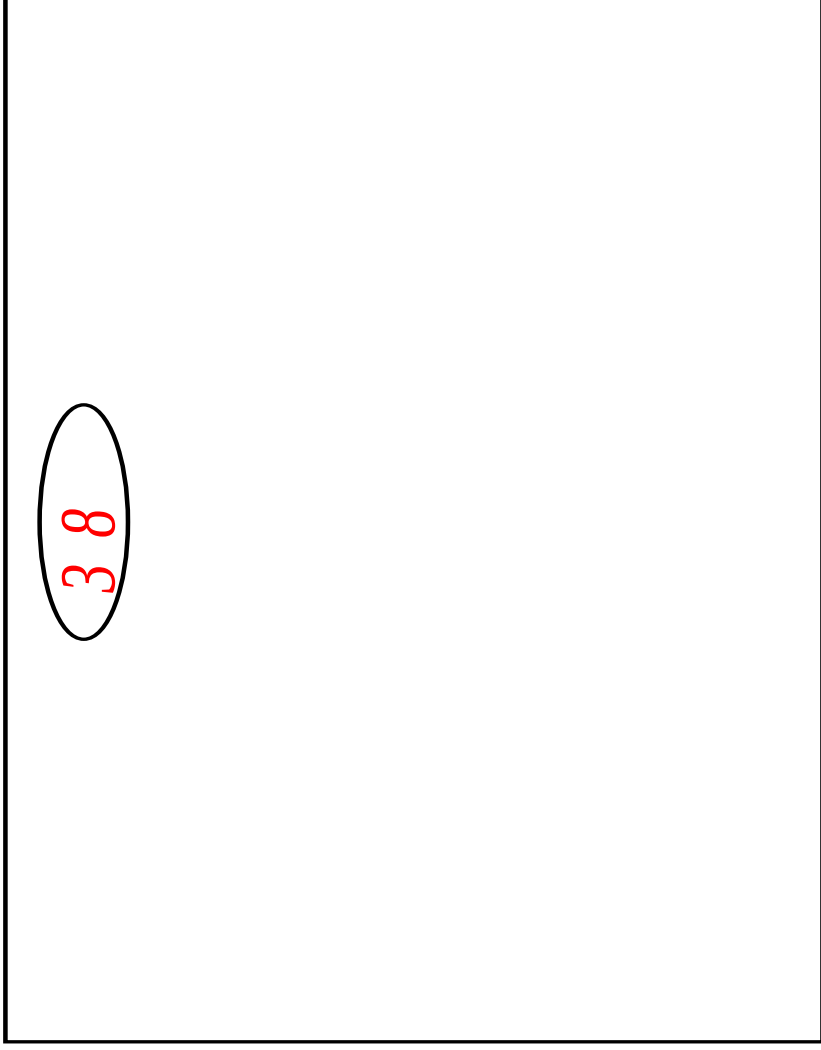
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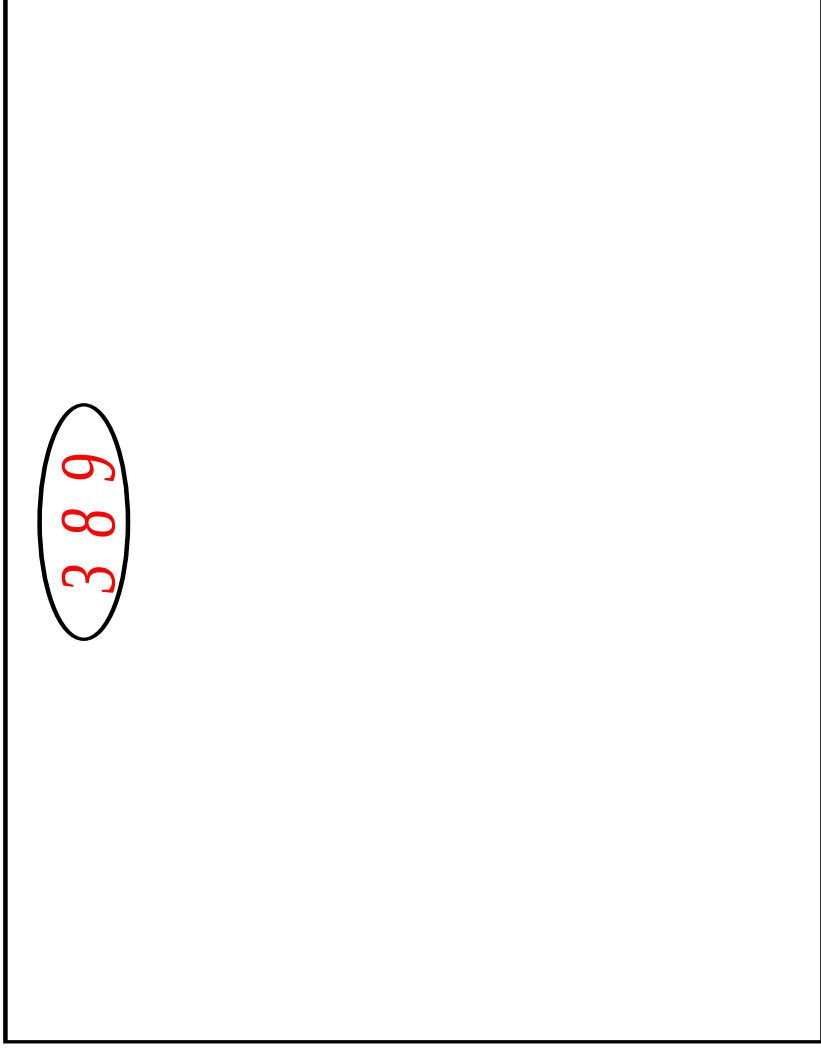
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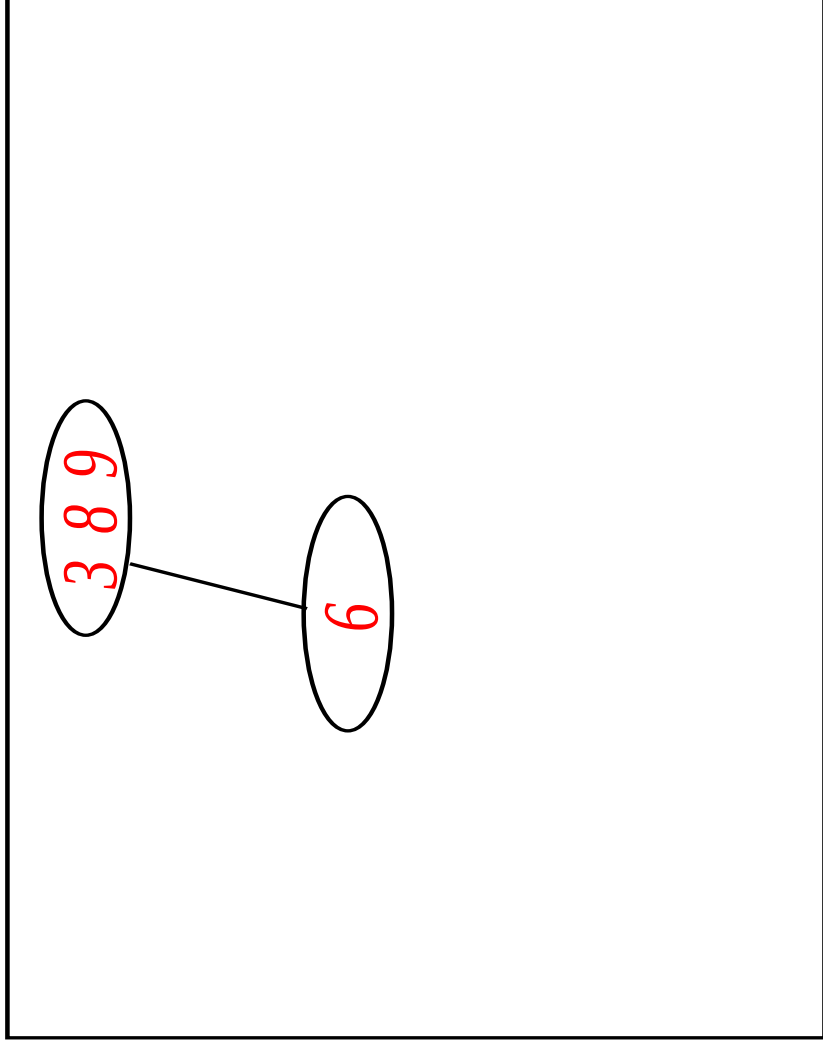
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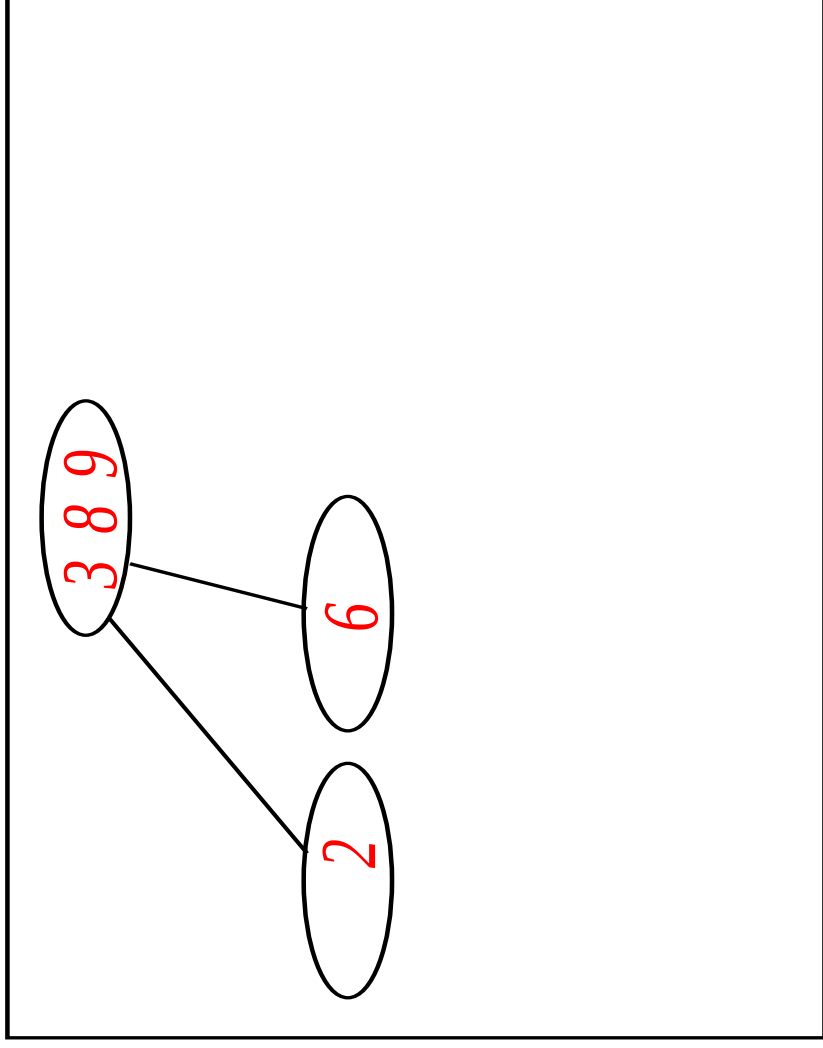
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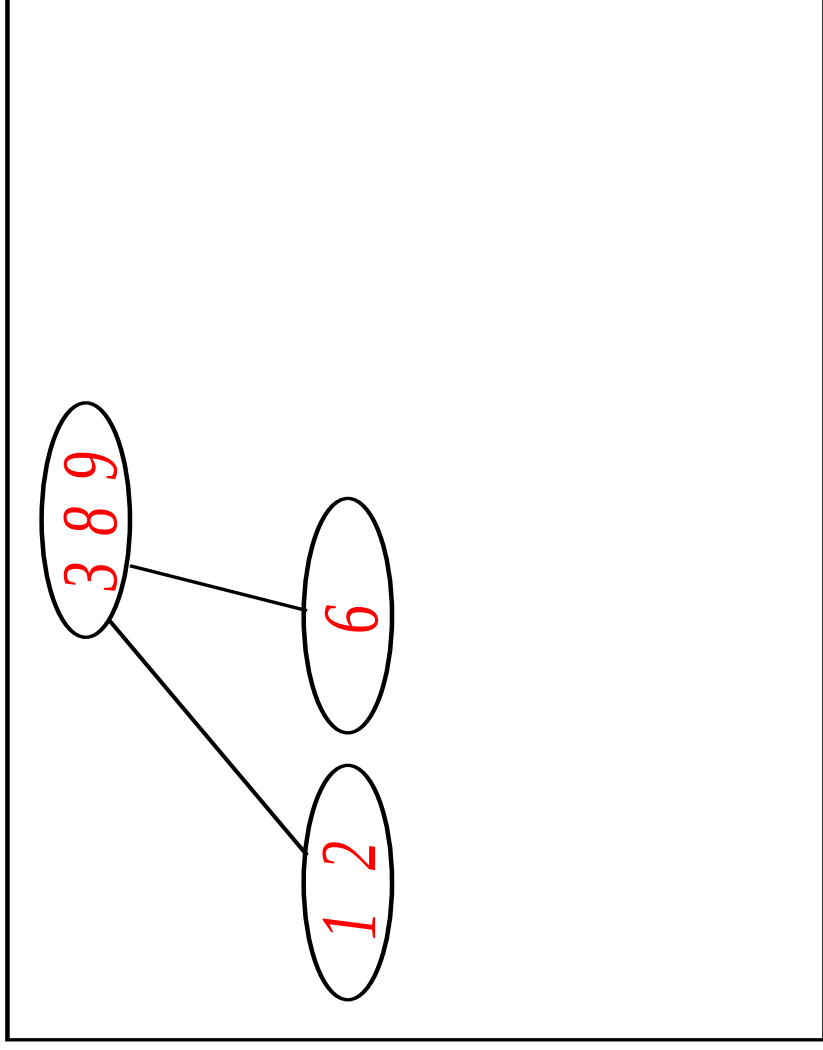
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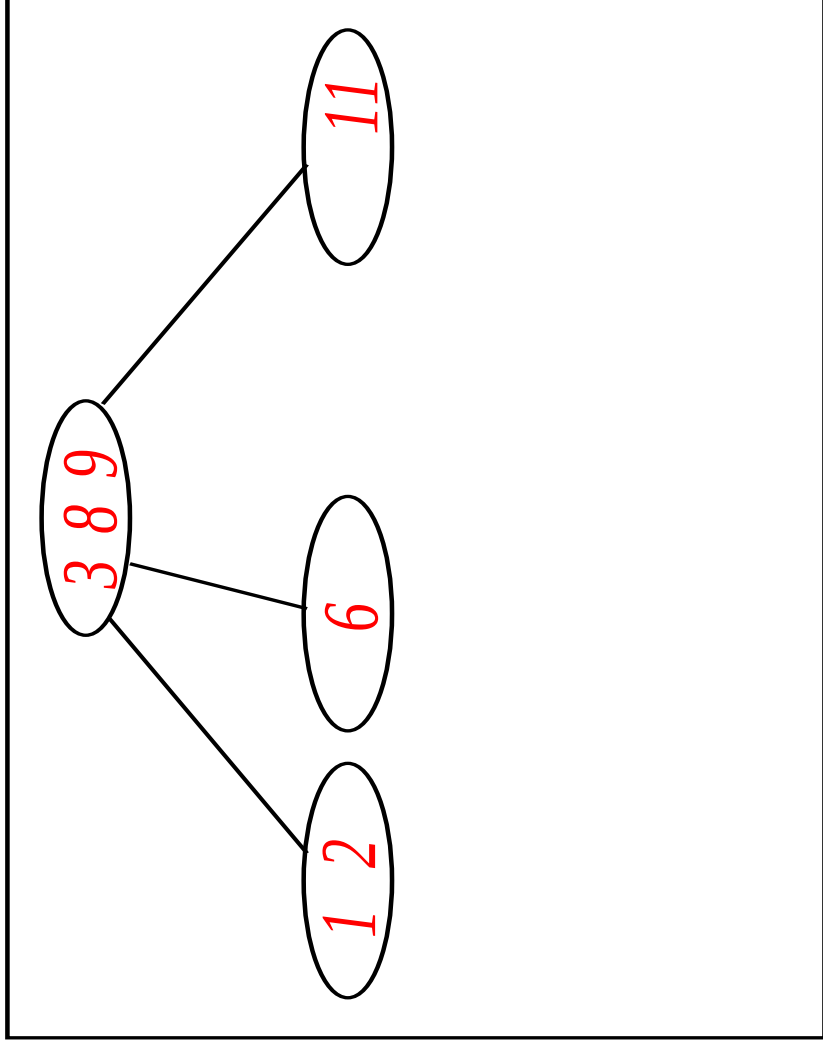
List of data: 8, 3, 9, 6, 2, 1, 11, 7, 10, 4, 5.



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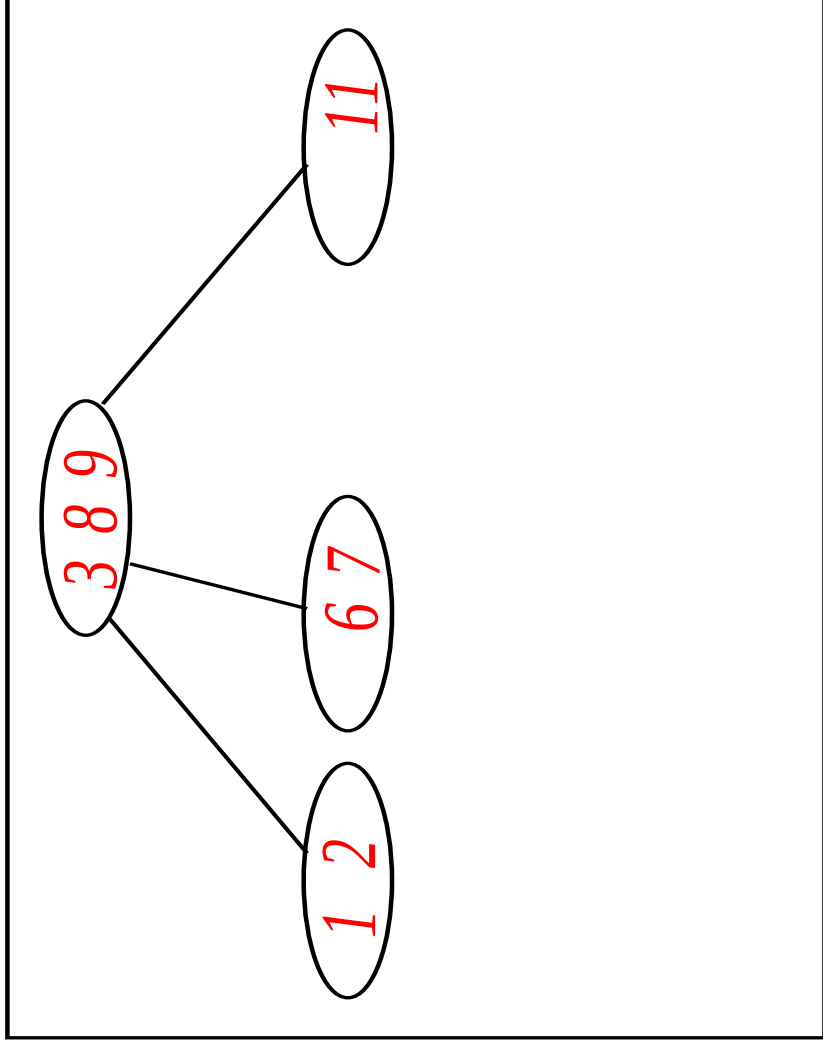
List of data: 8, 3, 9, 6, 2, 1, 11, 7, 10, 4, 5.



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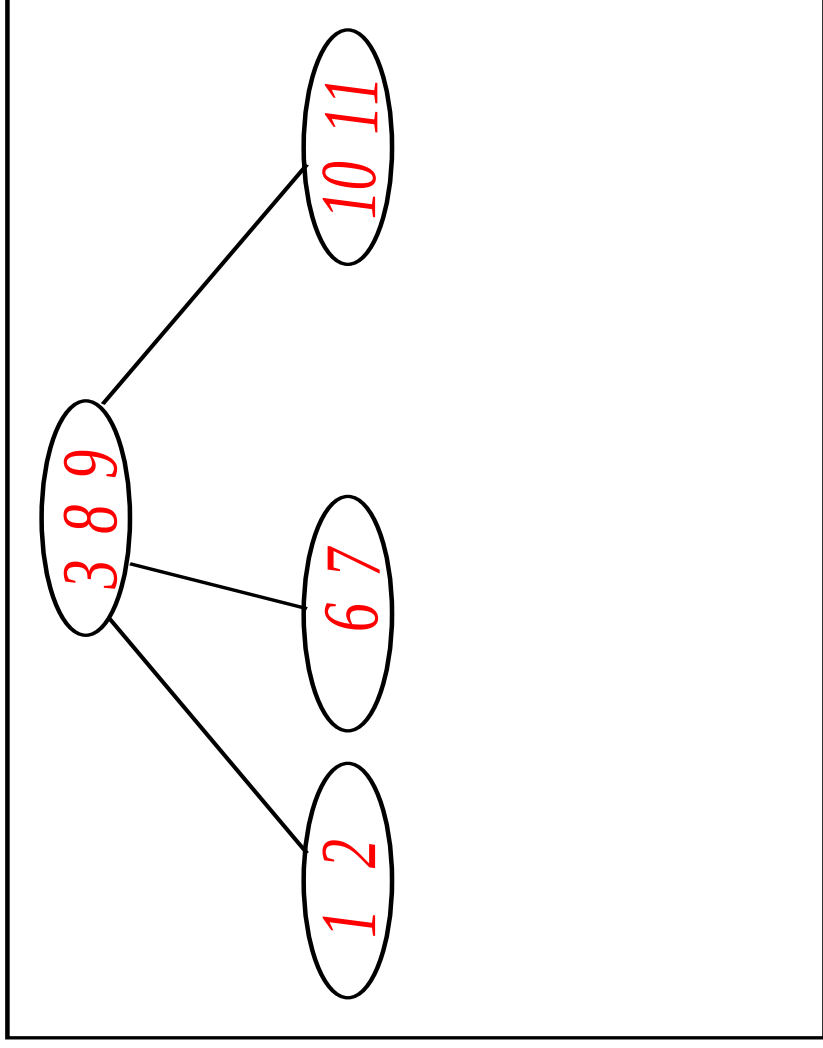
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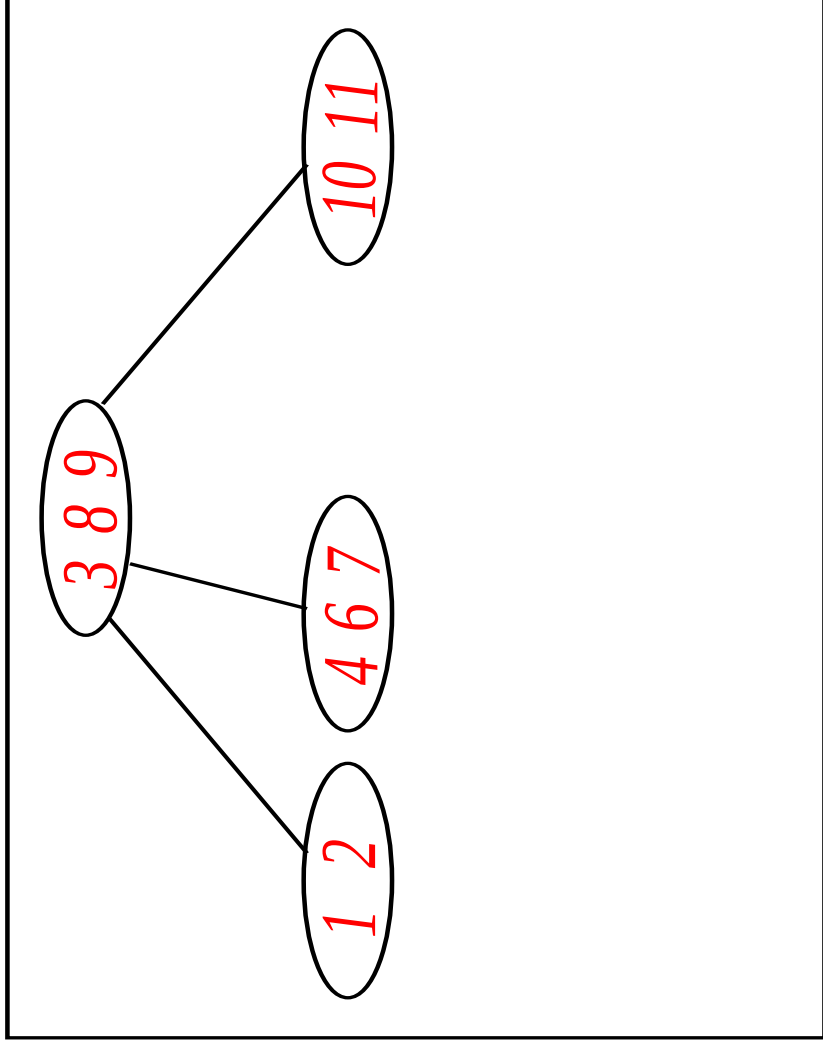
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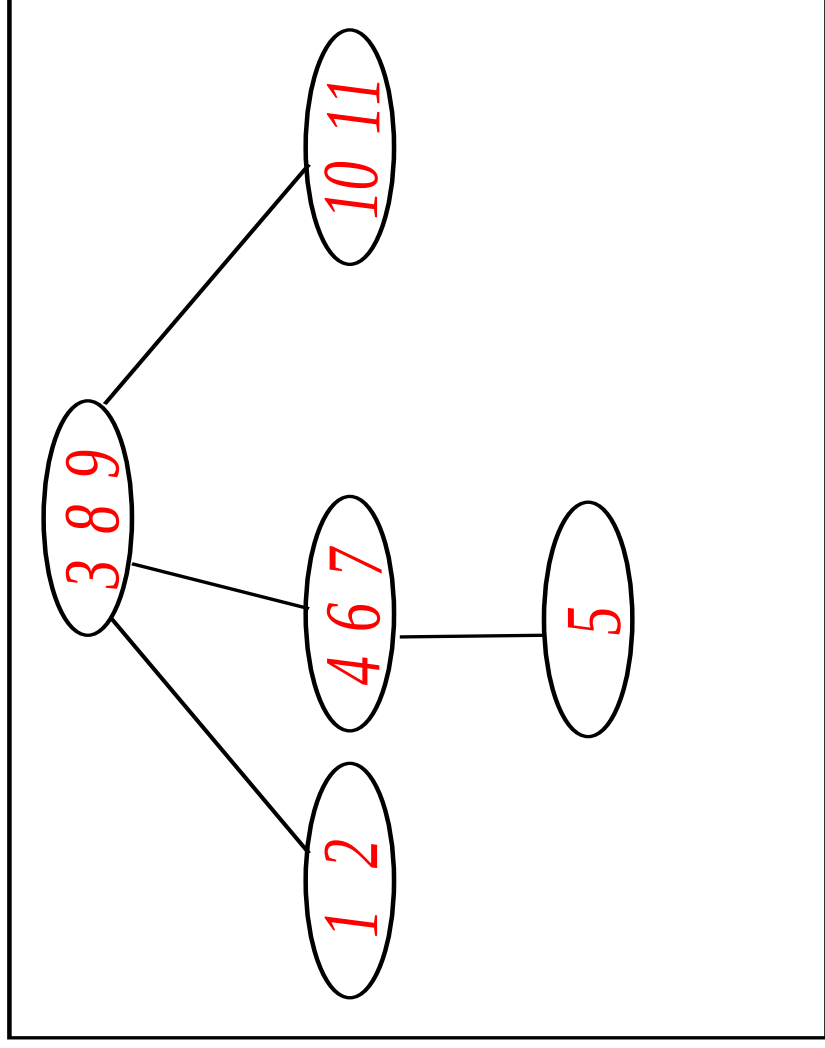
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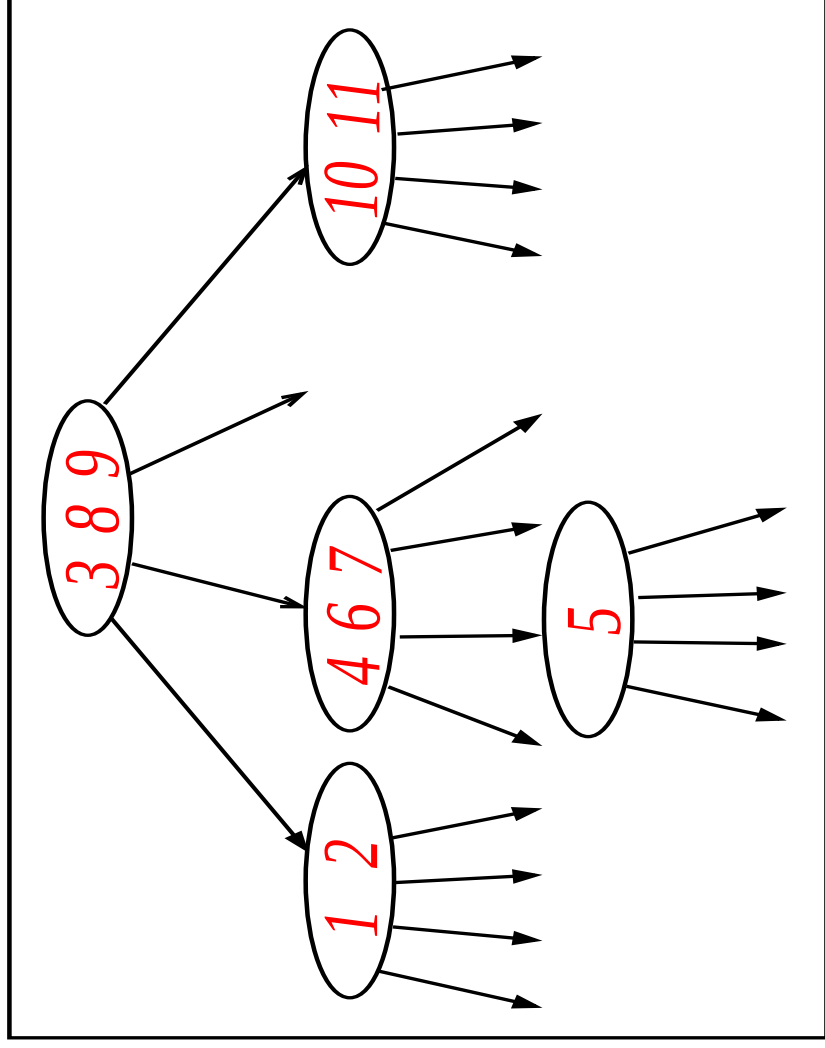
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m-ary search trees

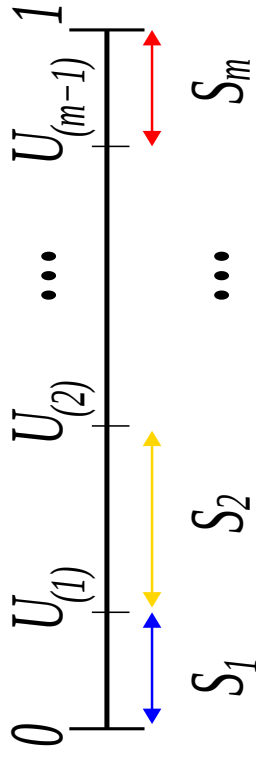
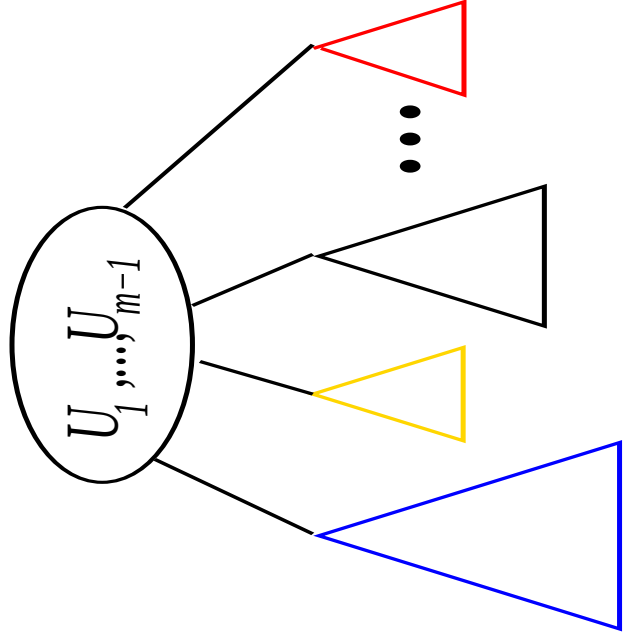
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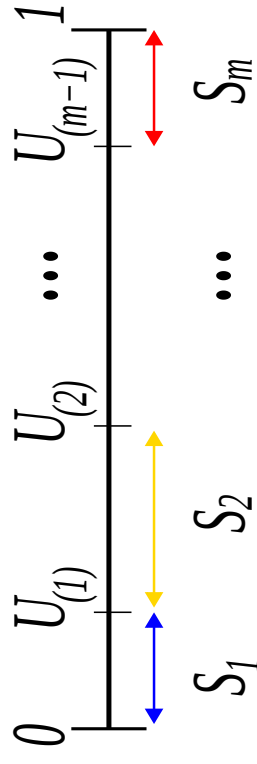
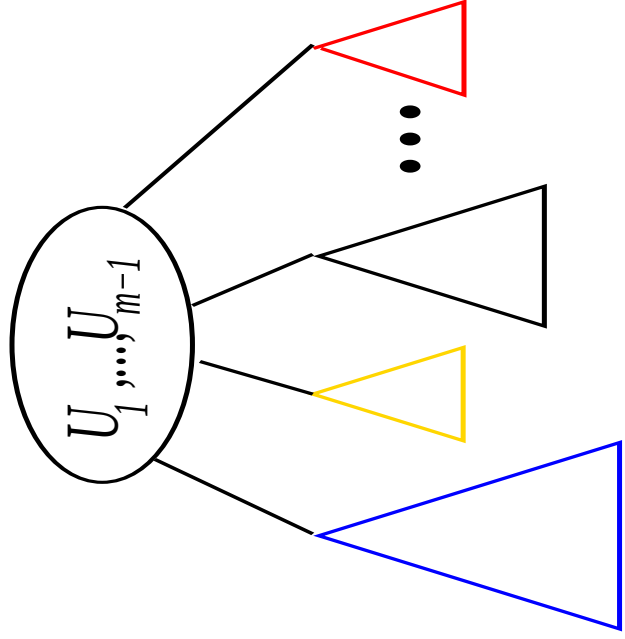
m-ary search tree

Data: $U_1, \dots, U_n$ i.i.d. $\text{unif}[0, 1]$
(or uniform permutation)



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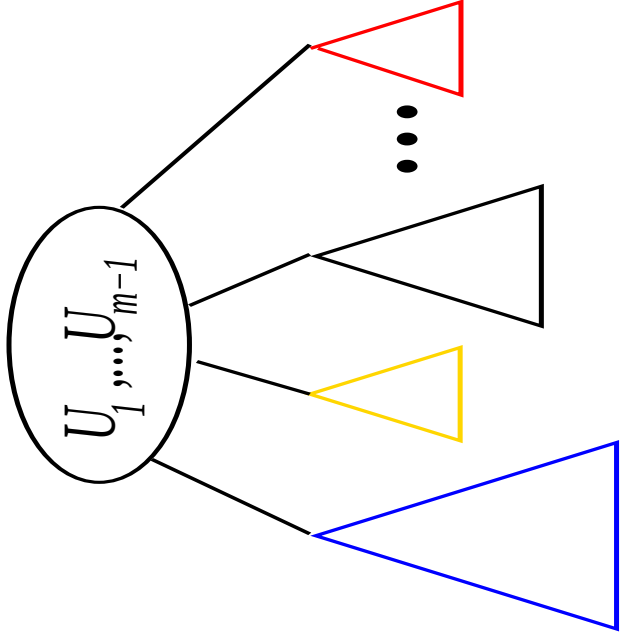


Space needed
(number of int. nodes):

$$X_n \stackrel{d}{=} \sum_{r=1}^m X_{I_r^{(n)}}^{(r)} + 1, \quad n \geq m.$$

$$X_0 = 0, X_1 = \dots = X_{m-1} = 1.$$

m-ary search tree

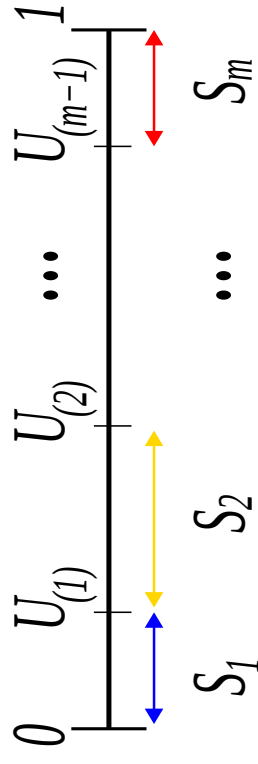


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(or uniform permutation)

Sizes of subtrees:

$$I^{(n)} \stackrel{d}{=} M(n - m + 1, S_1, \dots, S_m),$$

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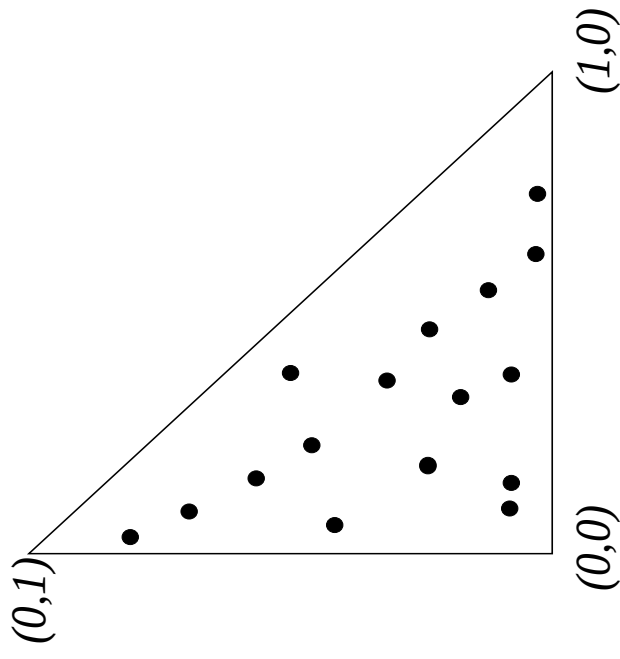


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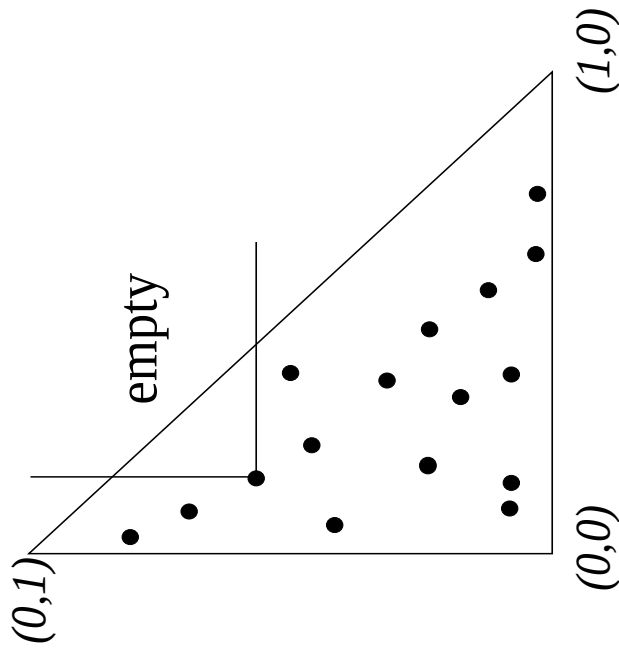
Maxima in right triangles

Data: $U_1, \dots, U_n$ indep. unif. in right triangle



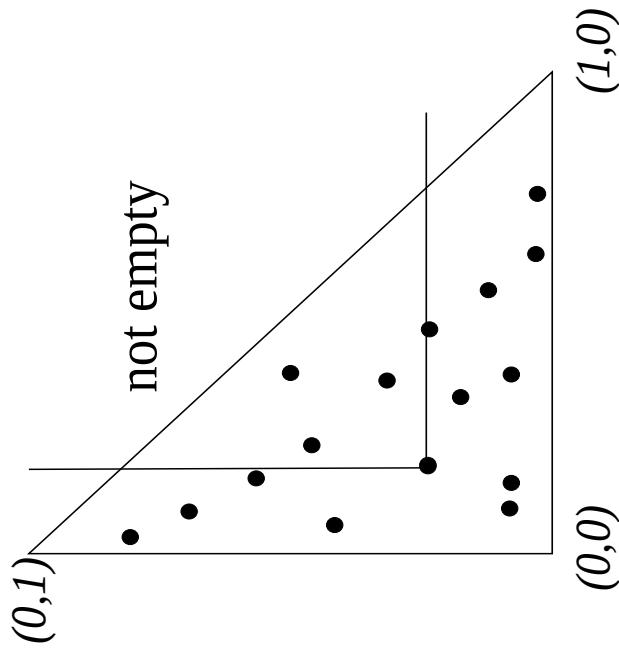
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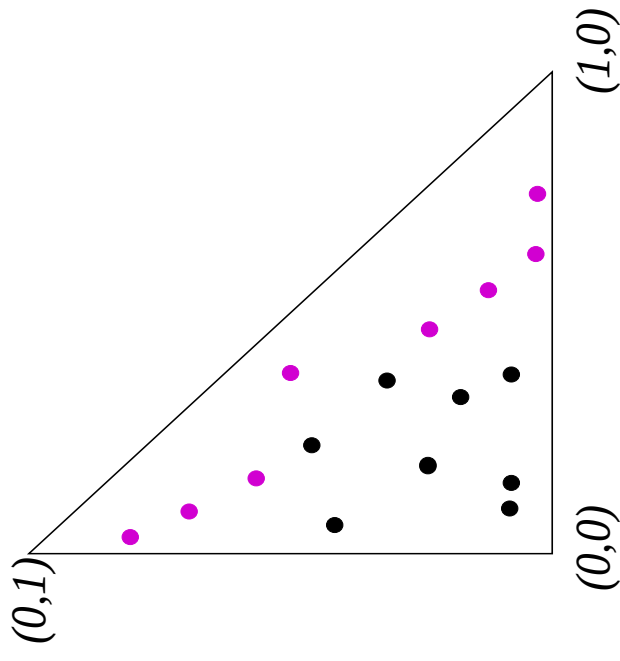
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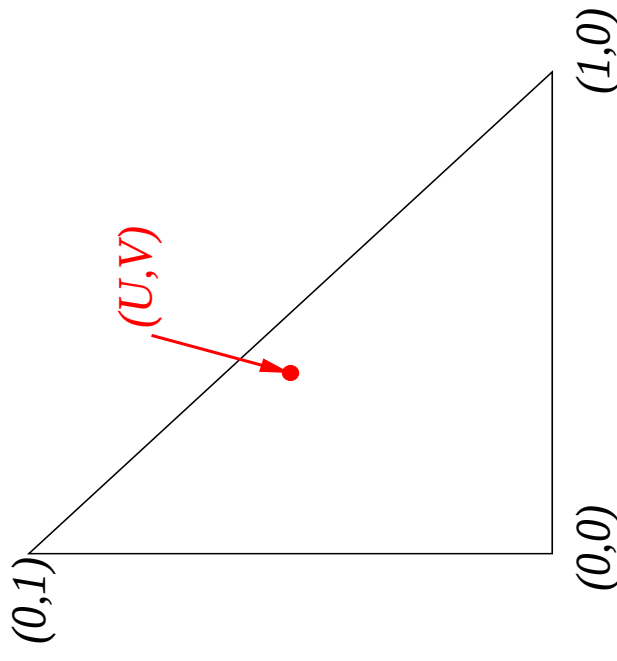
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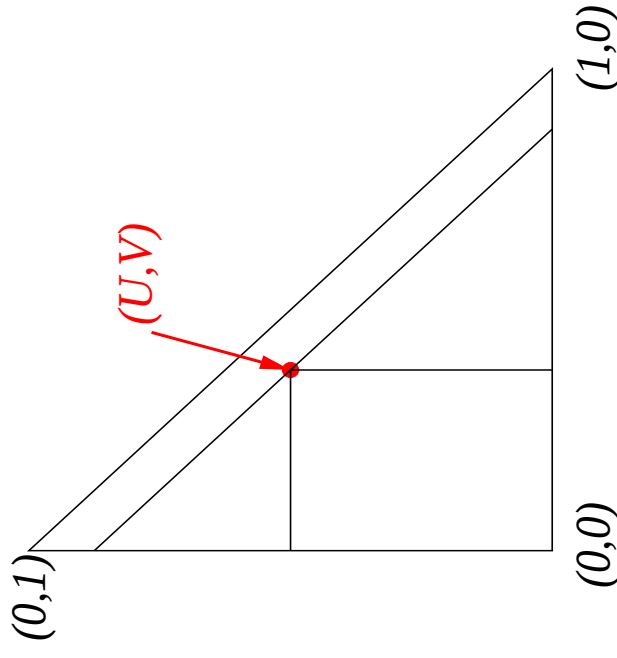
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(U, V) has maximal sum of coordinates.

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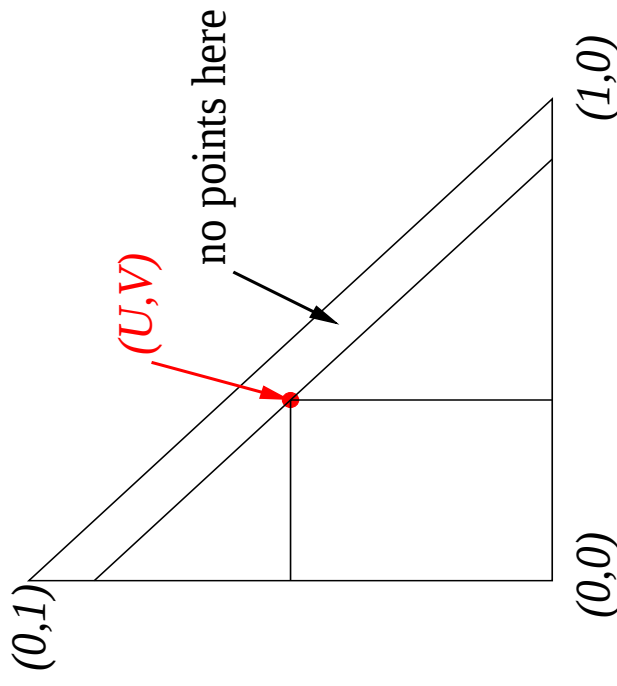
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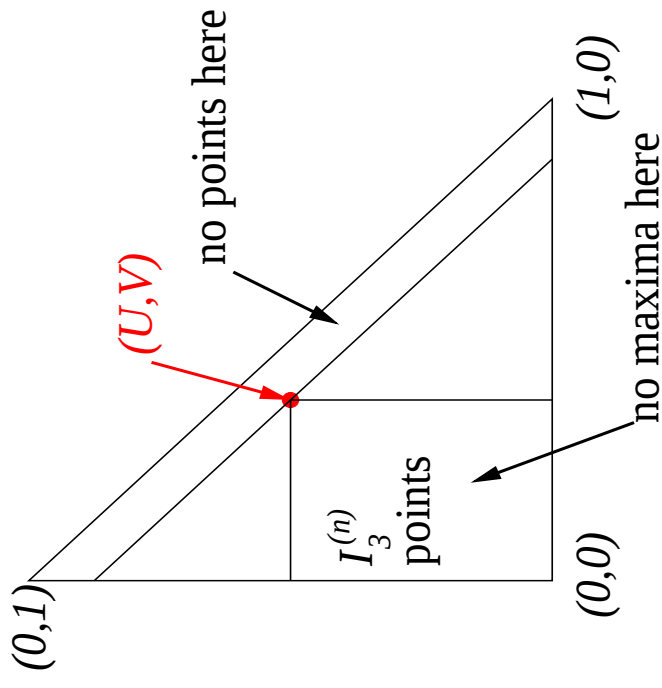
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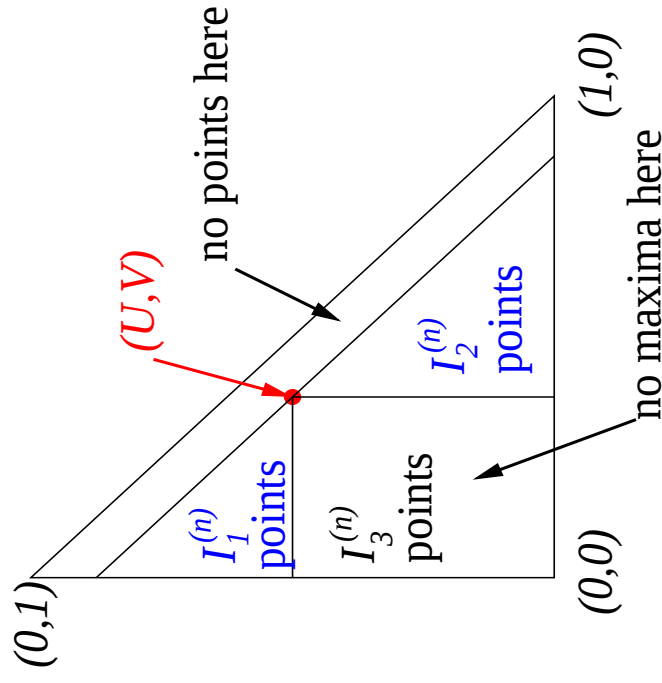
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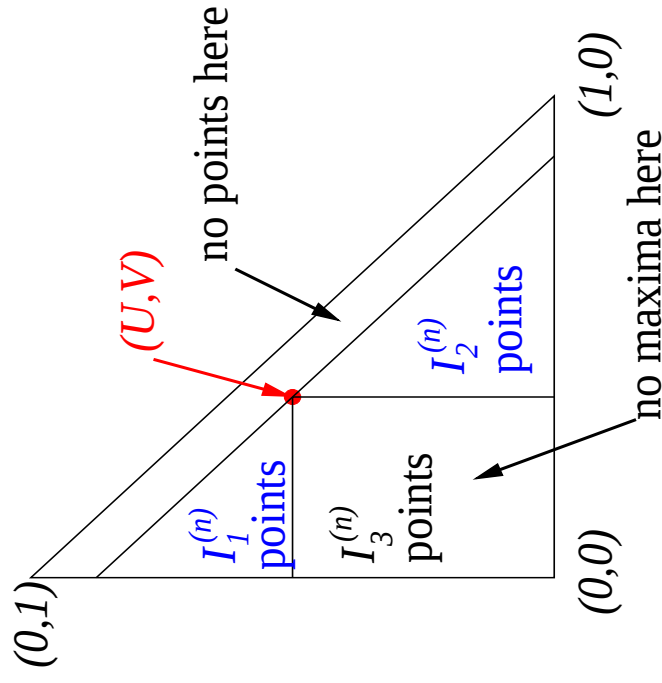
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(U, V) has maximal sum of coordinates.

$$X_n \stackrel{d}{=} X_{I_1^{(n)}}^{(1)} + X_{I_2^{(n)}}^{(2)} + 1, \quad n \geq 2.$$

General recursion

$$X_n \stackrel{\text{d}}{=} \sum_{r=1}^K A_r(n) X_{I_r(n)}^{(r)} + b_n, \quad n > n_0.$$

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- $K \geq 1$ Number of subproblems (also $K = K_n$).
- $X_n^{(r)} \stackrel{d}{=} X_n$ (recursive).
- $I^{(n)} = (I_1^{(n)}, \dots, I_K^{(n)})$ Sizes of subproblems.
- $\left(X_n^{(1)} \right), \dots, \left(X_n^{(K)} \right), (A_1(n), \dots, A_K(n), b_n, I^{(n)})$ independent.

Contraction method

Rösler (1991, 1992)

Rachev and Rüschendorf (1995)

$$X_n \stackrel{d}{=} \sum_{r=1}^K A_r(n) X_{I_r(n)}^{(r)} + b_n, \quad n > n_0.$$

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$$Y_n := \frac{X_n - \mu(n)}{\sigma(n)}.$$

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with

$$\begin{aligned} A_r^{(n)} &= \frac{\sigma(I_r^{(n)})}{\sigma(n)} A_r(n), \\ b^{(n)} &= \frac{1}{\sigma(n)} (b_n - \mu(n) + \sum_{r=1}^K A_r(n) \mu(I_r^{(n)})). \end{aligned}$$

Convergence

Idea:

$$\begin{array}{ccc}
 \gamma_n & \stackrel{\text{d}}{=} & \sum_{r=1}^K A_r^{(n)} \gamma_{I_r^{(n)}}^{(r)} + b^{(n)} \\
 \downarrow & & \downarrow \quad \downarrow \quad \downarrow \\
 \gamma & \stackrel{\text{d}}{=} & \sum_{r=1}^K A_r^* \gamma^{(r)} + b^*
 \end{array}$$

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 \downarrow & \downarrow \quad \downarrow & \downarrow \\
 \gamma & \sum_{r=1}^K A_r^* \gamma^{(r)} + b^* & \\
 & & \left. \begin{array}{cc} A_r^{(n)} \longrightarrow A_r^* \\ b^{(n)} \longrightarrow b^* \end{array} \right\} \Longrightarrow \gamma_n \longrightarrow \gamma.
 \end{array}$$

Convergence

Idea:

$$\begin{array}{ccc}
 \textcolor{red}{Y}_n & \stackrel{\text{d}}{=} & \sum_{r=1}^K A_r^{(n)} \textcolor{red}{Y}_{I_r^{(n)}}^{(r)} + \textcolor{blue}{b}^{(n)} \\
 \downarrow & & \downarrow \quad \downarrow \quad \downarrow \\
 \textcolor{red}{Y} & \stackrel{\text{d}}{=} & \sum_{r=1}^K A_r^* \textcolor{red}{Y}^{(r)} + \textcolor{blue}{b}^*
 \end{array}
 \qquad
 \left.
 \begin{array}{c}
 A_r^{(n)} \longrightarrow A_r^* \\
 \textcolor{blue}{b}^{(n)} \longrightarrow \textcolor{blue}{b}^*
 \end{array}
 \right\}
 \Longrightarrow
 \textcolor{red}{Y}_n \longrightarrow \textcolor{red}{Y}.$$

Limit map:

$$\begin{array}{ccc}
 \textcolor{green}{T} : \mathcal{M} & \longrightarrow & \mathcal{M} \\
 \textcolor{green}{v} & \longmapsto & \mathcal{L} \Big(\sum_{r=1}^K A_r^* \textcolor{green}{Z}^{(r)} + \textcolor{green}{b}^* \Big)
 \end{array}$$

with $(A_1^*, \dots, A_K^*, \textcolor{blue}{b}^*)$, $\textcolor{green}{Z}^{(1)}, \dots, \textcolor{green}{Z}^{(K)}$ independent, $\textcolor{green}{Z}^{(r)} \stackrel{\text{d}}{=} \textcolor{green}{v}$.

The minimal ℓ_p metric

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Definition: The minimal ℓ_p metric ($p \geq 1$ fixed) is given by

$$\ell_p : \mathcal{M}_p \times \mathcal{M}_p \rightarrow [0, \infty)$$

$$(\mu, \nu) \mapsto \inf\{\|X - Y\|_p : \mathcal{L}(X) = \mu, \mathcal{L}(Y) = \nu\}$$

The minimal ℓ_p metric II

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Well-known fact: For a $\text{unif}[0, 1]$ r.v. U we have

$$\mathcal{L}(F_X^{-1}(U)) = \mathcal{L}(X).$$

The minimal ℓ_p metric — optimal couplings

2nd step: Use the same $\text{unif}[0, 1]$ r.v. U for both, i.e.

$$X = F_\mu^{-1}(U), \quad Y = F_\nu^{-1}(U). \quad (1)$$

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Definition: A vector (X, Y) with $\mathcal{L}(X) = \mu$, $\mathcal{L}(Y) = \nu$ and

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Theorem: **Optimal coupling do always exist.** For $\mu, \nu \in \mathcal{M}_p$ optimal couplings are given by (4).

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Theorem: **Optimal coupling do always exist.** For $\mu, \nu \in \mathcal{M}_p$ optimal couplings are given by (4).

Corollary: We have

$$\ell_p(\mu, \nu) = \left(\int_0^1 |F_\mu^{-1}(u) - F_\nu^{-1}(u)|^p \, du \right)^{1/p}.$$

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Hence

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$\Rightarrow \mu_n \xrightarrow{\ell_p} \mathcal{L}(X) \in \mathcal{M}_p$. ♣.

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Corollary: For $\mu_n, \mu \in \mathcal{M}_p$ with $\ell_p(\mu_n, \mu) \rightarrow 0$:

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L^p convergence implies convergence in distribution.

Moreover

$$\left| \|\mu_n\|_p - \|\mu\|_p \right| = \left| \|X_n\|_p - \|X\|_p \right| \leq \|X_n - X\|_p \rightarrow 0 \quad \clubsuit$$

Lipschitz continuity on $(\mathcal{M}_p, \ell_p)$

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Theorem: Assume that $(A_1, \dots, A_k, b)$ are L^p -integrable r.v.,

$$\begin{aligned} T: \mathcal{M}_p &\rightarrow \mathcal{M}_p \\ \mu &\mapsto \mathcal{L} \left(\sum_{r=1}^k A_r Z^{(r)} + b \right), \end{aligned}$$

where $(A_1, \dots, A_k, b), Z^{(1)}, \dots, Z^{(k)}$ are indep. and $\mathcal{L}(Z^{(r)}) = \mu$.

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Then, for all $\mu, \nu \in \mathcal{M}_p$,

$$\ell_p(T(\mu), T(\nu)) \leq \left(\sum_{r=1}^k \|A_r\|_p \right) \ell_p(\mu, \nu).$$

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Then, for all $\mu, \nu \in \mathcal{M}_p$,

$$\ell_p(T(\mu), T(\nu)) \leq \left(\sum_{r=1}^k \|A_r\|_p \right) \ell_p(\mu, \nu).$$

If $\sum_{r=1}^k \|A_r\|_p < 1$ then T is a **contraction** on $(\mathcal{M}_p, \ell_p)$.

Proof

$$T(\mu) = \mathcal{L}\left(\sum_{r=1}^K A_r Z^{(r)} + b\right), \quad \mu \in \mathcal{M}_p$$

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$\|\mathsf{T}(\mu)\|_{\mathsf{p}} < \infty$, hence $\mathsf{T}(\mathcal{M}_{\mathsf{p}}) \subset \mathcal{M}_{\mathsf{p}}$.

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Let $(Z^{(1)}, w^{(1)}), \dots, (Z^{(K)}, w^{(K)})$ be i.i.d.
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$$\begin{aligned} \ell_p(T(\mu), T(\nu)) &= \inf\{\|X - Y\|_p : \mathcal{L}(X) = T(\mu), \mathcal{L}(Y) = T(\nu)\} \\ &\leq \left\| \sum_{r=1}^K A_r Z^{(r)} + \mathbf{b} - \left(\sum_{r=1}^K A_r W^{(r)} + \mathbf{b} \right) \right\|_p \end{aligned}$$

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Theorem: Assume $(A_1, \dots, A_k, b)$ are L^2 -integr. r.v. with $\mathbb{E} b = 0$,

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where $(A_1, \dots, A_k, b), Z^{(1)}, \dots, Z^{(K)}$ are indep. and $\mathcal{L}(Z^{(r)}) = \mu$.

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Convergence: Quickselect I

$$Y_n \stackrel{d}{=} \frac{I_n}{n} Y_{I_n} + \frac{n-1}{n}, \quad \mathcal{L}(I_n) = \text{unif}\{0, \dots, n-1\}$$

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$$\Delta(n) \leq \left\| \frac{I_n}{n} \Delta(I_n) \right\|_p + \frac{1 + \|Y\|_p}{n}.$$

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We obtain $\Delta(n) \rightarrow 0$.

(E.g., for $p = 1$ show $\Delta(n) \leq (C \log n)/n$ by induction.)

General theorem in $\mathcal{M}_p$

Let $(Y_n)_{n \geq 0}$ be L^p -integrable, $p \geq 1$, with (as before)

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Let $(Y_n)_{n \geq 0}$ be L^2 -integrable, with $\mathbb{E} Y_n = 0$ for all $n \geq 0$ and (as before)

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$$Y_n \stackrel{d}{=} \frac{\ln Y_{I_n}^*}{n} + \frac{n-1-\ln Y_{n-1-I_n}^{**}}{n} + b^{(n)},$$

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Convergence of coefficients and technical condition satisfied.

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In $(\mathcal{M}_2(0), \ell_2)$: **YES:**

$$\mathbb{E} \left[(A_1^*)^2 \right] + \mathbb{E} \left[(A_2^*)^2 \right] = \mathbb{E}U^2 + \mathbb{E}(1-U)^2 = \frac{2}{3} < 1.$$

Application: Central limit theorem

Let $W_1, W_2, \dots$ be i.i.d., L^p -integrable, $p \geq 2$,
with $\mathbb{E} W_1 = \mu$, $\text{Var}(W_1) = \sigma^2$.

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Then we have

$$Y_n := \frac{X_n - \mathbb{E} X_n}{\sqrt{\text{Var}(X_n)}}$$

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$$\begin{aligned} Y_n &:= \frac{X_n - \mathbb{E} X_n}{\sqrt{\text{Var}(X_n)}} = \frac{X_n - \mu n}{\sigma \sqrt{n}} \\ &\stackrel{d}{=} \sqrt{\frac{\lceil n/2 \rceil}{n}} Y_{\lceil n/2 \rceil}^* + \sqrt{\frac{\lfloor n/2 \rfloor}{n}} Y_{\lfloor n/2 \rfloor}^{**}. \end{aligned}$$

Application: Central limit theorem

Let $W_1, W_2, \dots$ be i.i.d., L^p -integrable, $p \geq 2$,
with $\mathbb{E} W_1 = \mu$, $\text{Var}(W_1) = \sigma^2$.

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Limit equation:

$$Y \stackrel{d}{=} \frac{1}{\sqrt{2}} Y^* + \frac{1}{\sqrt{2}} Y^{**}.$$

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with random $(A_1^*, \dots, A_K^*)$ with

$$\sum_{r=1}^K (A_r^*)^2 = 1 \quad \text{almost surely.}$$

The fundamental problem

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Zolotarev metric II

Spaces of probability measures

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Lemma: $\zeta_s, s > 0$, is finite on $\mathcal{M}_s(M_1, \dots, M_m) \times \mathcal{M}_s(M_1, \dots, M_m)$.
Furthermore, we have

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Exercise. ♣

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Short Notation: $\zeta_s(X, Y) := \zeta_s(\mathcal{L}(X), \mathcal{L}(Y))$.

Zolotarev metric: properties

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a) ζ_s is $(s, +)$ -ideal, i.e.,

for $c \neq 0$ and Z independent of (X, Y) we have

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c) If $X_1, \dots, X_K, Y_1, \dots, Y_K$ independent, then

$$\zeta_s \left(\sum_{r=1}^K X_r, \sum_{r=1}^K Y_r \right) \leq \sum_{r=1}^K \zeta_s(X_r, Y_r).$$

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b) Characteristic functions:

$$\varphi_{X_n}(t) = \mathbb{E} \exp(itX_n) = \mathbb{E} \cos(tX_n) + i\mathbb{E} \sin(tX_n).$$

Now, $x \mapsto \frac{1}{2|t|^s} \cos(tx)$ is in $\mathcal{F}_s$.

Hence, $\zeta_s(X_n, X) \rightarrow 0$ implies $|\varphi_{X_n}(t) - \varphi_X(t)| \rightarrow 0$ for all $t \in \mathbb{R}$.

Continuity theorem of Lévy implies $X_n \xrightarrow{d} X$.

Second assertion use $x \mapsto \frac{1}{s(s-1)\dots(s-m+1)}|x|^s$ in $\mathcal{F}_s$. ♣

c) Exercise (easy). ♣

Lipschitz in ζ_s

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The fundamental problem II

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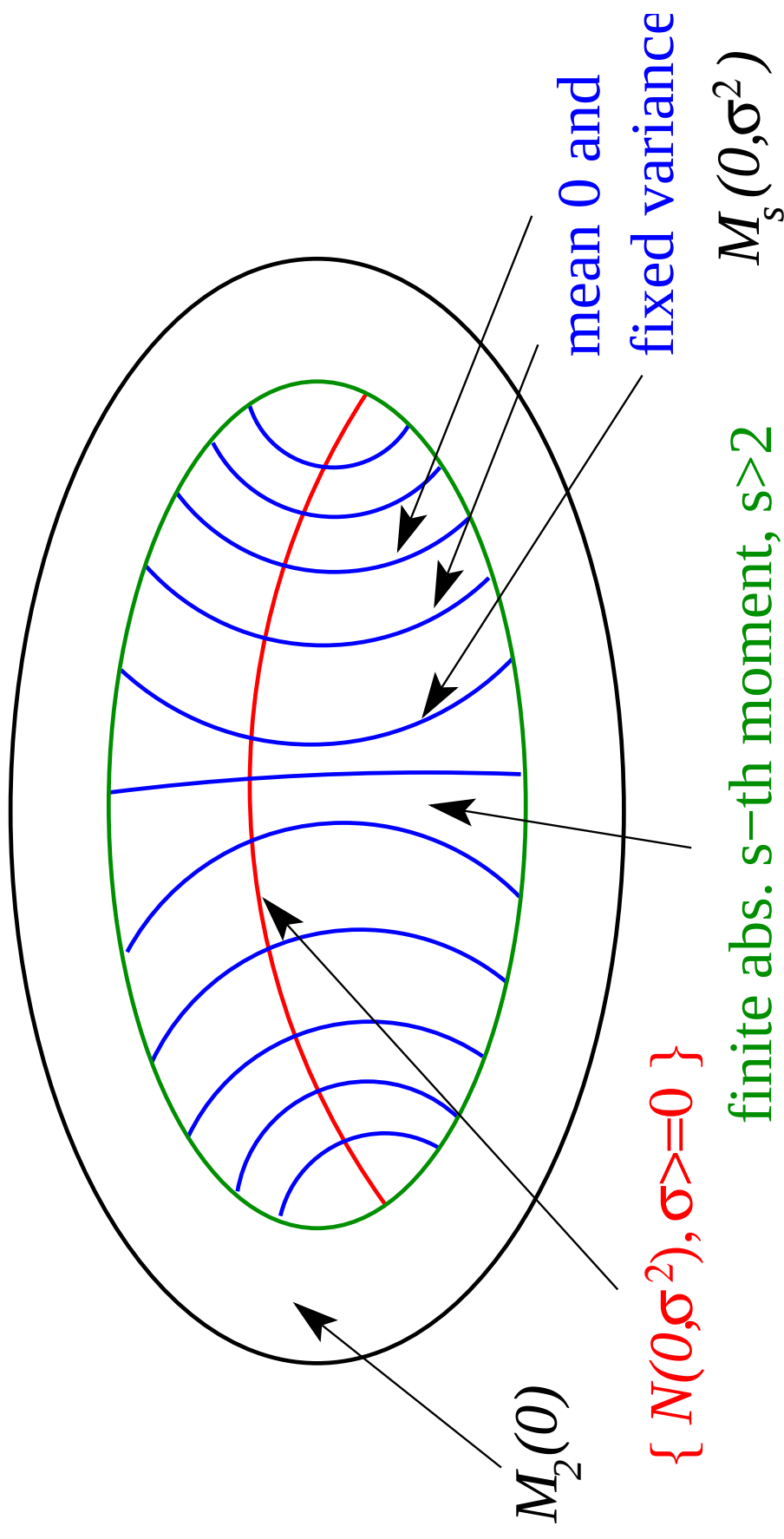
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$\Rightarrow \mathcal{N}(0, 1)$ unique fixed point of T in $\mathcal{M}_s(0, 1)$.

The work space



Central limit theorem II

$W_1, W_2, \dots$ i.i.d., L^s -integr., $s > 2$, with $\mathbb{E} W_1 = \mu$, $\text{Var}(W_1) = \sigma^2$.

$$X_n := \sum_{i=1}^n W_i \stackrel{d}{=} X_{\lceil n/2 \rceil}^* + X_{\lfloor n/2 \rfloor}^{**},$$

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$$\Rightarrow \zeta_s(Y_n, N) < \infty \text{ for all } n \geq 1.$$

Convergence

$$Y_n \stackrel{d}{=} \sqrt{\frac{\lceil n/2 \rceil}{n}} Y_{\lceil n/2 \rceil}^* + \sqrt{\frac{\lfloor n/2 \rfloor}{n}} Y_{\lfloor n/2 \rfloor}^{**},$$

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$$\begin{aligned} \zeta_s(Y_n, N) &= \zeta_s \left(\sqrt{\frac{\lceil n/2 \rceil}{n}} Y_{\lceil n/2 \rceil}^* + \sqrt{\frac{\lfloor n/2 \rfloor}{n}} Y_{\lfloor n/2 \rfloor}^{**}, \sqrt{\frac{\lceil n/2 \rceil}{n}} N^* + \sqrt{\frac{\lfloor n/2 \rfloor}{n}} N^{**} \right) \\ &\stackrel{c)}{\leq} \zeta_s \left(\sqrt{\frac{\lceil n/2 \rceil}{n}} Y_{\lceil n/2 \rceil}^*, \sqrt{\frac{\lceil n/2 \rceil}{n}} N^* \right) + \zeta_s \left(\sqrt{\frac{\lfloor n/2 \rfloor}{n}} Y_{\lfloor n/2 \rfloor}^{**}, \sqrt{\frac{\lfloor n/2 \rfloor}{n}} N^{**} \right) \\ &\stackrel{a)}{\leq} \left(\frac{\lceil n/2 \rceil}{n} \right)^{s/2} \zeta_s \left(Y_{\lceil n/2 \rceil}^*, N^* \right) + \left(\frac{\lfloor n/2 \rfloor}{n} \right)^{s/2} \zeta_s \left(Y_{\lfloor n/2 \rfloor}^{**}, N^{**} \right) \end{aligned}$$

Convergence II

With $\Delta(\mathfrak{n}) := \zeta_s(Y_{\mathfrak{n}}, \mathbf{N})$ we obtain

$$\Delta(\mathfrak{n}) \leq \left(\frac{\lceil n/2 \rceil}{\mathfrak{n}} \right)^{s/2} \Delta(\lceil n/2 \rceil) + \left(\frac{\lfloor n/2 \rfloor}{\mathfrak{n}} \right)^{s/2} \Delta(\lfloor n/2 \rfloor).$$

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A useful extension

Theorem:

$$X_n \stackrel{d}{=} \sum_{r=1}^K X_{I_r}^{(r)} + b_n, \quad n > n_0$$

with conditions as before and L^s -integrable, $2 < s \leq 3$.

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Then

$$\frac{X_n - \mathbb{E} X_n}{\sqrt{\text{Var}(X_n)}} \xrightarrow{d} \mathcal{N}(0, 1).$$