

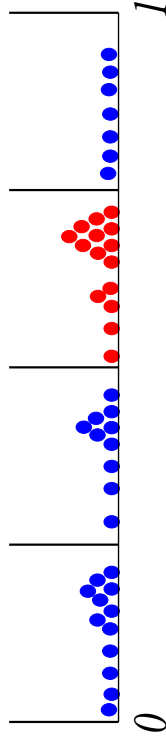
An Introduction to the Contraction Method

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Bucket Selection

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

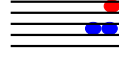
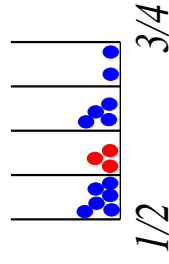
Problem: Find element with rank $k \in \{1, \dots, n\}$.



X_n : number of items distributed

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

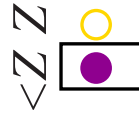
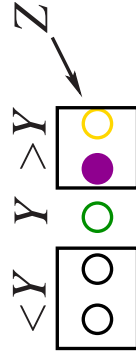
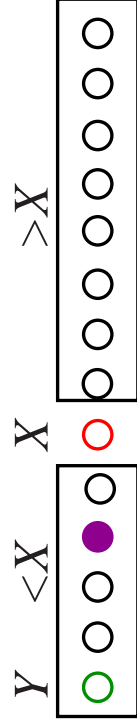
$I_n = \#$ items in relevant bucket



Quickselect

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

Problem: Find element with rank $k \in \{1, \dots, n\}$.



X_n : number of key comparisons

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

I_n = length of relevant sublist.

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

Quickselect: Analysis for $k = 1$

$$X_n \stackrel{d}{=} X_{I_n} + n - 1, \quad n \geq 2, \quad (X_0 = X_1 = 0).$$

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

I_n independent of X_0, \dots, X_{n-1} .

Scaling

$$Y_n := \frac{X_n}{n} \stackrel{d}{=} \frac{X_{I_n}}{n} + \frac{n-1}{n} = \frac{X_{I_n}}{n} + \frac{n-1}{n} = \underbrace{\frac{I_n}{n}}_{\rightarrow U} Y_{I_n} + \underbrace{\frac{n-1}{n}}_{\rightarrow 1}.$$

With $n \rightarrow \infty$:

$$\frac{X_n}{n} = Y_n \rightarrow Y \stackrel{d}{=} UY + 1$$

with U, Y independent and $U \stackrel{d}{=} \text{unif}[0, 1]$.

Quickselect: Analysis for $k = 1$

$$\frac{X_n}{n} = Y_n \rightarrow Y \stackrel{d}{=} UY + 1$$

with U, Y independent and $U \stackrel{d}{=} \text{unif}[0, 1]$.

$$\mathbb{E}Y = \mathbb{E}[UY] + 1 = \frac{1}{2}\mathbb{E}Y + 1 \Rightarrow \mathbb{E}Y = 2,$$

$$\mathbb{E}[Y^2] = \mathbb{E}[U^2Y^2 + 2UY + 1] = \frac{1}{3}\mathbb{E}[Y^2] + 3 \Rightarrow \mathbb{E}[Y^2] = \frac{9}{2},$$

$$\text{Var}(Y) = \frac{1}{2}.$$

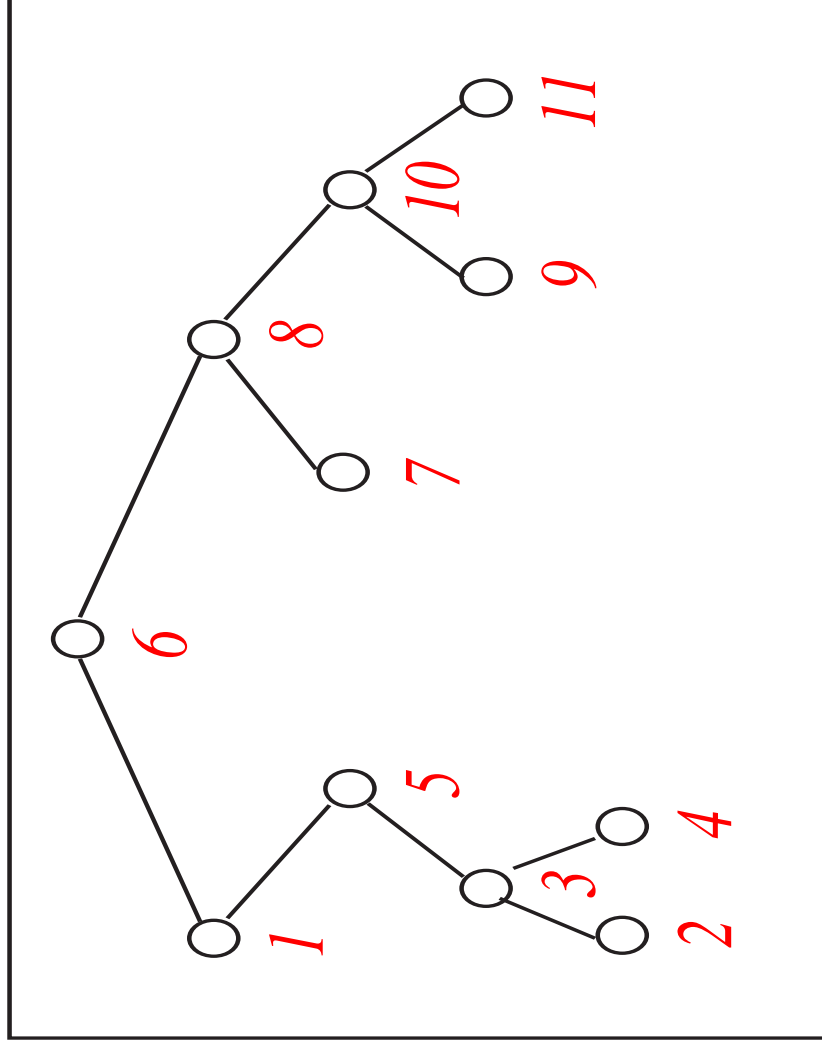
Hence, this suggests

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

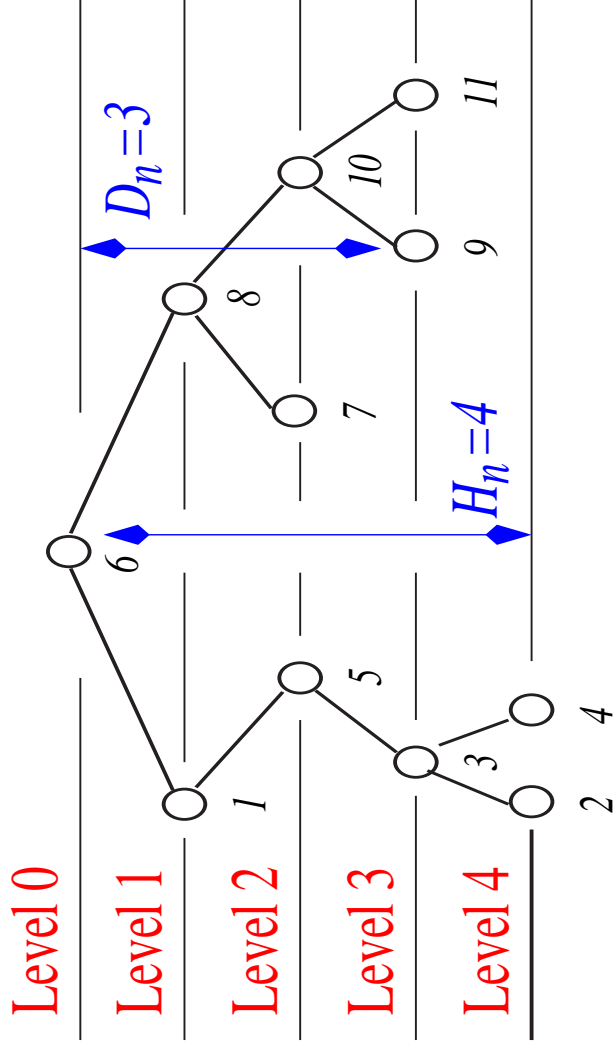
$$\text{Var}(X_n) = \text{Var}(nY_n) \sim n^2\text{Var}(Y) = \frac{1}{2}n^2.$$

Binary search tree

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



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

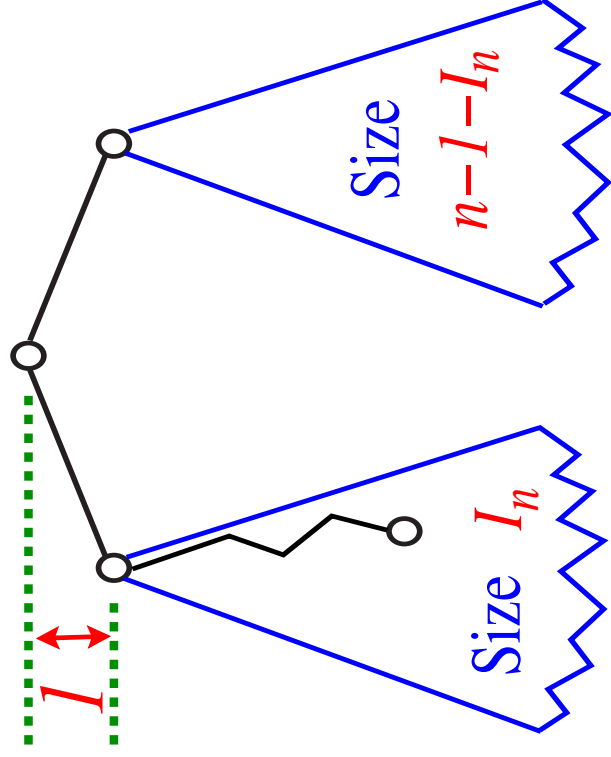
Random binary search tree

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 :

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

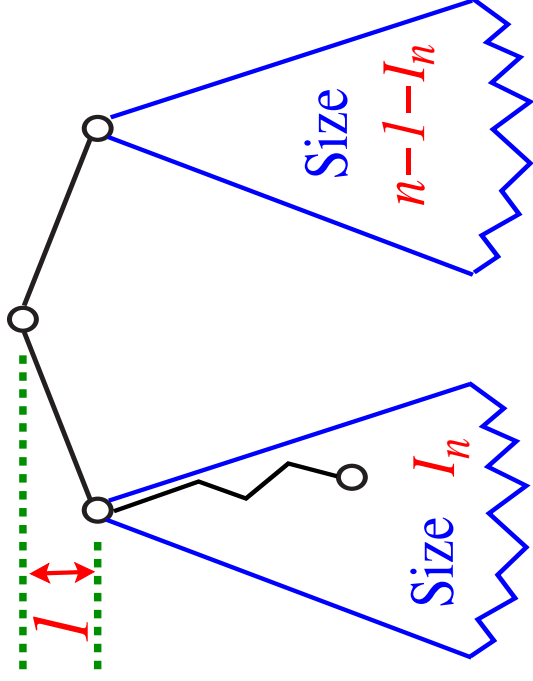
$Q_0^*, \dots, Q_{n-1}^*, Q_0^{**}, \dots, Q_{n-1}^{**}, I_n$
indep.,

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

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

This needs a proof!

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 $j \in \mathbb{N}$.

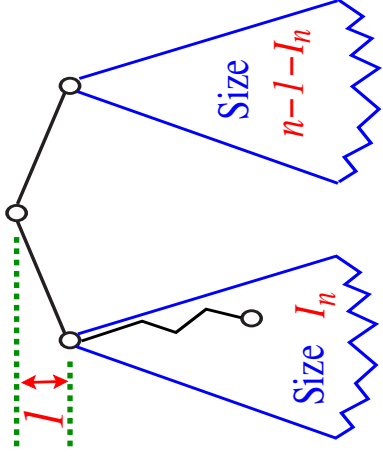
Show: For all $j \in \mathbb{N}, k \in \{0, \dots, n-1\}$:

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

[Total probability theorem yields:

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

Proof of the recurrence



To prove:

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

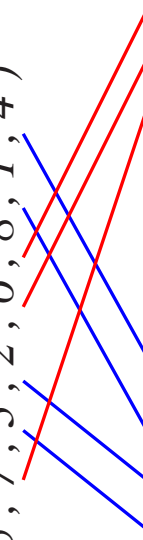
We show: Given $\{I_n = k\}$ we have:

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

π equiprobable permutation
in S_n .

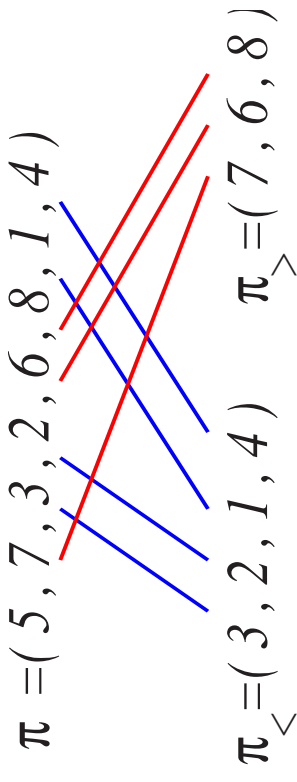
$$\pi = (5, 7, 3, 2, 6, 8, 1, 4)$$

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



Proof of the recurrence II

π equiprobable in S_n .



We show: Given $\pi_1 = k + 1$:

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

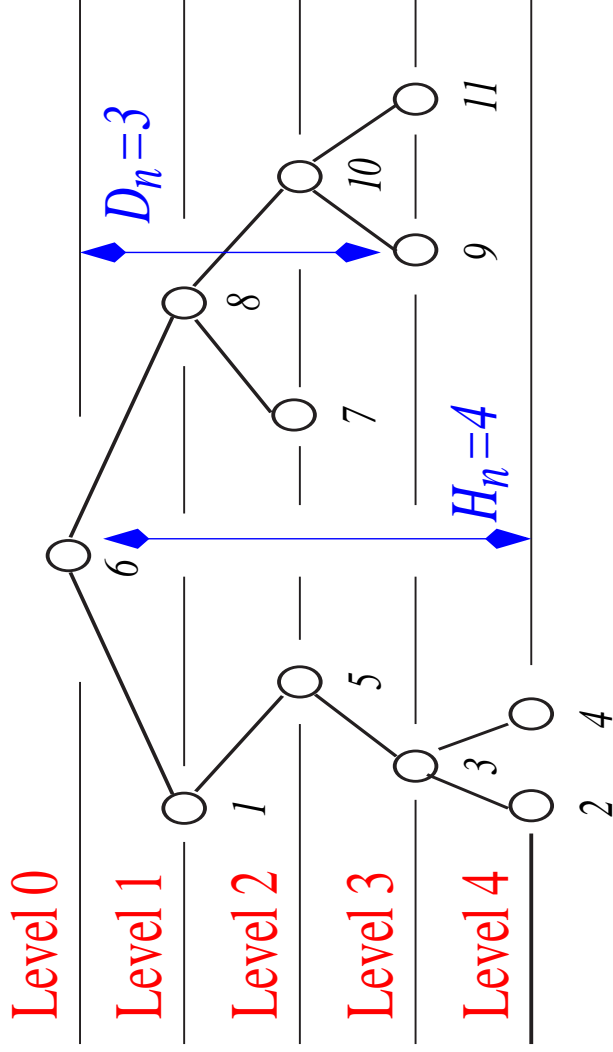
For arbitrary $\sigma \in S_k$:

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

Given $\pi_1 = k + 1$ hence $\pi_{<}$ equiprobable on S_k .

Second assertion similar.

Other BST recurrences



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

$$H_n \stackrel{d}{=} H_{I_n}^{(1)} \vee H_{n-1-I_n}^{(2)} + 1$$

$$D_n \stackrel{d}{=} \mathbf{1}_{A_n} D_{I_n} + \mathbf{1}_{A_n^c} D_{n-1-I_n} + 1$$

Expected internal path length

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

$$\begin{aligned} q_n := \mathbb{E} Q_n &= \mathbb{E} [Q_{I_n}^* + Q_{n-1-I_n}^{**}] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} \mathbb{E} [Q_k^* + Q_{n-1-k}^{**}] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} (q_k + q_{n-1-k} + n - 1) \\ &= n - 1 + \frac{2}{n} \sum_{k=0}^{n-1} q_k. \end{aligned}$$

Solves easily:

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

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

Rescaling

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

Scaling

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

Then

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

with

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

Rescaling II

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

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

$$= \frac{1}{n} (2I_n \log(I_n) + cI_n + 2(n-1-I_n) \log(n-1-I_n) + c(n-1-I_n) - 2n \log(n) - cn + o(n) + n - 1)$$

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

$$- 2(I_n + (n-1-I_n) + 1) \log(n) + o(n) + n - 1)$$

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

$$\rightarrow 2u \log(u) + 2(1-u) \log(1-u) + 1 =: g(u)$$

Rescaling: Summary

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

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

Then

$$Y_n = \underbrace{\frac{I_n}{n}}_{\rightarrow u} Y_{I_n}^* + \underbrace{\frac{n-1-I_n}{n}}_{\rightarrow 1-u} Y_{n-1-I_n}^{**} + \underbrace{b^{(n)}}_{\rightarrow g(u)}$$

with

$$g(u) = 2u \log(u) + 2(1-u) \log(1-u) + 1$$

Hence, this suggests

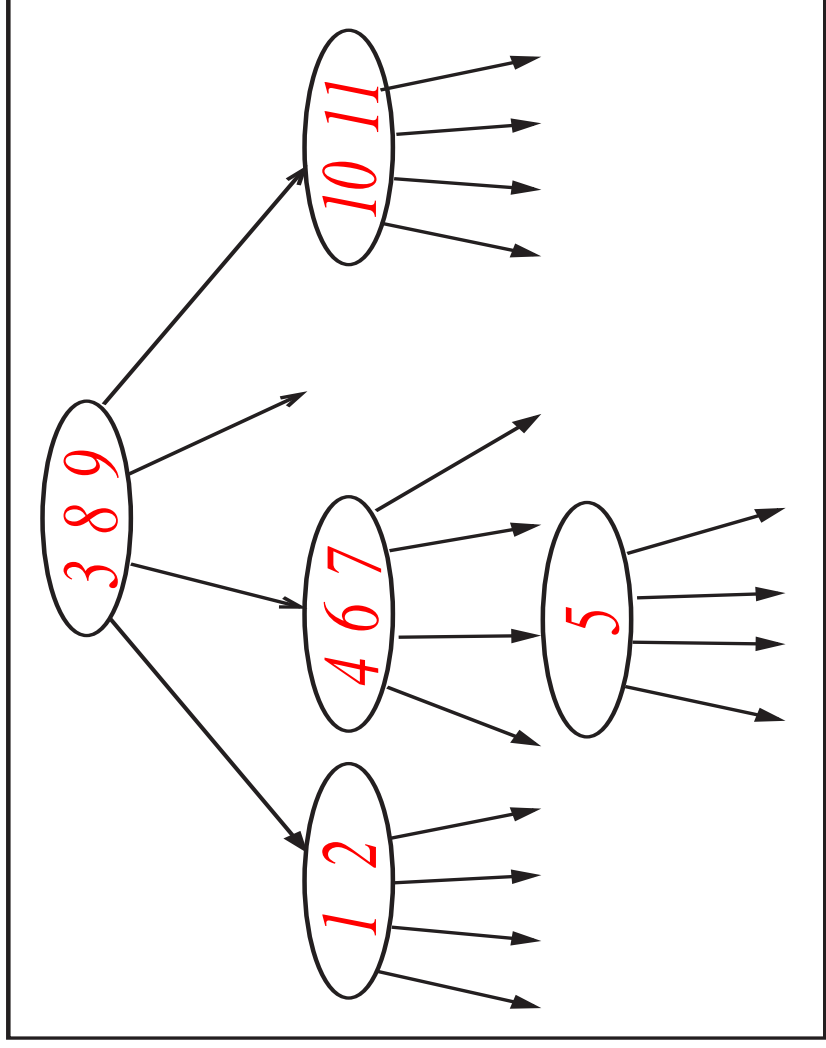
$$Y_n \rightarrow Y \stackrel{d}{=} uY^* + (1-u)Y^{**} + g(u),$$

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

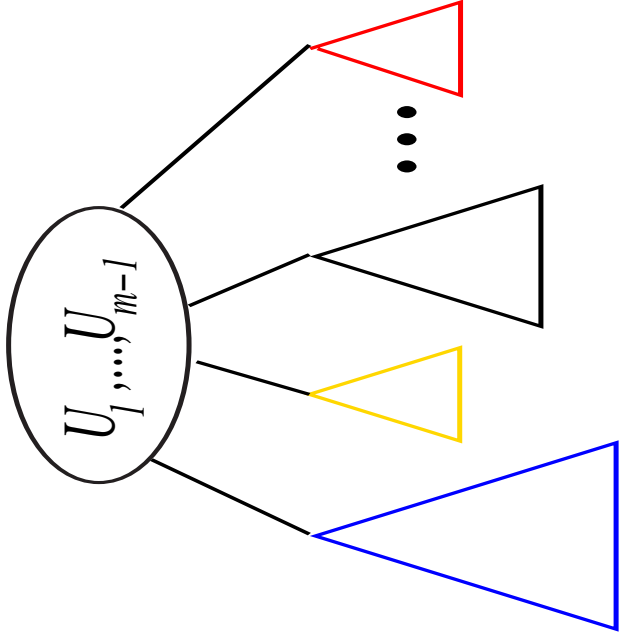
m-ary search trees

Example: $m = 4$

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



m-ary search tree

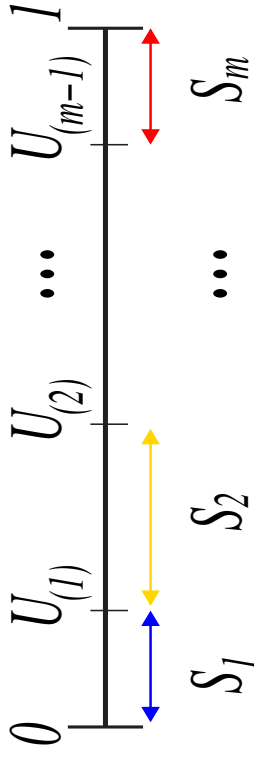


Data: U_1, \dots, U_n i.i.d. unif[0, 1]
 (or uniform permutation)

Sizes of subtrees:

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

Space needed
 (number of int. nodes):

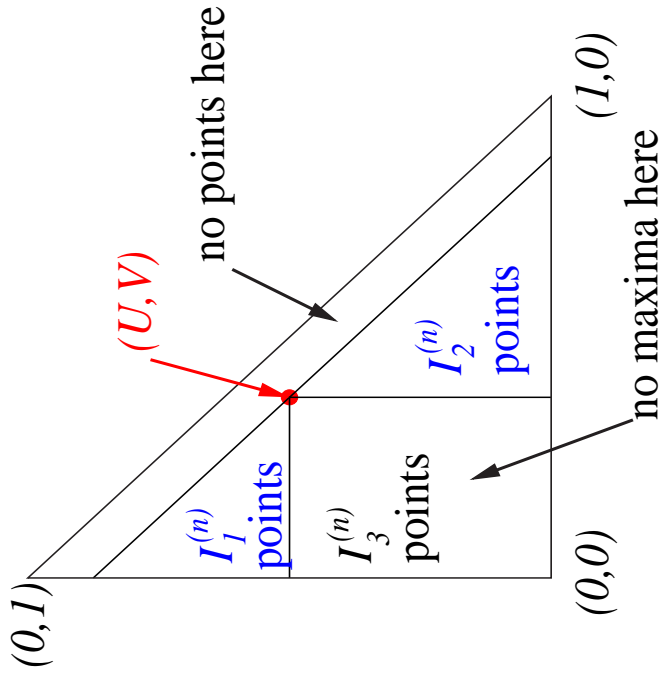


$$X_n \stackrel{d}{=} \sum_{r=1}^m X_{I_r^{(n)}}^{(r)} + 1, \quad n \geq m.$$

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

Maxima in right triangles

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



(U, V) has maximal sum of coordinates.

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

General recursion

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

- $K \geq 1$ 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.
- $(X_n^{(1)}, \dots, (X_n^{(K)}), (A_1(n), \dots, A_K(n), b_n, I^{(n)})$ independent.

Contraction method

Rösler (1991, 1992)
Rachev and Rüschendorf (1995)

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

Scaling

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

Then

$$Y_n \stackrel{d}{=} \sum_{r=1}^K A_r^{(n)} Y_{I_r(n)}^{(r)} + b^{(n)}, \quad n > n_0,$$

with

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

Convergence

Idea:

$$\begin{array}{ccc}
 \mathbf{Y}_n & \stackrel{d}{=} & \sum_{r=1}^K A_r^{(n)} \mathbf{Y}_{I_r^{(n)}} + \mathbf{b}^{(n)} \\
 \downarrow & & \downarrow \quad \downarrow \quad \downarrow \\
 \mathbf{Y} & \stackrel{d}{=} & \sum_{r=1}^K A_r^* \mathbf{Y}^{(r)} + \mathbf{b}^*
 \end{array}
 \quad \left. \begin{array}{l}
 A_r^{(n)} \rightarrow A_r^* \\
 \mathbf{b}^{(n)} \rightarrow \mathbf{b}^*
 \end{array} \right\} \Rightarrow \mathbf{Y}_n \rightarrow \mathbf{Y}.$$

Limit map:

$$\begin{array}{l}
 T: \mathcal{M} \rightarrow \mathcal{M} \\
 \nu \mapsto \mathcal{L} \left(\sum_{r=1}^K A_r^* Z^{(r)} + \mathbf{b}^* \right)
 \end{array}$$

with $(A_1^*, \dots, A_K^*, \mathbf{b}^*)$, $Z^{(1)}, \dots, Z^{(K)}$ independent, $Z^{(r)} \stackrel{d}{=} \nu$.

The minimal ℓ_p metric

\mathcal{M} : Space of probability measures on \mathbb{R} .

$$\mathcal{M}_p := \left\{ \mu \in \mathcal{M} : \|\mu\|_p := \left(\int |x|^p d\mu(x) \right)^{1/p} < \infty \right\}, \quad p \geq 1,$$

$$\mathcal{M}_2(0) := \left\{ \mu \in \mathcal{M}_2 : \mathbb{E} \mu := \int x d\mu(x) = 0 \right\}.$$

For r.v. X with distribution $\mathcal{L}(X) = \mu$ we have $\|X\|_p = \|\mu\|_p$.

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

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

How can X, Y with distributions μ, ν be constructed so that $\|X - Y\|_p$ is as small as possible?

1st Step: Construct X with $\mathcal{L}(X) = \mu$: Define distribution function of X and its (generalized) inverse by

$$F_X(x) := \mathbb{P}(X < x) = \mathcal{L}(X)((-\infty, x)), \quad x \in \mathbb{R},$$

$$F_X^{-1}(t) := \sup\{x \in \mathbb{R} : F(x) \leq t\}, \quad t \in \mathbb{R}.$$

Well-known fact: For a 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.

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

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

$$\ell_p(\mu, \nu) = \|X - Y\|_p$$

is called an **optimal coupling** of $\mu, \nu \in \mathcal{M}_p$.

Theorem: **Optimal coupling do always exist.** For $\mu, \nu \in \mathcal{M}_p$ optimal couplings are given by (1).

Corollary: We have

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

The minimal ℓ_p metric — properties

Corollary: (\mathcal{M}_p, ℓ_p) is a metric space.

Proof.

- $\ell_p(\mu, \nu) = 0 \Leftrightarrow \mu = \nu$: OK. symmetric: OK.
- **Triangle inequality:** For given $\mu, \nu, \rho \in \mathcal{M}_p$:

Define

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

with a unif[0, 1] r.v. U .

(X, Y) , (Y, Z) , (Y, Z) optimal coupl. of (μ, ν) , (ν, ρ) , and (ν, ρ) resp.

Hence

$$\ell_p(\mu, \nu) = \|X - Y\|_p \leq \|X - Z\|_p + \|Z - Y\|_p = \ell_p(\mu, \nu) + \ell_p(\nu, \rho). \quad \clubsuit$$

The minimal ℓ_p metric — properties

Corollary: (\mathcal{M}_p, ℓ_p) is complete.

Proof.

Cauchy sequence (μ_n) in \mathcal{M}_p .

Represent μ_n by r.v. X_n such that all pairs are opt. coupl.

$\|X_n - X_m\|_p = \ell_p(\mu_n, \mu_m) \Rightarrow (X_n)$ is Cauchy sequence in $L^p(\Omega, \mathcal{F}, \mathbb{P})$.

$L^p(\Omega, \mathcal{F}, \mathbb{P})$ is complete $\Rightarrow X_n \xrightarrow{L^p} X \in L^p(\Omega, \mathcal{F}, \mathbb{P})$.

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

The minimal ℓ_p metric — properties

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

$$\mu_n \xrightarrow{w} \mu, \quad \|\mu_n\|_p \rightarrow \|\mu\|_p \quad (n \rightarrow \infty).$$

Proof.

Choose $(X_n), X$ such that all pairs are optimal couplings.

Then $\|X_n - X\|_p \rightarrow 0$.

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, ℓ_p)

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

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

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

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, ℓ_p) .

Proof

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

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

Let $\mu, \nu \in \mathcal{M}_p$.

There exists **opt. coupl.** $(Z^{(1)}, W^{(1)})$ of μ and ν .

Let $(Z^{(1)}, W^{(1)}), \dots, (Z^{(K)}, W^{(K)})$ be i.i.d. and independent of $(A_1, \dots, A_K, \mathbf{b})$.

Then

$$T(\mu) = \mathcal{L} \left(\sum_{r=1}^K A_r Z^{(r)} + \mathbf{b} \right), \quad T(\nu) = \mathcal{L} \left(\sum_{r=1}^K A_r W^{(r)} + \mathbf{b} \right).$$

Proof

$$T(\boldsymbol{\mu}) = \mathcal{L} \left(\sum_{r=1}^K A_r Z^{(r)} + \mathbf{b} \right), \quad T(\boldsymbol{\nu}) = \mathcal{L} \left(\sum_{r=1}^K A_r W^{(r)} + \mathbf{b} \right).$$

Hence

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

Lipschitz on $(\mathcal{M}_2(0), \ell_2)$

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

$$T: \mathcal{M}_2(0) \rightarrow \mathcal{M}_2(0)$$

$$\mu \mapsto \mathcal{L} \left(\sum_{r=1}^k A_r Z^{(r)} + b \right),$$

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

Then, for all $\mu, \nu \in \mathcal{M}_2(0)$,

$$\ell_2(T(\mu), T(\nu)) \leq \left(\sum_{r=1}^k \mathbb{E} A_r^2 \right)^{1/2} \ell_2(\mu, \nu).$$

If $\sum_{r=1}^k \mathbb{E} A_r^2 < 1$ then T is a contraction on $(\mathcal{M}_2(0), \ell_2)$.

Convergence: Quickselect I

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

$$Y \stackrel{d}{=} UY + 1, \quad \mathcal{L}(U) = \text{unif}[0, 1].$$

We choose $I_n = \lfloor nU \rfloor$ and (Y_j, Y) as optimal couplings. Then

$$\begin{aligned} \Delta(n) &:= \ell_p(\mathcal{L}(Y_n), \mathcal{L}(Y)) \leq \left\| \frac{I_n Y_{I_n} + n - 1}{n} - (UY + 1) \right\|_p \\ &\leq \left\| \frac{I_n Y_{I_n} \pm I_n Y - UY}{n} \right\|_p + \frac{1}{n} \\ &\leq \left\| \frac{I_n}{n} (Y_{I_n} - Y) \right\|_p + \frac{1 + \|Y\|_p}{n}. \end{aligned}$$

Convergence: Quickselect II

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

Now,

$$\begin{aligned} \left\| \frac{I_n}{n} (Y_{I_n} - Y) \right\|_p^p &= \mathbb{E} \left[\left\| \frac{I_n}{n} (Y_{I_n} - Y) \right\|_p^p \right] = \frac{1}{n} \sum_{j=0}^{n-1} \binom{j}{n}^p \mathbb{E} [|Y_j - Y|^p] \\ &= \frac{1}{n} \sum_{j=0}^{n-1} \binom{j}{n}^p \ell_p^p(\mathcal{L}(Y_j), \mathcal{L}(Y)) = \left\| \frac{I_n}{n} \Delta(I_n) \right\|_p^p. \end{aligned}$$

Together

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

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)

$$Y_n \stackrel{d}{=} \sum_{r=1}^K A_r^{(n)} Y_{I_r^{(n)}}^{(r)} + b^{(n)}.$$

Assume that

$$\left(A_1^{(n)}, \dots, A_K^{(n)}, b^{(n)} \right) \xrightarrow{\ell^p} (A_1^*, \dots, A_K^*, b^*),$$

$$\sum_{r=1}^K \|A_r^*\|_p < 1,$$

$$\mathbb{E} \left[\mathbf{1}_{\{I_r^{(n)} \leq \ell\}} |A_r^{(n)}|^p \right] \xrightarrow{n \rightarrow \infty} 0, \quad \forall \ell \in \mathbb{N}, r = 1, \dots, K.$$

Then

$$\ell^p(\mathcal{L}(Y_n), \mathcal{L}(Y)) \xrightarrow{n \rightarrow \infty} 0,$$

where $\mathcal{L}(Y)$ is the fixed-point of T in \mathcal{M}_p .

General theorem in $\mathcal{M}_2(0)$

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

$$Y_n \stackrel{d}{=} \sum_{r=1}^K A_r^{(n)} Y_{I_r^{(n)}}^{(r)} + b^{(n)}.$$

Assume that

$$\left(A_1^{(n)}, \dots, A_K^{(n)}, b^{(n)} \right) \xrightarrow{\ell_2} (A_1^*, \dots, A_K^*, b^*),$$

$$\sum_{r=1}^K \mathbb{E} [(A_r^*)^2] < 1,$$

$$\mathbb{E} \left[\mathbf{1}_{\{I_r^{(n)} \leq \ell\}} |A_r^{(n)}|^2 \right] \xrightarrow{n \rightarrow \infty} 0, \quad \forall \ell \in \mathbb{N}, r = 1, \dots, K.$$

Then

$$\ell_2(\mathcal{L}(Y_n), \mathcal{L}(Y)) \xrightarrow{n \rightarrow \infty} 0,$$

where $\mathcal{L}(Y)$ is the fixed-point of T in $\mathcal{M}_2(0)$.

Application: Path length in BST

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

$$Y \stackrel{d}{=} UY^* + (1-U)Y^{**} + g(U).$$

Convergence of coefficients and technical condition satisfied.

Contraction: $A_1^* = U$, $A_2^* = 1 - U$.

In (\mathcal{M}_p, ℓ_p) : **NO** for all $p \geq 1$:

$$\|A_1^*\|_p + \|A_2^*\|_p \geq \|A_1^*\|_1 + \|A_2^*\|_1 = \mathbb{E}U + \mathbb{E}(1-U) = 1.$$

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

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

Application: Central limit theorem

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

Consider

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

Then we have

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

Limit equation:

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

Contraction properties

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

Contraction: $A_1^* = A_2^* = 1/\sqrt{2}$

In (\mathcal{M}_p, ℓ_p) : **NO** for all $p \geq 1$: $\|A_1^*\|_p + \|A_2^*\|_p = \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} = \sqrt{2} > 1$.

In $(\mathcal{M}_2(0), \ell_2)$: **NO**: $\mathbb{E}[(A_1^*)^2] + \mathbb{E}[(A_2^*)^2] = \frac{1}{2} + \frac{1}{2} = 1$.

Same problem for

$$Y \stackrel{d}{=} \sum_{r=1}^K A_r^* Y^{(r)},$$

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

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

The fundamental problem

$$Y \stackrel{d}{=} \sum_{r=1}^K A_r^* Y^{(r)}, \quad \sum_{r=1}^K (A_r^*)^2 = 1 \quad \text{a.s.}$$

Appears in CLT, space of m -ary search tree ($3 \leq m \leq 26$), number of maxima in right triangles, etc.

Easy to check:

$\mathcal{N}(0, \sigma^2)$ is solution for all $\sigma \geq 0$.

However

$\mathcal{N}(0, \sigma^2) \in \mathcal{M}_2(0)$ for all $\sigma \geq 0$.

Also $\mathcal{N}(0, \sigma^2) \in \mathcal{M}_p$ for all $p \geq 1$, $\sigma \geq 0$.

\implies no unique fixed-point in these spaces.

\implies no contraction in any metric in these spaces.

Zolotarev metric

$$\zeta_s(\mathcal{L}(X), \mathcal{L}(Y)) = \sup_{f \in \mathcal{F}_s} |\mathbb{E}[f(X) - f(Y)]|, \quad s > 0,$$

with

$$\mathcal{F}_s = \{f \in C^m(\mathbb{R}, \mathbb{R}) : |f^{(m)}(x) - f^{(m)}(y)| \leq |x - y|^\alpha\},$$

$$s = m + \alpha, \quad m \in \mathbb{N}_0, \quad \alpha \in (0, 1].$$

Functions in \mathcal{F}_s :

$$x \mapsto \frac{1}{2|t|^s} \cos(tx), \quad t \neq 0; \quad x \mapsto \frac{1}{s(s-1) \cdots (s-m+1)} |x|^s$$

$$\text{For } f \in \mathcal{F}_s: \quad x \mapsto f(x+z), \quad z \in \mathbb{R}; \quad x \mapsto \frac{1}{|c|^s} f(cx), \quad c \neq 0.$$

Zolotarev metric II

Spaces of probability measures

$$\begin{aligned}\mathcal{M}_s(M_1, \dots, M_m) &:= \left\{ \mu \in \mathcal{M}_s \mid \int x^k d\mu(x) = M_k, k = 1, \dots, m \right\} \\ &= \left\{ \mathcal{L}(X) \in \mathcal{M}_s \mid \mathbb{E}[X^k] = M_k, k = 1, \dots, m \right\}.\end{aligned}$$

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

$$\zeta_s(\mathcal{L}(X), \mathcal{L}(Y)) \leq \frac{\Gamma(1 + \alpha)}{\Gamma(1 + s)} (\mathbb{E}|X|^s + \mathbb{E}|Y|^s).$$

Proof.

Exercise. ♣

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

Zolotarev metric: properties

Lemma:

a) ζ_s is $(s, +)$ -ideal, i.e.,

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

$$\zeta_s(X + Z, Y + Z) \leq \zeta_s(X, Y), \quad \zeta_s(cX, cY) = |c|^s \zeta_s(X, Y).$$

b) If $\zeta_s(X_n, X) \rightarrow 0$, then $X_n \xrightarrow{d} X$ and $\|X_n\|_s \rightarrow \|X\|_s$.

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

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

Zolotarev metric: properties

Proof.

a) Note that $x \mapsto f(x+z)$ in \mathcal{F}_s for $f \in \mathcal{F}_s$.

$$\begin{aligned}\zeta_s(X+Z, Y+Z) &= \sup_{f \in \mathcal{F}_s} |\mathbb{E}[f(X+Z) - f(Y+Z)]| \\ &= \sup_{f \in \mathcal{F}_s} \left| \int \mathbb{E}[f(X+z) - f(Y+z)] d\mathbb{P}_Z(z) \right| \\ &\leq \int \sup_{f \in \mathcal{F}_s} |\mathbb{E}[f(X+z) - f(Y+z)]| d\mathbb{P}_Z(z) \\ &= \underbrace{\sup_{f \in \mathcal{F}_s} \mathbb{E}[|f(X) - f(Y)|]}_{= \zeta_s(X, Y)} \\ &= \zeta_s(X, Y).\end{aligned}$$

Second part similar with $x \mapsto |c|^{-s}f(cx)$ in \mathcal{F}_s for $f \in \mathcal{F}_s$. ♣

Zolotarev metric: properties

Proof.

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

Theorem: Assume (A_1, \dots, A_k, b) are L^s -integrable r.v. and

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

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

Then, for all $\mu, \nu \in \mathcal{M}_s(M_1, \dots, M_m)$,

$$\zeta_s(T(\mu), T(\nu)) \leq \left(\sum_{r=1}^k \mathbb{E} [|A_r|^s] \right) \zeta_s(\mu, \nu).$$

Lipschitz in ζ_s

Proof.

$Z^{(1)}, \dots, Z^{(K)}, W^{(1)}, \dots, W^{(K)}, (A_1, \dots, A_k, b)$ independent with $\mathcal{L}(Z^{(r)}) = \mu, \mathcal{L}(W^{(r)}) = \nu$.

Hence

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

Then

$$\zeta_s(T(\mu), T(\nu)) = \sup_{f \in \mathcal{F}_s} \mathbb{E} \left[\left| f\left(\sum_{r=1}^K A_r Z^{(r)} + b\right) - f\left(\sum_{r=1}^K A_r W^{(r)} + b\right) \right| \right]$$

Lipschitz in ζ_s

Denote $\sigma = \mathbb{P}_{(A_1, \dots, A_K, b)}$ and $\mathbf{x} = (\alpha_1, \dots, \alpha_K, \beta)$. Then

$$\begin{aligned}
 \zeta_s(T(\mu), T(\nu)) &= \sup_{f \in \mathcal{F}_s} \left| \mathbb{E} \left[f \left(\sum_{r=1}^K A_r Z^{(r)} + b \right) - f \left(\sum_{r=1}^K A_r W^{(r)} + b \right) \right] \right| \\
 &= \sup_{f \in \mathcal{F}_s} \left| \int \mathbb{E} \left[f \left(\sum_{r=1}^K \alpha_r Z^{(r)} + \beta \right) - f \left(\sum_{r=1}^K \alpha_r W^{(r)} + \beta \right) \right] d\sigma(\mathbf{x}) \right| \\
 &\leq \int \sup_{f \in \mathcal{F}_s} \left| \mathbb{E} \left[f \left(\sum_{r=1}^K \alpha_r Z^{(r)} + \beta \right) - f \left(\sum_{r=1}^K \alpha_r W^{(r)} + \beta \right) \right] \right| d\sigma(\mathbf{x}) \\
 &\leq \int \zeta_s \left(\sum_{r=1}^K \alpha_r W^{(r)} + \beta, \sum_{r=1}^K \alpha_r Z^{(r)} + \beta \right) d\sigma(\mathbf{x}) \\
 &\stackrel{a),c)}{\leq} \int \sum_{r=1}^K \zeta_s \left(\alpha_r W^{(r)}, \alpha_r Z^{(r)} \right) d\sigma(\mathbf{x}) \stackrel{a)}{\leq} \int \sum_{r=1}^K |\alpha_r|^s \zeta_s \left(W^{(r)}, Z^{(r)} \right) d\sigma(\mathbf{x}) \\
 &= \left(\sum_{r=1}^K \mathbb{E} [|A_r|^s] \right) \zeta_s(\mu, \nu). \quad \clubsuit
 \end{aligned}$$

The fundamental problem II

$$T: \mathcal{M} \rightarrow \mathcal{M}, \quad \mu \mapsto \mathcal{L} \left(\sum_{r=1}^K A_r^* Z^{(r)} \right), \quad \text{with } \sum_{r=1}^K (A_r^*)^2 = 1 \quad \text{a.s.},$$

where $(A_1^*, \dots, A_K^*), Z^{(1)}, \dots, Z^{(K)}$ independent and $\mathcal{L}(Z^{(r)}) = \mu$.

$$\text{Excluded case: } \mathbb{P} \left(\bigcup_{r=1}^K \{A_r^* = 1\} \right) = 1. \quad (*)$$

Known: $\{\mathcal{N}(0, \sigma^2) \mid \sigma \geq 0\}$ are fixed-points.

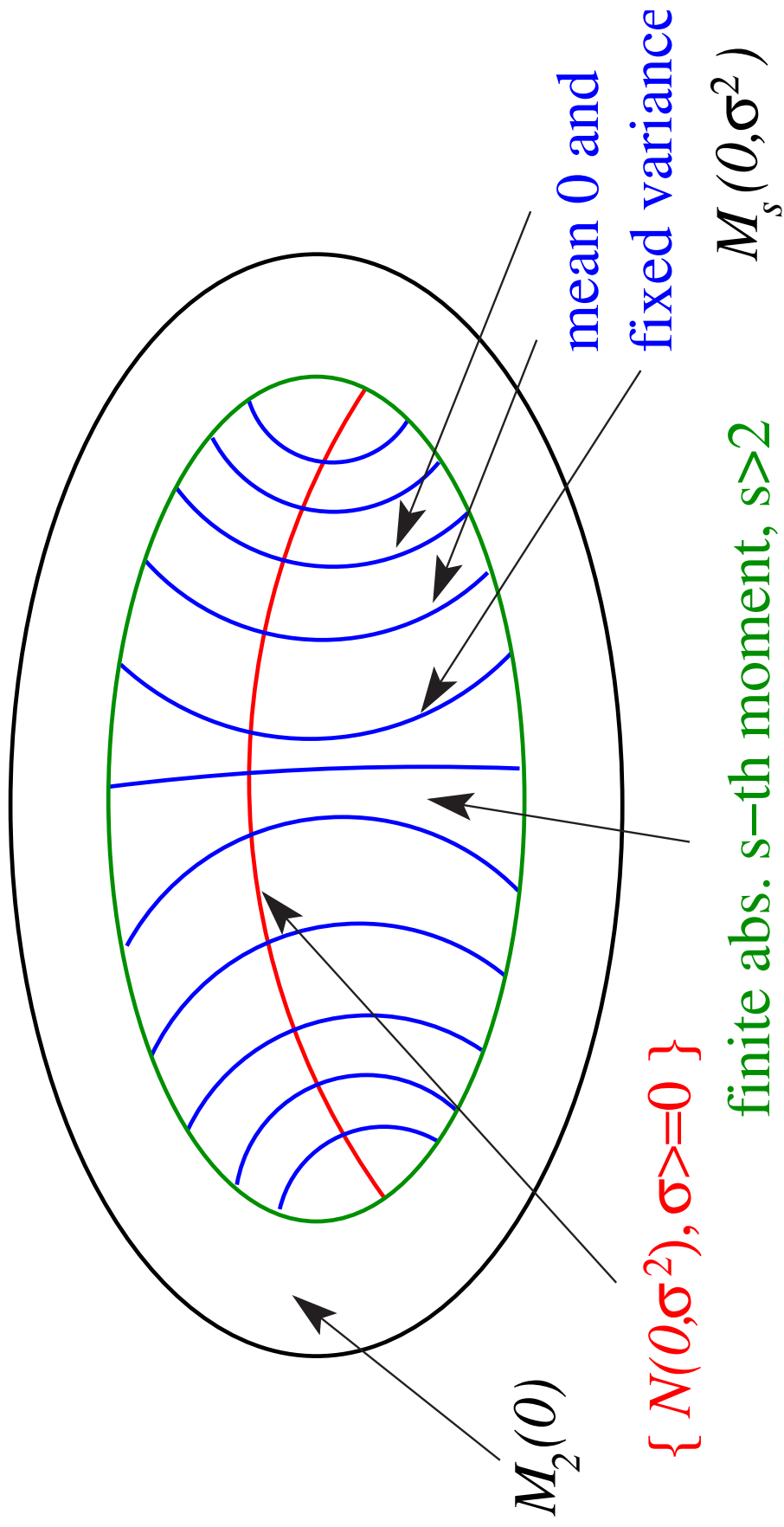
We have $T(\mathcal{M}_s(0, 1)) \subseteq \mathcal{M}_s(0, 1)$ for $2 < s \leq 3$.

(*) implies $\sum_{r=1}^K \mathbb{E} [(A_r^*)^s] < 1$.

Previous Lemma: T contraction on $(\mathcal{M}_s(0, 1), \zeta_s)$, $2 < s \leq 3$.

$\Rightarrow \mathcal{N}(0, 1)$ 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 $\mathbb{E}W_1 = \mu$, $\text{Var}(W_1) = \sigma^2$.

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

$$Y_n := \frac{X_n - \mu n}{\sigma \sqrt{n}} \stackrel{d}{=} \sqrt{\frac{\lceil n/2 \rceil}{n}} Y_{\lceil n/2 \rceil}^* + \sqrt{\frac{\lceil n/2 \rceil}{n}} Y_{\lceil n/2 \rceil}^{**}.$$

Consider $2 < s \leq 3$. For $\mathcal{L}(N) = \mathcal{N}(0, 1)$ we have

$$\mathbb{E}Y_n^k = \mathbb{E}N^k \text{ for } k = 1, 2 \quad \text{and} \quad \mathbb{E}|Y_n^k|^s, \mathbb{E}|N|^s < \infty.$$

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

Convergence

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

$$N \stackrel{d}{=} \sqrt{\frac{\lfloor n/2 \rfloor}{n}} N^* + \sqrt{\frac{\lfloor n/2 \rfloor}{n}} N^{**}.$$

Thus

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

Convergence II

With $\Delta(n) := \zeta_s(Y_n, N)$ we obtain

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

Easy: $\Delta(n) \leq \zeta_s(Y_1, N)$ for all $n \geq 1$ (by induction).

With $L := \limsup_{n \rightarrow \infty} \Delta(n)$ we obtain

$$L \leq 2^{1-s/2}L, \quad \text{hence } L = 0.$$

A useful extension

Theorem:

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

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

$$\mathbb{E} X_n = \mu(n) + o(\sigma(n)), \quad \text{Var}(X_n) = \sigma^2(n) + o(\sigma^2(n))$$

and

$$\left(\frac{\sigma(I_1^{(n)})}{\sigma(n)}, \dots, \frac{\sigma(I_K^{(n)})}{\sigma(n)} \right) \xrightarrow{\ell_s} (A_1^*, \dots, A_K^*), \quad \sum_{r=1}^K (A_r^*)^2 = 1 \text{ a.s. with } (*),$$

$$b_n + \mu(n) - \sum_{r=1}^K \mu(I_r^{(n)}) = o(\sigma(n)) \text{ in } L_s.$$

Then

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