

RANDOM TREES – AN ANALYTIC APPROACH

Michael Drmota

Institute of Discrete Mathematics and Geometry

TU Wien

A 1040 Wien, Austria

michael.drmota@tuwien.ac.at

<http://www.dmg.tuwien.ac.at/drmota/>

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References

Books

Michael Drmota,

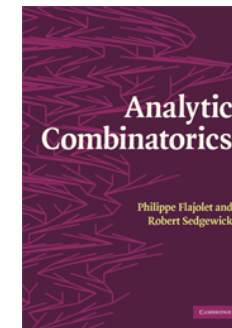
Random Trees, Springer, Wien-New York, 2009.



Philippe Flajolet and Robert Sedgewick,

Analytic Combinatorics, Cambridge University Press, 2009.

(<http://algo.inria.fr/flajolet/Publications/books.html>)



Asymptotic analysis of random objects

Levels of complexity:

1. Asymptotic enumeration
2. Distribution of (shape) parameters
3. Asymptotic shape (= continuous limiting object)

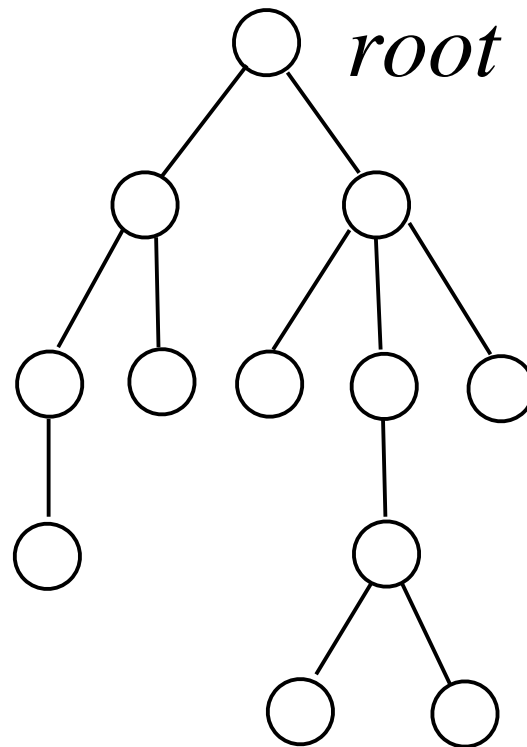
Contents 1

I. COMBINATORIAL RANDOM TREES

- Catalan trees and Cayley trees
- Functional equations and algebraic singularities
- A combinatorial central limit theorem
- The degree distribution of random trees

Random Trees

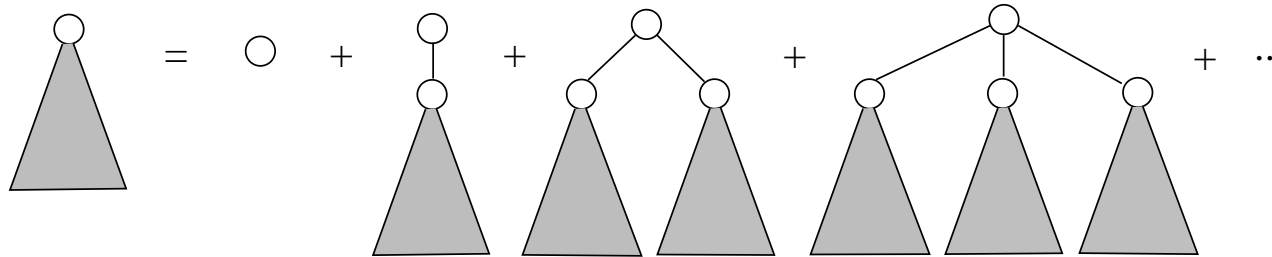
Catalan trees



rooted, ordered (or plane) tree

Random Trees

Catalan trees. g_n = number of Catalan trees of size n ; $G(x) = \sum_{n \geq 1} g_n x^n$



$$G(x) = x(1 + G(x) + G(x)^2 + \dots) = \frac{x}{1 - G(x)}$$

$$G(x) = \frac{1 - \sqrt{1 - 4x}}{2} \implies g_n = \frac{1}{n} \binom{2n - 2}{n - 1} \sim \frac{4^{n-1}}{\sqrt{\pi} \cdot n^{3/2}}$$

(Catalan numbers)

Random Trees

Catalan trees with singularity analysis (to be discussed later)

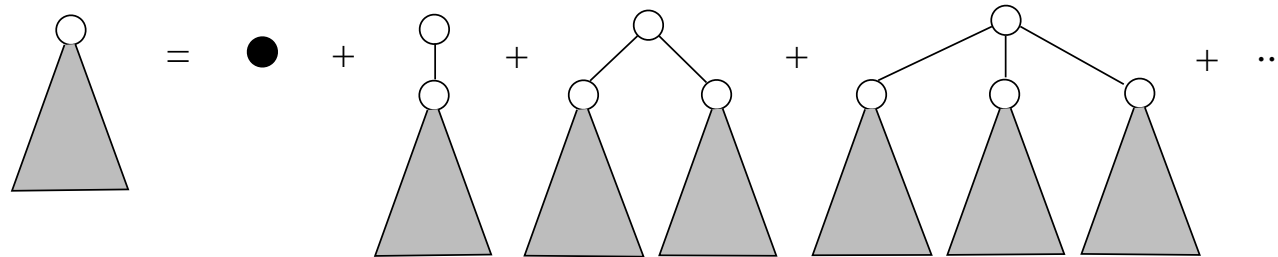
$$G(x) = \frac{1 - \sqrt{1 - 4x}}{2} = \frac{1}{2} - \frac{1}{2}\sqrt{1 - 4x}$$

$$\implies g_n \sim -\frac{1}{2} \cdot \frac{4^n n^{-3/2}}{\Gamma(-\frac{1}{2})} = \frac{4^{n-1}}{\sqrt{\pi} \cdot n^{3/2}}$$

Random Trees

Number of leaves of Catalan trees

$g_{n,k}$ = number of Catalan trees of size n with k leaves.



$$G(x, u) = xu + x(G(x, u) + G(x, u)^2 + \dots) = xu + \frac{xG(x, u)}{1 - G(x, u)}$$

$$\implies G(x, u) = \frac{1}{2} \left(1 + (u - 1)x - \sqrt{1 - 2(u + 1)x + (u - 1)^2 x^2} \right)$$

$$\implies g_{n,k} = \frac{1}{n} \binom{n}{k} \binom{n-1}{k} \sim \frac{4^n}{\pi n^2} \exp\left(-\frac{(k - \frac{n}{2})^2}{\frac{1}{4}n}\right) \quad \text{for } k \approx \frac{n}{2}$$

Random Trees

Number of leaves of Catalan trees

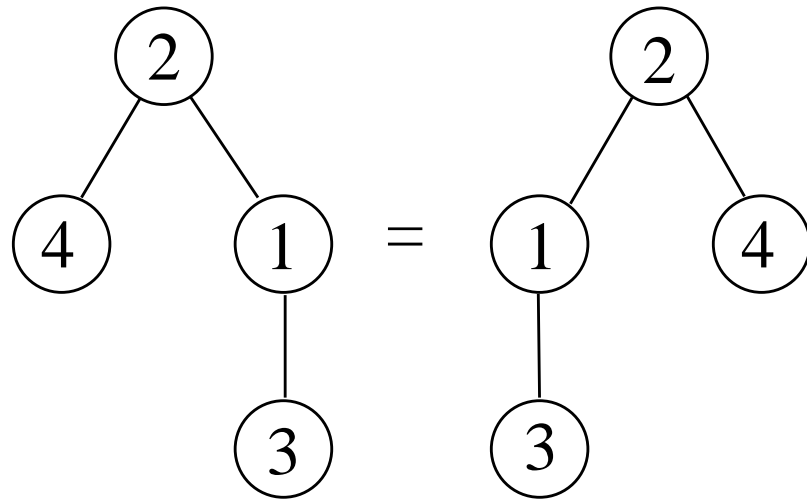
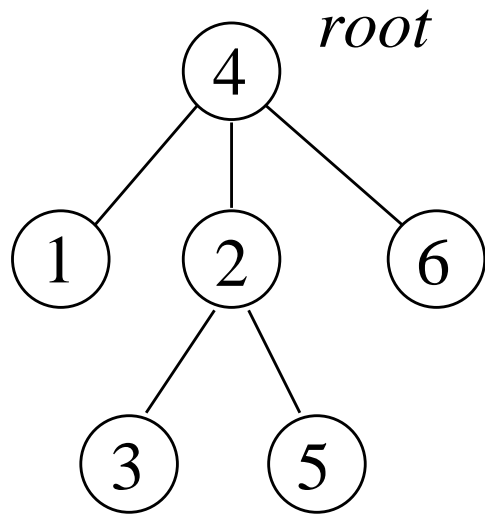
$$G(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}$$

for certain analytic function $g(x, u)$, $h(x, u)$, and $\rho(u)$.

$$\implies g_{n,k} = ???$$

Random Trees

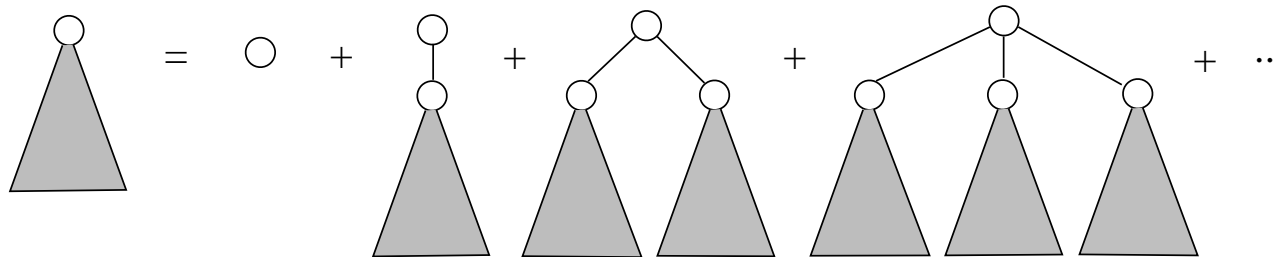
Cayley Trees:



labelled, rooted, unordered (or non-plane) tree

Random Trees

Cayley Trees. r_n = number of Cayley trees of size n ; $R(x) = \sum_{n \geq 1} r_n \frac{x^n}{n!}$



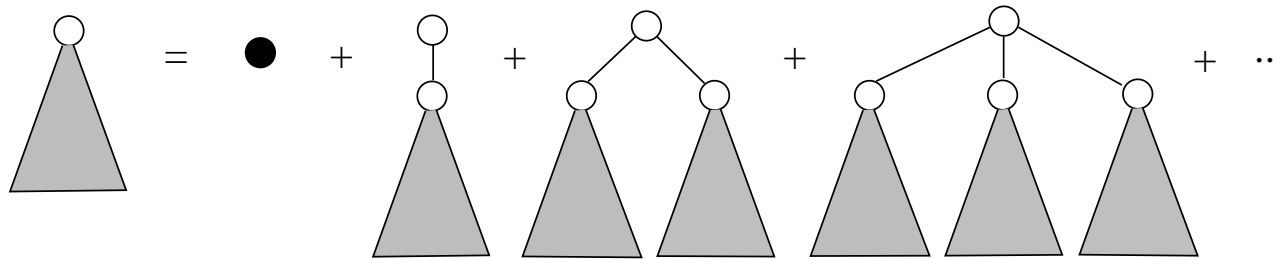
$$R(x) = x \left(1 + R(x) + \frac{R(x)^2}{2!} + \frac{R(x)^3}{3!} + \dots \right) = x e^{R(x)}$$

\implies $r_n = n^{n-1}$... by Lagrange inversion

Random Trees

Number of leaves of Cayley trees

$r_{n,k}$ = number of Cayley trees of size n with k leaves.



$$R(x, u) = xu + x \left(R(x, u) + \frac{R(x, u)^2}{2!} + \frac{R(x, u)^3}{3!} + \dots \right) = xe^{R(x, u)} + x(u - 1)$$

$$\implies R(x, u) = ???$$

Functional equations

Catalan trees: $G(x, u) = xu + xG(x, u)/(1 - G(x, u))$

Cayley trees: $R(x, u) = xe^{R(x, u)} + x(u - 1)$

Recursive structure leads to functional equation for gen. func.:

$$A(x, u) = \Phi(x, u, A(x, u))$$

Functional equations

Linear functional equation: $\Phi(x, u, a) = \Phi_0(x, u) + a\Phi_1(x, u)$

$$\implies A(x, u) = \frac{\Phi_0(x, u)}{1 - \Phi_1(x, u)}$$

Usually these kinds of generating functions are easy to handle, since they are explicit.

Functional equations

Non-linear functional equations: $\Phi_{aa}(x, u, a) \neq 0$.

Suppose that $A(x, u) = \Phi(x, u, A(x, u))$, where $\Phi(x, u, a)$ has a power series expansion at $(0, 0, 0)$ with non-negative coefficients and $\Phi_{aa}(x, u, a) \neq 0$.

Let $x_0 > 0$, $a_0 > 0$ (inside the region of convergence) satisfy the system of equations:

$$a_0 = \Phi(x_0, 1, a_0), \quad 1 = \Phi_a(x_0, 1, a_0).$$

Then there exists analytic function $g(x, u)$, $h(x, u)$, and $\rho(u)$ such that locally

$$A(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}.$$

Functional equations

Idea of the Proof.

Set $F(x, u, a) = \Phi(x, u, a) - a$. Then we have

$$F(x_0, 1, a_0) = 0$$

$$F_a(x_0, 1, a_0) = 0$$

$$F_x(x_0, 1, a_0) \neq 0$$

$$F_{aa}(x_0, 1, a_0) \neq 0.$$

Weierstrass preparation theorem implies that there exist analytic functions $H(x, u, a)$, $p(x, u)$, $q(x, u)$ with $H(x_0, 1, a_0) \neq 0$, $p(x_0, 1) = q(x_0, 1) = 0$ and

$$F(x, u, a) = H(x, u, a) \left((a - a_0)^2 + p(x, u)(a - a_0) + q(x, u) \right).$$

Functional equations

$$F(x, u, a) = 0 \iff \boxed{(a - a_0)^2 + p(x, u)(a - a_0) + q(x, u) = 0}.$$

Consequently

$$\begin{aligned} A(x, u) &= a_0 - \frac{p(x, u)}{2} \pm \sqrt{\frac{p(x, u)^2}{4} - q(x, u)} \\ &= \boxed{g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}}, \end{aligned}$$

where we write

$$\frac{p(x, u)^2}{4} - q(x, u) = K(x, u)(x - \rho(u))$$

which is again granted by the Weierstrass preparation theorem and we set

$$g(x, u) = a_0 - \frac{p(x, u)}{2} \quad \text{and} \quad h(x, u) = \sqrt{-K(x, u)\rho(u)}.$$

Random Trees

Catalan Trees $G(x, u) = xu + \frac{xG(x, u)}{1-G(x, u)}$

$$\implies G(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}$$

$$G(x, 1) = G(x) = g(x, 1) - h(x, 1) \sqrt{1 - \frac{x}{\rho(1)}}, \quad \rho(1) = \frac{1}{4}$$

Cayley Trees $T(x, u) = xe^{T(x, u)} + x(u - 1)$

$$\implies T(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}$$

$$T(x, 1) = T(x) = g(x, 1) - h(x, 1) \sqrt{1 - \frac{x}{\rho(1)}}, \quad \rho(1) = \frac{1}{e}$$

Algebraic Singularities

Singular expansion

$$\begin{aligned} A(x) &= g(x) - h(x) \sqrt{1 - \frac{x}{\rho}} \\ &= \left(g_0 + g_1(x - \rho) + g_2(x - \rho)^2 + \dots \right) \\ &\quad + \left(h_0 + h_1(x - \rho) + h_2(x - \rho)^2 + \dots \right) \sqrt{1 - \frac{x}{\rho}} \\ &= a_0 + a_1 \left(1 - \frac{x}{\rho} \right)^{\frac{1}{2}} + a_2 \left(1 - \frac{x}{\rho} \right)^{\frac{2}{2}} + a_3 \left(1 - \frac{x}{\rho} \right)^{\frac{3}{2}} + \dots \\ &= a_0 + a_1 \left(1 - \frac{x}{\rho} \right)^{\frac{1}{2}} + a_2 \left(1 - \frac{x}{\rho} \right) + O \left(\left(1 - \frac{x}{\rho} \right)^{\frac{3}{2}} \right) \end{aligned}$$

Algebraic Singularities

Singular expansion

$$\begin{aligned} A(x) &= \boxed{g(x) - h(x)\sqrt{1 - \frac{x}{\rho}}} \\ &= (g_0 + g_1(x - \rho) + g_2(x - \rho)^2 + \dots) \\ &\quad + (h_0 + h_1(x - \rho) + h_2(x - \rho)^2 + \dots) \sqrt{1 - \frac{x}{\rho}} \\ &= a_0 + a_1 \left(1 - \frac{x}{\rho}\right)^{\frac{1}{2}} + a_2 \left(1 - \frac{x}{\rho}\right)^{\frac{2}{2}} + a_3 \left(1 - \frac{x}{\rho}\right)^{\frac{3}{2}} + \dots \\ &= a_0 + a_1 \boxed{\left(1 - \frac{x}{\rho}\right)^{\frac{1}{2}}} + a_2 \left(1 - \frac{x}{\rho}\right) + \boxed{O\left(\left(1 - \frac{x}{\rho}\right)^{\frac{3}{2}}\right)} \end{aligned}$$

Algebraic Singularities

Singularity Analysis

Lemma 1 *Suppose that*

$$y(x) = \left(1 - \frac{x}{x_0}\right)^{-\alpha}.$$

Then

$$y_n = (-1)^n \binom{-\alpha}{n} x_0^{-n} = \frac{n^{\alpha-1}}{\Gamma(\alpha)} x_0^{-n} + \mathcal{O}\left(n^{\alpha-2} x_0^{-n}\right).$$

Remark: This asymptotic expansion is uniform in α if α varies in a compact region of the complex plane.

Algebraic Singularities

Singularity Analysis

Lemma 2 (Flajolet and Odlyzko) *Let*

$$y(x) = \sum_{n \geq 0} y_n x^n$$

be analytic in a region

$$\Delta = \{x : |x| < x_0 + \eta, |\arg(x - x_0)| > \delta\},$$

$$x_0 > 0, \eta > 0, 0 < \delta < \pi/2.$$

Suppose that for some real α

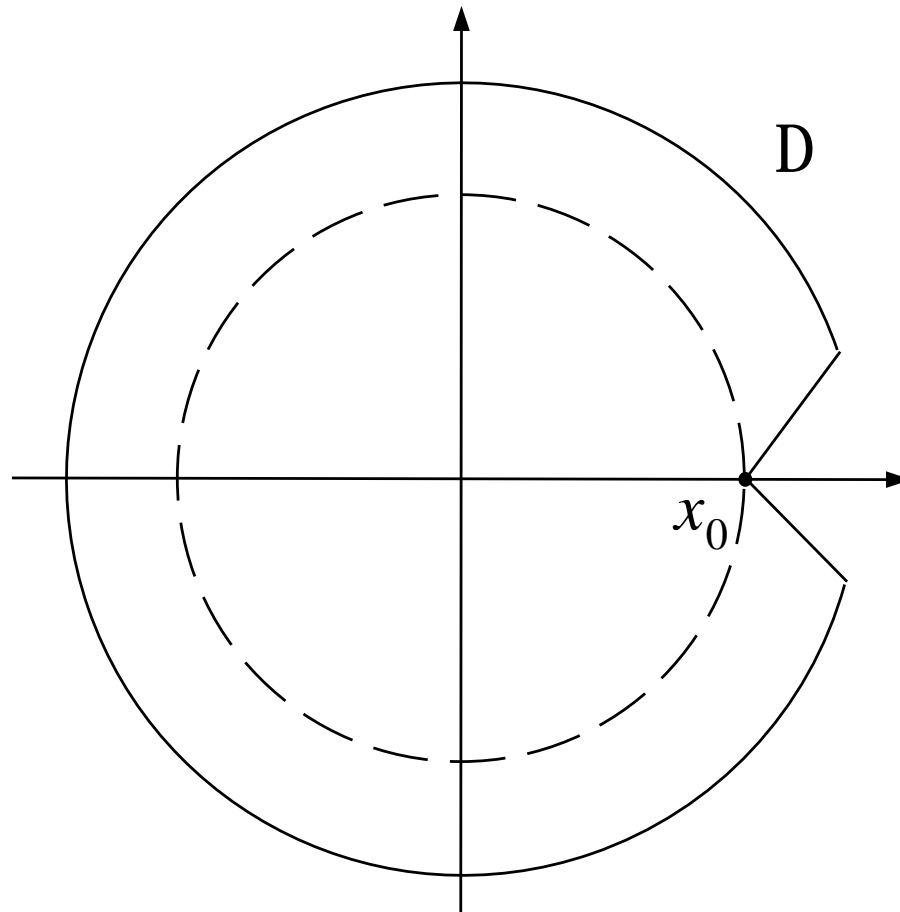
$$y(x) = \mathcal{O}\left((1 - x/x_0)^{-\alpha}\right) \quad (x \in \Delta).$$

Then

$$y_n = \mathcal{O}\left(x_0^{-n} n^{\alpha-1}\right).$$

Algebraic Singularities

Δ -region



Algebraic Singularities

Singularity Analysis

Suppose that

$$\begin{aligned} A(x) &= g(x) - h(x) \sqrt{1 - \frac{x}{\rho}} \\ &= a_0 + a_1 \left(1 - \frac{x}{\rho}\right)^{\frac{1}{2}} + a_2 \left(1 - \frac{x}{\rho}\right) + O\left(\left(1 - \frac{x}{\rho}\right)^{\frac{3}{2}}\right) \end{aligned}$$

for $x \in \Delta$ then

$$a_n = [x^n] A(x) = \frac{h(\rho)}{2\sqrt{\pi}} \rho^{-n} n^{-\frac{3}{2}} \left(1 + O\left(\frac{1}{n}\right)\right).$$

Algebraic Singularities

Singularity Analysis

Suppose that

$$\begin{aligned} A(x, u) &= g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}} \\ &= a_0(u) + a_1(u) \left(1 - \frac{x}{\rho(u)}\right)^{\frac{1}{2}} + a_2(u) \left(1 - \frac{x}{\rho(u)}\right) + O\left(\left(1 - \frac{x}{\rho(u)}\right)^{\frac{3}{2}}\right) \end{aligned}$$

for $x \in \Delta = \Delta(u)$ then

$$a_n(u) = [x^n] A(x, u) = \frac{h(\rho(u), u)}{2\sqrt{\pi}} \rho(u)^{-n} n^{-\frac{3}{2}} \left(1 + O\left(\frac{1}{n}\right)\right).$$

Probabilistic Model

a_n ... number of objects of size n

$a_{n,k}$... number of objects of size n , where a certain **parameter** has value k

If all objects of size n are considered to be **equally likely** then the parameter can be considered as a random variable X_n with distribution

$$\mathbb{P}\{X_n = k\} = \frac{a_{nk}}{a_n}.$$

Probabilistic Model

Generating functions and the probability generating function

$$A(x, u) = \sum_{n,k} a_{n,k} x^n u^k$$

$$\begin{aligned} \Rightarrow \boxed{\mathbb{E} u^{X_n}} &= \sum_{k \geq 0} \mathbb{P}\{X_n = k\} u^k \\ &= \sum_{k \geq 0} \frac{a_{nk}}{a_n} u^k \\ &= \boxed{\frac{[x^n] A(x, u)}{[x^n] A(x, 1)}} = \frac{a_n(u)}{a_n} \end{aligned}$$

Probabilistic Model

Generating functions and the probability generating function

$$A(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}$$

$$\begin{aligned} \Rightarrow \boxed{\mathbb{E} u^{X_n}} &= \frac{[x^n] A(x, u)}{[x^n] A(x, 1)} \\ &= \frac{\frac{h(\rho(u), u)}{2\sqrt{\pi}} \rho(u)^{-n} n^{-\frac{3}{2}} \left(1 + O\left(\frac{1}{n}\right)\right)}{\frac{h(\rho(1), 1)}{2\sqrt{\pi}} \rho(1)^{-n} n^{-\frac{3}{2}} \left(1 + O\left(\frac{1}{n}\right)\right)} \\ &= \boxed{\frac{h(\rho(u), u)}{h(\rho(1), 1)} \left(\frac{\rho(1)}{\rho(u)}\right)^n \left(1 + O\left(\frac{1}{n}\right)\right)}. \end{aligned}$$

Probabilistic Model

Quasi-Power Theorem (Hwang)

Let X_n be a sequence of random variables with the property that

$$\mathbb{E} u^{X_n} = A(u) \cdot B(u)^{\lambda_n} \cdot \left(1 + O\left(\frac{1}{\phi_n}\right) \right)$$

holds uniformly in a complex neighborhood of $u = 1$, $\lambda_n \rightarrow \infty$ and $\phi_n \rightarrow \infty$, and $A(u)$ and $B(u)$ are analytic functions in a neighborhood of $u = 1$ with $A(1) = B(1) = 1$. Set

$$\mu = B'(1) \quad \text{and} \quad \sigma^2 = B''(1) + B'(1) - B'(1)^2.$$

$$\implies \mathbb{E} X_n = \mu \lambda_n + O(1 + \lambda_n/\phi_n), \quad \mathbb{V} X_n = \sigma^2 \lambda_n + O(1 + \lambda_n/\phi_n),$$

$$\frac{X_n - \mathbb{E} X_n}{\sqrt{\mathbb{V} X_n}} \xrightarrow{d} N(0, 1) \quad (\sigma^2 \neq 0).$$

Probabilistic Model

Sums of independent random variables

$X_n = \xi_1 + \xi_2 + \cdots + \xi_n$, where ξ_j are i.i.d.

$$B(u) = \mathbb{E} u^{\xi_j}$$

$$\begin{aligned} \implies \mathbb{E} u^{X_n} &= \mathbb{E} u^{\xi_1 + \xi_2 + \cdots + \xi_n} \\ &= \mathbb{E} u^{\xi_1} \cdot \mathbb{E} u^{\xi_2} \cdots \mathbb{E} u^{\xi_n} \\ &= B(u)^n. \end{aligned}$$

Probabilistic Model

COMBINATORIAL CENTRAL LIMIT THEOREM

Suppose that a sequence of random variables X_n has distribution

$$\mathbb{P}\{X_n = k\} = \frac{a_{nk}}{a_n},$$

where the generating function $A(x, u) = \sum_{n,k} a_{n,k} x^n u^k$ satisfies a functional equation of the form $A(x, u) = \Phi(x, u, A(x, u))$, where $\Phi(x, u, a)$ has a power series expansion at $(0, 0, 0)$ with non-negative coefficients and $\Phi_{aa}(x, u, a) \neq 0$.

Let $x_0 > 0$, $a_0 > 0$ (inside the region of convergence) satisfy the system of equations:

$$a_0 = \Phi(x_0, 1, a_0), \quad 1 = \Phi_a(x_0, 1, a_0).$$

Probabilistic Model

COMBINATORIAL CENTRAL LIMIT THEOREM (cont.)

Set

$$\mu = \frac{\Phi_u}{x_0 \Phi_x},$$
$$\sigma^2 = \mu + \mu^2 + \frac{1}{x_0 \Phi_x^3 \Phi_{aa}} \left(\Phi_x^2 (\Phi_{aa} \Phi_{uu} - \Phi_{au}^2) - 2\Phi_x \Phi_u (\Phi_{aa} \Phi_{xu} - \Phi_{ax} \Phi_{au}) \right. \\ \left. + \Phi_u^2 (\Phi_{aa} \Phi_{xx} - \Phi_{ax}^2) \right),$$

(where all partial derivatives are evaluated at the point $(x_0, a_0, 1)$)

Then we have

$$\mathbb{E} X_n = \mu n + O(1)$$

and

$$\text{Var} X_n = \sigma^2 n + O(1)$$

and if $\sigma^2 > 0$ then

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

Random Trees

Leaves in Catalan trees

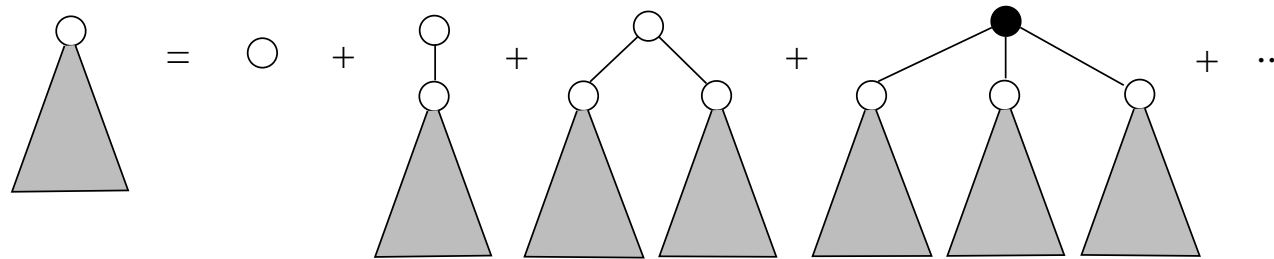
The number of leaves in Catalan trees of size n satisfy a **central limit theorem** with mean $\sim \frac{1}{2}n$ and variance $\sim \frac{1}{8}n$

Leaves in Cayley trees

The number of leaves in Cayley trees of size n satisfy a **central limit theorem** with mean $\sim \frac{1}{e}n$ and variance $\sim \left(\frac{1}{e^2} + \frac{1}{e}\right)n$

Random Trees

Nodes of out-degree d in Catalan trees



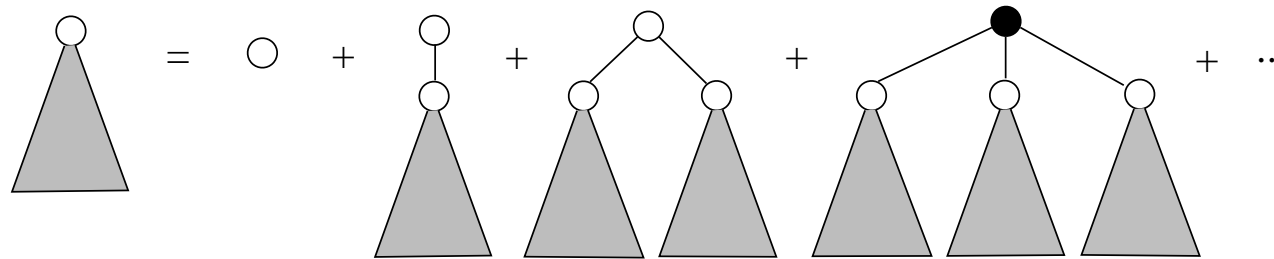
$$G(x, u) = \frac{x}{1 - G(x, u)} + x(u - 1)G(x, u)^d$$

The number $X_n^{(d)}$ of nodes with out-degree d in Catalan trees of size n satisfy a **central limit theorem** with mean $\sim \mu_d n$ and variance $\sim \sigma_d^2 n$, where

$$\mu_d = \frac{1}{2^{d+1}} \quad \text{and} \quad \sigma_d^2 = \frac{1}{2^{d+1}} + \frac{1}{2^{2(d+1)}} - \frac{(d-1)^2}{2^{2d+3}}.$$

Random Trees

Nodes of out-degree d in Cayley trees



$$R(x, u) = xe^{R(x, u)} + x(u-1) \frac{R(x, u)^d}{d!}$$

The number of nodes with out-degree d in Cayley trees of size n satisfy a **central limit theorem** with mean $\sim \mu_d n$ and variance $\sim \sigma_d^2 n$, where

$$\mu_d = \frac{1}{e d!} \quad \text{and} \quad \sigma_d^2 = \frac{1 + (d-1)^2}{e^2 (d!)^2} + \frac{1}{e d!}$$

Random Trees

Degree distribution for Catalan trees

$p_{n,d}$... probability that a random node in a random Catalan tree of size n has out-degree d :

$$\mathbb{E} X_n^{(d)} = n p_{n,d}$$

$$p_d := \lim_{n \rightarrow \infty} p_{n,d} = \frac{1}{2^{d+1}} = \mu_d$$

Probability generating function of the out-degree distribution:

$$p(w) := \sum_{d \geq 0} p_d w^d = \frac{1}{2-w}$$

Random Trees

Degree distribution for Cayley trees

$p_{n,d}$... probability that a random node in a random Cayley tree of size n has out-degree d :

$$\mathbb{E} X_n^{(d)} = n p_{n,d}$$

$$p_d := \lim_{n \rightarrow \infty} p_{n,d} = \frac{1}{e d!} = \mu_d$$

Probability generating function of the out-degree distribution:

$$p(w) := \sum_{d \geq 1} p_d w^d = e^{w-1}$$

Contents 2

I. COMBINATORIAL RANDOM TREES

- Maximum degree
- Unrooted trees

II. PATTERN COUNTS IN RANDOM TREES

- Pattern in trees
- Systems of functional equations

Random Trees

Maximum degree

Δ_n ... maximum out-degree

$X_n^{(>d)} = X_n^{(d+1)} + X_n^{(d+2)} + \dots$... number of nodes of out-degree $> d$.

$$\Delta_n > d \iff X_n^{(>d)} > 0$$

Random Trees

First moment method

X ... a discrete random variable on non-negative integers.

$$\implies \boxed{\mathbb{P}\{X > 0\} \leq \min\{1, \mathbb{E} X\}}$$

Proof

$$\mathbb{E} X = \sum_{k \geq 0} k \mathbb{P}\{X = k\} \geq \sum_{k \geq 1} \mathbb{P}\{X = k\} = \mathbb{P}\{X > 0\}.$$

Random Trees

Second moment method

X is a non-negative random variable with finite second moment.

$$\implies \boxed{\mathbb{P}\{X > 0\} \geq \frac{(\mathbb{E} X)^2}{\mathbb{E}(X^2)}}$$

Proof

$$\mathbb{E} X = \mathbb{E} (X \cdot \mathbf{1}_{[X>0]}) \leq \sqrt{\mathbb{E}(X^2)} \sqrt{\mathbb{E}(\mathbf{1}_{[X>0]}^2)} = \sqrt{\mathbb{E}(X^2)} \sqrt{\mathbb{P}\{X > 0\}}.$$

Random Trees

Tail estimates and expected value

- $\mathbb{P}\{\Delta_n > d\} \leq \min\{1, \mathbb{E} X_n^{(>d)}\}$

- $\mathbb{P}\{\Delta_n > d\} \geq \frac{(\mathbb{E} X_n^{(>d)})^2}{\mathbb{E} (X_n^{(>d)})^2}$

$$\implies \mathbb{P}\{\Delta_n \leq d\} \leq 1 - \frac{(\mathbb{E} X_n^{(>d)})^2}{\mathbb{E} (X_n^{(>d)})^2} = \frac{\text{Var } X_n^{(>d)}}{\mathbb{E} (X_n^{(>d)})^2}$$

- $\mathbb{E} \Delta_n = \sum_{d \geq 0} \mathbb{P}\{\Delta_n > d\}$

Random Trees

Maximum degree of Catalan trees

$$\mathbb{E} X_n^{(>d)} \sim \frac{n}{2^{d+1}}, \quad \text{Var} (X_n^{(>d)})^2 \sim n \left(\frac{1}{2^{d+1}} + \frac{1}{2^{2(d+1)}} - \frac{(d-1)^2}{2^{2d+3}} \right)$$

$$\implies \mathbb{P}\{\Delta_n > d\} \leq \min \left\{ 1, \frac{n}{2^{d+1}} \right\},$$

$$\begin{aligned} \mathbb{P}\{\Delta_n \leq d\} &= 1 - \mathbb{P}\{\Delta_n > d\} \\ &\leq \frac{1}{n} \frac{1}{\frac{1}{2^{d+1}} + \frac{1}{2^{2(d+1)}} - \frac{(d-1)^2}{2^{2d+3}}} \sim \frac{2^{d+1}}{n} \end{aligned}$$

$$\implies \boxed{\Delta_n \text{ is concentrated at } \log_2 n + O(1)}$$

Random Trees

Maximum degree of Catalan trees (Carr, Goh and Schmutz)

$$\mathbb{P}\{\Delta_n \leq k\} = \exp\left(-2^{-(k - \log_2 n + 1)}\right) + o(1)$$

$$\mathbb{E} \Delta_n = \log_2 n + O(1)$$

Random Trees

Unrooted trees

p_n ... number of different embeddings of **unrooted** trees of size n in the plane, $P(x) = \sum_{n \geq 1} p_n x^n$:

$$P(x) = x \sum_{k \geq 0} Z_{\mathfrak{C}_k}(G(x), G(x^2), \dots, G(x^k)) - \frac{1}{2}G(x)^2 + \frac{1}{2}G(x^2),$$

where $G(x) = x/(1 - G(x)) = (1 - \sqrt{1 - 4x})/2$ and

$$Z_{\mathfrak{C}_k}(x_1, x_2, \dots, x_k) = \frac{1}{k} \sum_{d|k} \varphi(d) x_d^{k/d}$$

is the cycle index of the cyclic group \mathfrak{C}_k of k elements

Random Trees

Unrooted trees

Cancellation of the $\sqrt{1-4x}$ -term:

$$G(x) = \frac{1 - \sqrt{1 - 4x}}{2} \implies P(x) = a_0 + a_2(1 - 4x) + \frac{1}{6}(1 - 4x)^{3/2} + \dots$$

$$\implies \boxed{p_n = \frac{1}{8\sqrt{\pi}} 4^n n^{-5/2} (1 + O(n^{-1}))}$$

Random Trees

Degree distribution of unrooted trees

$X_n^{(d)}$... number of nodes of degree d in trees of size n

$$\begin{aligned} P(x, u) = & x \sum_{k \neq d} Z_{\mathfrak{C}_k}(G(x, u), G(x^2, u^2), \dots, G(x^k, u^k)) \\ & + xu Z_{\mathfrak{C}_d}(G(x, u), G(x^2, u^2), \dots, G(x^d, u^d)) \\ & - \frac{1}{2}G(x, u)^2 + \frac{1}{2}G(x^2, u^2), \end{aligned}$$

where

$$G(x, u) = \frac{x}{1 - G(x, u)} + x(u - 1)G(x, u)^{d-1}.$$

Random Trees

Degree distribution of unrooted trees

Cancellation of the $\sqrt{1-4x}$ -term:

$$G(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}$$

$$\implies P(x, u) = a_0(u) + a_2(u) \left(1 - \frac{x}{\rho(u)}\right) + a_3(u) \left(1 - \frac{x}{\rho(u)}\right)^{\frac{3}{2}} + \dots$$

$\implies X_n^{(d)}$ satisfies a **central limit theorem** with mean $\sim \mu_{d-1}n$ and variance $\sim \sigma_{d-1}^2 n$, where

$$\mu_d = \frac{1}{2^{d+1}} \quad \text{and} \quad \sigma_d^2 = \frac{1}{2^{d+1}} + \frac{1}{2^{2(d+1)}} - \frac{(d-1)^2}{2^{2d+3}}.$$

Random Trees

Degree distribution of unrooted trees

$p_{n,d}$... probability that a random node in a tree of size n has degree d :

$$\mathbb{E} X_n^{(d)} = n p_{n,d}$$

$$p_d = \lim_{n \rightarrow \infty} p_{n,d} = \mu_{d-1} = \frac{1}{2^d}$$

Probability generating function of the degree distribution:

$$p(w) = \sum_{d \geq 1} p_d w^d = \frac{w}{2-w}$$

Random Trees

Maximum degree for unrooted trees

Δ_n ... maximum degree of unrooted trees of size n

Δ_n is concentrated at $\log_2 n$

$$\mathbb{E} \Delta_n = \log_2 n + O(1)$$

Random Trees

Unrooted labelled trees

$t_n = r_n/n = n^{n-2}$... number of different **unrooted** labelled trees of

size n :
$$T(x) = \sum_{n \geq 1} t_n \frac{x^n}{n!} :$$

$$T(x) = xe^{R(x)} - \frac{1}{2}R(x)^2 = R(x) - \frac{1}{2}R(x)^2,$$

where $R(x) = xe^{R(x)}$ (note that $T'(x) = R(x)/x$)

Cancellation of the $\sqrt{1-ex}$ -term:

$$R(x) = g(x) - h(x)\sqrt{1-ex} \implies T(x) = a_0 + a_2(1-4x) + \frac{1}{6}(1-ex)^{3/2} + \dots$$

Random Trees

Degree distribution of unrooted labelled trees

$X_n^{(d)}$... number of nodes of degree d in trees of size n

$$T(x, u) = xe^{R(x, u)} + x(u-1) \frac{R(x, u)^d}{d!} - \frac{1}{2} R(x, u)^2,$$

where

$$R(x, u) = xe^{R(x, u)} + x(u-1) \frac{R(x, u)^{d-1}}{(d-1)!}.$$

Random Trees

Degree distribution of unrooted labelled trees

Cancellation of the $\sqrt{1-4x}$ -term:

$$R(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}$$

$$\implies T(x, u) = a_0(u) + a_2(u) \left(1 - \frac{x}{\rho(u)}\right) + a_3(u) \left(1 - \frac{x}{\rho(u)}\right)^{\frac{3}{2}} + \dots$$

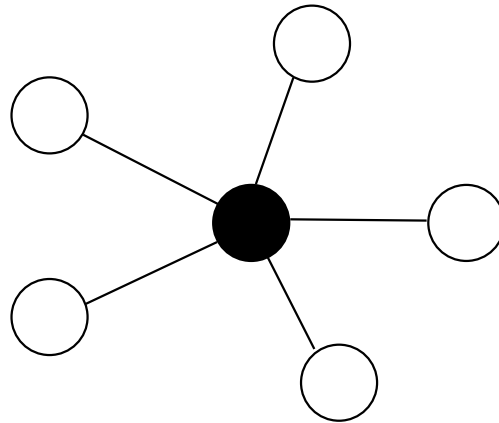
$\implies X_n^{(d)}$ satisfies a **central limit theorem** with mean $\sim \mu_{d-1}n$ and variance $\sim \sigma_{d-1}^2 n$, where

$$\mu_d = \frac{1}{e d!} \quad \text{and} \quad \sigma_d^2 = \frac{1 + (d-1)^2}{e^2 (d!)^2} + \frac{1}{e d!}$$

(Note again that $\frac{\partial}{\partial x} T(x, u) = R(x, u)/x$)

Random Trees

Star pattern

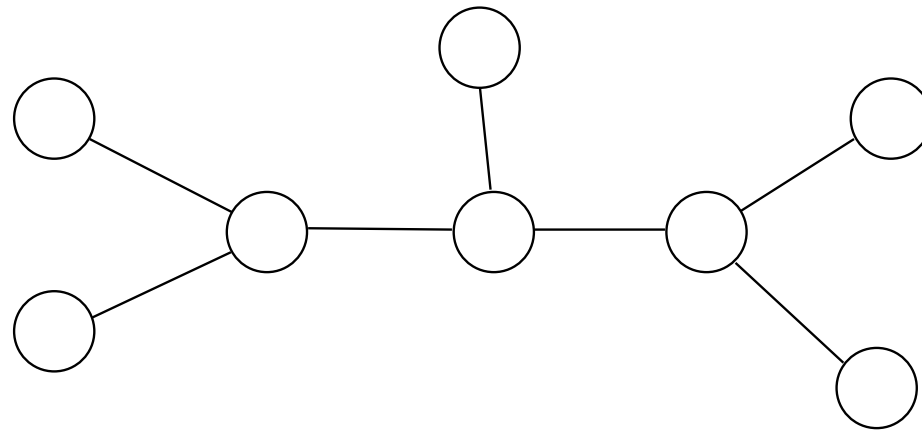


$$d = 5$$

$X_n^{(d)}$ = number of nodes of degree d in trees of size n
= number of star pattern with d rays in trees of size n

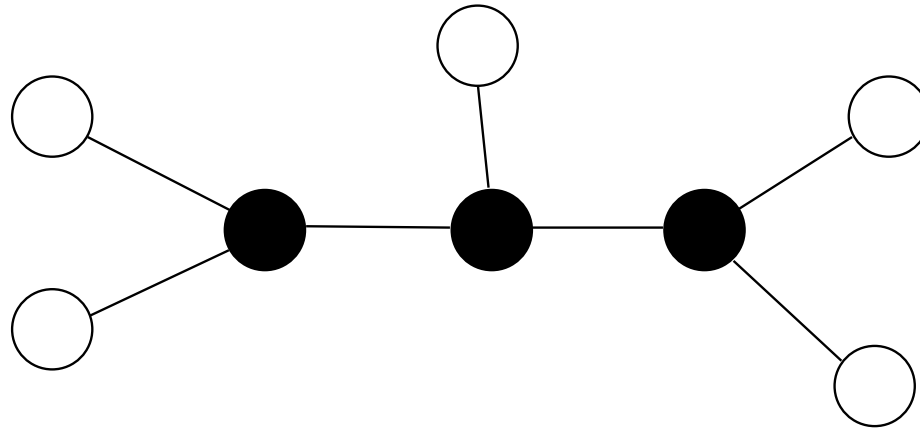
Patterns in Trees

Pattern \mathcal{M}



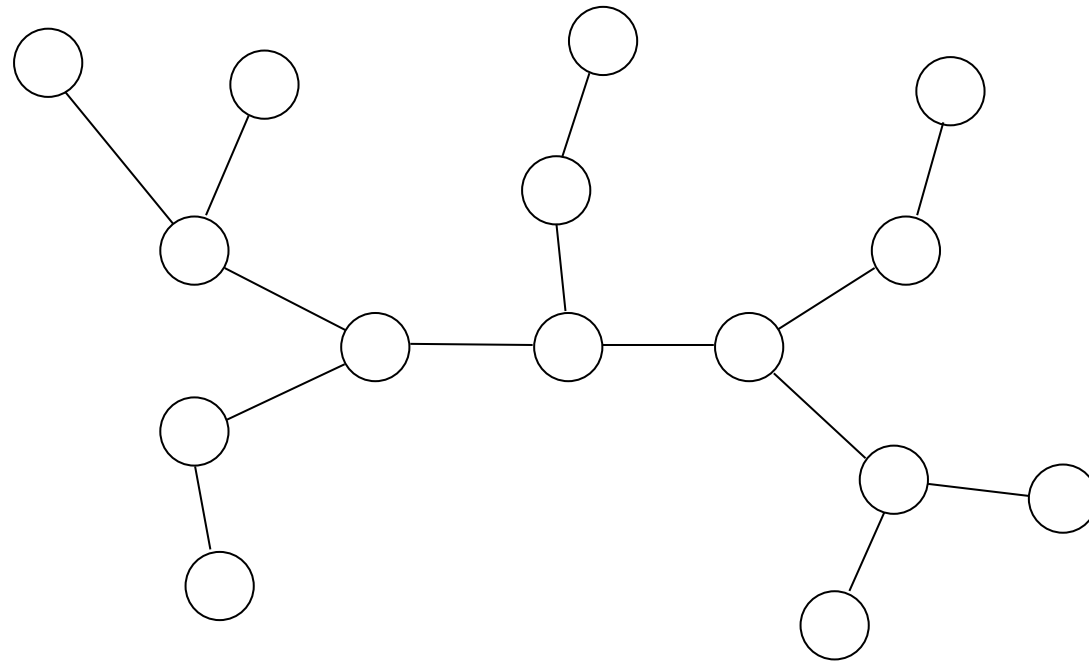
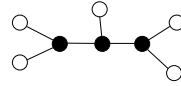
Patterns in Trees

Pattern \mathcal{M}



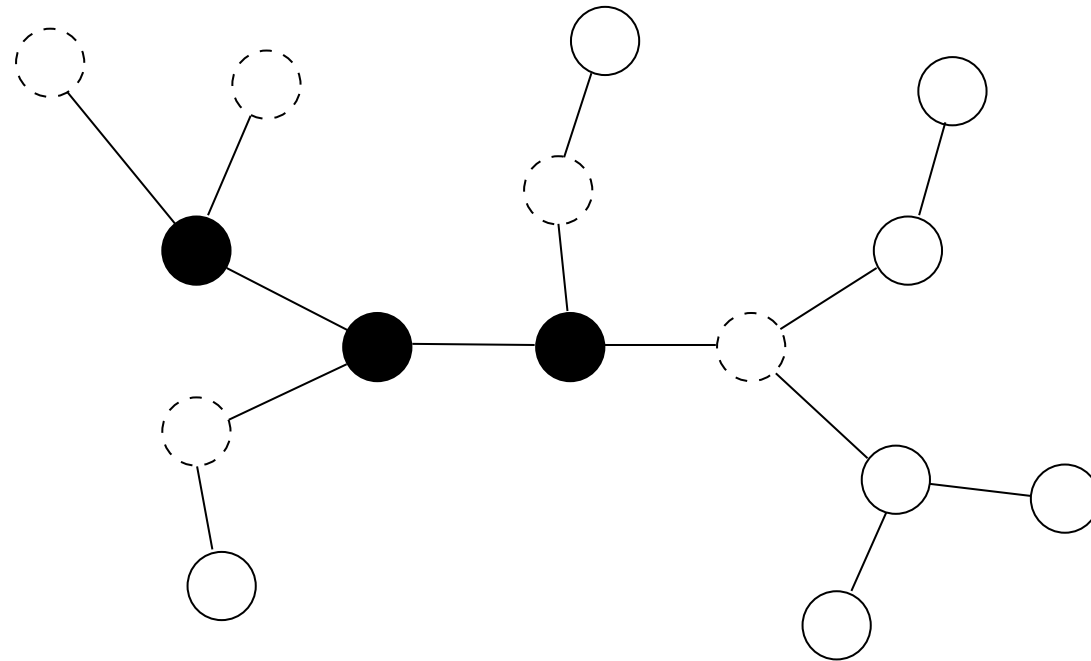
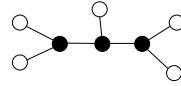
Patterns in Trees

Occurrence of a pattern \mathcal{M}



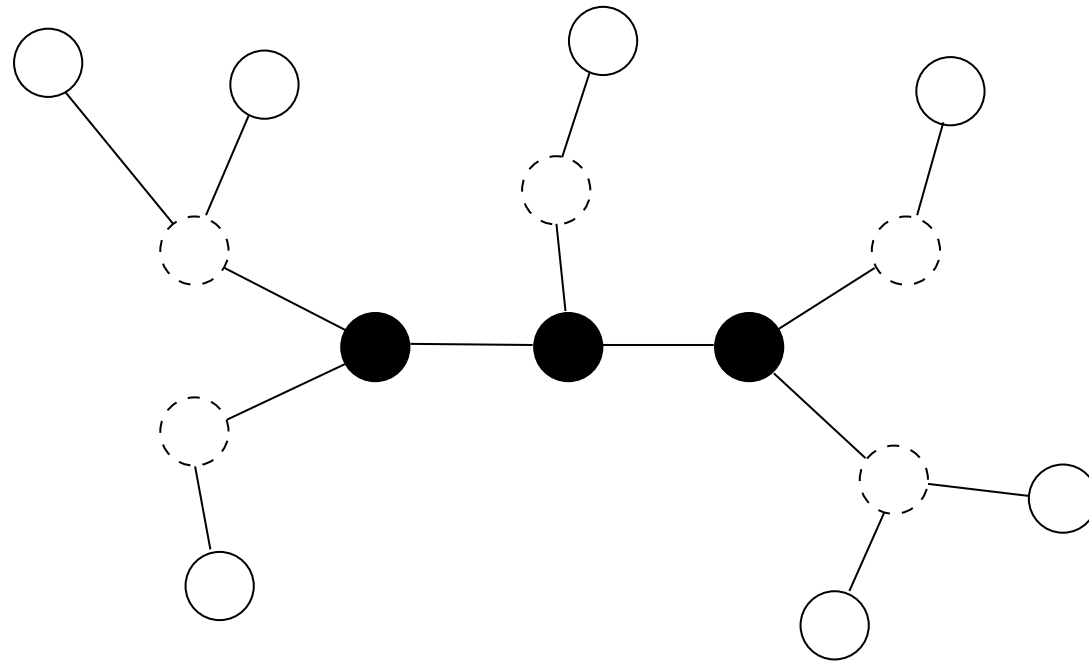
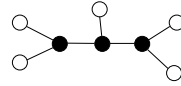
Patterns in Trees

Occurrence of a pattern \mathcal{M}



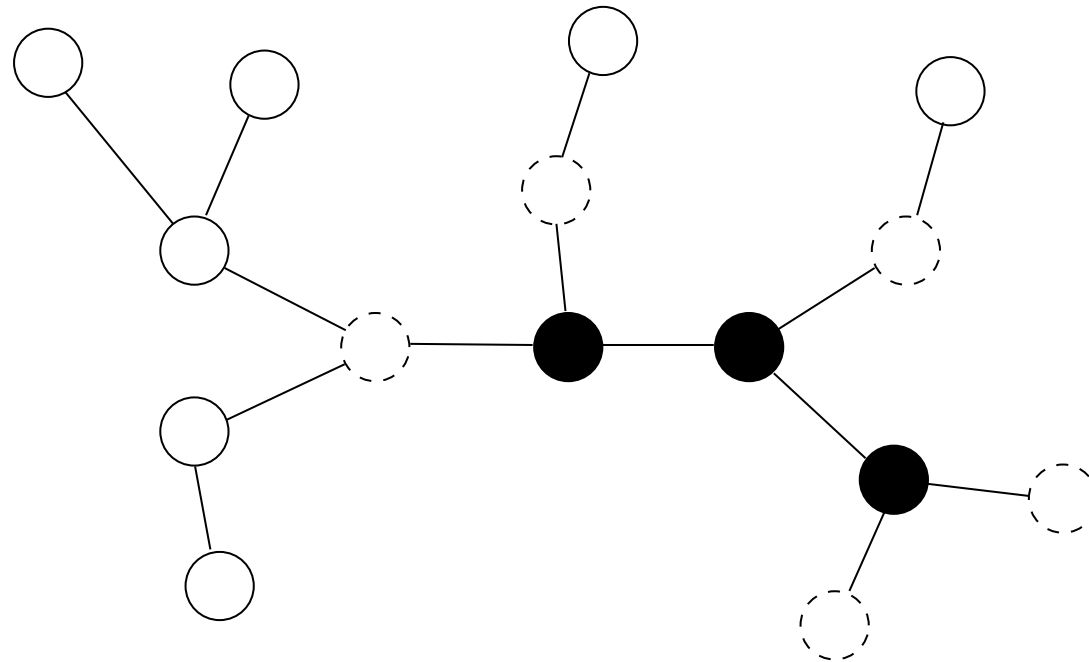
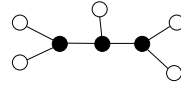
Patterns in Trees

Occurrence of a pattern \mathcal{M}



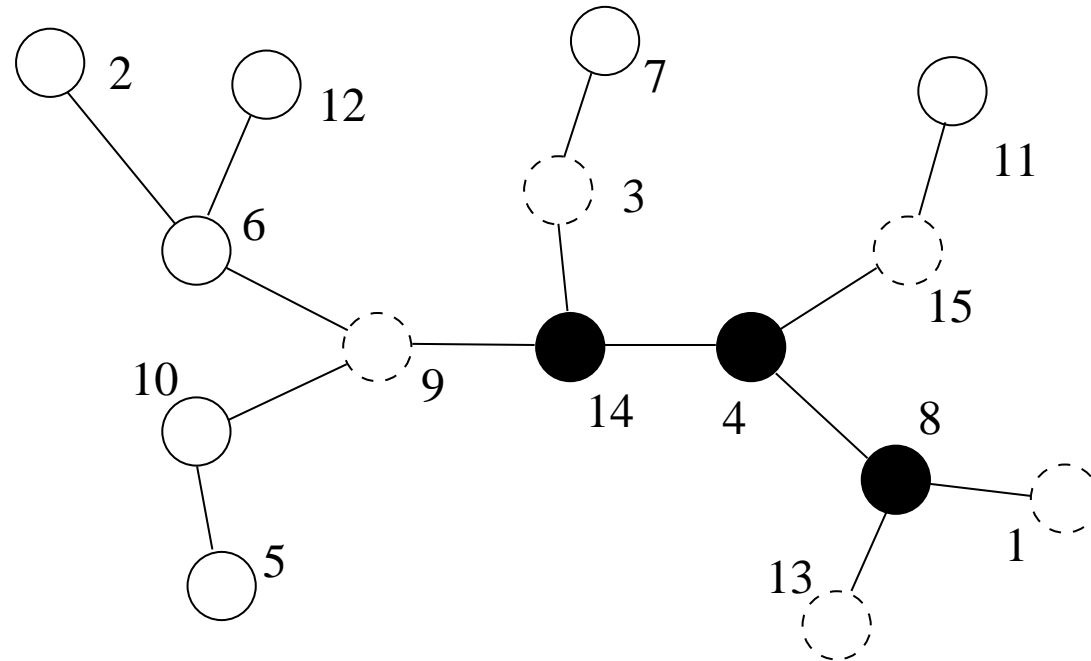
Patterns in Trees

Occurrence of a pattern \mathcal{M}



Patterns in Trees

Occurrence of a pattern \mathcal{M}  in a labelled tree



Patterns in Trees

Cayley's formula

$r_n = n^{n-1}$... number of **rooted** labelled trees with n nodes

$t_n = n^{n-2}$... number of labelled trees with n nodes

Generating functions

$$R(x) = \sum_{n \geq 1} r_n \frac{x^n}{n!}:$$

$$R(x) = xe^{R(x)}$$

$$T(x) = \sum_{n \geq 1} t_n \frac{x^n}{n!}:$$

$$T(x) = R(x) - \frac{1}{2}R(x)^2$$

(Note that $xT'(x) = R(x)$ so that we also have $T(x) = \int R(x)/x dx$.)

Patterns in Trees

Theorem

\mathcal{M} ... be a given finite tree.

X_n ... number of occurrences of \mathcal{M} in a labelled tree of size n

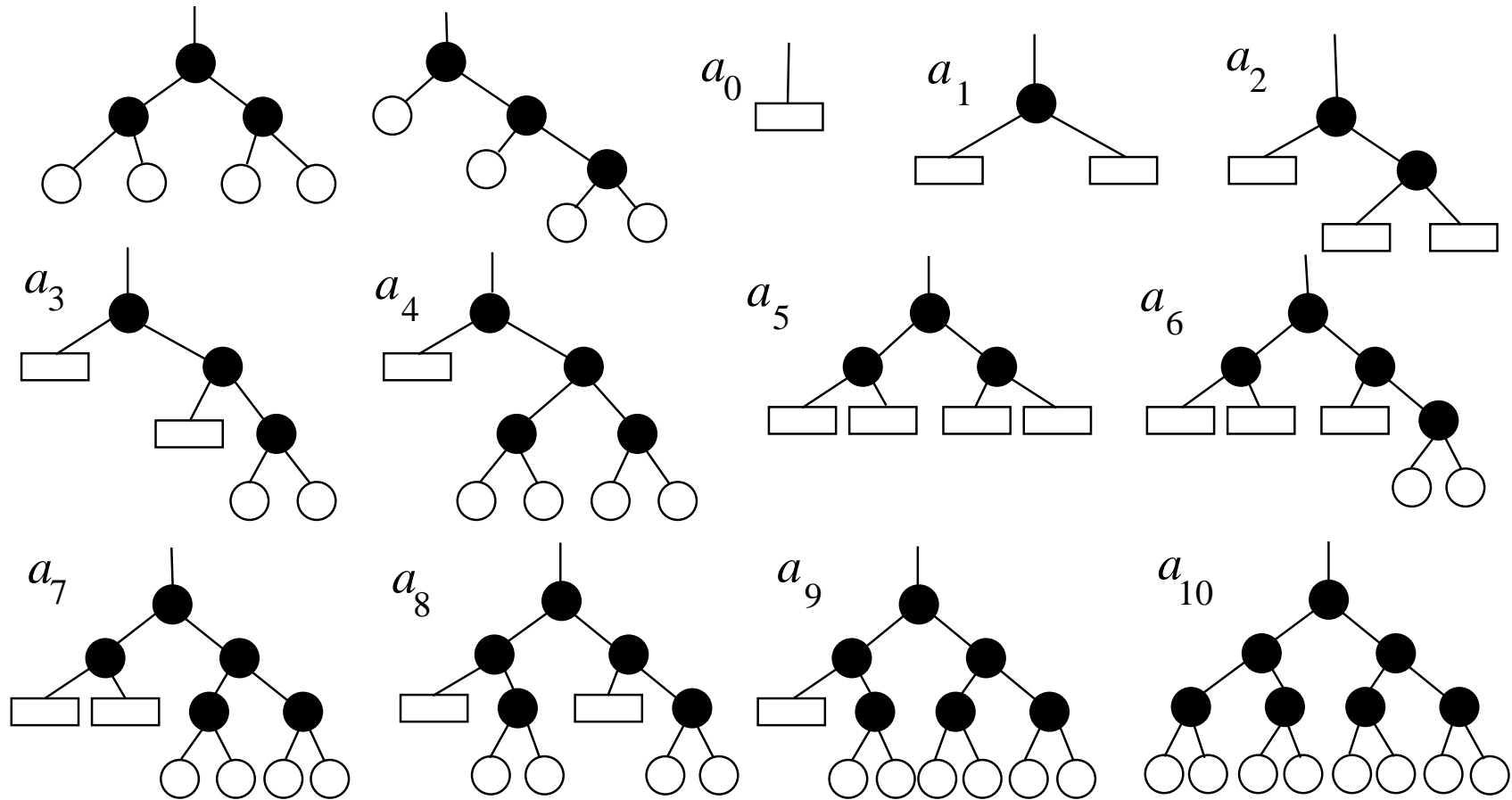
$\implies X_n$ satisfies a **central limit theorem** with

$$\mathbb{E} X_n \sim \mu n \quad \text{and} \quad \mathbb{V} X_n \sim \sigma^2 n.$$

$\mu > 0$ and $\sigma^2 \geq 0$ depend on the pattern \mathcal{M} and can be computed explicitly and algorithmically and can be represented as polynomials (with rational coefficients) in $1/e$.

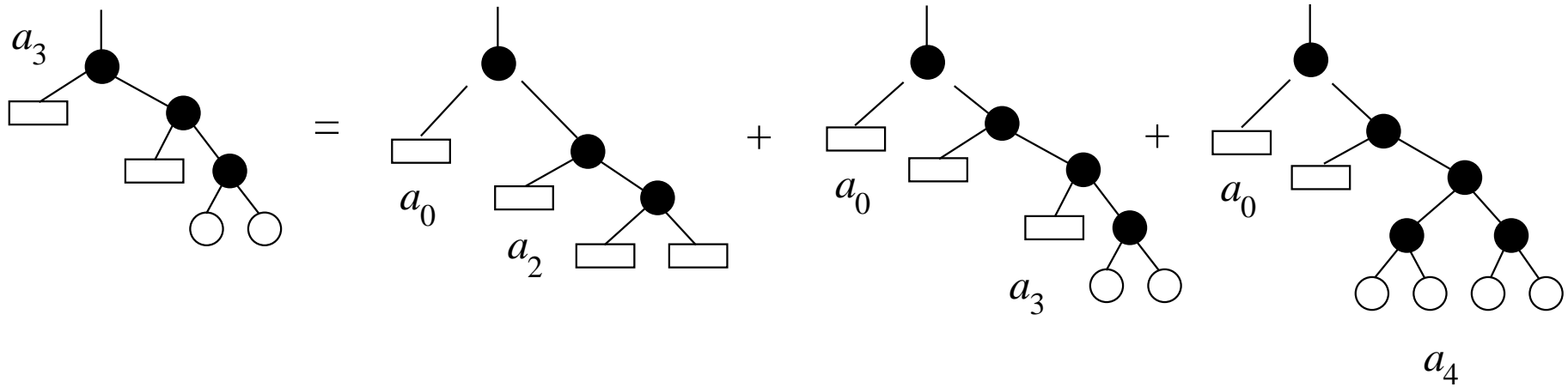
Patterns in Trees

Partition of trees in classes (\square ... out-degree different from 2)



Patterns in Trees

Recurrences $A_3 = xA_0A_2 + xA_0A_3 + xA_0A_4$

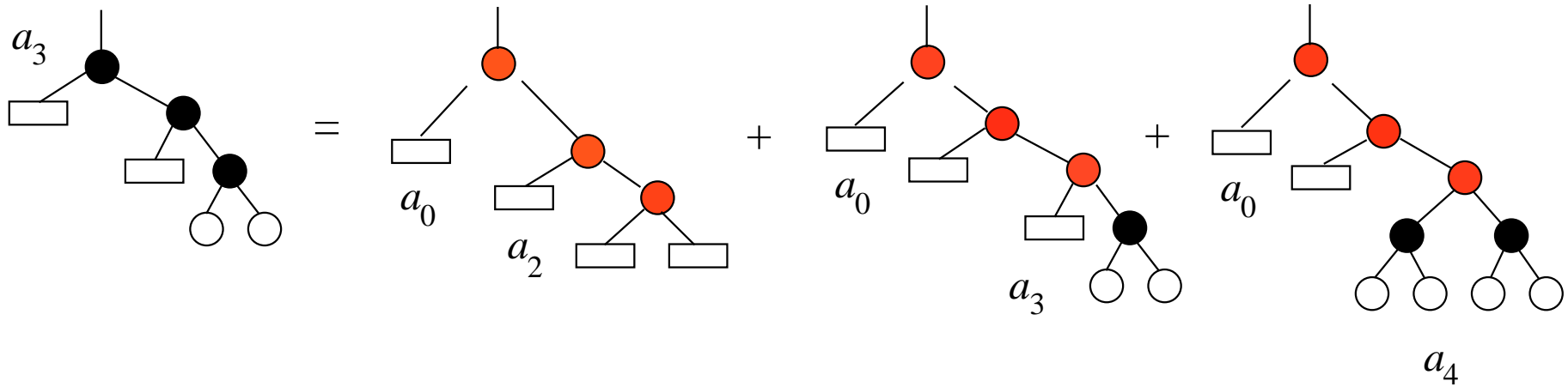


$$A_j(x) = \sum_{n,k} a_{j;n} \frac{x^n}{n!}$$

$a_{j;n}$... number of trees of size n in class j

Patterns in Trees

Recurrences $A_3 = xuA_0A_2 + xuA_0A_3 + xuA_0A_4$



$$A_j(x, u) = \sum_{n,k} a_{j;n,k} \frac{x^n}{n!} u^k$$

$a_{j;n,k}$... number of trees of size n in class j with k occurrences of \mathcal{M}

Patterns in Trees

$$A_0 = A_0(x, u) = x + x \sum_{i=0}^{10} A_i + x \sum_{n=3}^{\infty} \frac{1}{n!} \left(\sum_{i=0}^{10} A_i \right)^n,$$

$$A_1 = A_1(x, u) = \frac{1}{2} x A_0^2,$$

$$A_2 = A_2(x, u) = x A_0 A_1,$$

$$A_3 = A_3(x, u) = x A_0 (A_2 + A_3 + A_4) u,$$

$$A_4 = A_4(x, u) = x A_0 (A_5 + A_6 + A_7 + A_8 + A_9 + A_{10}) u^2,$$

$$A_5 = A_5(x, u) = \frac{1}{2} x A_1^2 u,$$

$$A_6 = A_6(x, u) = x A_1 (A_2 + A_3 + A_4) u^2,$$

$$A_7 = A_7(x, u) = x A_1 (A_5 + A_6 + A_7 + A_8 + A_9 + A_{10}) u^3,$$

$$A_8 = A_8(x, u) = \frac{1}{2} x (A_2 + A_3 + A_4)^2 u^3,$$

$$A_9 = A_9(x, u) = x (A_2 + A_3 + A_4) (A_5 + A_6 + A_7 + A_8 + A_9 + A_{10}) u^4,$$

$$A_{10} = A_{10}(x, u) = \frac{1}{2} x (A_5 + A_6 + A_7 + A_8 + A_9 + A_{10})^2 u^5.$$

Systems of Functional equations

COMBINATORIAL CENTRAL LIMIT THEOREM II

Suppose that a sequence of random variables X_n has distribution

$$\mathbb{P}[X_n = k] = \frac{a_{nk}}{a_n},$$

where the generating function $A(x, u) = \sum_{n,k} a_{n,k} x^n u^k$ is given by

$$A(x, u) = \Psi(x, u, A_1(x, u), \dots, A_r(x, u))$$

for an analytic function Ψ and the generating functions

$$A_1(x, u) = \sum_{n,k} a_{1;n,k} u^k x^n, \dots, A_r(x, u) = \sum_{n,k} a_{r;n,k} u^k x^n$$

satisfy a **system of non-linear equations**

$$A_j(x, u) = \Phi_j(x, u, A_1(x, u), \dots, A_r(x, u)), \quad (1 \leq j \leq r).$$

Systems of Functional equations

COMBINATORIAL CENTRAL LIMIT THEOREM II (cont.)

Suppose that at least one of the functions $\Phi_j(x, u, a_1, \dots, a_r)$ is non-linear in a_1, \dots, a_r and they all have a power series expansion at $(0, 0, 0)$ with non-negative coefficients.

Let $x_0 > 0$, $\mathbf{a}_0 = (a_{0,0}, \dots, a_{r,0}) > 0$ (inside the region of convergence) satisfy the system of equations: $(\Phi = (\Phi_1, \dots, \Phi_r))$

$$\boxed{\mathbf{a}_0 = \Phi(x_0, 1, \mathbf{a}_0), \quad 0 = \det(\mathbb{I} - \Phi_{\mathbf{a}}(x_0, 1, \mathbf{a}_0)}$$

such that the spectral radius of the Jacobian $\Phi_{\mathbf{a}}$ equals 1. Suppose further, that the **dependency graph** of the system $\mathbf{a} = \Phi(x, u, \mathbf{a})$ is **strongly connected** (which means that no subsystem can be solved before the whole system).

Systems of Functional equations

COMBINATORIAL CENTRAL LIMIT THEOREM II (cont.)

Then there exists analytic function $g_j(x, u)$, $h_j(x, u)$, and $\rho(u)$ (that is **independent of j**) such that locally

$$A_j(x, u) = g_j(x, u) - h_j(x, u) \sqrt{1 - \frac{x}{\rho(u)}}.$$

and consequently (for some $g(x, u)$, $h(x, u)$)

$$A(x, u) = g(x, u) - h(x, u) \sqrt{1 - \frac{x}{\rho(u)}}.$$

Consequently the random variable X_n satisfies a **central limit theorem** with

$$\boxed{\mathbb{E} X_n \sim n\mu} \quad \text{and} \quad \boxed{\text{Var} X_n \sim n\sigma^2},$$

where μ and σ^2 can be computed.

Patterns in Trees

Final Result for $\mathcal{M} =$ 

Central limit theorem with

$$\mu = \frac{5}{8e^3} = 0.0311169177\dots$$

and

$$\sigma^2 = \frac{20e^3 + 72e^2 + 84e - 175}{32e^6} = 0.0764585401\dots$$

Contents 3

III. CONTINUOUS LIMITING OBJECTS

- Weak Convergence
- The Depth-First-Search of Rooted Trees
- The Continuum Random Tree
- The Profile of Galton-Watson trees
- Scaling Limit of Series-Parallel Graphs

Asymptotics on Random Discrete Objects

Levels of complexity:

1. Asymptotic enumeration
2. Distribution of (shape) parameters
3. Asymptotic shape (= continuous limiting object)

Weak Convergence

$X_n, X \dots$ (real) random variables:

$$\boxed{X_n \xrightarrow{d} X} \quad :\Leftrightarrow \quad \boxed{\lim_{n \rightarrow \infty} \mathbb{P}\{X_n \leq x\} = \mathbb{P}\{X \leq x\}}$$

for all points of continuity
of $F_X(x) = \mathbb{P}\{X \leq x\}$

$$\Leftrightarrow \quad \boxed{\lim_{n \rightarrow \infty} \mathbb{E} G(X_n) = \mathbb{E} G(X)}$$

for all **bounded** continuous
functionals $G : \mathbb{R} \rightarrow \mathbb{R}$

$$\Leftrightarrow \quad \boxed{\lim_{n \rightarrow \infty} \mathbb{E} e^{itX_n} = \mathbb{E} e^{itX}}$$

for all real t
(Levy's criterion)

Weak Convergence

Polish space: (S, d) ... complete, separable, metric space

Examples: \mathbb{R} , \mathbb{R}^k , $C[0, 1]$, $\mathcal{M}_0(X)$ (probability measures on X)

S -valued random variable: $X : \Omega \rightarrow S$... measurable function

$S = \mathbb{R}$: random variable

$S = \mathbb{R}^k$: k -dimensional random vector

$S = C[0, 1]$: **stochastic process** $(X(t), 0 \leq t \leq 1)$

$S = \mathcal{M}_0(X)$: random measure

Weak Convergence

Definition

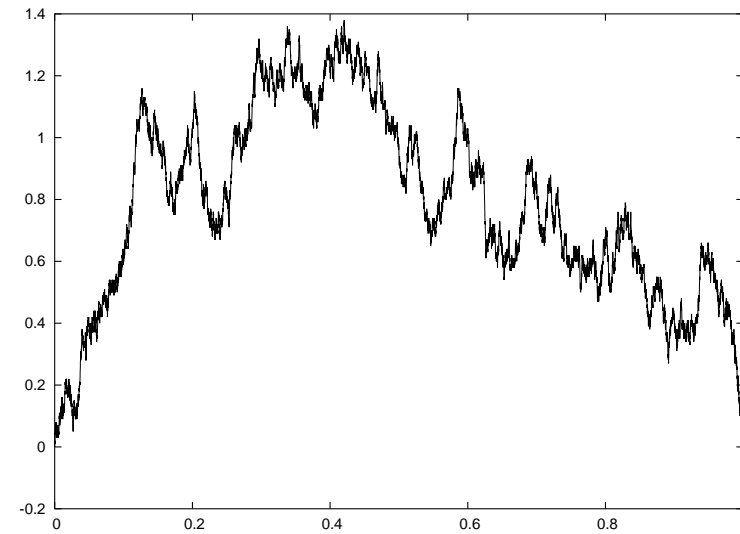
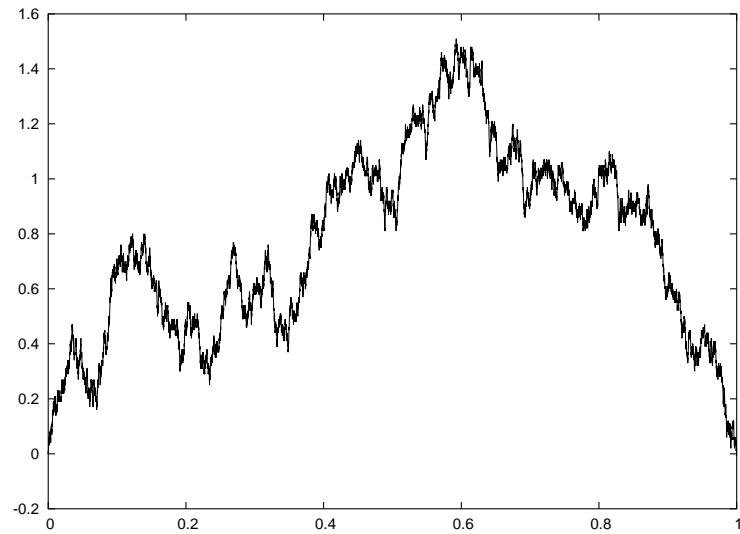
$X_n, X : \Omega \rightarrow S$... S -valued random variables $((S, d)$... Polish space)

$$\boxed{X_n \xrightarrow{d} X} \quad :\Leftrightarrow \quad \boxed{\lim_{n \rightarrow \infty} \mathbb{E} G(X_n) = \mathbb{E} G(X)}$$

for all **bounded** continuous
functionals $G : S \rightarrow \mathbb{R}$

Weak Convergence

Stochastic process: random function



Weak Convergence

Stochastic process

$X_n : \Omega \rightarrow C[0, 1]$ sequence of stochastic processes, $X : \Omega \rightarrow C[0, 1]$

- $X_n \xrightarrow{d} X \implies F(X_n) \xrightarrow{d} F(X)$ for all continuous $F : S \rightarrow S'$.
- $X_n \xrightarrow{d} X \implies X_n(t_0) \xrightarrow{d} X(t_0)$ for all fixed $t_0 \in [0, 1]$.
- $X_n \xrightarrow{d} X \implies (X_n(t_1), \dots, X_n(t_k)) \xrightarrow{d} (X(t_1), \dots, X(t_k))$
for all $k \geq 1$ and all fixed $t_1, \dots, t_k \in [0, 1]$.

The converse statement is not necessarily true, one needs **tightness**.

Weak Convergence

Stochastic process

$X_n : \Omega \rightarrow C[0, 1]$ sequence of stochastic processes, $X : \Omega \rightarrow C[0, 1]$

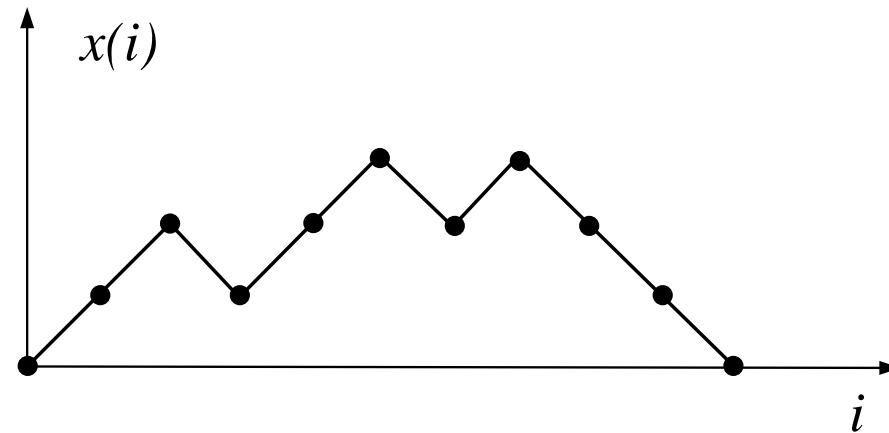
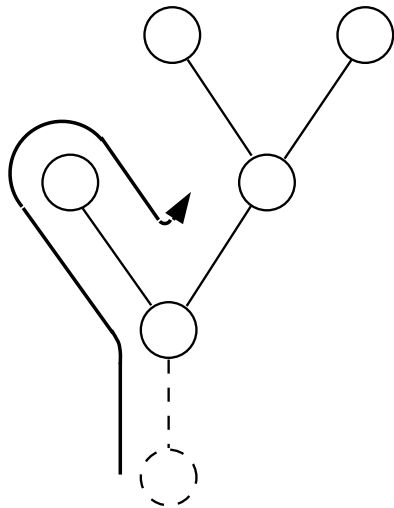
1. $(X_n(t_1), \dots, X_n(t_k)) \xrightarrow{d} (X(t_1), \dots, X(t_k))$
for all $k \geq 1$ and all fixed $t_1, \dots, t_k \in [0, 1]$
2. $\mathbb{E}(|X_n(0)|^\beta) \leq C$
for some constant $C > 0$ and an exponent $\beta > 0$
3. $\mathbb{E}(|X_n(t) - X_n(s)|^\beta) \leq C|t - s|^\alpha$ for all $s, t \in [0, 1]$
for some constant $C > 0$ and exponents $\alpha > 1$ and $\beta > 0$.

Then

$$\boxed{(X_n(t), 0 \leq t \leq 1) \xrightarrow{d} (X(t), 0 \leq t \leq 1)}.$$

Depth-First-Search

Rooted trees and discrete excursions



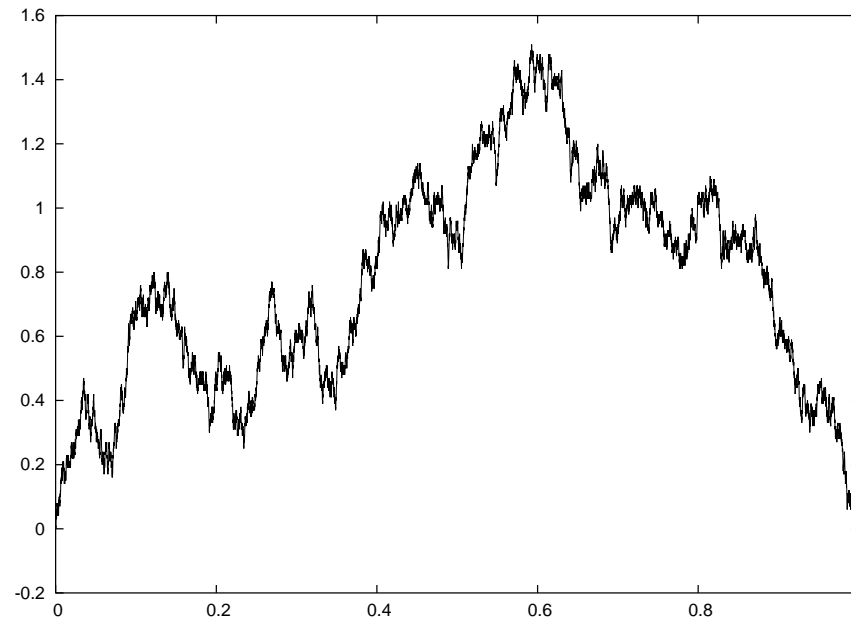
Bijection between

Catalan trees \longleftrightarrow Dyck paths

random trees of size n \longleftrightarrow **random** Dyck paths of length $2n$

Depth-First-Search

Brownian excursion ($e(t), 0 \leq t \leq 1$)



Rescaled Brownian motion between 2 zeros.

Random function in $C[0, 1]$.

Depth-First-Search

Kaigh's Theorem

$(X_n(t), 0 \leq t \leq 2n)$... random Dyck path of length $2n$.

$$\implies \left(\frac{1}{\sqrt{2n}} X_n(2nt), 0 \leq t \leq 1 \right) \xrightarrow{d} (2e(t), 0 \leq t \leq 1).$$

Remark. This theorem also holds for more general random walks with independent increments conditioned to be an excursion.

Real Trees

T ... tree, \mathcal{T} ... embedding of T into the plane \mathbb{R}^2

$\implies \mathcal{T}$ is a metric space (and a **real tree** in the following sense):

Definition

A metric space (\mathcal{T}, d) is a **real tree** if the following two properties hold for every $x, y \in \mathcal{T}$.

1. There is a unique isometric map $h_{x,y} : [0, d(x, y)] \rightarrow \mathcal{T}$ such that $h_{x,y}(0) = x$ and $h_{x,y}(d(x, y)) = y$.

2. If q is a continuous injective map from $[0, 1]$ into \mathcal{T} with $q(0) = x$ and $q(1) = y$ then

$$q([0, 1]) = h_{x,y}([0, d(x, y)]).$$

A rooted real tree (\mathcal{T}, d) is a real tree with a distinguished vertex $r = r(\mathcal{T})$ called the root.

Real Trees

Two real trees (\mathcal{T}_1, d_1) , (\mathcal{T}_2, d_2) are **equivalent** if there is a root-preserving isometry that maps \mathcal{T}_1 onto \mathcal{T}_2 .

\mathbb{T} ... set of all equivalence classes of rooted compact real trees.

Gromov-Hausdorff Distance $d_{\text{GH}}(\mathcal{T}_1, \mathcal{T}_2)$ of two real trees $\mathcal{T}_1, \mathcal{T}_2$ is the infimum of the Hausdorff distance of all isometric embeddings of $\mathcal{T}_1, \mathcal{T}_2$ into the same metric space.

Hausdorff distance: $\delta_{\text{Haus}}(X, Y) = \max \left\{ \sup_{x \in X} \inf_{y \in Y} d(x, y), \sup_{y \in Y} \inf_{x \in X} d(x, y) \right\}$

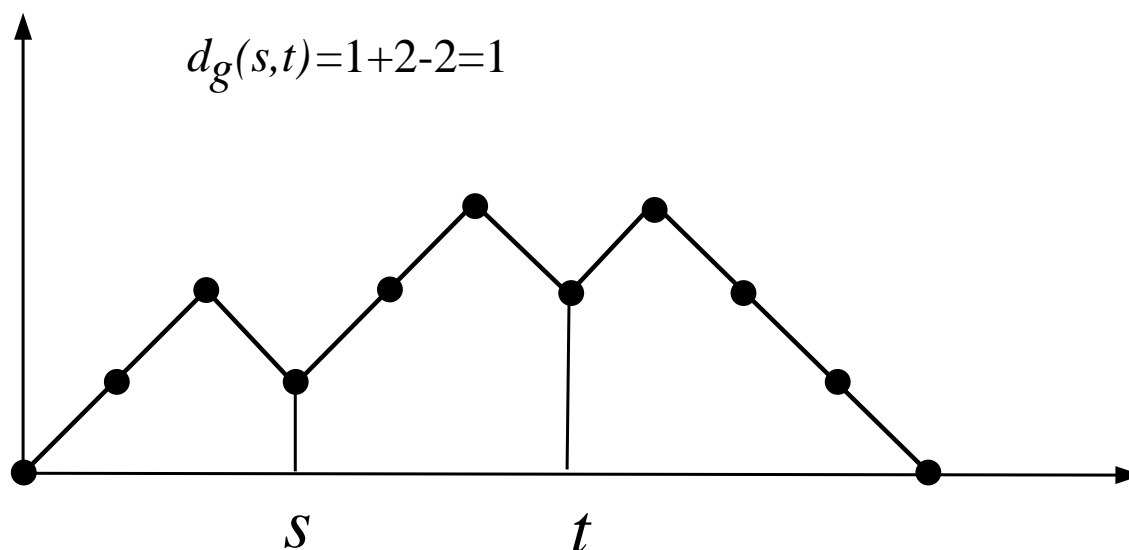
Theorem

The metric space $(\mathbb{T}, d_{\text{GH}})$ is a Polish space.

Real Trees

$g : [0, 1] \rightarrow [0, \infty)$... continuous, ≥ 0 , $g(0) = g(1) = 0$

$$d_g(s, t) = g(s) + g(t) - 2 \inf_{\min\{s, t\} \leq u \leq \max\{s, t\}} g(u)$$



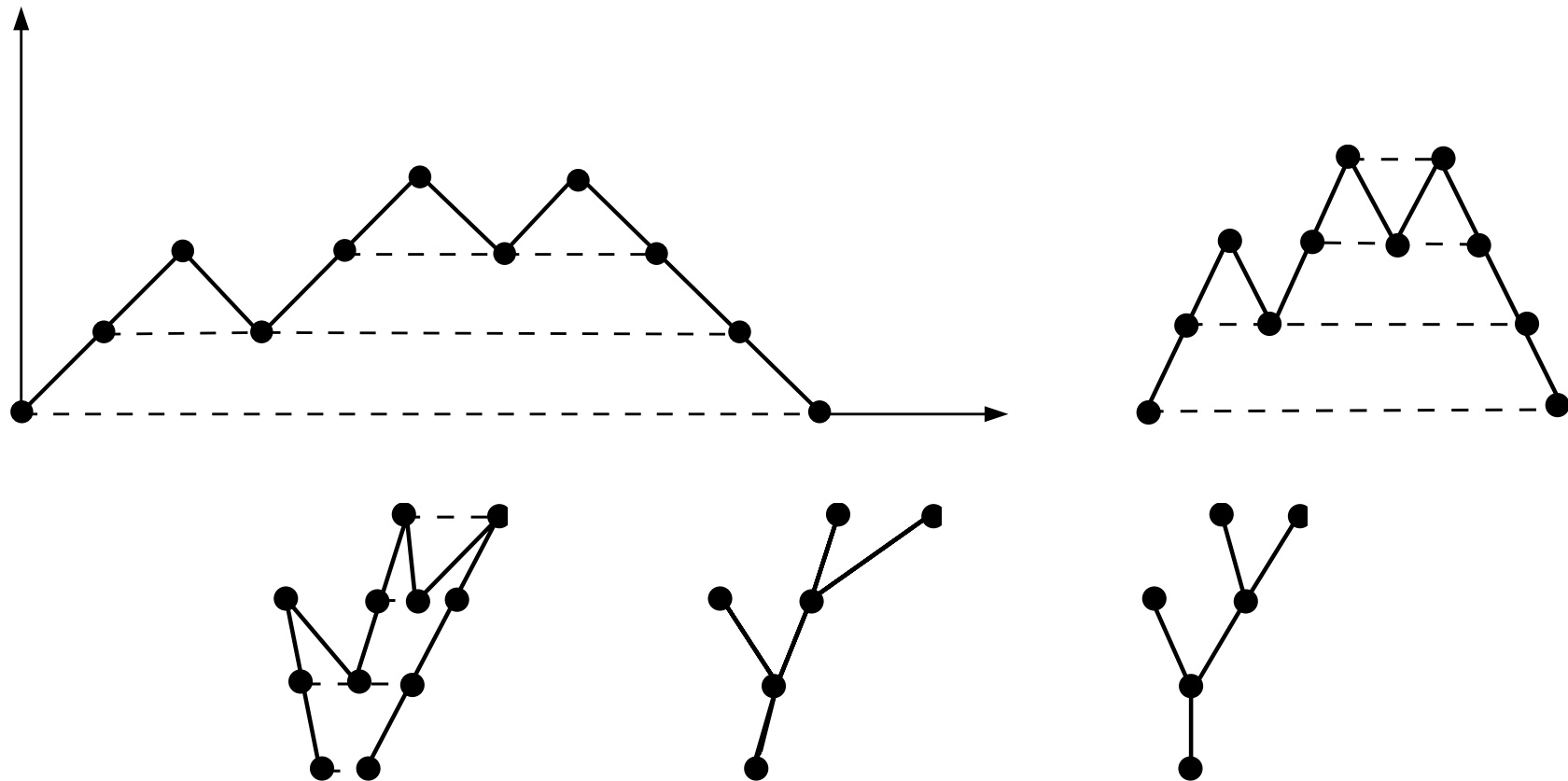
$$s \sim t \iff d_g(s, t) = 0$$

$$\mathcal{T}_g = [0, 1] / \sim$$

$\implies (\mathcal{T}_g, d_g)$ is a compact real tree.

Real Trees

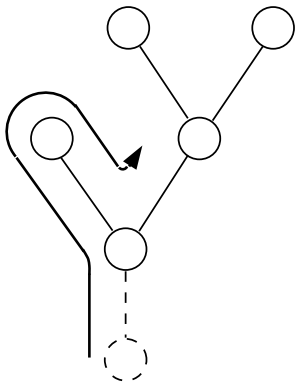
Construction of a real tree \mathcal{T}_g



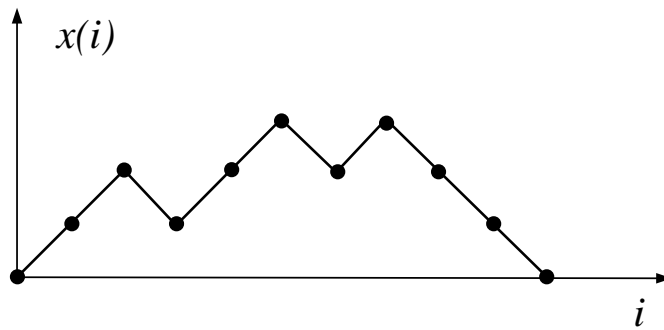
The mapping $C[0, 1] \rightarrow \mathbb{T}$, $g \mapsto \mathcal{T}_g$ is **continuous**.

Real Trees

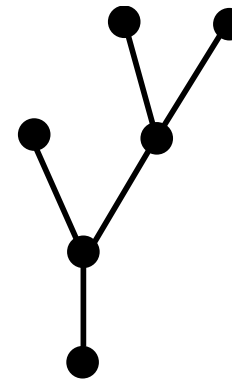
Catalan trees as real trees



T_n



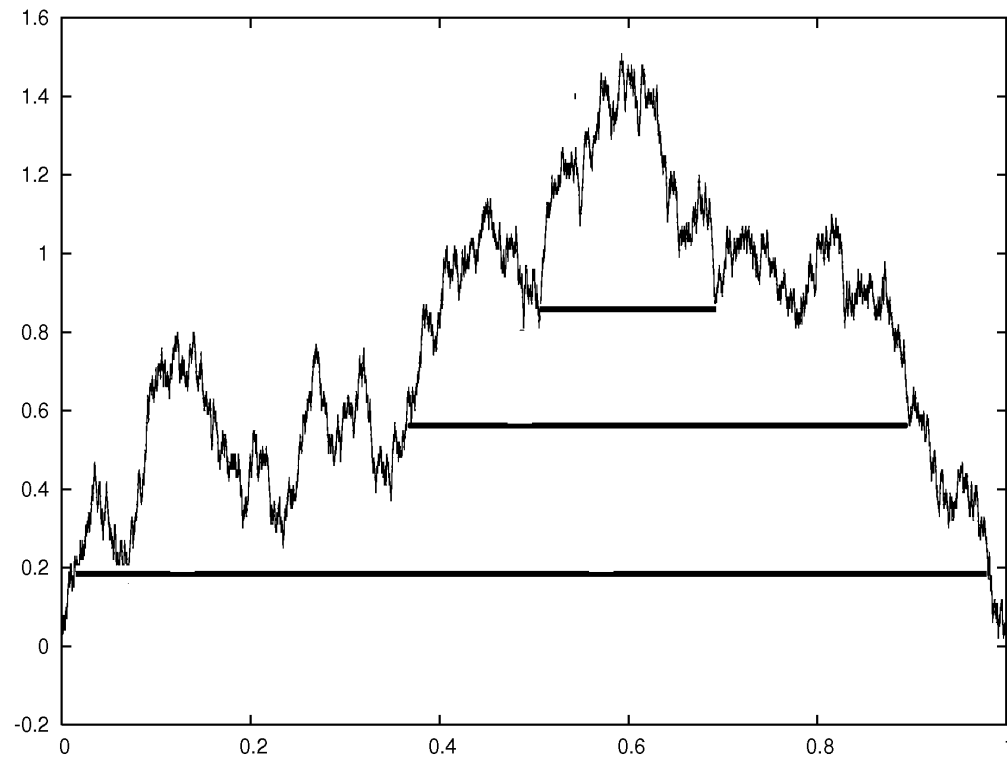
$X_n = X_{T_n}$



\mathcal{T}_{X_n}

Real Trees

Continuum random tree \mathcal{T}_{2e} (with Brownian excursion $e(t)$)



Real Trees

Theorem

$(X_n(t), 0 \leq t \leq 2n)$... random Dyck paths of length $2n$
or the depth-first-search process of Catalan trees of size n .

$$\implies \boxed{\frac{1}{\sqrt{2n}} \mathcal{T}_{X_n} \xrightarrow{d} \mathcal{T}_{2e}}$$

In other words...

Scaled Catalan trees (interpreted as “real trees”) converge weakly to the continuum random tree.

Galton-Watson Trees

Galton-Watson branching process

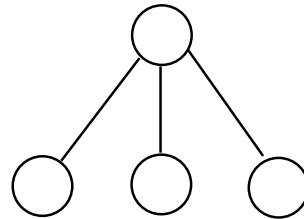
ξ ... offspring distribution, $\varphi_k = \mathbb{P}\{\xi = k\}$, $\varphi_0 > 0$



Galton-Watson Trees

Galton-Watson branching process

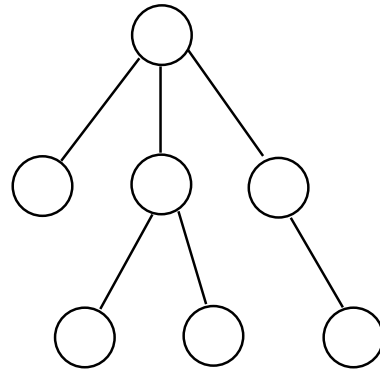
ξ ... offspring distribution, $\varphi_k = \mathbb{P}\{\xi = k\}$, $\varphi_0 > 0$



Galton-Watson Trees

Galton-Watson branching process

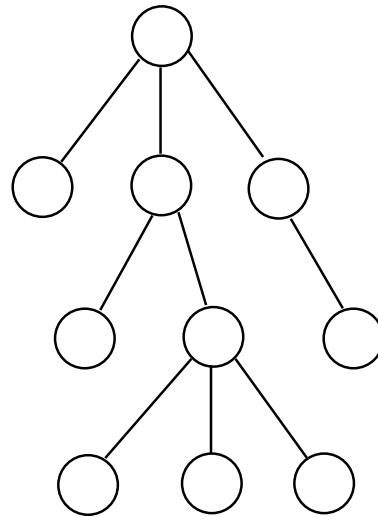
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Galton-Watson Trees

Galton-Watson branching process

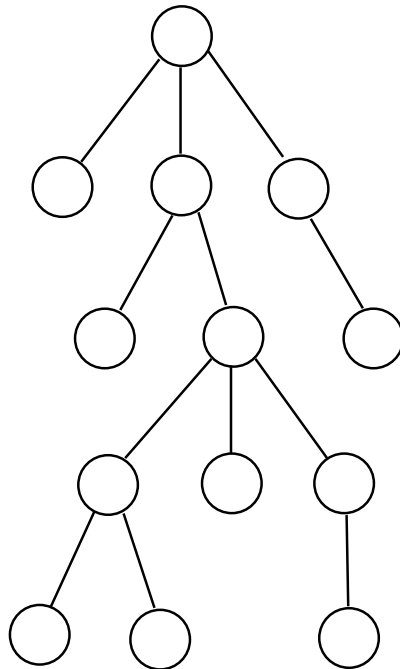
ξ ... offspring distribution, $\varphi_k = \mathbb{P}\{\xi = k\}$, $\varphi_0 > 0$



Galton-Watson Trees

Galton-Watson branching process

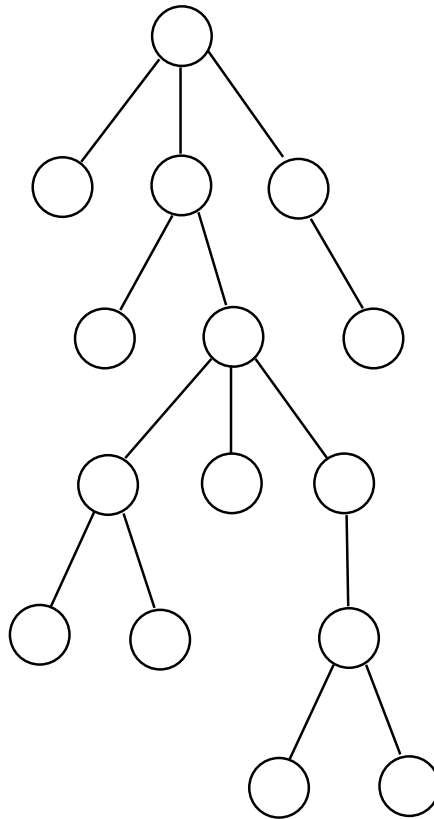
ξ ... offspring distribution, $\varphi_k = \mathbb{P}\{\xi = k\}$, $\varphi_0 > 0$



Galton-Watson Trees

Galton-Watson branching process

ξ ... offspring distribution, $\varphi_k = \mathbb{P}\{\xi = k\}$, $\varphi_0 > 0$



Galton-Watson Trees

Galton-Watson branching process. $(Z_k)_{k \geq 0}$

$Z_0 = 1$, and for $k \geq 1$

$$Z_k = \sum_{j=1}^{Z_{k-1}} \xi_j^{(k)},$$

where the $(\xi_j^{(k)})_{k,j}$ are iid random variables distributed as ξ .

Z_k ... number of nodes in k -th generation

$Z = Z_0 + Z_1 + Z_2 + \dots$... total progeny

Galton-Watson Trees

Generating functions

$$y_n = \mathbb{P}\{Z = n\}, \quad y(x) = \sum_{n \geq 1} y_n x^n$$

$$\Phi(w) = \mathbb{E} w^\xi = \sum_{k \geq 0} \varphi_k w^k$$

$$\implies \boxed{y(x) = x \Phi(y(x))}$$

Conditioned Galton-Watson tree

GW-branching process conditioned on the total progeny $Z = n$.

Galton-Watson Trees

Example. $\mathbb{P}\{\xi = k\} = 2^{-k-1}$, $\Phi(w) = 1/(2 - w)$

\implies all trees of size n have the same probability

\implies conditioned GW-tree of size n is the same model as the **Catalan tree model** (with the uniform distribution on trees of size n)

Example. $\Phi(w) = \frac{1}{2}(1 + w)^2$: **binary trees** with n internal nodes.

Example. $\Phi(w) = \frac{1}{3}(1 + w + w^2)$: **Motzkin trees**

Example. $\Phi(w) = e^{w-1}$: **Cayley trees**

Galton-Watson Trees

General assumption: $\mathbb{E} \xi = 1$, $0 < \text{Var} \xi = \sigma^2 < \infty$

Theorem (Aldous)

$X_n(t)$... depth-first-search of conditioned GW-trees of size n

$$\implies \left(\frac{\sigma}{2\sqrt{n}} X_n(2nt), 0 \leq t \leq 1 \right) \xrightarrow{d} (e(t), 0 \leq t \leq 1).$$

Corollary

$$\frac{\sigma}{\sqrt{n}} \mathcal{T}_{X_n} \xrightarrow{d} \mathcal{T}_{2e}$$

Galton-Watson Trees

Corollary H_n ... height of conditioned GW-trees of size n :

$$\implies \boxed{\frac{1}{\sqrt{n}}H_n \xrightarrow{d} \frac{2}{\sigma} \max_{0 \leq t \leq 1} e(t)}$$

Remark. Distribution function of $\max_{0 \leq t \leq 1} e(t)$:

$$\mathbb{P}\left\{\max_{0 \leq t \leq 1} e(t) \leq x\right\} = 1 - 2 \sum_{k=1}^{\infty} (4x^2k^2 - 1)e^{-2x^2k^2}$$

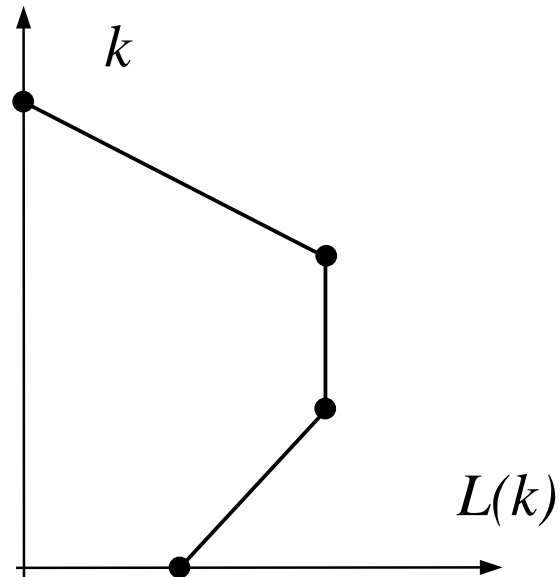
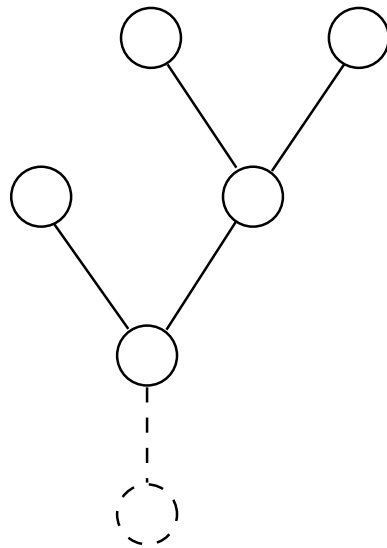
Galton-Watson Trees

Profile

$L_T(k)$... number of nodes at distance k from the root

$(L_T(k))_{k \geq 0}$... profile of T

$(L_T(s), s \geq 0)$... linearly interpolated profile of T



Galton-Watson Trees

Value distribution

$$\mu_T = \frac{1}{|T|} \sum_{k \geq 0} L_T(k) \delta_k$$

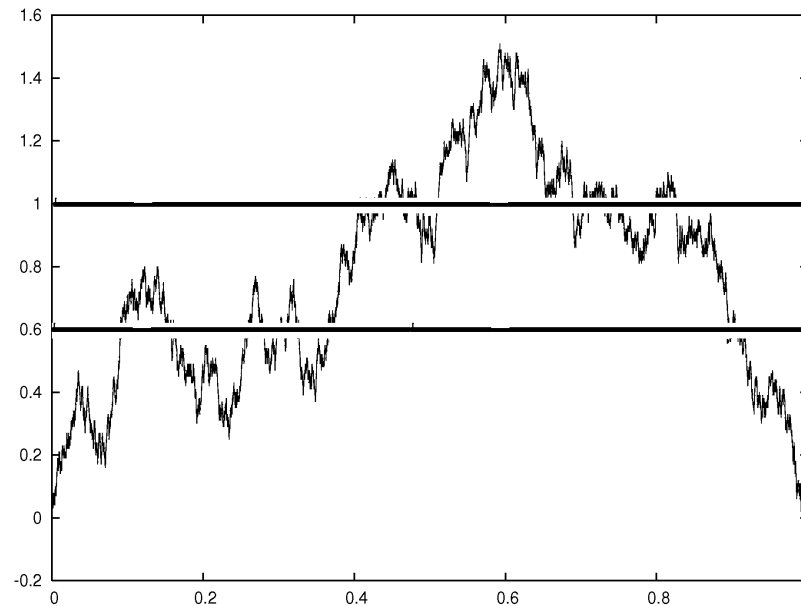
δ_x ... δ -distribution concentrated at x

Galton-Watson Trees

Occupation measure: random measure on \mathbb{R}

$$\mu(A) = \int_0^1 \mathbf{1}_A(e(t)) dt$$

measure how long $e(t)$ stays in set A



Galton-Watson Trees

Theorem (Aldous)

$(L_n(k), k \geq 0)$... random profile of conditioned GW-trees of size n

$$\implies \boxed{\frac{1}{n} \sum_{k \geq 0} L_n(k) \delta_{(\sigma/2)k/\sqrt{n}} \xrightarrow{d} \mu}$$

Galton-Watson Trees

Local time of the Brownian excursion: random density of μ

$$l(s) = \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int_0^1 \mathbf{1}_{[s, s+\varepsilon]}(e(t)) dt$$

Theorem (D.+Gittenberger)

$(L_n(s), s \geq 0)$... random profile of conditioned GW-trees of size n

$$\implies \left(\frac{1}{\sqrt{n}} L_n(s\sqrt{n}), s \geq 0 \right) \xrightarrow{d} \left(\frac{\sigma}{2} l \left(\frac{\sigma}{2} s \right), s \geq 0 \right)$$

Proof with asymptotics on generating functions (very involved)!!!

Galton-Watson Trees

Width

$$W = \max_{k \geq 0} L(k) = \max_{t \geq 0} L(t),$$

maximal number of nodes in a level.

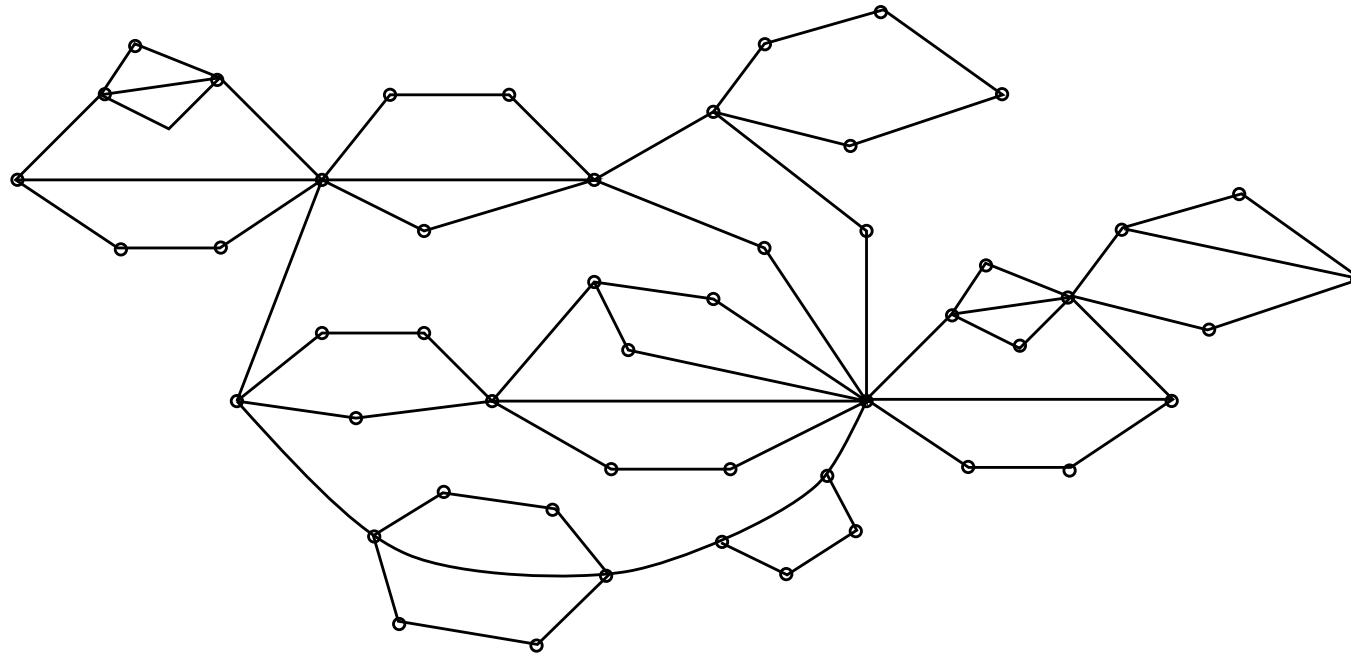
Corollary

$$\frac{1}{\sqrt{n}} W_n \xrightarrow{d} \frac{\sigma}{2} \sup_{0 \leq t \leq 1} l(t)$$

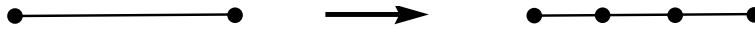
Remark. $\sup_{t \geq 0} l(t) = 2 \sup_{0 \leq t \leq 1} e(t)$ (in distribution)

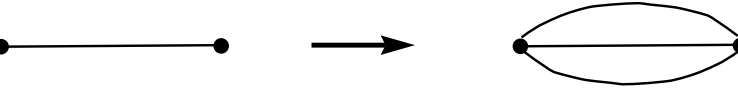
Series-Parallel Graphs

Connected Series-Parallel Graphs



Series-parallel extension of a tree (or no K_4 as a minor)

Series-extension: 

Parallel-extension: 

Scaling Limit of Series Parallel Graphs

A typical series-parallel graph of size n has $\approx c_1 n$ 2-connected components that form a **tree**

The 2-connected components do not scale in distribution, their expected size is finite and they behave *almost) independent and identically distributed*.

*So, series-parallel graphs look **tree-like**.*

Scaling Limit of Series Parallel Graphs

Theorem (Panagiotou, Stufler, and Weller)

C_n ... connected, vertex labelled series-parallel graphs with n vertices

$$\frac{c}{\sqrt{n}} C_n \xrightarrow{d} \mathcal{T}_{2e}$$

for some constant $c > 0$.

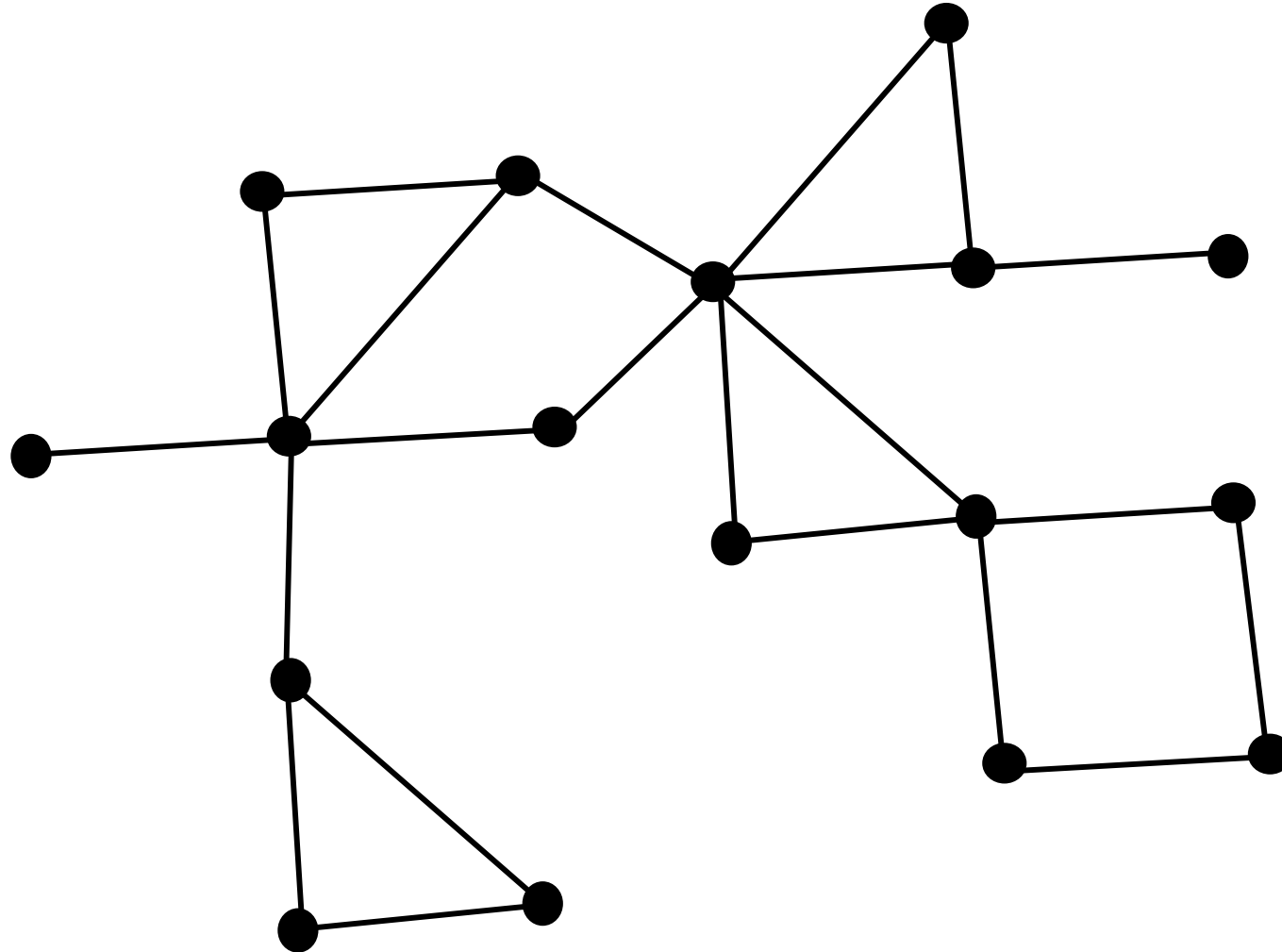
Remark. The same result holds for so-called **subcritical graph classes** like cacti-graphs, outerplanar graphs etc. In all these graph classes the diameter is of order \sqrt{n} .

Contents 4

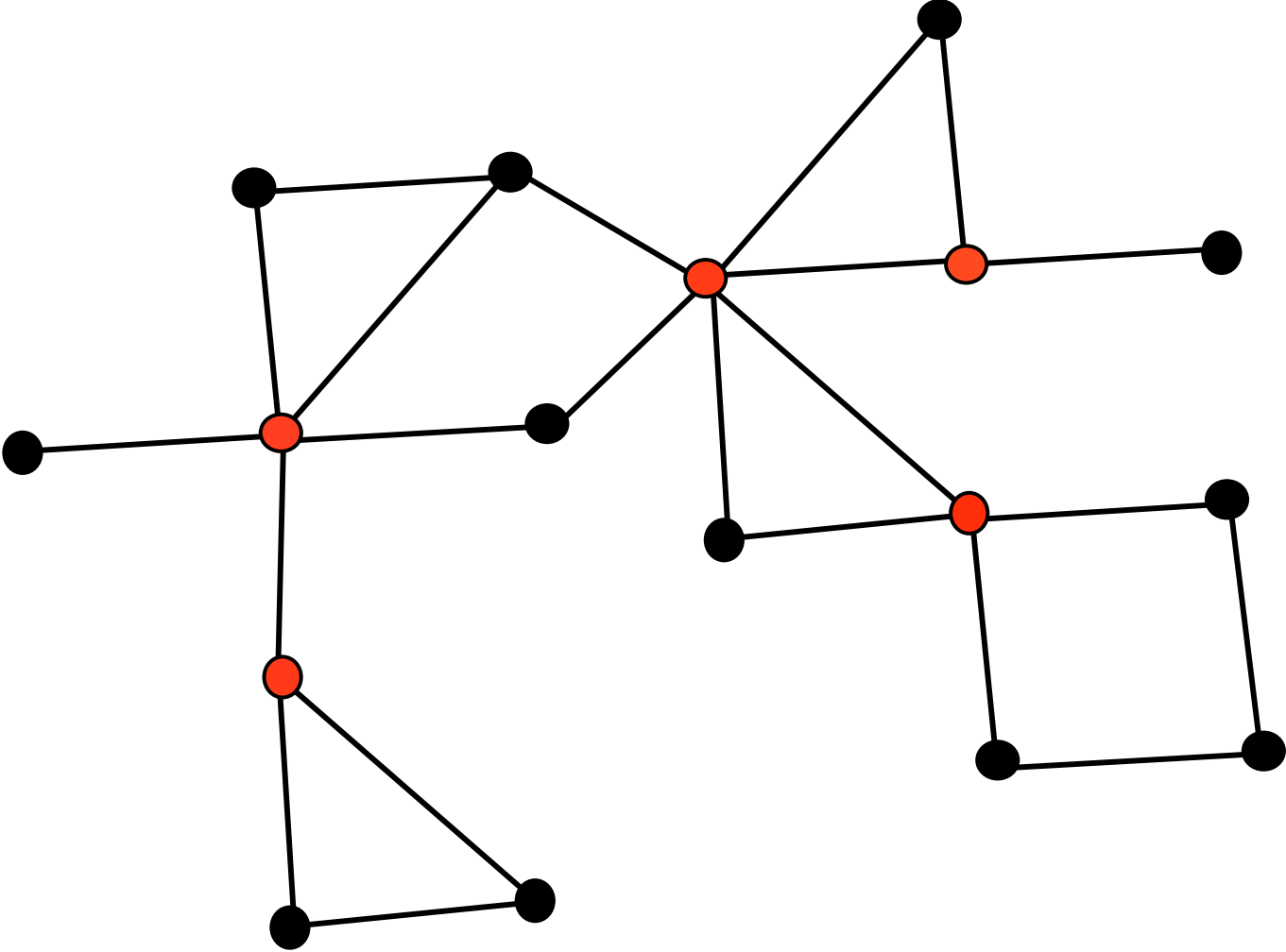
IV. SUBGRAPH COUNTS IN SERIES PARALLEL GRAPHS

- *Sub-critical graph classes*
- *Asymptotic counting of sub-critical graph classes*
- *Series parallel graphs are sub-critical*
- *Subgraph counting*
- *A combinatorial CLT for infinite systems*

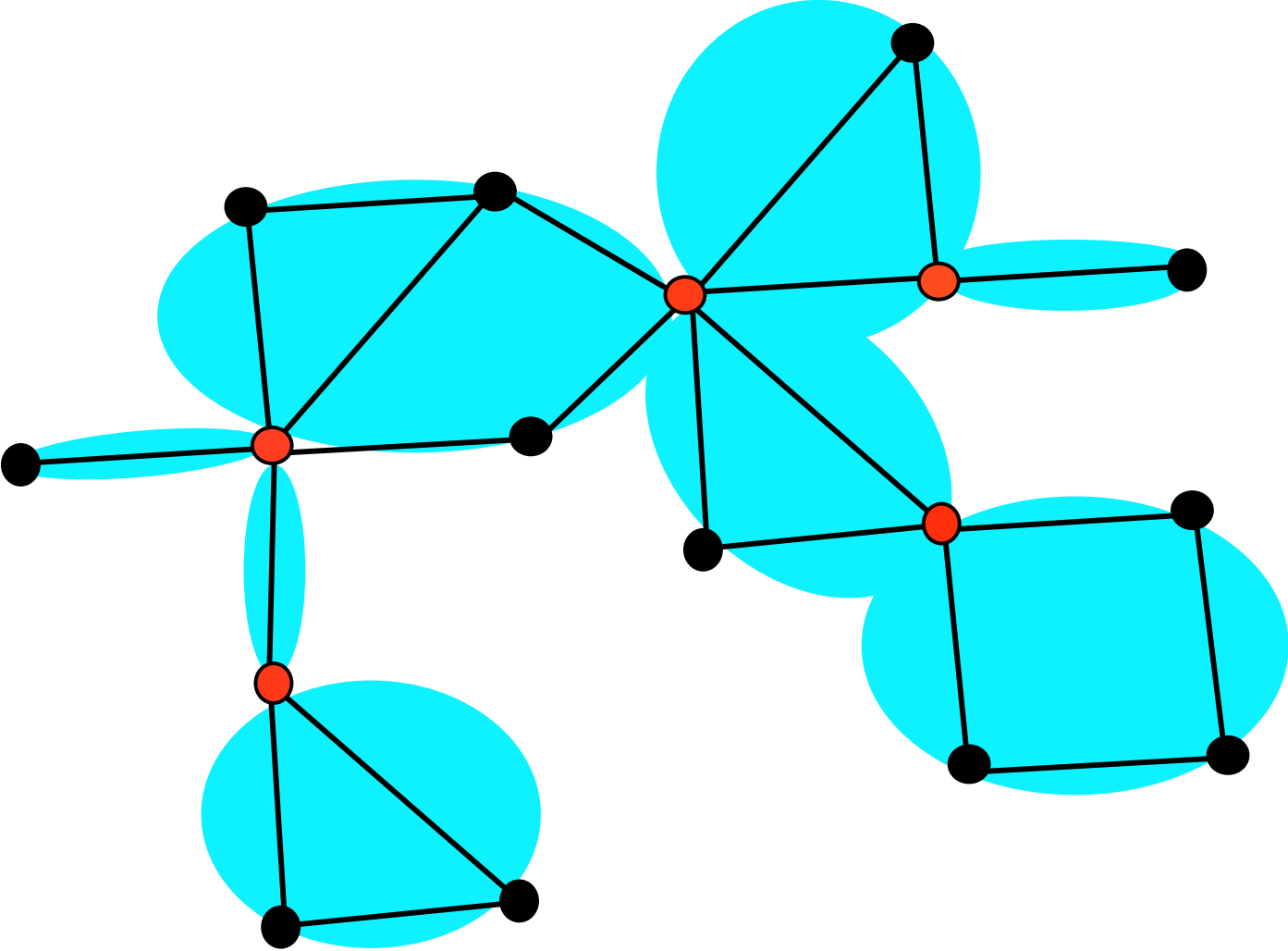
Block-Decomposition



Block-Decomposition



Block-Decomposition



Block-Decomposition

block: *2-connected component (= maximal 2-connected subgraph)*

Block-stable graph class \mathcal{G} : *\mathcal{G} contains the one-edge graph and $G \in \mathcal{G}$ if and only if all blocks of G are contained in \mathcal{G} .*

Equivalently, the 2-connected graphs of \mathcal{G} and the one-edge graph generate all graphs of \mathcal{G} .

Examples: *Planar graphs, series-parallel graphs, minor-closed graph classes etc.*

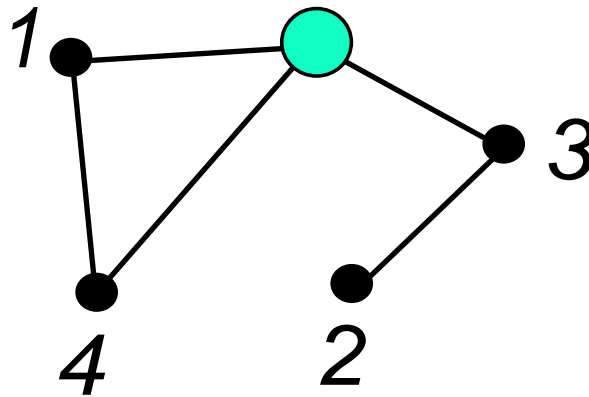
$B(x)$... GF for 2-connected graphs in \mathcal{G}

$C(x)$... GF for connected graphs in \mathcal{G}

[We will consider here only connected graphs]

Generating Functions for Block-Decomposition

Vertex-rooted graphs: *one vertex (the root) is distinguished (and usually discounted, that is, it gets no label)*

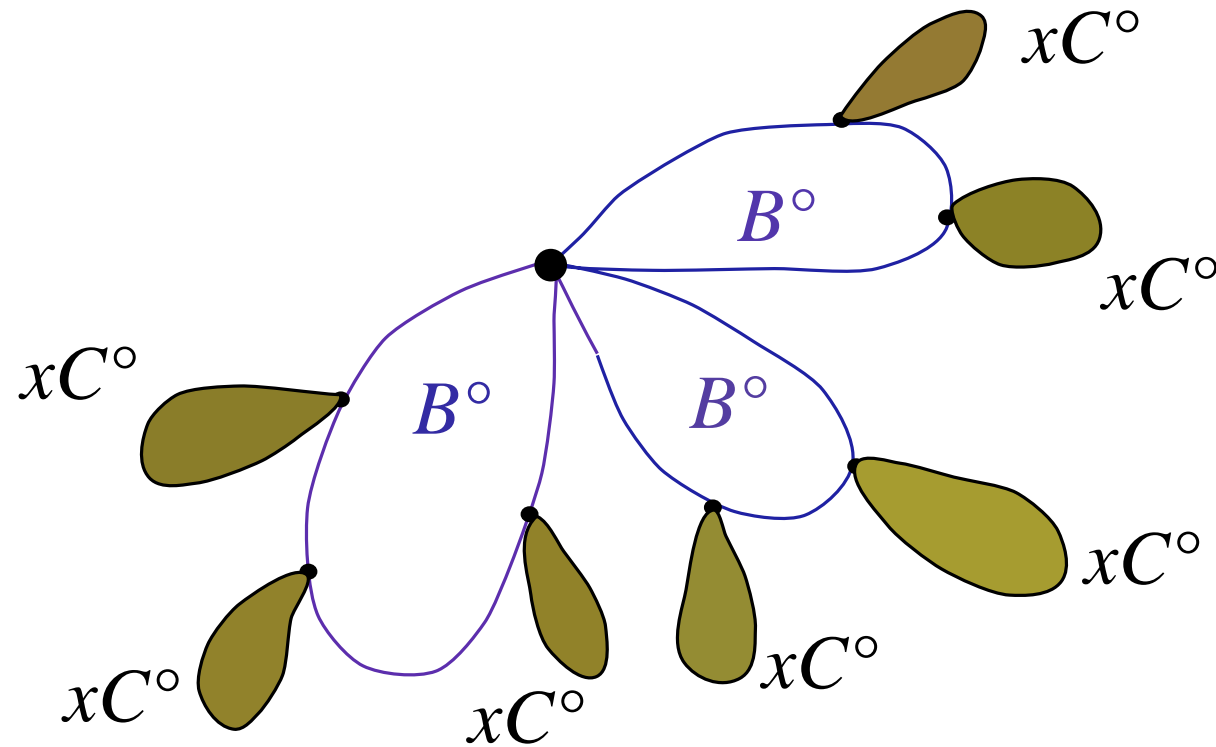


Generating function: *(in den labelled case)*

$$G^\bullet(x) = G'(x)$$

Generating Functions for Block-Decomposition

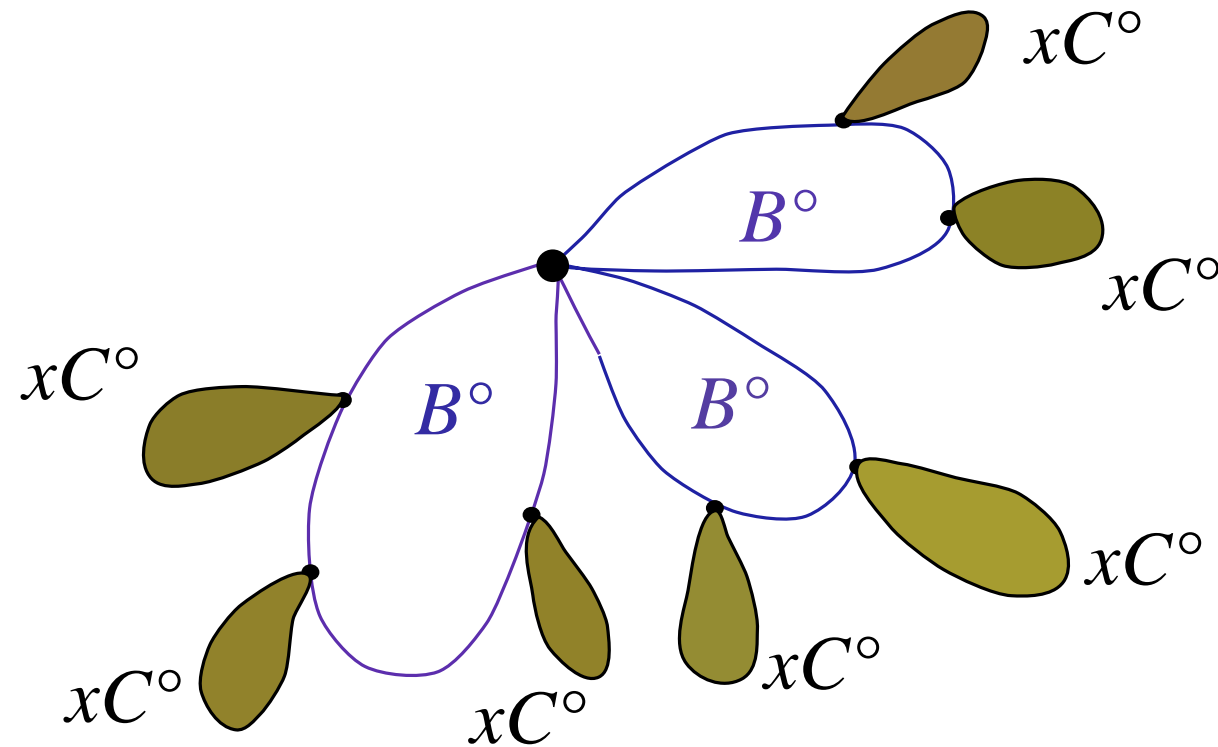
(in the labelled case)



$$C^\bullet(x) = e^{B^\bullet(xC^\bullet(x))}$$

Generating Functions for Block-Decomposition

(in the labelled case)



$$\frac{\partial C(x, y)}{\partial x} = \exp \left(\frac{\partial B}{\partial x} \left(x \frac{\partial C(x, y)}{\partial x}, y \right) \right)$$

Labelled Trees

Rooted Trees:

$$B^\bullet(x) = x$$



$R(x) = xC^\bullet(x)$... *generating function of **rooted, labelled trees***

$$\boxed{C^\bullet(x) = e^{B^\bullet(xC^\bullet(x))}} \implies \boxed{R(x) = xe^{R(x)}}$$

Remark: $T(x)$... *GF for unrooted labelled trees:*

$$T(x)' = \frac{1}{x}R(x) \implies T(x) = R(x) - \frac{1}{2}R(x)^2$$

Outerplanar Graphs

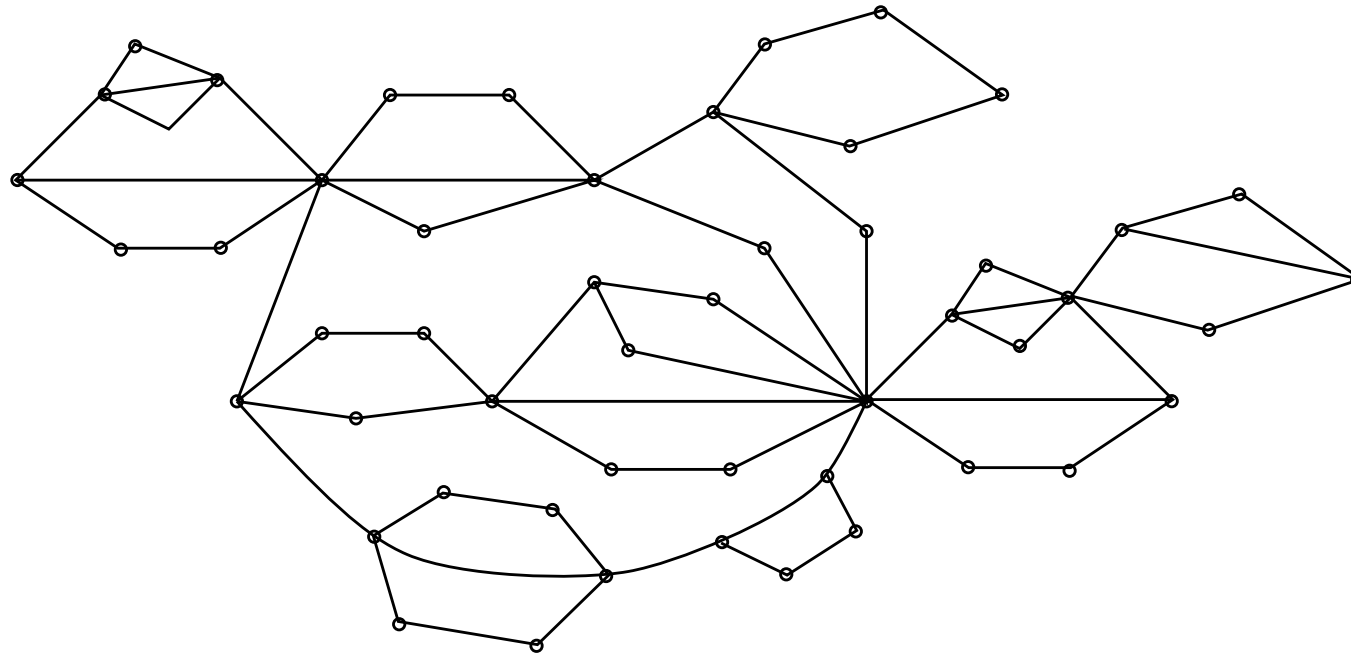
Generating functions

$$\boxed{C^\bullet(x) = e^{B^\bullet(xC^\bullet(x))}},$$


$$B^\bullet(x) = \frac{1 + 5x - \sqrt{1 - 6x + x^2}}{8}.$$

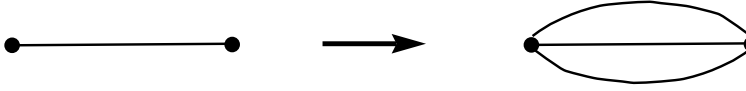
2-connected outerplanar graphs = dissections of the n -gon

Series-Parallel Graphs



Series-parallel extension of a tree (if we restrict to connected graphs)

Series-extension: 

Parallel-extension: 

Series-Parallel Graphs

Equivalent Definitions

- $Ex(K_4)$
- $tree-width \leq 2$
- *nested ear decomposition (if connected)*

Series-Parallel Graphs

Generating functions

$$\boxed{\frac{\partial C(x, y)}{\partial x} = \exp \left(\frac{\partial B}{\partial x} \left(x \frac{\partial C(x, y)}{\partial x}, y \right) \right)},$$

$$\frac{\partial B(x, y)}{\partial y} = \frac{x^2}{2} e^{S(x, y)},$$

$$S(x, y) = \frac{x(P(x, y) + y)^2}{1 - x(P(x, y) + y)},$$

$$P(x, y) = (e^{S(x, y)} - 1 - S(x, y)) + y(e^{S(x, y)} - 1).$$

Sub-critical Graphs

Repetition: Functional equations

Suppose that $A(x) = \Phi(x, A(x))$, where $\Phi(x, a)$ has a power series expansion at $(0, 0)$ with non-negative coefficients and $\Phi_{aa}(x, a) \neq 0$.

Let $x_0 > 0$, $a_0 > 0$ (inside the region of convergence of Φ) satisfy the system of equations:

$$a_0 = \Phi(x_0, a_0), \quad 1 = \Phi_a(x_0, a_0).$$

Then there exists analytic function $g(x), h(x)$ such that locally

$$A(x) = g(x) - h(x) \sqrt{1 - \frac{x}{x_0}}.$$

Remark. If there is no x_0, a_0 inside the region of convergence of Φ then the singular behaviour of Φ determines the singular behaviour of $A(x)$!!!

Sub-critical Graphs

$$A(x) = xC^\bullet(x), \quad \Phi(x, a) = xe^{B^\bullet(a)}, \quad \boxed{xC^\bullet(x) = xe^{B^\bullet(xC^\bullet(x))}}$$

$$\implies \boxed{A(x) = \Phi(x, A(x))}$$

A block-stable graph class is called **sub-critical** if the system (note that $B^\bullet(x) = B'(x)$)

$$a_0 = x_0 e^{B'(a_0)}, \quad 1 = x_0 e^{B'(a_0)} B''(a_0)$$

has positive solutions x_0, a_0 inside the region of convergence of $\Phi(x, a) = xe^{B^\bullet(a)}$. In particular we get a **squareroot singularity** for $C^\bullet(x)$.

This means that “ a_0 is smaller than the radius of convergence η of B^\bullet ”.

Eliminating x_0 leads to $\boxed{a_0 B''(a_0) = 1}$ or that

$$\boxed{\eta B''(\eta) > 1}$$

where η is the radius of convergence of $B(x)$.

Sub-critical Graphs

- **Trees** *are sub-critical*
- **Outerplanar graphs** *are sub-critical*
- **Series-parallel graphs** *are sub-critical*

Sub-critical Graphs

Lemma. *Suppose that $B(x)$ has radius of convergence $\eta \in (0, \infty]$.*

$$\lim_{x \rightarrow \eta} B''(x) = \infty \implies \text{sub-critical.}$$

Corollary *If $B^\bullet(x) = B'(x)$ is entire or has a squareroot singularity:*

$$B^\bullet(x) = g(x) - h(x) \sqrt{1 - \frac{x}{\eta}},$$

*then we are in the **sub-critical** case.*

*This applies for **outerplanar** and **series-parallel** graphs.*

Sub-critical Graphs

What does “**sub-critical**” mean?

In a sub-critical graph class the average size of the 2-connected components is bounded.

*⇒ This leads to a **tree like structure**.*

*⇒ The **law of large numbers** should apply so that we can expect **universal behaviors** that are independent of the the precise structure of 2-connected components.*

Sub-critical Graphs

Universal properties

- Asymptotic enumeration:

Labelled case:

$$c_n \sim c n^{-5/2} \rho^{-n} n!$$

Unlabelled case:

$$c_n \sim c n^{-5/2} \rho^{-n}$$

($c > 0$, ρ ... radius of convergence of $C(z)$)

[D.+Fusy+Kang+Kraus+Rue 2011]

Sub-critical Graphs

- Asymptotic enumeration:

$$C^\bullet(x) = e^{B^\bullet(xC^\bullet(x))}$$

$$\longrightarrow xC^\bullet(x) = xC'(x) = g(x) - h(x)\sqrt{1 - \frac{x}{\rho}}$$

$$\longrightarrow [x^n]xC'(x) = \frac{nc_n}{n!} \sim cn^{-3/2}\rho^{-n}$$

$$\longrightarrow \boxed{c_n \sim cn^{-5/2}\rho^{-n}n!}.$$

Additive Parameters in Subcritical Graph Classes

Theorem 1 [D.+Fusy+Kang+Kraus+Rue]

X_n ... number of **edges** / number of **blocks** / number of **cut-vertices**
/ number of **vertices of degree k**

$$\implies \boxed{\frac{X_n - \mu n}{\sqrt{n}} \rightarrow N(0, \sigma^2)}$$

with $\mu > 0$ and $\sigma^2 \geq 0$.

Remark. There is an easy to check “combinatorial condition” that ensures $\sigma^2 > 0$.

Additive Parameters in Subcritical Graph Classes

Proof Methods:

Refined versions of the functional equation $C^\bullet(x) = e^{B^\bullet(xC^\bullet(x))}$,
+ singularity analysis (**always squareroot singularity**)

E.g: number of edges:

$$C^\bullet(x, y) = e^{B^\bullet(xC^\bullet(x,y), y)}$$

or number of 2-connected components:

$$C^\bullet(x, y) = e^{yB^\bullet(xC^\bullet(x,y))}$$

$$\longrightarrow C^\bullet(x, y) = g(x, y) - h(x, y) \sqrt{1 - \frac{x}{\rho(y)}}$$

$$\longrightarrow [x^n]C^\bullet(x, y) \sim c(y)\rho(y)^{-n}n^{-3/2}$$

+ application of Quasi-Power-Theorem (by Hwang).

Graph Limits

\mathcal{T}_e ... continuum random tree (CRT)

Theorem [Panagiotou+Stufler+Weller]

\mathcal{C} ... sub-critical graph class of connected graphs

$$\implies \boxed{\frac{c}{\sqrt{n}} \mathcal{C}_n \rightarrow \mathcal{T}_e}$$

with respect to the Gromov-Hausdorff metric, where $c > 0$ is a constant.

Corollary. The diameter D_n as well as a typical distance in a sub-critical graph is of order \sqrt{n} .

Subgraph Counting

Theorem [D.+Ramos+Rue]

\mathcal{G} ... sub-critical graph class, $H \in \mathcal{G}$ fixed.

$X_n^{(H)}$... number of occurrences of H as a subgraph in graphs of size n

$$\implies \boxed{\frac{X_n^{(H)} - \mu n}{\sqrt{n}} \rightarrow N(0, \sigma^2)}$$

with $\mu > 0$ and $\sigma^2 \geq 0$.

Remark. The proof is easy if H is 2-connected = **additive parameter!!!**

Subgraph Counting

$H = P_2$... path of length 2

$B_j^\bullet(w_1, w_2, w_3, \dots; u)$ generating function of blocks in \mathcal{G} , where the root has degree j , where w_i counts the number of non-root vertices of degree i , and where u counts the number of occurrences of $H = P_2$.

$C_j^\bullet(x, u)$... generating function of connected rooted graphs in \mathcal{G} , where the root vertex has degree j , where x counts the number of (all) vertices and u the number of occurrences of $H = P_2$.

Subgraph Counting

System of infinite number of equations

$$C_j^\bullet(x, u) = \sum_{s \geq 0} \frac{1}{s!} \sum_{j_1 + \dots + j_s = j} u^{\sum_{i_1 < i_2} j_{i_1} j_{i_2}} \\ \times \prod_{i=1}^s B_{j_i}^\bullet \left(x \sum_{\ell_1 \geq 0} u^{\ell_1} C_{\ell_1}^\bullet(x, u), x \sum_{\ell_2 \geq 0} u^{2\ell_2} C_{\ell_2}^\bullet(x, u), \dots; u \right), \\ (j \geq 0)$$

$$C_j^\bullet(x, 1) = \sum_{s \geq 0} \frac{1}{s!} \sum_{j_1 + \dots + j_s = j} \prod_{i=1}^s B_{j_i}^\bullet(x C^\bullet(x), x C^\bullet(x), \dots; 1)$$

$$C^\bullet(x) = \sum_{\ell \geq 0} C_\ell^\bullet(x, 1)$$

Subgraph Counting

System of infinite number of equations

Suppose that $\mathbf{A}(z) = (A_j(z))_{j \geq 0} = \Phi(z, \mathbf{A}(z))$ is a **positive, non-linear, infinite and strongly connected** system such that the **Jacobian** $\Phi_{\mathbf{a}}(z, \mathbf{a})$ is **compact** for $z > 0$ and $\mathbf{a} > \mathbf{0}$.

Let $z_0 > 0$, $\mathbf{a}_0 = (a_{j,0})_{j \geq 0}$ (inside the region of convergence) satisfy the system of equations:

$$\mathbf{a}_0 = \Phi(z_0, \mathbf{a}_0), \quad r(\Phi_{\mathbf{a}}(z_0, \mathbf{a}_0)) = 1,$$

where $r(\cdot)$ denotes the spectral radius.

Then there exists analytic function $g_j(z), h_j(z) \neq 0$ such that locally

$$A_j(z) = g_j(z) - h_j(z) \sqrt{1 - \frac{z}{z_0}}.$$

with $g_j(z_0) = a_{j,0}$ and $h_j(z_0) > 0$.

Infinite Systems of Functional Equations

COMBINATORIAL CENTRAL LIMIT THEOREM III

Suppose that $\mathbf{A}(z, u) = (A_j(z, u))_{j \geq 0} = \Phi(z, u, \mathbf{A}(z, u))$ is a **positive, non-linear, infinite and strongly connected** system such that the **Jacobian $\Phi_{\mathbf{a}}(z, 1, \mathbf{a})$ is compact** for $z > 0$ and $\mathbf{a} > 0$.

Let $z_0 > 0$, $\mathbf{a}_0 = (a_{j,0})_{j \geq 0}$ (inside the region of convergence) satisfy the system of equations:

$$\mathbf{a}_0 = \Phi(z_0, 1, \mathbf{a}_0), \quad r(\Phi_{\mathbf{a}}(z_0, 1, \mathbf{a}_0)) = 1,$$

where $r(\cdot)$ denotes the spectral radius.

Then there exists analytic function $g_j(z, u), h_j(z, u) \neq 0$ and $\rho(u)$ such that locally

$$A_j(z, u) = g_j(z, u) - h_j(z, u) \sqrt{1 - \frac{z}{\rho(u)}}.$$

with $g_j(z_0, 1) = a_{j,0}$, $h_j(z_0, 1) > 0$, and $\rho(1) = z_0$.

Infinite Systems of Functional Equations

COMBINATORIAL CENTRAL LIMIT THEOREM III (cont.)

Suppose that $A(z, u) = \Psi(z, u, (A_j(z, u))_{j \geq 0})$, where Ψ is analytic with non-negative coefficients.

$$\begin{aligned} \implies \quad A(z, u) &= g(z, u) - h(z, u) \sqrt{1 - \frac{z}{\rho(u)}} \\ \longrightarrow \quad [z^n] A(z, u) &\sim C(u) \rho(u)^{-n} n^{-3/2} \end{aligned}$$

Consider the random variable X_n given by

$$\mathbb{P}\{X_n = k\} = \frac{a_{nk}}{a_n},$$

where $a_{n,k} = [z^n u^k] A(z, u)$ and $a_n = [z^n] A(z, 1)$. Then X_n satisfies a **central limit theorem** with $\mathbb{E} X_n \sim \mu n$ and $\text{Var} X_n \sim \sigma^2 n$.

Subgraph Counting

Special case of infinite system

$$A_j = \Phi_j(z, u, A_0, A_1, \dots), \quad j \geq 0,$$

with

$$\Phi_j(z, 1, A_0, A_1, \dots) = \tilde{\Phi}_j(z, A_0 + A_1 + \dots),$$

so that $A = A_0 + A_1 + \dots$ satisfies

$$A = \tilde{\Phi}(z, A),$$

where

$$\tilde{\Phi}(z, A) = \sum_{j \geq 0} \tilde{\Phi}_j(z, A) = \sum_{j \geq 0} \Phi(z, 1, A_0, A_1, \dots)$$

$$\implies \frac{\partial \Phi_j}{\partial a_i}(z, 1, \mathbf{a}) \quad \text{does not depend on } i$$

$$\implies \Phi_{\mathbf{a}}(z, 1, \mathbf{a}) \quad \text{is compact}$$

Thank You!