

PROFILES OF m -ARY SEARCH TREES

Michael Drmota

(joint work with Svante Janson and Ralph Neininger)

Institut für Diskrete Mathematik und Geometrie,
TU Wien

michael.drmota@tuwien.ac.at

www.dmg.tuwien.ac.at/drmota/

Summery

- Quicksort
- Probabilistic Model
- Profile of Trees
- Function Spaces
- Contraction Method
- Cauchy Integral

Quicksort

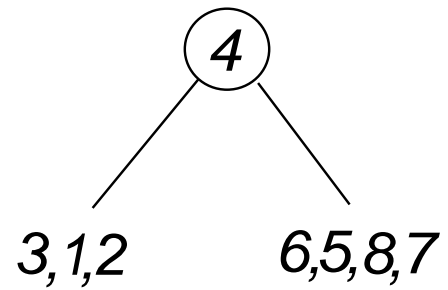
Sorting of data

4,6,3,5,1,8,2,7

Quicksort

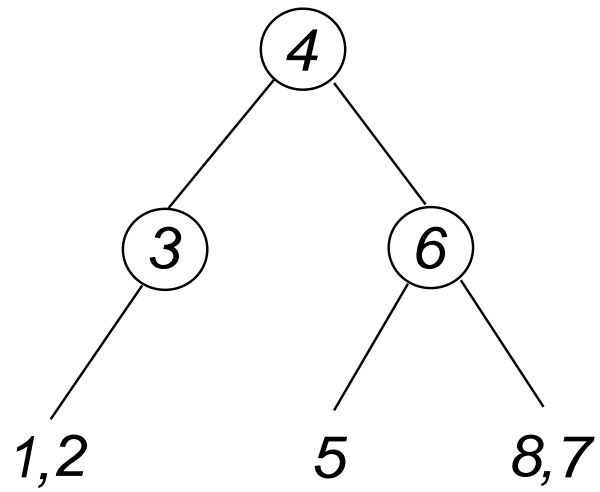
Sorting of data

6,3,5,1,8,2,7



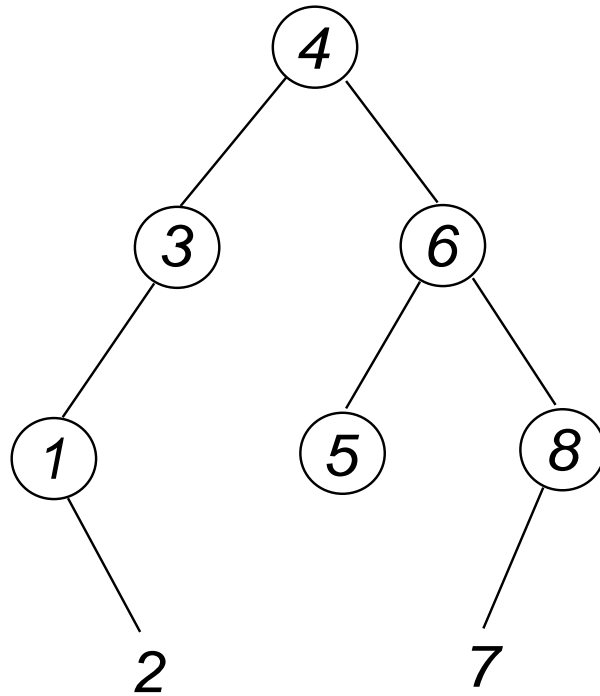
Quicksort

Sorting of data



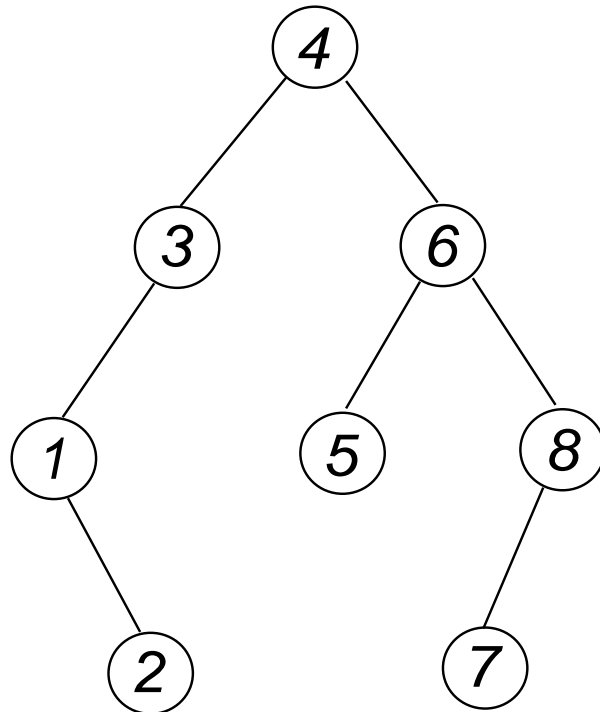
Quicksort

Sorting of data



Quicksort

Sorting of data



Binary Search Tree

Storing of data

4,6,3,5,1,8,2,7

Binary Search Tree

Storing of data

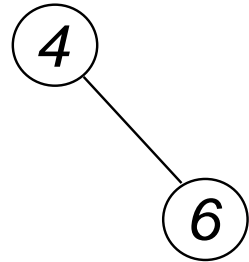
6,3,5,1,8,2,7

4

Binary Search Tree

Storing of data

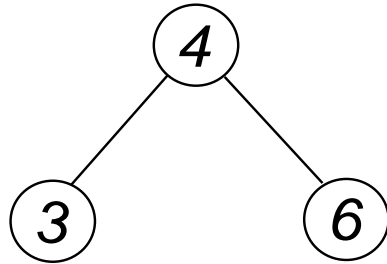
3,5,1,8,2,7



Binary Search Tree

Storing of data

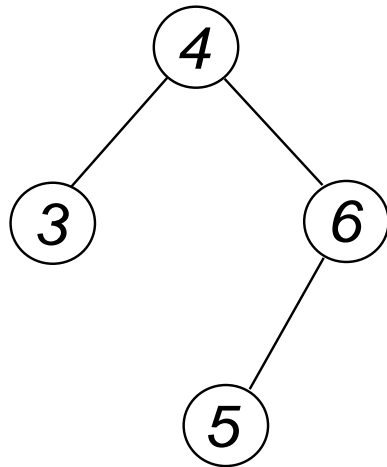
5,1,8,2,7



Binary Search Tree

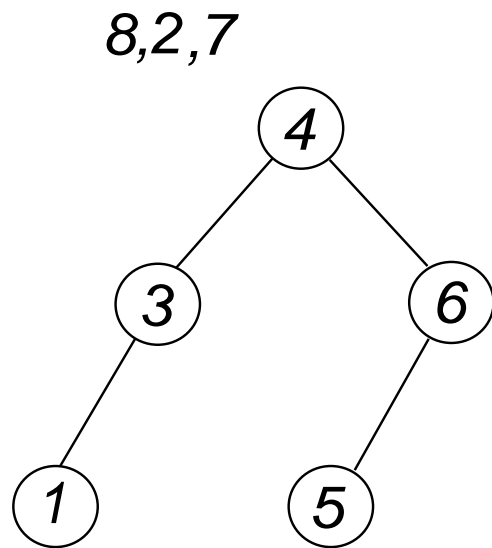
Storing of data

1,8,2,7



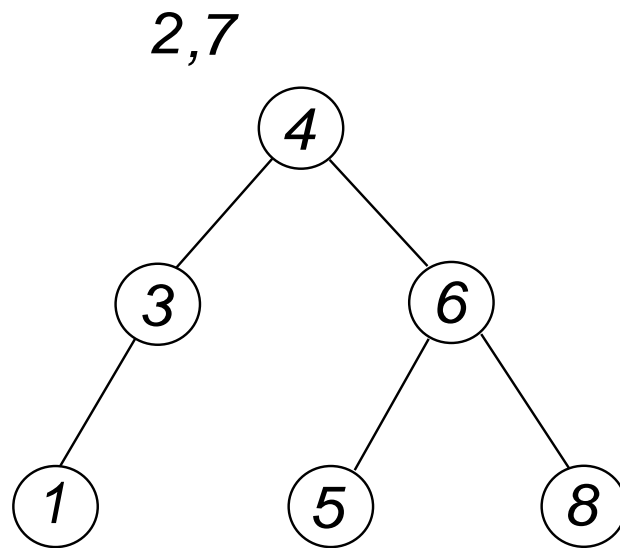
Binary Search Tree

Storing of data



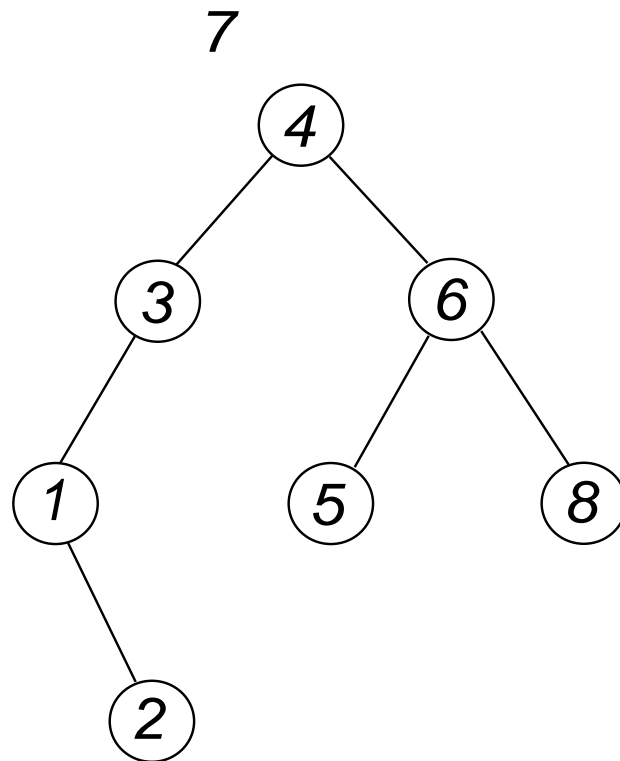
Binary Search Tree

Storing of data



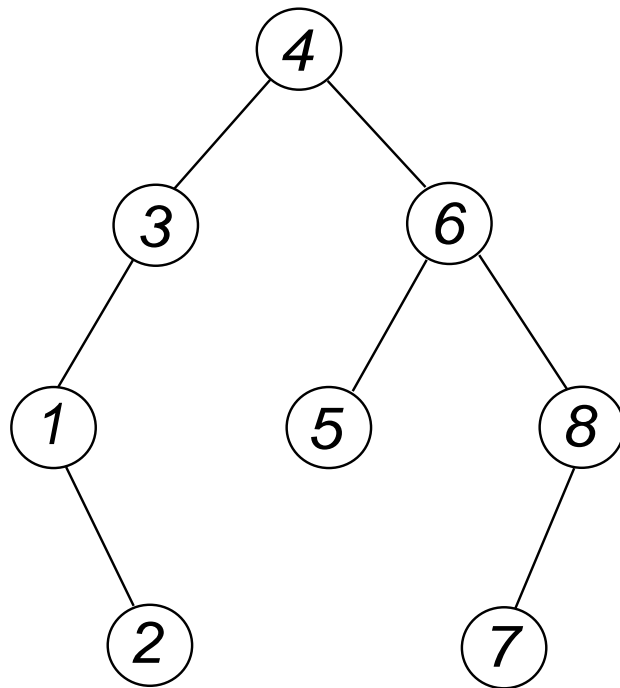
Binary Search Tree

Storing of data



Binary Search Tree

Storing of data



Quicksort

Median of 3

4,6,3,5,1,8,2,7

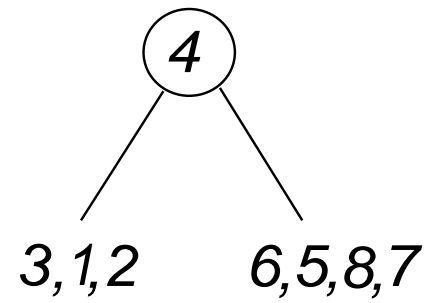
Quicksort

Median of 3

↓
4,6,3,5,1,8,2,7

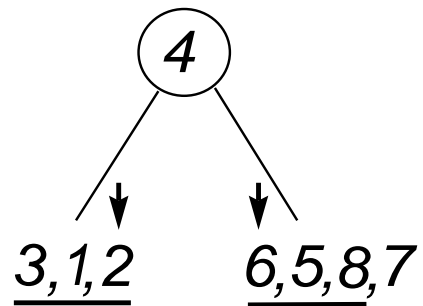
Quicksort

Median of 3



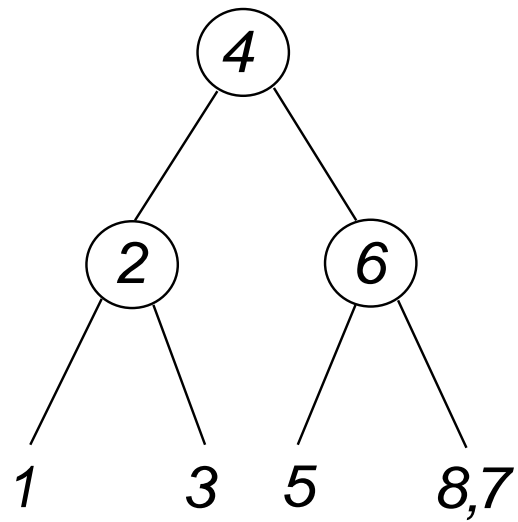
Quicksort

Median of 3



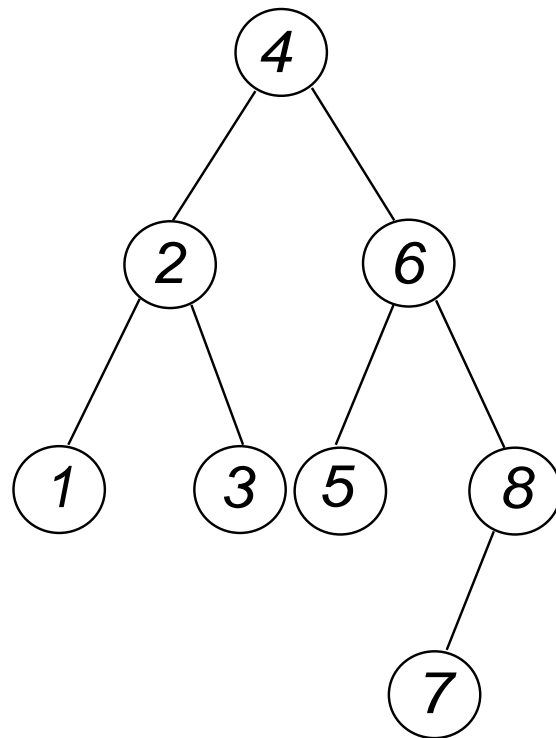
Quicksort

Median of 3



Quicksort

Median of 3



Probabilistic Model

Every permutation on the data $\{1, 2, \dots, n\}$ ist equally likely

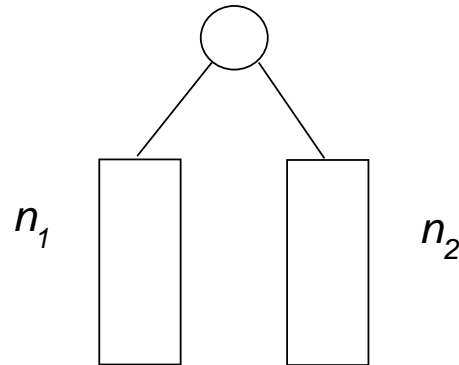
→ probability distribution on binary (m -ary) trees of size n

→ all tree parameters are **random variables**

Probabilistic Model

Recursive structure

Subtrees have the same structure:
($n = n_1 + n_2 + 1$).



Splitting probabilities: p_{n_1, n_2}

Quicksort:
$$p_{n_1, n_2} = \frac{1}{n}$$

Median of 3:
$$p_{n_1, n_2} = \frac{n_1 n_2}{\binom{n}{3}}$$

Probabilistic Model

General Model

$m \geq 2, t \geq 0$... given integers n keys (data)

- If $n \geq m$, we **randomly select** $m - 1$ **pivots** $x_1 < x_2 < \dots < x_{m-1}$.
- The pivots are stored in the **root**.
- The remaining $n - m + 1$ keys are divided into m **subsets** I_1, \dots, I_m :
$$I_1 := \{x_i : x_i < x_1\}, I_2 := \{x_i : x_1 < x_i < x_2\}, \dots, I_m := \{x_i : x_{m-1} < x_i\}.$$
- Apply this procedure **recursively** to I_1, I_2, \dots, I_m .

Probabilistic Model

General Splitting Probabilities

$\mathbf{V}_n = (V_{n,1}, V_{n,2}, \dots, V_{n,m})$.. random splitting vector

$V_{n,k} := |I_k|$... number of keys in the k th subset
(= the number of nodes in the k th subtree of the root)

$$V_{n,1} + V_{n,2} + \dots + V_{n,m} = n - (m - 1) = n + 1 - m$$

$$\mathbb{P}\{\mathbf{V}_n = (n_1, \dots, n_m)\} = \frac{\binom{n_1}{t} \dots \binom{n_m}{t}}{\binom{n}{mt+m-1}}$$

$$(n_1 + n_2 + \dots + n_m = n - m + 1)$$

Quicksort: $m = 2, t = 0$

Median of 3: $m = 2, t = 1$

Probabilistic Model

One-dimensional projection

$$\mathbb{P}\{V_{n,j} = \ell\} = \frac{\binom{\ell}{t} \binom{n-\ell-1}{(m-1)t+m-2}}{\binom{n}{mt+m-1}}$$

Profile of Trees

Parameters of interest:

- **Depth** of a random node: D_n
- **Internal path length**: I_n (sum of all distances to the root)
- **Height** H_n
- **Profile** $X_{n,k}$ (number of nodes at depth k)

Remark:

Number of comparisons in Quicksort = internal path length I_n

Profile of Trees

Relations to the profile $X_{n,k}$:

- $\Pr\{D_n = k\} = \frac{1}{n} \mathbf{E} X_{n,k}$
- $I_n = \sum_{k \geq 0} k X_{n,k}$
- $H_n = \max\{k \geq 0 : X_{n,k} > 0\}$
- The profile describes the **shape** of the tree.

Profile of Trees

Recursive relation:

$$X_{n,k} \stackrel{d}{=} X_{V_{n,1},k-1}^{(1)} + X_{V_{n,2},k-1}^{(2)} + \cdots + X_{V_{n,m},k-1}^{(m)}$$

$(X_{n,k}^{(j)})_{k \geq 0}, j = 1, \dots, m$... independent copies of $X_{n,k}$

Profile of Trees

Expected Profile

$$F(\theta) := \frac{t!}{m(mt + m - 1)!} (\theta + t)(\theta + t + 1) \cdots (\theta + mt + m - 2),$$

$\lambda_1(z)$, $\lambda_2(z)$, ..., $\lambda_{(m-1)(t+1)}(z)$... roots of $F(\theta) = z$:

$$\Re(\lambda_1(z)) \geq \Re(\lambda_2(z)) \geq \dots$$

$\beta(\alpha) > 0$... defined by $\beta(\alpha)\lambda'_1(\beta(\alpha)) = \alpha$.

$$\alpha = \alpha_0 := \left(\frac{1}{t+1} + \frac{1}{t+2} + \cdots + \frac{1}{(t+1)m-1} \right)^{-1}$$

Profile of Trees

Expected Profile

$$k = \alpha \log n$$

- $0 < \alpha = k / \log n < \alpha_0$:

$$\mathbf{E} X_{n,k} \sim (m - 1)m^k.$$

- $\alpha = k / \log n > \alpha_0$:

$$\mathbf{E} X_{n,k} \sim \frac{E(\beta(\alpha))n^{\lambda_1(\beta(\alpha)) - \alpha \log(\beta(\alpha)) - 1}}{\sqrt{2\pi(\alpha + \beta(\alpha)^2 \lambda_1''(\beta(\alpha))) \log n}}$$

for some continuous function $E(z)$

Note: $m^k = n^{\alpha \log m}$

Profile of Trees

Expected Profile

$$\alpha_{\max} := \left(\frac{1}{t+2} + \frac{1}{t+3} + \dots + \frac{1}{(t+1)m} \right)^{-1}$$

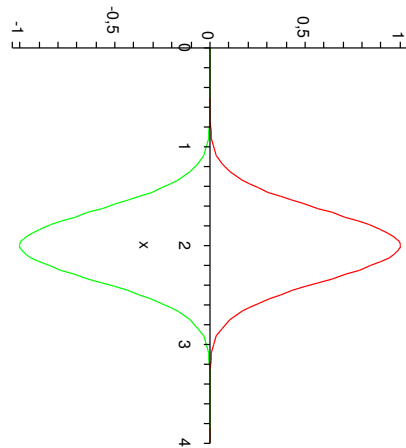
$$\mathbf{E} X_{n,k} \sim \frac{n}{\sqrt{2\pi(\alpha_{\max} + \lambda_1''(1)) \log n}} \exp \left(-\frac{(k - \alpha_{\max} \log n)^2}{2(\alpha_{\max} + \lambda_1''(1)) \log n} \right)$$

(\implies CLT for depth D_n)

Profile of Trees

The average profile: $m = 2, t = 0$ (special case)

$$\mathbf{E} X_{n,k} = \frac{n}{\sqrt{4\pi \log n}} \left(e^{-\frac{(k-2 \log n)^2}{4 \log n}} + \mathcal{O}\left(\frac{1}{\sqrt{\log n}}\right) \right).$$



Profile of Trees

Theorem 1 $m \geq 2, t \geq 0 \dots$ given integers

$(X_{n,k})_{k \geq 0} \dots$ random profile

$$I = \{\beta > 0 : 1 < \lambda_1(\beta^2) < 2\lambda_1(\beta) - 1\},$$

$$I' = \{\beta \lambda_1'(\beta) : \beta \in I\}$$

$$\boxed{\beta(\alpha) \lambda_1'(\beta(\alpha)) = \alpha.}$$

\implies

$$\boxed{\left(\frac{X_{n, \lfloor \alpha \log n \rfloor}}{\mathbf{E} X_{n, \lfloor \alpha \log n \rfloor}}, \alpha \in I' \right) \xrightarrow{d} (Y(\beta(\alpha)), \alpha \in I')}$$

in $D(I')$ (Skorohod topology).

Profile of Trees

Random analytic functions

$B \subseteq \mathbb{C}$, $(I \subseteq B)$ $Y(z)$... random analytic function on B

$$Y(z) \stackrel{d}{=} zV_1^{\lambda_1(z)-1}Y^{(1)}(z) + zV_2^{\lambda_1(z)-1}Y^{(2)}(z) \dots + zV_m^{\lambda_1(z)-1}Y^{(m)}(z)$$

$Y^{(j)}(z)$... independent copies of $Y(z)$

$\mathbf{V} = (V_1, V_2, \dots, V_m)$... random vector supported on the simplex

$\Delta = \{(s_1, \dots, s_m) : s_j \geq 0, s_1 + \dots + s_m = 1\}$ with density

$$f(s_1, \dots, s_m) = \frac{((t+1)^m - 1)!}{(t!)^m} (s_1 \cdots s_m)^t.$$

$\mathbf{V}, Y^{(1)}(z), \dots, Y^{(m)}(z)$... independent.

Profile of Trees

Profile Polynomials

$$W_n(z) := \sum_k X_{n,k} z^k$$

$$X_{n,k} \stackrel{d}{=} X_{V_{n,1},k-1}^{(1)} + X_{V_{n,2},k-1}^{(2)} + \cdots + X_{V_{n,m},k-1}^{(m)},$$

\implies

$$W_n(z) \stackrel{d}{=} zW_{V_{n,1}}^{(1)}(z) + zW_{V_{n,2}}^{(2)}(z) + \cdots + zW_{V_{n,m}}^{(m)}(z) + m - 1$$

for $n \geq m$

Profile of Trees

Profile Polynomials

Theorem 2 B ... complex region, $(1/m, \beta(\alpha_+)) \in B$,
 $\lambda_1(\beta(\alpha_+)) - \alpha_+ \log(\beta(\alpha_+)) - 1 = 0$.

\implies

$$\left(\frac{W_n(z)}{\mathbf{E} W_n(z)}, z \in B \right) \xrightarrow{d} (Y(z), z \in B)$$

in $\mathcal{H}(B)$.

Remark Th. 2 \implies Th. 1

Profile of Trees

Profile Polynomials

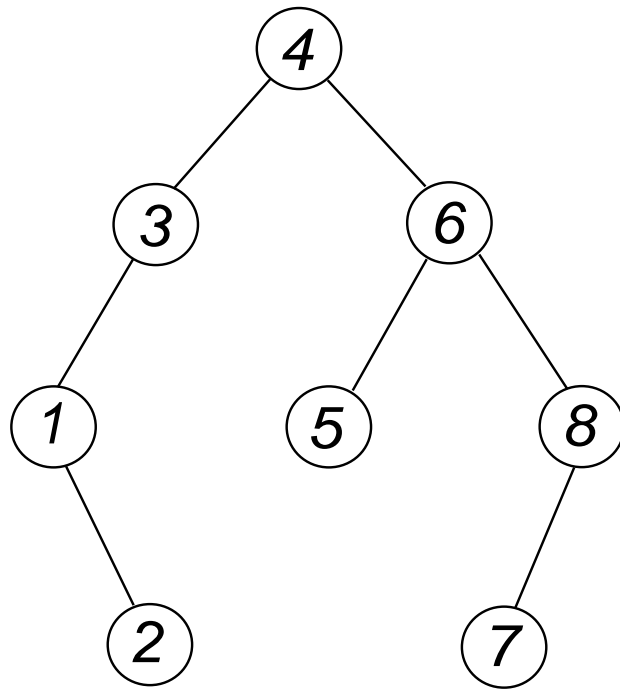
$(X_{n,k})$... random profile

$\implies W_n(z) := \sum_{k \geq 0} X_{n,k} z^k$... random analytic function

$\implies \frac{W_n(z)}{\mathbf{E} W_n(z)}$... random analytic function

Profile of Trees

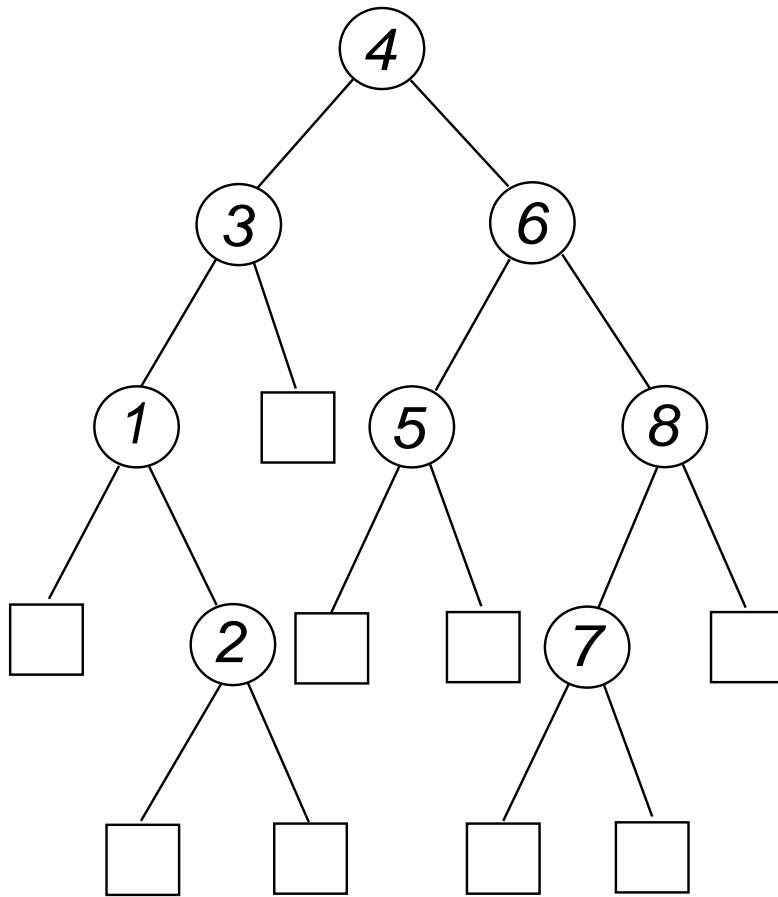
External profile of binary search trees



adding free places

Profile of Trees

External profile of binary search trees



□ ... free place

Profile of Trees

External profile of binary search trees

$Y_{n,k}$ = number of free (= external) nodes at level k ,

$X_{n,k}$ = number of (internal) nodes at level k .

$$X_{n,k} = \sum_{j>k} 2^{k-j} Y_{n,j}$$

Profile of Trees

External profile of binary search trees

$$W_n^{\text{ext}}(z) = \sum_{k \geq 0} Y_{n,k} z^k$$

Lemma 1 *The normalized external profile polynomials*

$$M_n(z) = \frac{W_n^{\text{ext}}(z)}{\mathbf{E} W_n^{\text{ext}}(z)}$$

*constitute a **martingale** with respect to the natural filtration that is induced by the growing tree process (T_n) .*

Remark $\mathbf{E} W_n^{\text{ext}}(z) = \binom{2z+n-1}{n}$

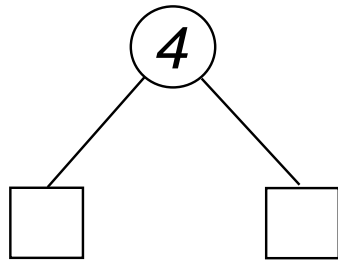
Profile of Trees

Growing tree process



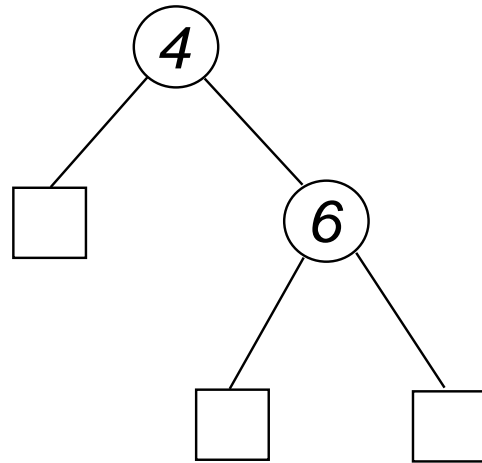
Profile of Trees

Growing tree process



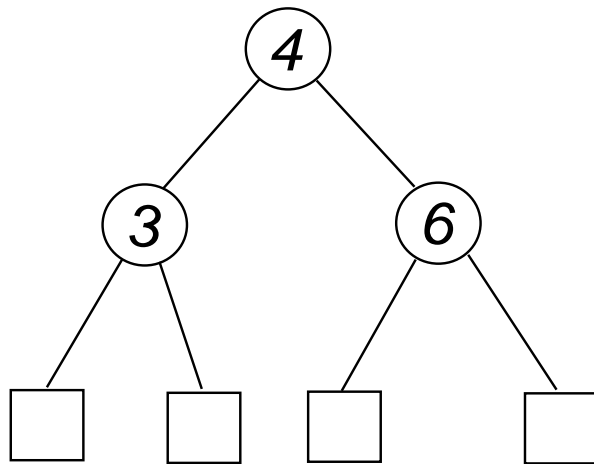
Profile of Trees

Growing tree process



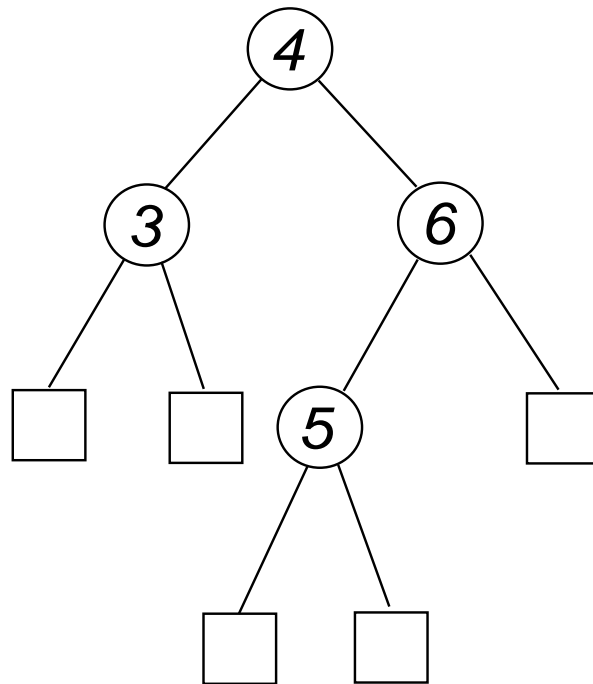
Profile of Trees

Growing tree process



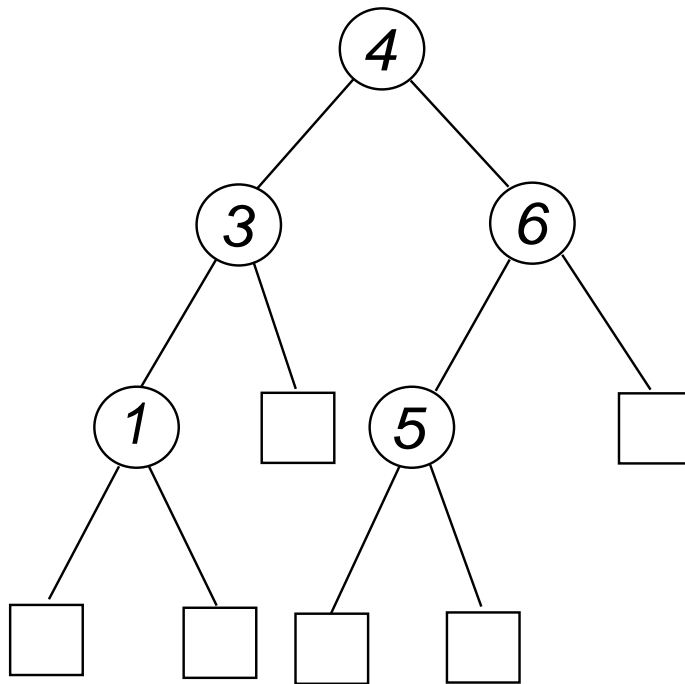
Profile of Trees

Growing tree process



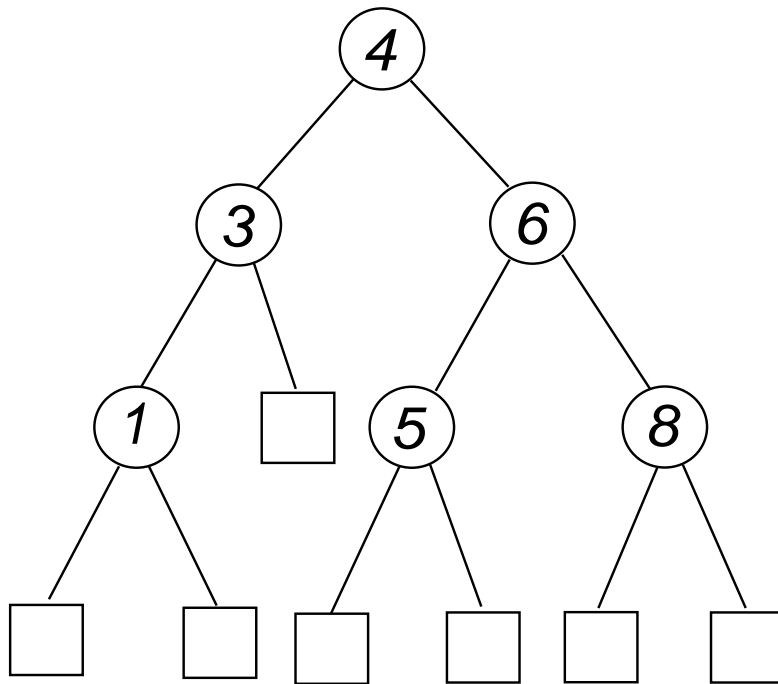
Profile of Trees

Growing tree process



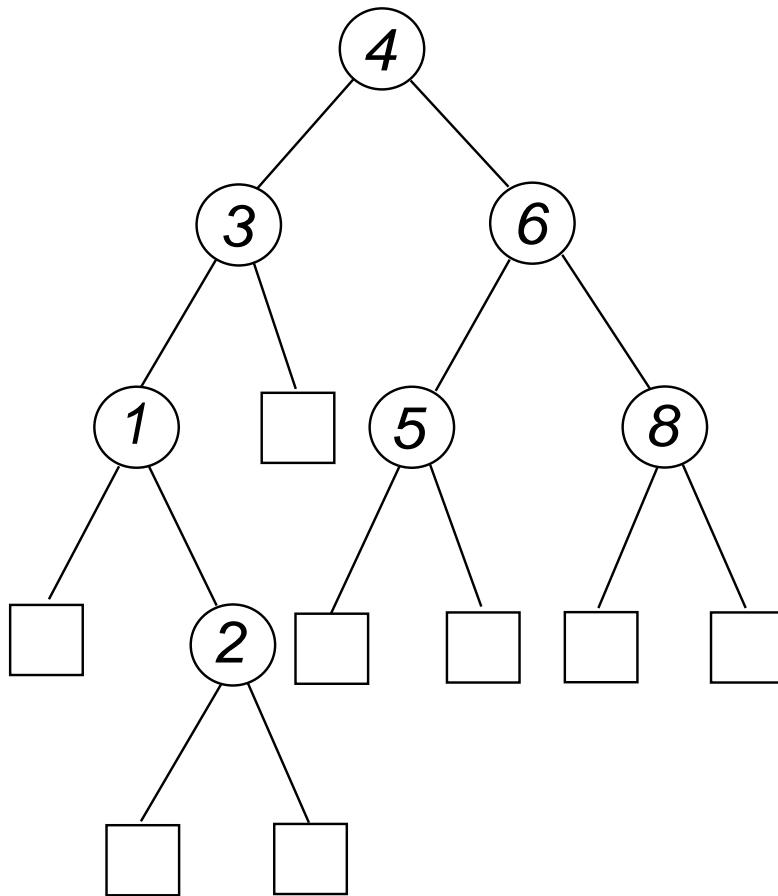
Profile of Trees

Growing tree process



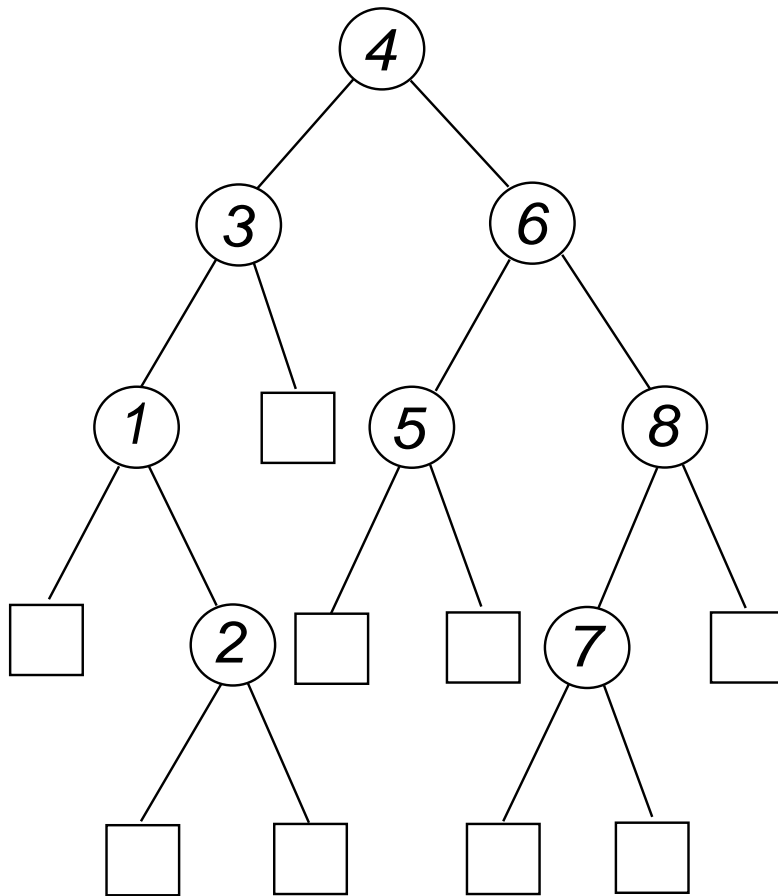
Profile of Trees

Growing tree process



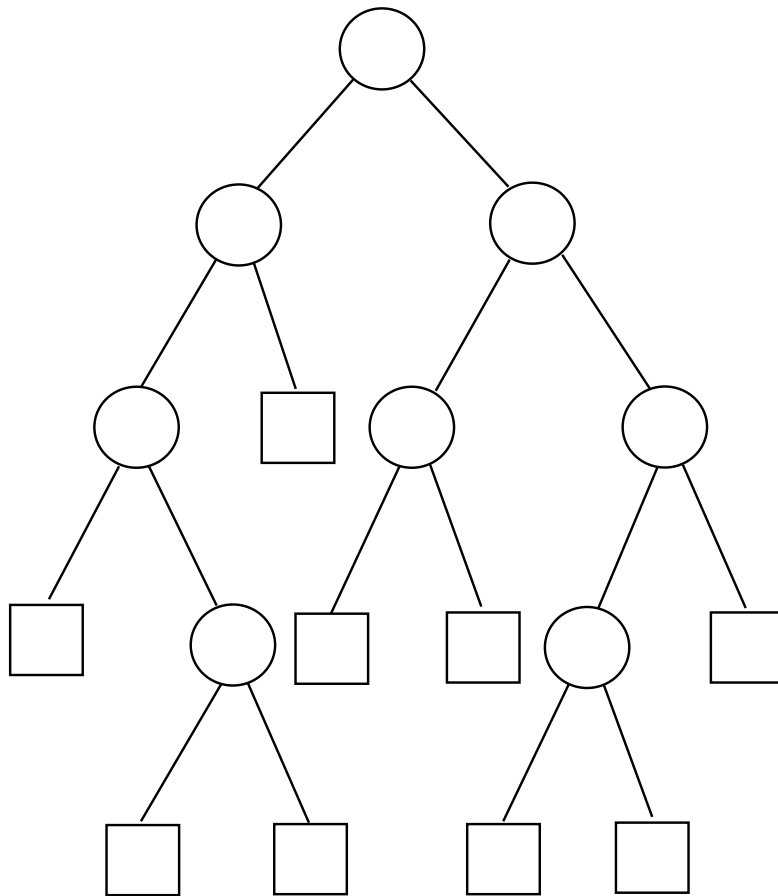
Profile of Trees

Growing tree process



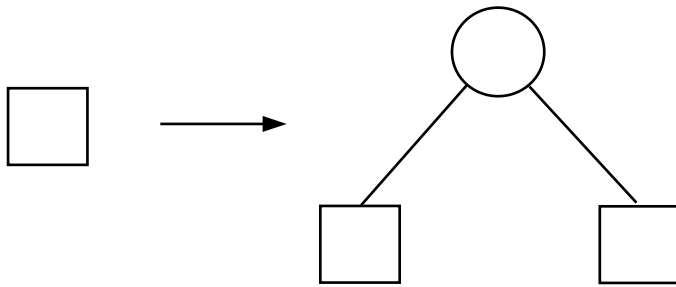
Profile of Trees

Growing tree process



Profile of Trees

Growing tree process



Unfortunately this procedure only works for $m = 2$ and $t = 0$.
In all other cases there is **no** corresponding **martingale**.

Function Spaces

$D \subseteq \mathbb{C}$... complex domain

- $\mathcal{H}(D)$... space of **all analytic functions on D** with the topology of uniform convergence on compact sets.

This topology can be defined by the family of seminorms $f \mapsto \sup_K |f|$, where K ranges over the compact subsets of D .

$\mathcal{H}(D)$ is a **Fréchet space**, i.e. a locally convex space with a topology that can be defined by a complete metric, and it has (by Montel's theorem on normal families) the property that every closed bounded subset is compact.

The topology is separable (for example, by regarding $\mathcal{H}(D)$ as a subspace of $C_0^\infty(D)$).

- $\mathcal{B}(D)$... **Bergman space** of all **square integrable analytic functions on D** , equipped with the norm given by $\|f\|_{\mathcal{B}(D)}^2 = \int_D |f(z)|^2 dm(z)$, where m is the two-dimensional Lebesgue measure.

$\mathcal{B}(D)$ is a **separable Hilbert space** since it can be regarded as a closed subspace of $L^2(\mathbb{R}^2)$.

Function Spaces

Lemma 2 *The embedding $\mathcal{B}(D) \rightarrow \mathcal{H}(D)$ is continuous.*

\implies

Convergence in distribution in $\mathcal{B}(D)$ implies convergence in $\mathcal{H}(D)$.

Lemma 3 *$D' \subset D$ subdomain of D .*

Then the restriction mappings $\mathcal{H}(D) \rightarrow \mathcal{H}(D')$ and $\mathcal{B}(D) \rightarrow \mathcal{B}(D')$ are continuous.

\implies

Convergence in distribution in $\mathcal{H}(D)$ or $\mathcal{B}(D)$ implies convergence (of the restrictions) in $\mathcal{H}(D')$ or $\mathcal{B}(D')$, respectively.

Function Spaces

Local – global convergence

Theorem 3 $D \subseteq \mathbb{C}$... complex domain,
 (W_n) ... sequence of random analytic functions on D .

For all $x \in D$, there is an open subdomain D_x with $x \in D_x \subset D$ and a random analytic function Z_x on D_x such that, as $n \rightarrow \infty$,

$$W_n \xrightarrow{d} Z_x \quad \text{in } \mathcal{H}(D_x).$$

\implies There exists a random analytic function Z on D such that

$$W_n \xrightarrow{d} Z \quad \text{in } \mathcal{H}(D)$$

and the restriction $Z|_{D_x} \stackrel{d}{=} Z_x$ for every x .

Function Spaces

Zolotarev metric

B ... Banach space, $s = m + \alpha > 0$... real number, $m \in \mathbb{Z}$, $0 < \alpha \leq 1$

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

X, Y ... random variables with values in B ,

$\mathcal{L}(X), \mathcal{L}(Y)$... laws of X, Y

Zolotarev metric ζ_s :

$$\zeta_s(X, Y) := \zeta_s(\mathcal{L}(X), \mathcal{L}(Y)) := \sup_{f \in \mathcal{F}_s} |\mathbf{E}(f(X) - f(Y))|.$$

Remark: $\mathbf{E} \|X\|^s < \infty$, $\mathbf{E} \|Y\|^s < \infty$, and $\mathbf{E} X^{\otimes k} = \mathbf{E} Y^{\otimes k}$ for all $k \leq m$
 $\implies \zeta_s(X, Y) < \infty$. ($\langle g, \mathbf{E} X^{\otimes k} \rangle = \mathbf{E} g(X, X, \dots, X)$, g multilinear)

Function Spaces

Zolotarev metric

$$\mathbf{z} = (z_1, \dots, z_m), \quad z_k \in B^{\otimes k}, \quad k = 1, \dots, m,$$

$$\mathcal{P}_{s,\mathbf{z}}(B) := \{\mathcal{L}(X) : \mathbf{E} \|X\|^s < \infty, \mathbf{E} X^{\otimes k} = z_k, k = 1, \dots, m\},$$

ζ_s is finite on each $\mathcal{P}_{s,\mathbf{z}}(B)$, and is also a **semi-metric** on $\mathcal{P}_{s,\mathbf{z}}(B)$.

Function Spaces

Zolotarev metric

Theorem 4 H ... separable Hilbert space, $s > 0$

$\mathbf{z} = (z_1, \dots, z_m)$, $z_k \in B^{\otimes k}$, $k = 1, \dots, m$,

$\implies \zeta_s$ is a **complete metric** on the set $\mathcal{P}_{s,\mathbf{z}}(H)$

X_n, X ... H -valued random variables with $\mathcal{L}(X), \mathcal{L}(Y) \in \mathcal{P}_{s,\mathbf{z}}(H)$:

$$\boxed{\zeta_s(X_n, X) \rightarrow 0 \implies X_n \xrightarrow{d} X}.$$

Remark: X_n is tight

Function Spaces

Minimal L^s -metric ℓ_s

$X, Y \dots$ random variables with values in a Banach space B

$$\mathbf{E} \|X\|^s, \mathbf{E} \|Y\|^s < \infty$$

$$\ell_s(X, Y) := \inf\{(\mathbf{E} \|X' - Y'\|^s)^{(1/s) \wedge 1} : \mathcal{L}(X') = \mathcal{L}(X), \mathcal{L}(Y') = \mathcal{L}(Y)\}.$$

Function Spaces

Minimal L^s -metric ℓ_s

Lemma 4 $\mathcal{L}(X), \mathcal{L}(Y) \in \mathcal{P}_{s,z}(B)$

$s > 1 \implies$

$$\zeta_s(X, Y) \leq \left((\mathbf{E} \|X\|^s)^{1-1/s} + (\mathbf{E} \|Y\|^s)^{1-1/s} \right) \ell_s(X, Y).$$

$0 < s \leq 1 \implies$

$$\zeta_s(X, Y) \leq \ell_s(X, Y).$$

Contraction Method

Fixed point equation

$\mathcal{P}(H)$... set of all probability distributions on Hilbert space H .

A_1^*, \dots, A_m^* ... random linear operators in H ,

b^* ... random variable in H

$$T : \mathcal{P}(H) \rightarrow \mathcal{P}(H)$$

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

$$\mathcal{L}(Z^{(r)}) = \mathcal{L}(Z) \text{ for } r = 1, \dots, m,$$

$(A_1^*, \dots, A_m^*, b^*), Z^{(1)}, \dots, Z^{(m)}$ independent.

Contraction Method

Fixed point equation

A ... linear operator A in H

$$\|A\|_{\text{op}} := \sup_{\|x\|=1} \|Ax\|$$

s -integrable means $\mathbf{E} \|A\|_{\text{op}}^s < \infty$.

Lemma 5 A_1^*, \dots, A_m^*, b^* be as above and s -integrable for some $0 < s \leq 2$.

$$0 < s \leq 1 \implies T(\mathcal{P}_s) \subseteq \mathcal{P}_s.$$

$$1 < s \leq 2 \text{ and } \mathbf{E} b^* = 0 \implies T(\mathcal{P}_{s,0}) \subseteq \mathcal{P}_{s,0}.$$

Contraction Method

Contraction

Lemma 6 Let A_1^*, \dots, A_m^*, b^* be and s -integrable for some $0 < s \leq 2$:
 $\mathbf{E} \|A_r^*\|_{\text{op}}^s < \infty$.

Further assume that

$$\mathbf{E} \sum_{r=1}^m \|A_r^*\|_{\text{op}}^s < 1.$$

$0 < s \leq 1 \implies$ the restriction of T to \mathcal{P}_s is a **strict contraction**.

$1 < s \leq 2, \mathbf{E} b^* = 0 \implies$ the restriction of T to $\mathcal{P}_{s,0}$ is a **strict contraction**.

Remark: $\zeta_s(A(X), A(Y)) \leq \|A\|_{\text{op}}^s \zeta_s(X, Y)$

Contraction Method

A recurrence:

$(X_n)_{n \geq 0}$... sequence of random variables in H

$$X_n \stackrel{d}{=} \sum_{r=1}^m A_r^{(n)} \left(X_{I_r^{(n)}}^{(r)} \right) + b^{(n)}, \quad n \geq n_0,$$

$A_r^{(n)}$... random linear operators in H ,

$b^{(n)}$... random variable in H ,

$I^{(n)} = (I_1^{(n)}, \dots, I_m^{(n)})$.. random integer vector ($I_r^{(n)} \in \{0, \dots, n\}$),

$(X_j^{(1)}), \dots, (X_j^{(m)}), (A_1^{(n)}, \dots, A_m^{(n)}, b^{(n)}, I^{(n)})$... independent,

$\mathcal{L}(X_j^{(r)}) = \mathcal{L}(X_j)$ for all r and j .

Contraction Method

Theorem 5 $(X_n)_{n \geq 0}$... sequence of random variables in H (as above)

All quantities being s -integrable for some $0 < s \leq 2$.

(For $1 < s \leq 2$ assume further $\mathbf{E} X_n = 0$)

There exists A_1^*, \dots, A_m^*, b^* with

$$\mathbf{E} \|A_r^{(n)} - A_r^*\|_{\text{op}}^s \rightarrow 0, \quad \mathbf{E} \|b^{(n)} - b^*\|^s \rightarrow 0,$$

$$\mathbf{E} \sum_{r=1}^m \|A_r^*\|_{\text{op}}^s < 1,$$

$$\mathbf{E} \left[\mathbf{1}_{\{I_r^{(n)} \leq \ell\}} \|A_r^{(n)}\|_{\text{op}}^s \right] \rightarrow 0 \quad (r, \ell \geq 1 \text{ integers})$$

Let $\mathcal{L}(X)$ be the unique fixed point of T in $\mathcal{P}_s(H)$ for $0 < s \leq 1$ and in $\mathcal{P}_{s,0}(H)$ for $1 < s \leq 2$.

$$\implies \boxed{\zeta_s(X_n, X) \rightarrow 0} \quad n \rightarrow \infty.$$

Contraction Method

Analytic functions

$(X_n)_{n \geq 0}$... sequence of random analytic functions in a domain $D \subseteq \mathbf{C}$

$$\boxed{X_n \stackrel{d}{=} \sum_{r=1}^m A_r^{(n)} \cdot X_{I_r^{(n)}}^{(r)} + b^{(n)}}, \quad n \geq n_0, \quad (1)$$

$A_1^{(n)}, \dots, A_m^{(n)}, b^{(n)}$... random analytic functions in D ,

$I^{(n)} = (I_1^{(n)}, \dots, I_m^{(n)})$... random integers vector ($I_r^{(n)} \in \{0, \dots, n\}$);

$\mathcal{L}(X_j^{(r)}) = \mathcal{L}(X_j)$ for all r and j ,

$(A_1^{(n)}, \dots, A_m^{(n)}, b^{(n)}, I^{(n)}), (X_j^{(1)}), \dots, (X_j^{(m)})$... independent.

Contraction Method

Contraction for random analytic functions

$$T : \mathcal{P}(\mathcal{H}(\tilde{D})) \rightarrow \mathcal{P}(\mathcal{H}(\tilde{D}))$$
$$\mathcal{L}(Z) \mapsto \mathcal{L} \left(\sum_{r=1}^m A_r^* \cdot Z^{(r)} + b^* \right),$$

A_1^*, \dots, A_m^*, b^* ... random analytic functions in \tilde{D} ,

$\mathcal{L}(Z^{(r)}) = \mathcal{L}(Z)$ for $r = 1, \dots, m$,

$(A_1^*, \dots, A_m^*, b^*), Z^{(1)}, \dots, Z^{(m)}$... independent.

Contraction Method

Theorem 6 $0 < s \leq 2$, (X_n) as above with $\mathbf{E} X_n(z) = 0$
 $A_r^{(n)}$, $b^{(n)}$... analytic funct. of $z \in D$, $\mathbf{E} |A_r^{(n)}(z)|^s < \infty$, $\mathbf{E} |b^{(n)}(z)|^s < \infty$

Suppose that there exist A_1^*, \dots, A_m^* and b^* in D and a connected subset $\Delta \subseteq D$ such that for each $x \in \Delta$ there exists a neighbourhood $U_x \subseteq D$ of x and a number $s(x) \leq s$ with

$$\sup_{z \in U_x} \mathbf{E} |A_r^{(n)}(z) - A_r^*(z)|^{s(x)} \rightarrow 0, \quad \sup_{z \in U_x} \mathbf{E} |b^{(n)}(z) - b^*(z)|^{s(x)} \rightarrow 0,$$

$$\sup_{z \in U_x} \mathbf{E} |A_r^*(z)|^{s(x)} < \infty,$$

$$\mathbf{E} \sum_{r=1}^m |A_r^*(x)|^{s(x)} < 1,$$

$$\sup_{z \in U_x} \mathbf{E} \left[\mathbf{1}_{\{I_r^{(n)} \leq \ell\}} |A_r^{(n)}(z)|^{s(x)} \right] \rightarrow 0,$$

Let $\mathcal{L}(X)$ bet the fixed point of the map T .

$$\implies \boxed{X_n \xrightarrow{d} X} \quad \text{in } \mathcal{H}(\tilde{D})$$

for some domain $\tilde{D} \subseteq D$ with $\Delta \subseteq \tilde{D}$

Contraction Method

Proof of Th. 2

$W_n(z)$... random profile polynomials

$$X_n(z) := \frac{W_n(z) - \mathbf{E} W_n(z)}{\mathbf{E} W_n(z)} = \frac{W_n(z)}{\mathbf{E} W_n(z)} - 1$$

$$\implies X_n(z) \stackrel{d}{=} \sum_{r=1}^m z \frac{G_{V_{n,r}}(z)}{G_n(z)} X_{V_{n,r}}^{(r)} + \frac{1}{G_n(z)} \left(m - 1 - G_n(z) + z \sum_{r=1}^m G_{V_{n,r}}(z) \right).$$

Notation: $G_n(z) := \mathbf{E} W_n(z)$,

Contraction Method

Proof of Th. 2

$$I_r^{(n)} = V_{n,r}$$

$$A_r^{(n)} = z \frac{G_{V_{n,r}}(z)}{G_n(z)},$$

$$b^{(n)} = \frac{1}{G_n(z)} \left(m - 1 - G_n(z) + z \sum_{r=1}^m G_{V_{n,r}}(z) \right)$$

Cauchy Integral

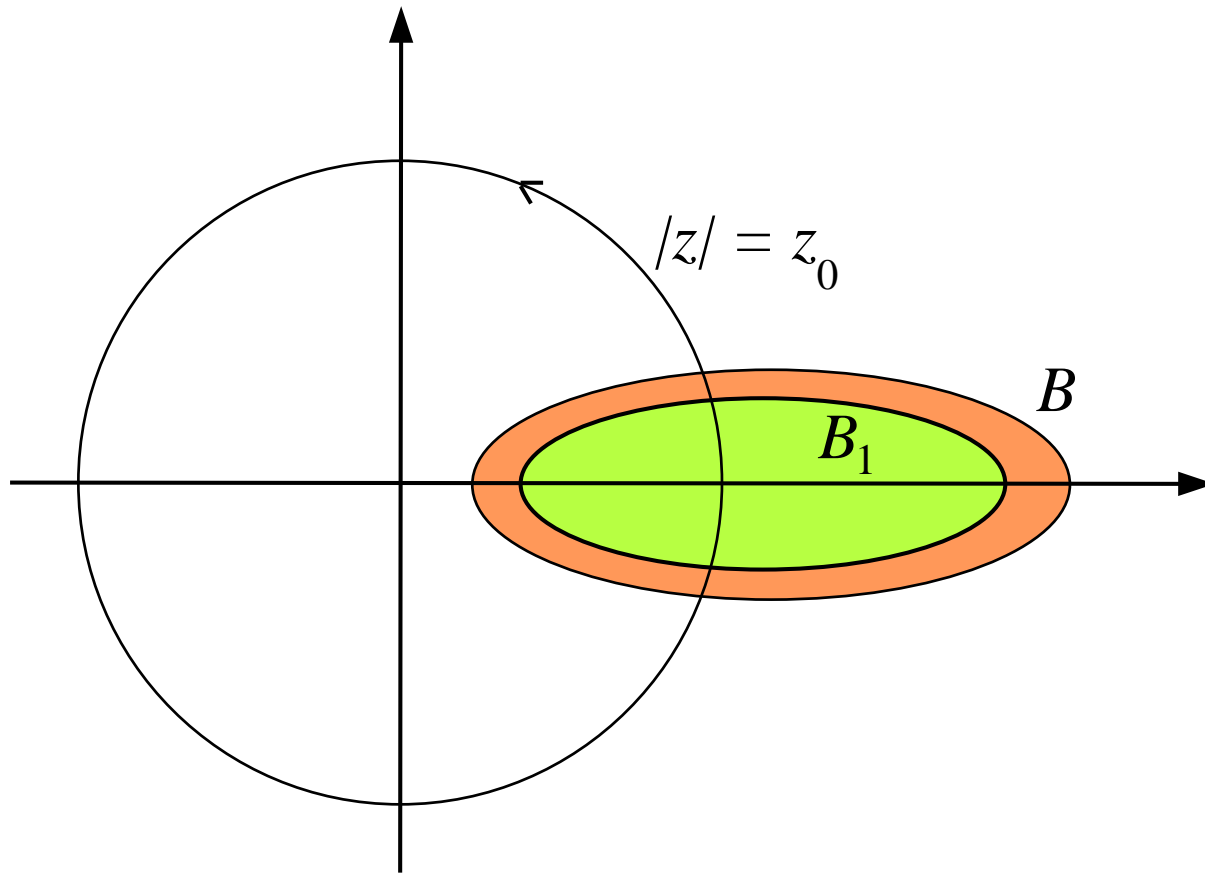
Th. 2 \implies Th. 1

Idea:

$$X_{nk} = \frac{1}{2\pi i} \int_{|z|=z_0} W_n(z) z^{-k-1} dz$$

$$\begin{aligned} \implies \boxed{\frac{X_{nk}}{\mathbf{E} X_{nk}}} &= \frac{1}{\mathbf{E} X_{nk}} \frac{1}{2\pi i} \int_{|z|=z_0} \frac{W_n(z)}{\mathbf{E} W_n(z)} \mathbf{E} W_n(z) z^{-k-1} dz \\ &= \frac{1}{\mathbf{E} X_{nk}} \frac{1}{2\pi i} \int_{|z|=\beta(\alpha), z \in B_1} \boxed{\frac{W_n(z)}{\mathbf{E} W_n(z)}} \mathbf{E} W_n(z) z^{-k-1} dz \\ &\quad + \frac{1}{\mathbf{E} X_{nk}} \frac{1}{2\pi i} \int_{|z|=z_0, z \notin B_1} W_n(z) z^{-k-1} dz. \end{aligned}$$

Cauchy Integral



Cauchy Integral

Continuous operators

$$(k = \lfloor \alpha \log n \rfloor)$$

$$T_n(G)(\alpha) = \frac{1}{\mathbf{E} X_{n, \lfloor \alpha \log n \rfloor}} \frac{1}{2\pi i} \int_{|z|=\beta(\alpha), z \in B_1} G(z) \mathbf{E} W_n(z) z^{-\lfloor \alpha \log n \rfloor - 1} dz, \alpha \in I'_c.$$

$\implies T_n(W_n(z)/\mathbf{E} W_n(z))$ is an approximation to

$$\left(X_{n, \lfloor \alpha \log n \rfloor} / \mathbf{E} X_{n, \lfloor \alpha \log n \rfloor}, \alpha \in I'_c \right).$$

Cauchy Integral

Notation: $\|f\|_E := \sup_E |f|$

Lemma 7

1. *The operators T_n are uniformly continuous with respect to the supremum norm:*

$$\|T_n(F) - T_n(G)\|_{I'_c} \leq C \cdot \|F - G\|_{B_1}.$$

(for some constant $C > 0$ depending on I_c and B_1).

2. $F_n \rightarrow F$ uniformly on $B_1 \implies T_n(F_n) \rightarrow F$ uniformly on I'_c .

Proof method: Saddle point method

Cauchy Integral

Application of Lemma 7:

$$(W_n(z)/\mathbf{E} W_n(z), z \in B_1) \rightarrow (Y(z), z \in B_1)$$

$$\implies T_n(W_n(z)/\mathbf{E} W_n(z)) \rightarrow Y(\beta(\alpha)), \alpha \in I'_c$$

Cauchy Integral

Lemma 8 $I_c \subseteq I = \{\beta > 0 : 1 < \lambda_1(\beta^2) < 2\lambda_1(\beta) - 1\}$ a compact interval.

$$\implies \sup_{\alpha \in I'_c} \left| \frac{\frac{1}{2\pi} \int_{|z|=\beta(\alpha), z \notin B_1} W_n(z) z^{-\lfloor \alpha \log n \rfloor - 1} dz}{\mathbf{E} X_{n, \lfloor \alpha \log n \rfloor}} \right| \xrightarrow{p} 0$$

as $n \rightarrow \infty$.

Proof method: Second moment estimates.

Cauchy Integral

Th. 2 \implies Th. 1:

$$\left(\frac{X_{n, \lfloor \alpha \log n \rfloor}}{\mathbf{E} X_{n, \lfloor \alpha \log n \rfloor}}, \alpha \in I'_c \right) = T_n(W_n(z)/\mathbf{E} W_n(z)) + o_p(1)$$
$$\xrightarrow{d} (Y(\beta(\alpha)) \alpha \in I'_c)$$

Thank You!