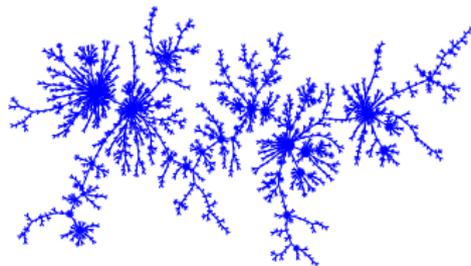


Scaling limits of large random trees

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Université Paris-Dauphine & ENS Paris



based on joint works with Grégory MIERMONT, Jim PITMAN, Robin STEPHENSON
and Matthias WINKEL

- 1 Introduction: scaling limits of Galton-Watson trees
- 2 Scaling limits of Markov branching trees
- 3 Applications: Galton-Watson trees, combinatorial trees, sequences of random trees built recursively

Scaling limits: a basic example

I.i.d sequence of *centered* random variables $X_i \in \{-1, 1\}$:

$-1, 1, 1, 1, 1, -1, -1, -1, 1, -1, -1, 1, 1, 1, 1, \dots$

Centered random walk: $S_n = X_1 + \dots + X_n$

- How does S_n behave when n is large ?
- (1) what is the growth rate ?
 - (2) what is the limit after rescaling ?

Scaling limits: a basic example

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Central limit theorem:

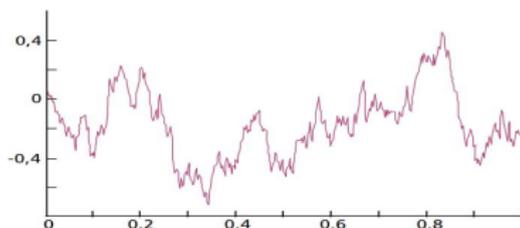
$$\frac{S_n}{\sqrt{n}} \xrightarrow{\text{law}} \mathcal{N}(0, 1)$$

Functional version:

DONSKER'S theorem (51):

$$\left(\frac{S_{[nt]}}{\sqrt{n}}, t \in [0, 1] \right) \xrightarrow{\text{law}} (B(t), t \in [0, 1])$$

where B is a standard Brownian motion



Scaling limits: a basic example

More generally:

- **invariance principle**: random walks with i.i.d. centered increments $X_i, i \geq 1$ with **finite variance** σ^2 converge at speed $\sigma\sqrt{n}$ towards a standard **Brownian motion**

Scaling limits: a basic example

More generally:

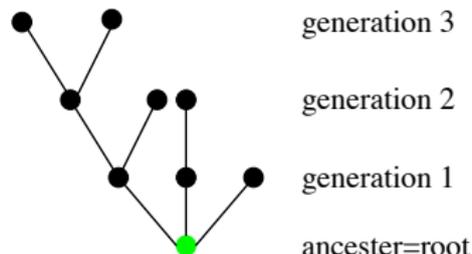
- **invariance principle:** random walks with i.i.d. centered increments $X_i, i \geq 1$ with **finite variance** σ^2 converge at speed $\sigma\sqrt{n}$ towards a standard **Brownian motion**
- **however:** random walks with i.i.d. increments $X_i, i \geq 1$ such that

$$\mathbb{P}(X_1 > x) \sim C_1 x^{-\alpha} \quad \text{and} \quad \mathbb{P}(X_1 < -x) \sim C_2 x^{-\alpha} \quad \text{for some } \alpha \in (0, 2)$$

with $C_1 + C_2 > 0$ (and centered if $\alpha \in [1, 2)$) converge at speed $n^{1/\alpha}$ towards an **α -stable Lévy process**.

Scaling limits: the example of Galton-Watson trees

Galton-Watson processes are introduced in 1873 to study the extinction of family names



η : proba. on $Z_+ = \{0, 1, 2, \dots\}$ (offspring distribution), such that $\eta(1) < 1$, with mean m

Extinction probability = 1 in subcritical ($m < 1$) and critical ($m = 1$) cases
 $\in [0, 1)$ in supercritical cases ($m > 1$)

Large Galton-Watson trees

Assumption: η has mean 1 and finite variance $\sigma^2 < \infty$

T_n^{GW} : GW tree conditioned to have n nodes (total progeny = n)

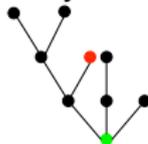
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- ▶ height (=generation) of a node chosen uniformly at random, $H_n^{(U)}$:

$$\frac{H_n^{(U)}}{\sqrt{n}} \xrightarrow{\text{law}} \frac{R}{\sigma},$$



where R has a Rayleigh distribution: $\mathbb{P}(R > x) = \exp(-x^2/2)$ (MEIR & MOON 78)

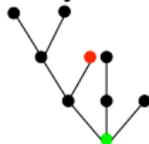
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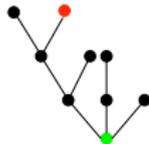
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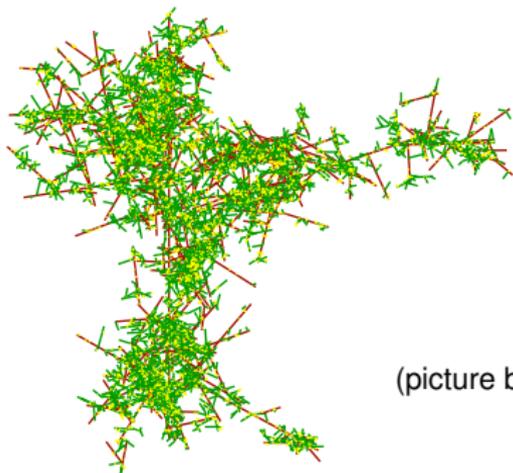
- ▶ height of the tree (=last generation), H_n :

$$\frac{H_n}{\sqrt{n}} \xrightarrow{\text{law}} \frac{2W}{\sigma},$$



the distribution of W being that of the maximum of a Brownian excursion with length 1 (KOLCHIN 86)

Universal limit: the Brownian tree \mathcal{T}_{Br}



(picture by G. Miermont)

ALDOUS 93:

$$\frac{\mathcal{T}_n^{\text{GW}}}{\sqrt{n}} \xrightarrow{\text{law}} \frac{2}{\sigma} \mathcal{T}_{\text{Br}}$$

Compact (random) *real tree*, i.e. compact metric space with the tree property:

$$\forall x, y \in \mathcal{T}_{\text{Br}} \exists! \text{ path from } x \text{ to } y$$

almost surely: - binary tree

- self-similar

- Hausdorff dimension = 2

- the set of leaves is dense in the tree

Topology on the set of compact metric spaces

For A, B compact subsets of a metric space (Z, d_Z)

$$d_{\text{Hausdorff}}(A, B) = \inf\{\epsilon > 0 : A \subset B_\epsilon \text{ and } B \subset A_\epsilon\},$$

where $A_\epsilon = \{x \in Z : d_Z(A, x) \leq \epsilon\}$

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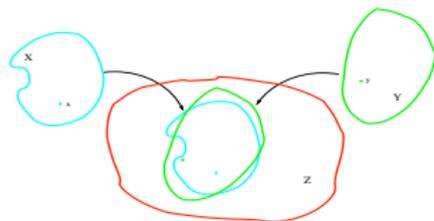
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Gromov-Hausdorff distance: let (X, x) , (Y, y) be compact, pointed, metric spaces

$$d_{\text{GH}}((X, x), (Y, y)) := \inf(d_{\text{Hausdorff}}(\varphi_1(X), \varphi_2(Y)) \vee d_Z(\varphi_1(x), \varphi_2(y)))$$

the infimum being on all isometric embeddings $\varphi_1 : X \hookrightarrow Z$ and $\varphi_2 : Y \hookrightarrow Z$ into a same metric space (Z, d_Z) .



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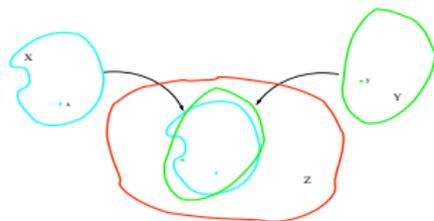
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(X, x) and (Y, y) are *equivalent* if $\exists \varphi$ isometry: $Y = \varphi(X), y = \varphi(x)$

d_{GH} : distance on the set of equivalence classes

ALDOUS 93:

$$\frac{T_n^{\text{GW}}}{\sqrt{n}} \xrightarrow[\text{GH}]{\text{law}} \frac{2}{\sigma} \mathcal{T}_{\text{Br}},$$

jointly with the convergence of the **uniform probability on the nodes** of T_n^{GW} towards a **probability measure on the leaves** of \mathcal{T}_{Br}

Examples: scaling limits of combinatorial trees

- $T_n^{(\text{ord})}$: uniform among *ordered* trees with n nodes, rooted

$$\frac{T_n^{(\text{ord})}}{\sqrt{n}} \xrightarrow{\text{law}} \mathcal{T}_{\text{Br}}$$

since $T_n^{(\text{ord})} \sim T_n^{\text{GW}}$ with a Geometric(1/2) offspring distribution

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- $T_n^{(\text{lab})}$: uniform among *labelled* trees with n nodes, rooted

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- $T_n^{(\text{P})}$: uniform among *non-ordered, non-labelled* trees with n nodes, rooted.
Problem: it is not a conditioned Galton-Watson tree !

Large Galton-Watson trees: when the variance is infinite

Assume: $\eta(k) = \mathbb{P}(\text{to have } k \text{ children}) \underset{k \rightarrow \infty}{\sim} Ck^{-\beta-1}, 1 < \beta < 2$

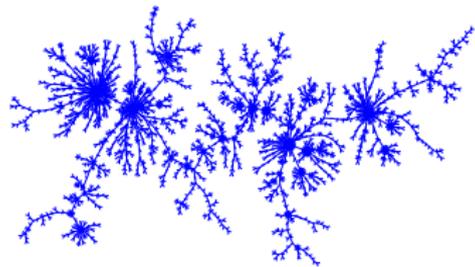
Then (DUQUESNE 03):

$$\frac{T_n^{\text{GW}}}{n^{1-1/\beta}} \xrightarrow[\text{GH}]{\text{law}} C^{-1/\beta} \mathcal{T}_\beta$$

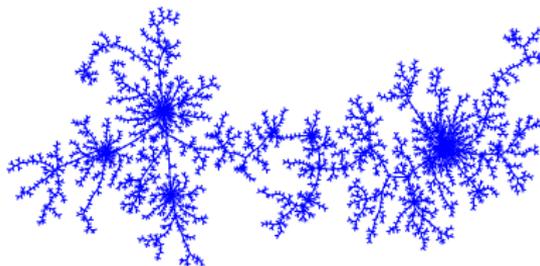
jointly with the convergence of the uniform probability on the nodes of T_n^{GW} towards a probability measure on the leaves of \mathcal{T}_β

- ▶ “smaller” trees
- ▶ \mathcal{T}_β belongs to the family of **stable Lévy trees** (introduced by DUQUESNE, LE GALL, LE JAN 98):
 - each branching vertex branches in an infinite, countable number of subtrees
 - self-similar, with Hausdorff dimension $= 1 + \frac{1}{\beta - 1}$ (DUQUESNE-LE GALL 05, H.-MIERMONT 04)

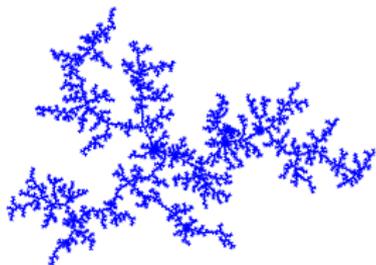
Simulations



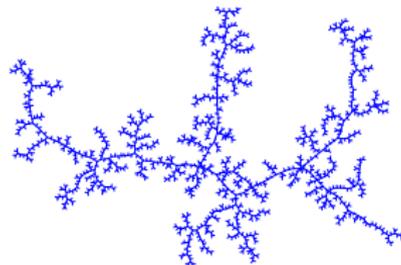
$$\beta = 1.2$$



$$\beta = 1.5$$



$$\beta = 1.9$$



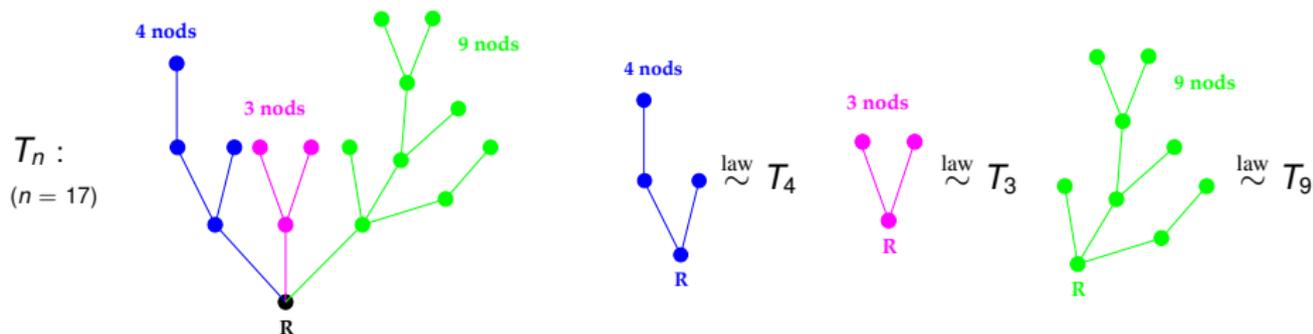
$$\beta = 2 \text{ (Brownian tree)}$$

(simulations by N. Curien)

Markov branching trees

$(T_n, n \geq 1)$: T_n random rooted tree with n nodes

Markov branching property:



Conditional on “the root of T_n has k children-trees with size $n_1 \geq \dots \geq n_k$ ”, these trees are independent, with respective distributions those of T_{n_1}, \dots, T_{n_k}

Similar definition for sequences of trees indexed by the number of **leaves**

Ex.: Galton-Watson trees conditioned to have n nodes (respectively n leaves);
sequences of trees built recursively

Scaling limits of Markov branching trees

$(T_n, n \geq 1)$ Markov Branching sequence indexed by *leaves*

Behavior when n is large ?

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Behavior when n is large ?

Let $q_n(n_1, \dots, n_p) := \mathbb{P}(\text{the root of } T_n \text{ has } p \text{ children-trees with } n_1 \geq \dots \geq n_p \text{ leaves})$

and $\mathcal{S}^\downarrow = \{s_1 \geq s_2 \geq \dots \geq 0 : \sum_{i \geq 1} s_i = 1\}$

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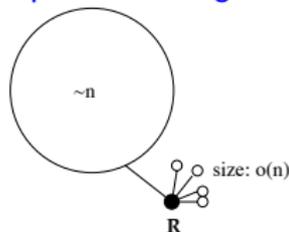
Assumption: \forall bounded continuous $f : \mathcal{S}^\downarrow \rightarrow \mathbb{R}$,

$$n^\gamma \sum_{(n_1, \dots, n_p) \text{ partition of } n} q_n(n_1, \dots, n_p) \left(1 - \frac{n_1}{n}\right) f\left(\frac{n_1}{n}, \dots, \frac{n_p}{n}, 0, \dots\right) \xrightarrow{n \rightarrow \infty} \int_{\mathcal{S}^\downarrow} (1 - s_1) f(\mathbf{s}) \nu(d\mathbf{s}),$$

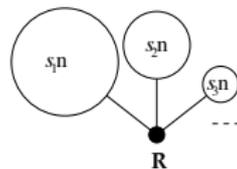
with $\gamma > 0$ and ν a non-trivial σ -finite measure on \mathcal{S}^\downarrow such that $\int_{\mathcal{S}^\downarrow} (1 - s_1) \nu(d\mathbf{s}) < \infty$.

Informally: **macroscopic branchings are rare:**

with proba. ~ 1 :



with proba. $\sim \frac{\nu(d\mathbf{s})}{n^\gamma}$:



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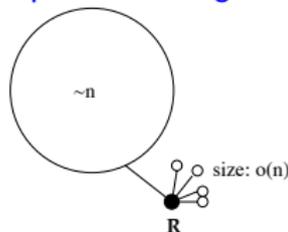
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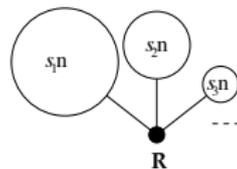
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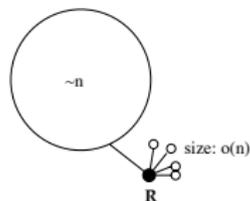


BROUTIN-DEVROYE-MCLEISH-DE LA SALLE 08: macroscopic branchings at each step: height of $T_n \sim c \log(n)$ + no interesting scaling limits

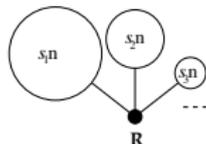
Scaling limits of Markov branching trees

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Theorem (H.-MIERMONT 11)

Under the previous assumption, \exists random compact real tree $\mathcal{T}_{\gamma, \nu}$ s.t.

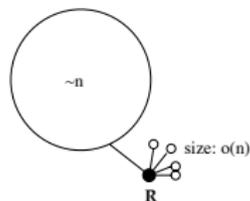
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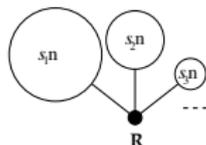
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Similar result for sequences of trees indexed by the number of nodes, with $\gamma \in (0, 1]$.

Outline of proof:

- 1 height of a random leaf
- 2 scaling limit of the tree spanned by k random leaves (finite dimensional cv)
- 3 tightness criterion

Markov branching trees: the limiting tree

- $\mathcal{T}_{\gamma, \nu}$ is **self-similar**, with Hausdorff dimension $\max(1, 1/\gamma)$ (H.-MIERMONT 04)
 γ : index of self-similarity
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$$\nu(\mathbf{s}_1 \in dx) = \frac{\sqrt{2}}{\sqrt{\pi} x^{3/2} (1-x)^{3/2}}, \quad 1/2 < x < 1$$

and all stable trees \mathcal{T}_{β} , $1 < \beta < 2$ when $\gamma = 1 - 1/\beta$ and $\nu = \nu_{\beta}$

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- the **height of a typical leaf** in $\mathcal{T}_{\gamma, \nu}$ writes

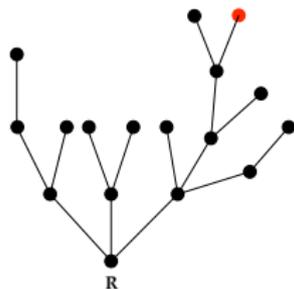
$$\int_0^{\infty} \exp(-\gamma \xi_r) dr,$$

for ξ a subordinator (increasing Lévy process) with Laplace exponent

$$\phi(\lambda) = \int_{\mathcal{S}^{\downarrow}} \sum_i (1 - s_i^{\lambda}) s_i \nu(d\mathbf{s}) \quad (\mathbb{E}[\exp(-\lambda \xi_t)] = \exp(-t\phi(\lambda)))$$

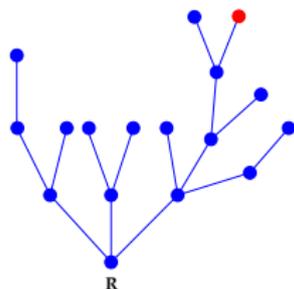
Proof: a Markov chain in the Markov branching sequence of trees

First step: height of a leaf chosen uniformly at random



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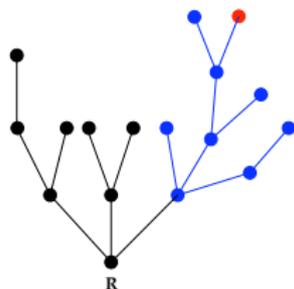


$$X_n(0) = 9$$

$X_n(k)$: size of the sub-tree above generation k containing the marked leaf

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First step: height of a leaf chosen uniformly at random

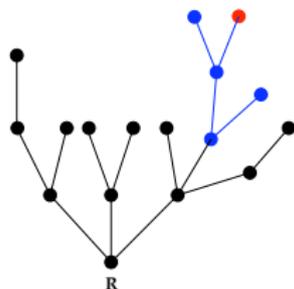


$$X_n(0) = 9, X_n(1) = 5$$

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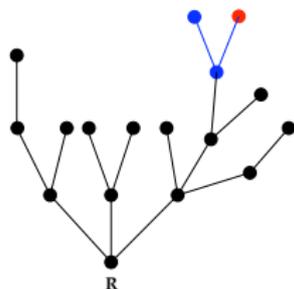


$$X_n(0) = 9, X_n(1) = 5, X_n(2) = 3$$

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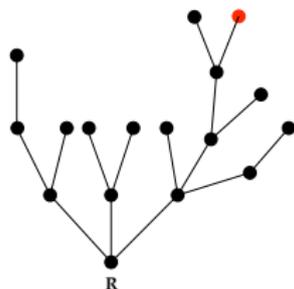


$$X_n(0) = 9, X_n(1) = 5, X_n(2) = 3, X_n(3) = 2$$

$X_n(k)$: size of the sub-tree above generation k containing the marked leaf

Proof: a Markov chain in the Markov branching sequence of trees

First step: height of a leaf chosen uniformly at random



$$X_n(0) = 9, X_n(1) = 5, X_n(2) = 3, X_n(3) = 2, X_n(4) = 1$$

$X_n(k)$: size of the sub-tree above generation k containing the marked leaf

It is a Markov chain!

A_n = absorption time at 1 = height of the marked leaf

Proof: a Markov chain in the Markov branching sequence of trees

Let:

- ξ be a subordinator with Laplace exponent

$$\phi(\lambda) = \int_{S^{\downarrow}} \sum_{i \geq 1} (1 - s_i^\lambda) s_i \nu(d\mathbf{s}), \quad \lambda \geq 0,$$

- ρ be the **acceleration of time**: $\rho(t) = \inf \{u \geq 0 : \int_0^u \exp(-\gamma \xi_r) dr \geq t\}$
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Proposition (H.-MIERMONT 11)

Under our assumption of rare macroscopic branchings,

$$\left(\frac{X_n([n^\gamma t])}{n}, t \geq 0 \right) \xrightarrow{\text{law}} (X_\infty(t), t \geq 0),$$

for the Skorokhod topology on the set $\mathbb{D}([0, \infty), [0, \infty))$, and

$$\frac{A_n}{m^\gamma} \xrightarrow{\text{law}} \int_0^\infty \exp(-\gamma \xi_r) dr.$$

Application 1: Galton-Watson trees

- ▶ we recover Aldous' results on critical Galton-Watson trees conditioned by nodes

$$\frac{T_n^{\text{GW}}}{\sqrt{n}} \xrightarrow{\text{law}} \frac{2}{\sigma} \mathcal{T}_{\text{Br}} \quad \text{if } \sigma^2 < \infty$$

and its extension by Duquesne to infinite variance cases

- ▶ RIZZOLO 11: $T_n^{\text{GW,L}}$ critical GW with finite variance σ^2 , n leaves

$$\frac{T_n^{\text{GW,L}}}{\sqrt{n}} \xrightarrow{\text{law}} \frac{2}{\sigma \eta(0)} \mathcal{T}_{\text{Br}}$$

where $\eta(0)$: proba. of having 0 child (see also KORTCHEMSKI 11)

Application 2: combinatorial trees

- $T_n^{(\text{ord})}$: uniform among *ordered* trees with n nodes, rooted

$$\frac{T_n^{(\text{ord})}}{\sqrt{n}} \xrightarrow{\text{law}} \mathcal{T}_{\text{Br}}$$

- $T_n^{(\text{lab})}$: uniform among *labelled*, trees with n nodes, rooted

$$\frac{T_n^{(\text{lab})}}{\sqrt{n}} \xrightarrow{\text{law}} 2\mathcal{T}_{\text{Br}}$$

- $T_n^{(\text{P})}$: uniform among *non-ordered, non-labelled* trees with n nodes, rooted
BROUTIN-FLAJOLET 08 : height DRMOTA-GITTENBERGER 10 : profile

Corollary (H.-MIERMONT 11) :

$$\frac{T_n^{(\text{P})}}{\sqrt{n}} \xrightarrow{\text{law}} c_{\text{P}} \mathcal{T}_{\text{Br}}, \quad c_{\text{P}} \sim 1.491.$$

Analog result for a uniform rooted tree with n non-ordered, non-labelled leaves and with **at most m children** per node (up to a constant c_m)

Application 3: sequences of trees built recursively

- Rémy's algorithm (85): $T_R(n)$ rooted, binary, n labelled leaves $n \geq 1$

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start with $T_R(1)$:



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$T_R(2)$:



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$T_R(2)$:

at each step :

- choose a leaf at random
- attach in its "middle" a new edge



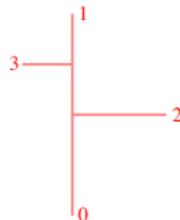
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$T_R(3)$:

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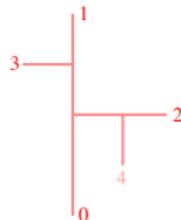
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$T_R(4)$:

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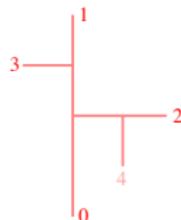
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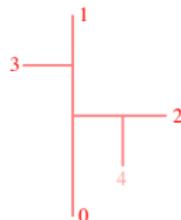
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$$\frac{T_R(n)}{\sqrt{n}} \xrightarrow{\text{law}} 2\sqrt{2}\mathcal{T}_{\text{Br}} \quad (\text{actually: a.s. cv})$$

- Ford's trees (05): $T_\alpha(n)$ rooted, binary, n labelled leaves, $n \geq 1$ ($\alpha \in (0, 1)$)

weight : $\circ 1 - \alpha$ on each edge-leaf
 $\circ \alpha$ on each internal edges

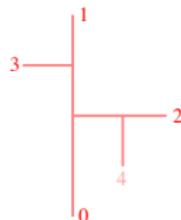
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$$\frac{T_\alpha(n)}{n^\alpha} \xrightarrow{\text{law}} \mathcal{T}_{\alpha, \nu_\alpha}$$

$$\nu_\alpha(\mathbf{s}_1 \in dx) = \frac{\mathbf{1}_{\{1/2 \leq x \leq 1\}}}{\Gamma(1-\alpha)} (\alpha(x(1-x))^{-\alpha-1} + (2-4\alpha)(x(1-x))^{-\alpha}) dx$$

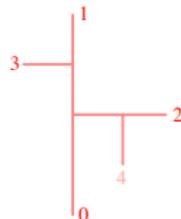
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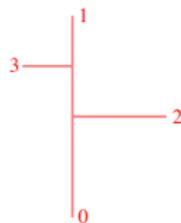
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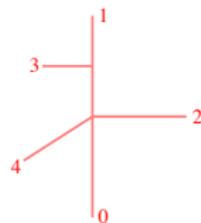
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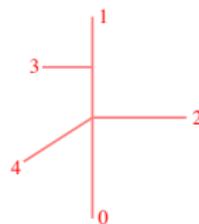


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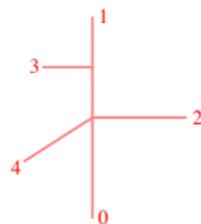
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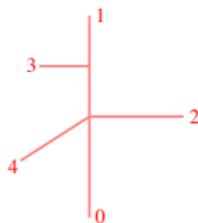
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- **Chen-Ford-Winkel trees** (09): different weights on edges and nodes (2 parameters); cv in proba.
- **k -ary trees** (work in progress with R. Stephenson): at each step, add $k - 1$ edges (instead of 1) in Rémy's algorithm $\rightarrow T_k(n)$ rooted, $1+(n-1)(k-1)$ leaves

$$\frac{T_k(n)}{n^{1/k}} \xrightarrow{\text{Proba}} \mathcal{T}_{1/k, \nu_k}$$

where $\nu_k(d\mathbf{s}) = \frac{(k-1)!}{k(\Gamma(\frac{1}{k}))^{(k-1)}} \prod_{i=1}^k s_i^{-(1-1/k)} \left(\sum_{i=1}^k \frac{1}{1-s_i} \right) \mathbf{1}_{\{s_k=1-s_1-\dots-s_{k-1}\}} ds_1 \dots ds_{k-1}$.