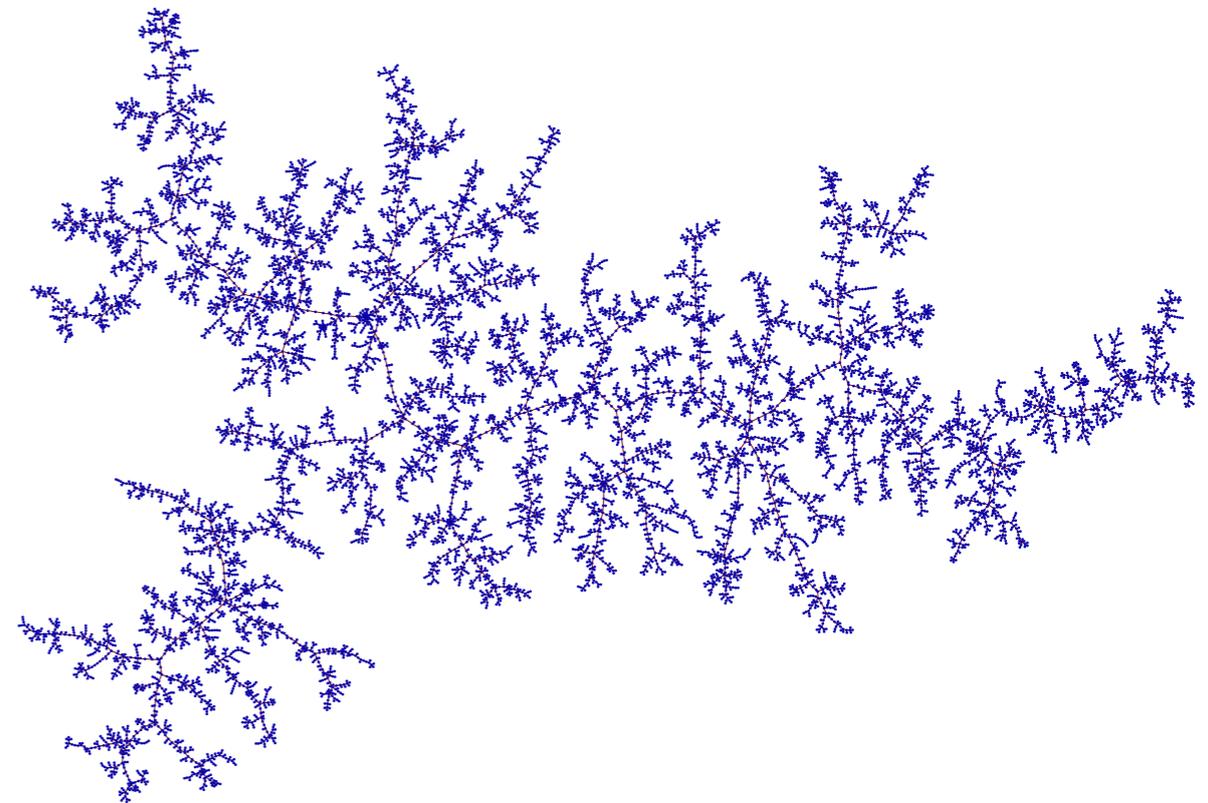
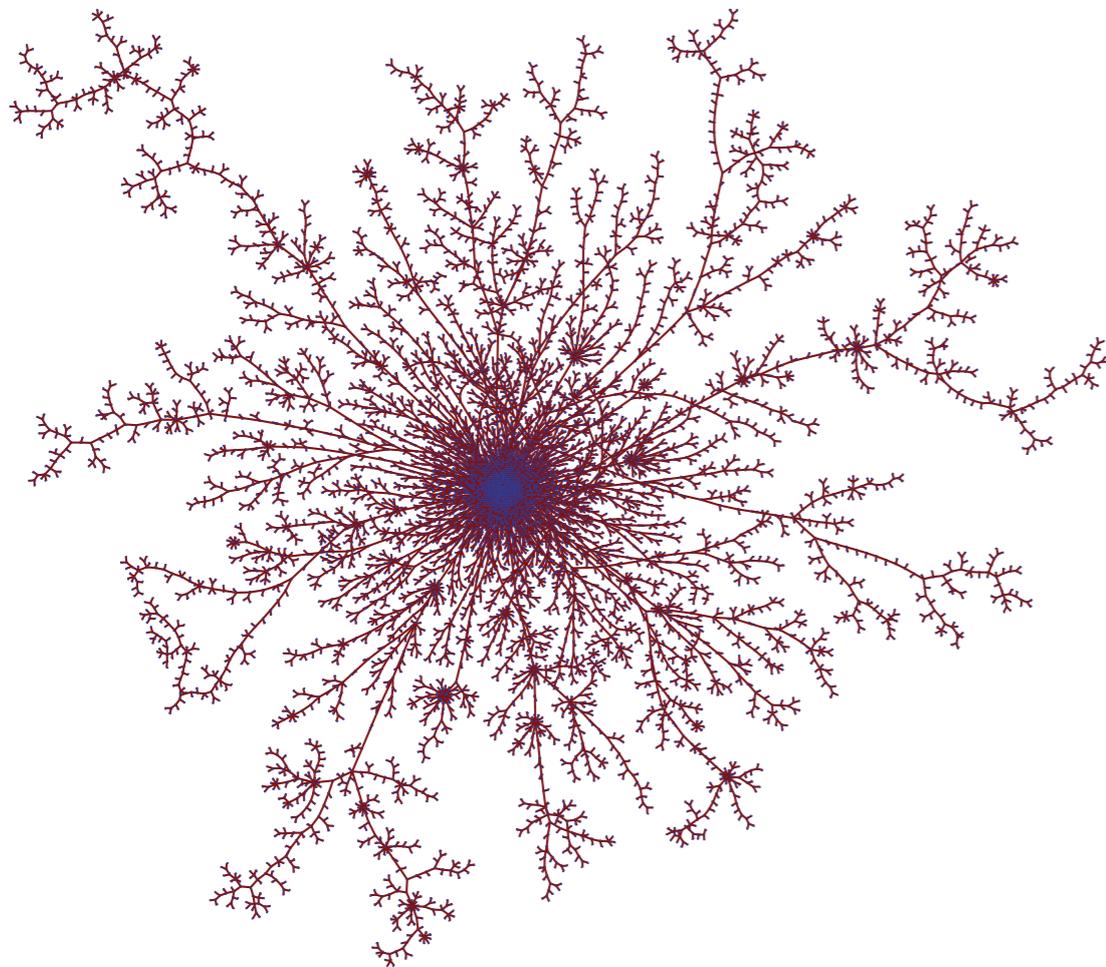
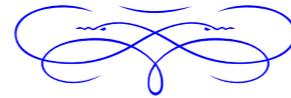


Limits of large random discrete structures



Igor Kortchemski
CNRS & École polytechnique



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↪ A possibility to study \mathcal{X}_n is to find a limiting object X such that $X_n \rightarrow X$ as $n \rightarrow \infty$.

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- *Universality:* if $(Y_n)_{n \geq 1}$ is another sequence of objects converging towards X , then X_n and Y_n share approximately the same properties for n large.

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- ↪ *In what space do the objects live?* Here, a metric space (Z, d) which will be complete and separable (there exists a dense countable subset).
- ↪ *What is the sense of the convergence when the objects are random?* Here, convergence in distribution:

$$\mathbb{E} [F(X_n)] \xrightarrow[n \rightarrow \infty]{} \mathbb{E} [F(X)]$$

for every continuous bounded function $F : Z \rightarrow \mathbb{R}$.

Outline

I. MODELS CODED BY TREES

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Stack triangulations (Albenque, Marckert)

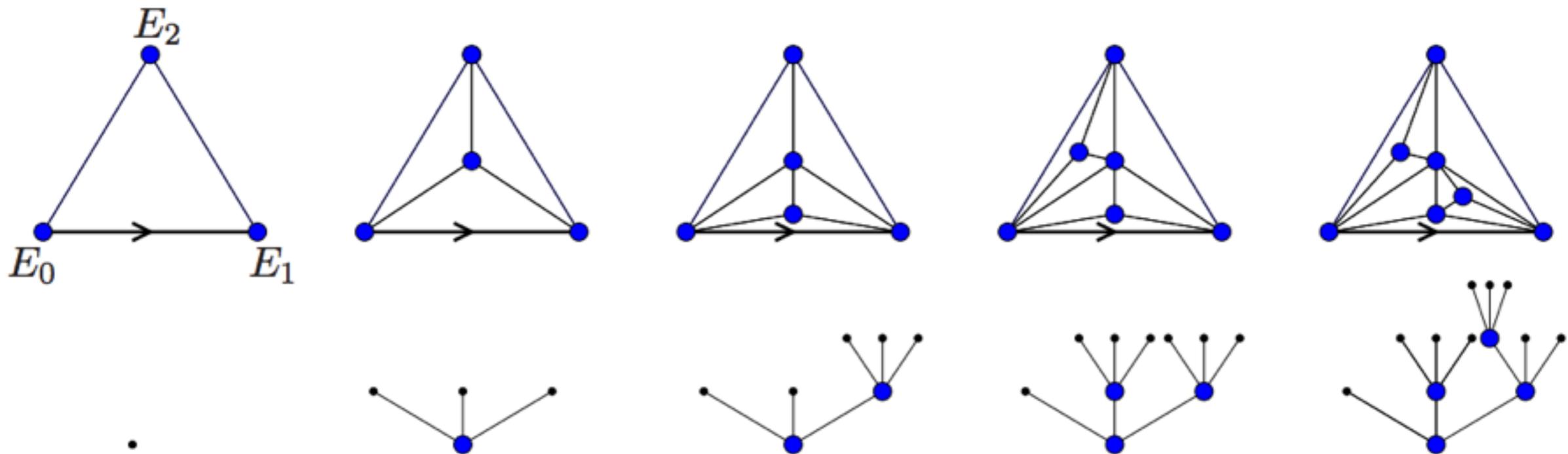


Figure 8: Construction of the ternary tree associated with an history of a stack-triangulation

Dissections (Curien, \mathcal{K} .)

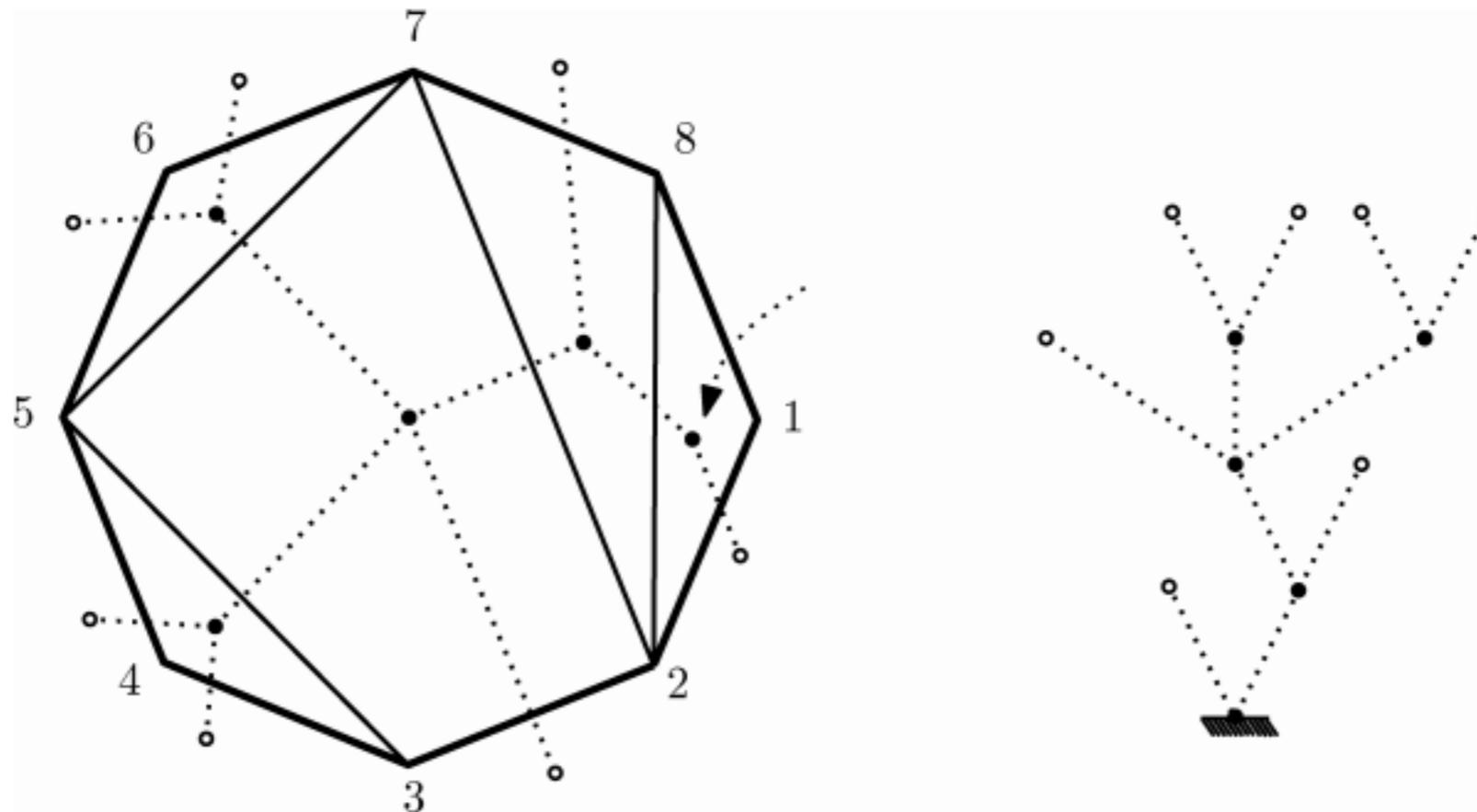


Fig. 4. The dual tree of a dissection of P_8 , note that the tree has 7 leaves.

Maps (Schaeffer)

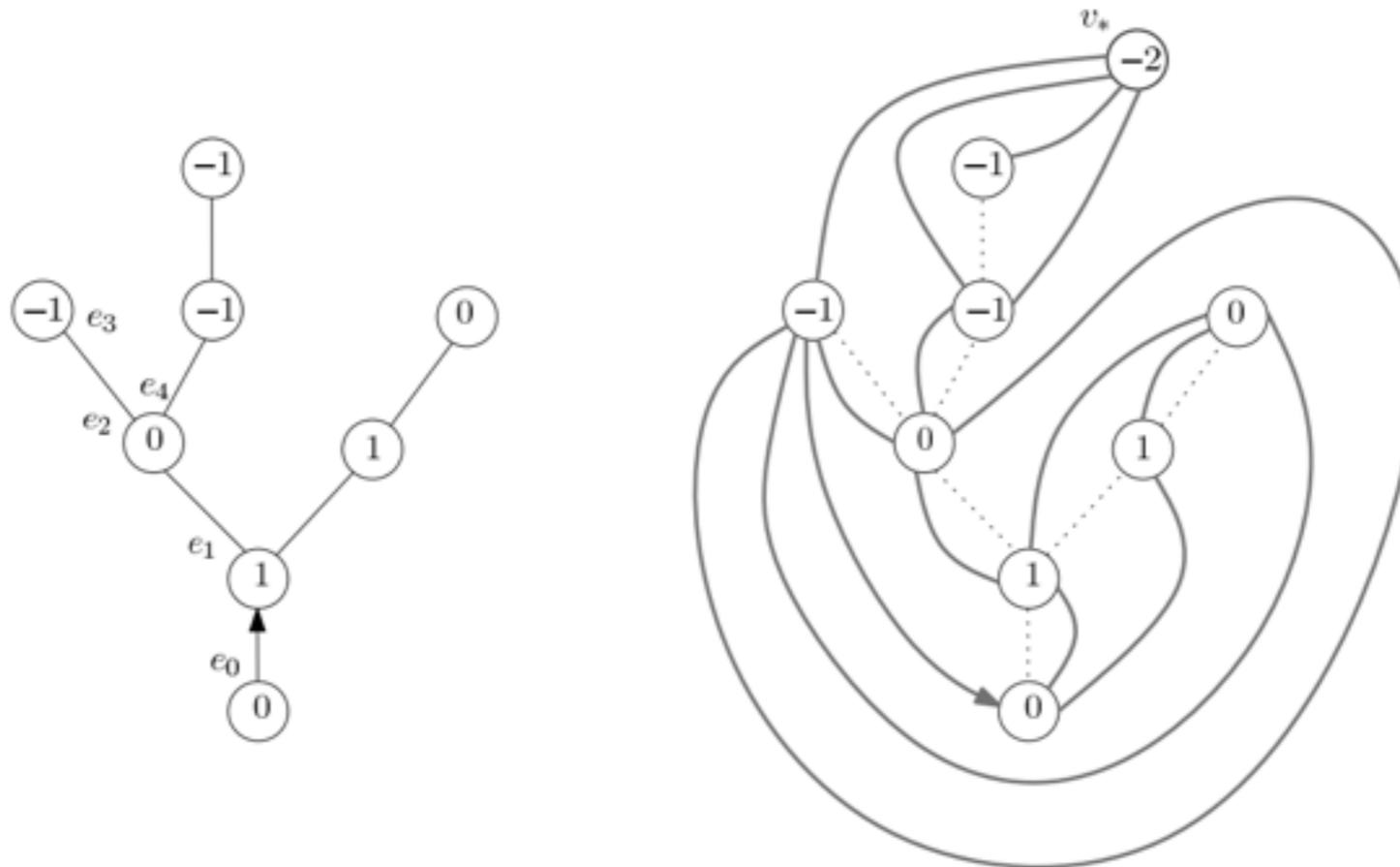
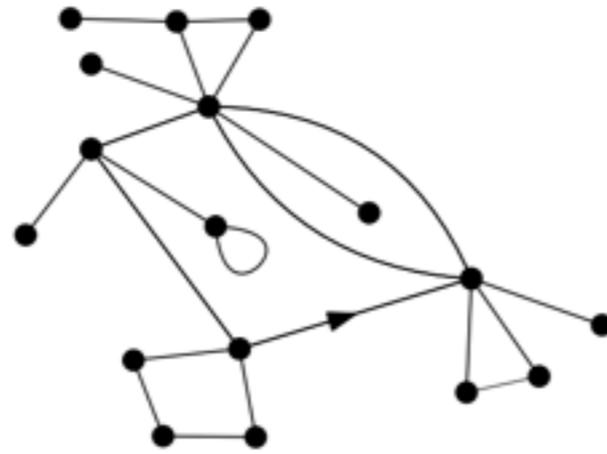
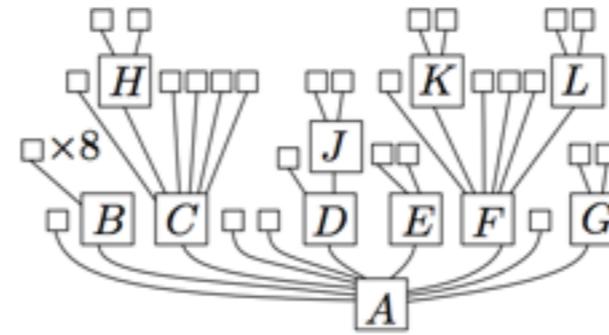
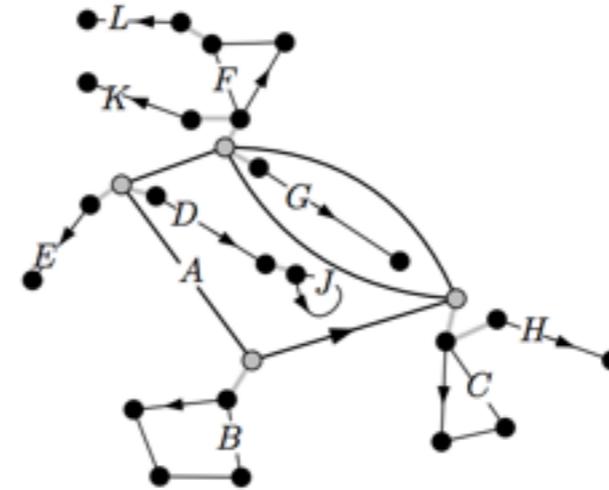
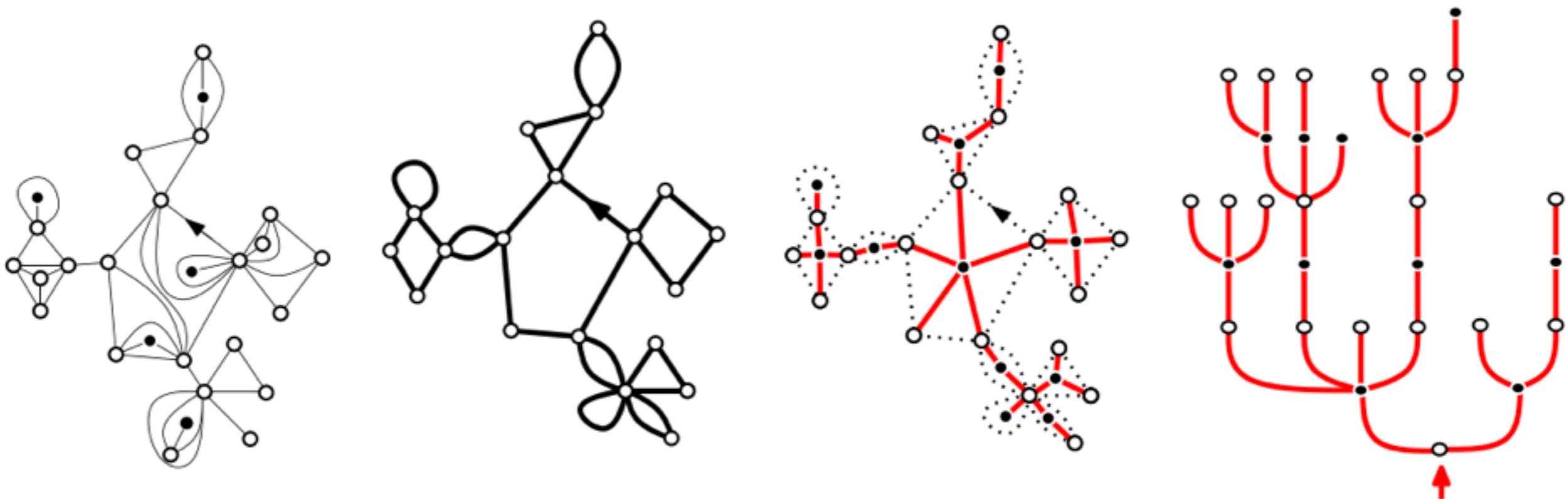
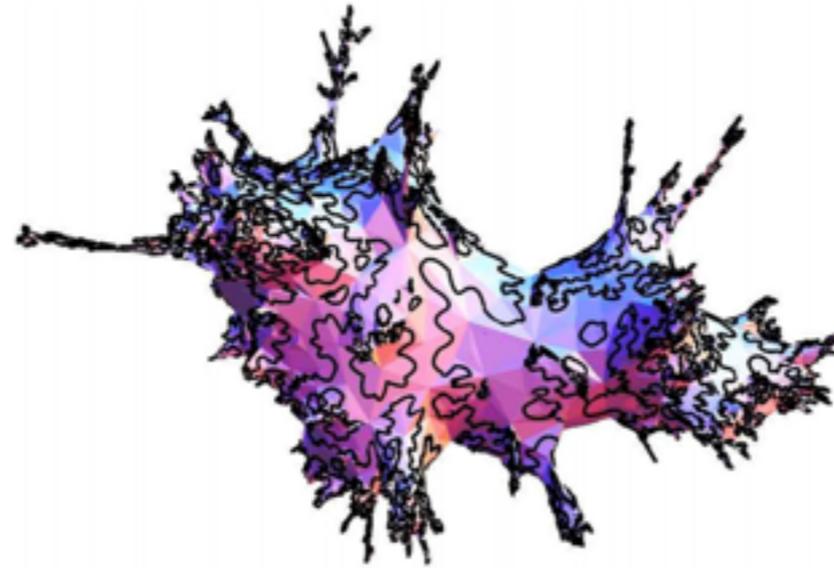


FIGURE 6. Illustration of the Cori-Vauquelin-Schaeffer bijection, in the case $\epsilon = 1$. For instance, e_3 is the successor of e_0 , e_2 the successor of e_1 , and so on.

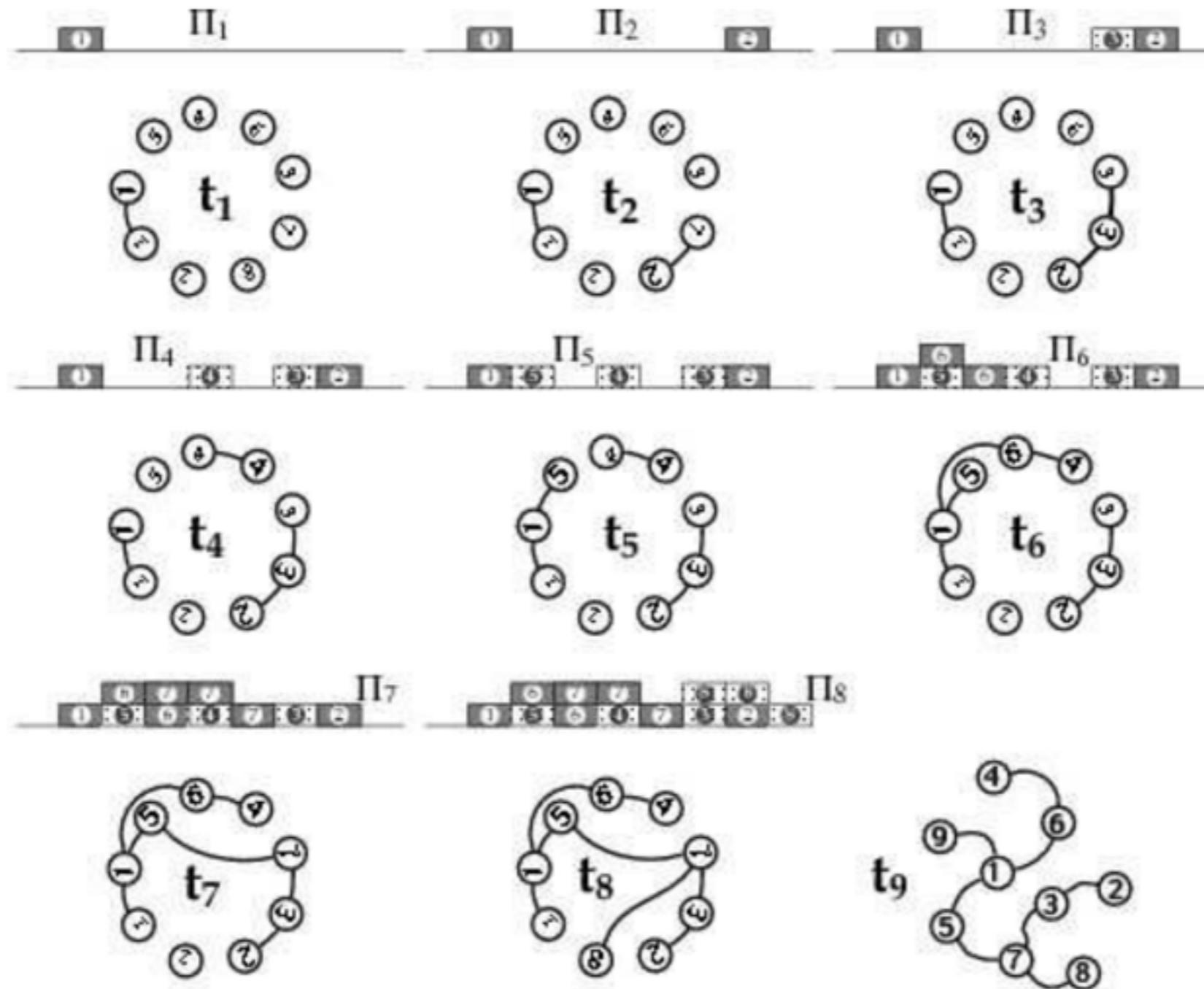
Maps (Addario-Berry)

(A) A map M .(B) The tree T_M . Tiny squares represent trivial blocks.(C) The decomposition of M into blocks. Blocks are joined by grey lines according to the tree structure. Root edges of blocks are shown with arrows.(D) The correspondence between blocks and nodes of T_M . Non-trivial blocks receive the alphabetical label (from A through L) of the corresponding node.

Maps with percolation (Curien, \mathcal{K} .)



Parking functions (Chassaing, Louchard)



I. MODELS CODED BY TREES

II. LOCAL LIMITS OF BGW TREES



III. SCALING LIMITS OF BGW TREES

Recall that in a BGW tree, every individual has a random number of children (independently of each other) distributed according to μ (offspring distribution).

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What does a large BGW tree look like, near the root?

Local limits: critical case

Let μ be a **critical** offspring distribution. Let \mathcal{T}_n be a BGW tree conditioned on having n vertices.

Theorem (Kesten '87, Janson '12, Abraham & Delmas '14)

The convergence

$$\mathcal{T}_n \xrightarrow[n \rightarrow \infty]{(d)} \mathcal{T}_\infty$$

holds in distribution for the local topology, where \mathcal{T}_∞ is the infinite BGW tree conditioned to survive.

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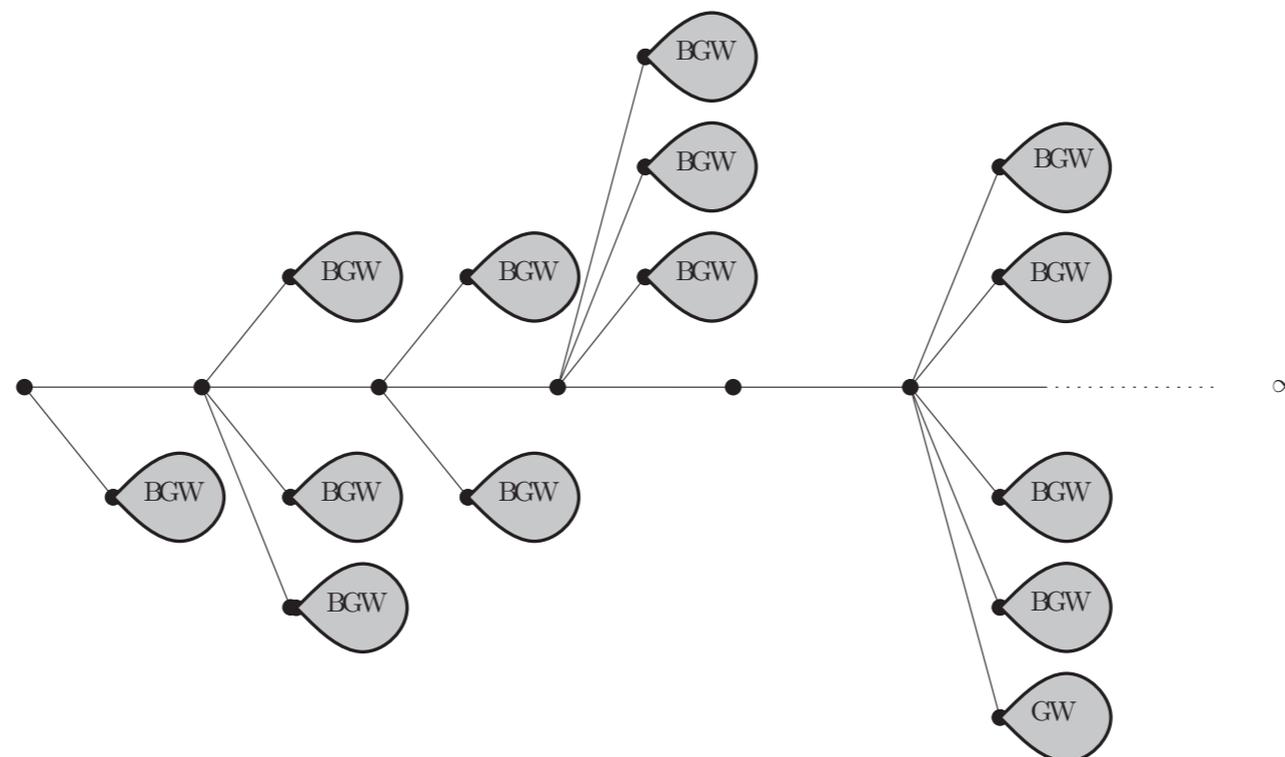
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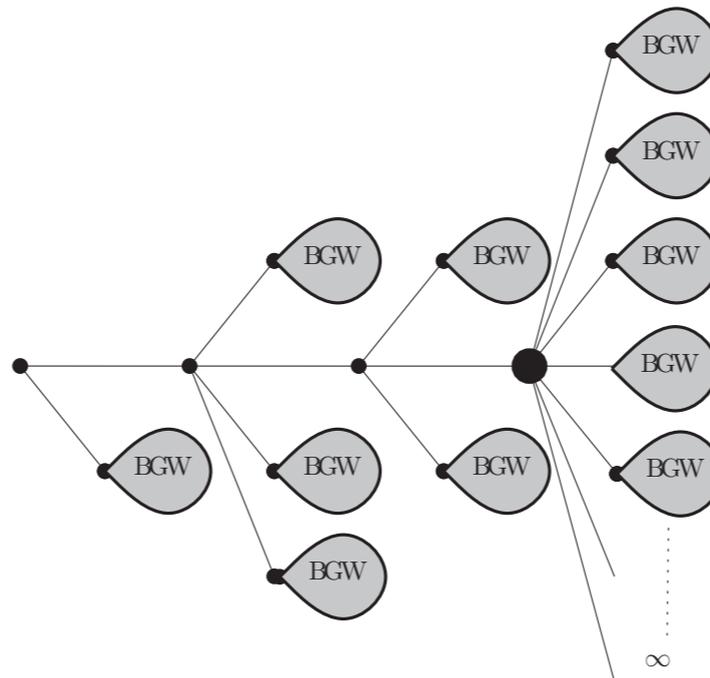
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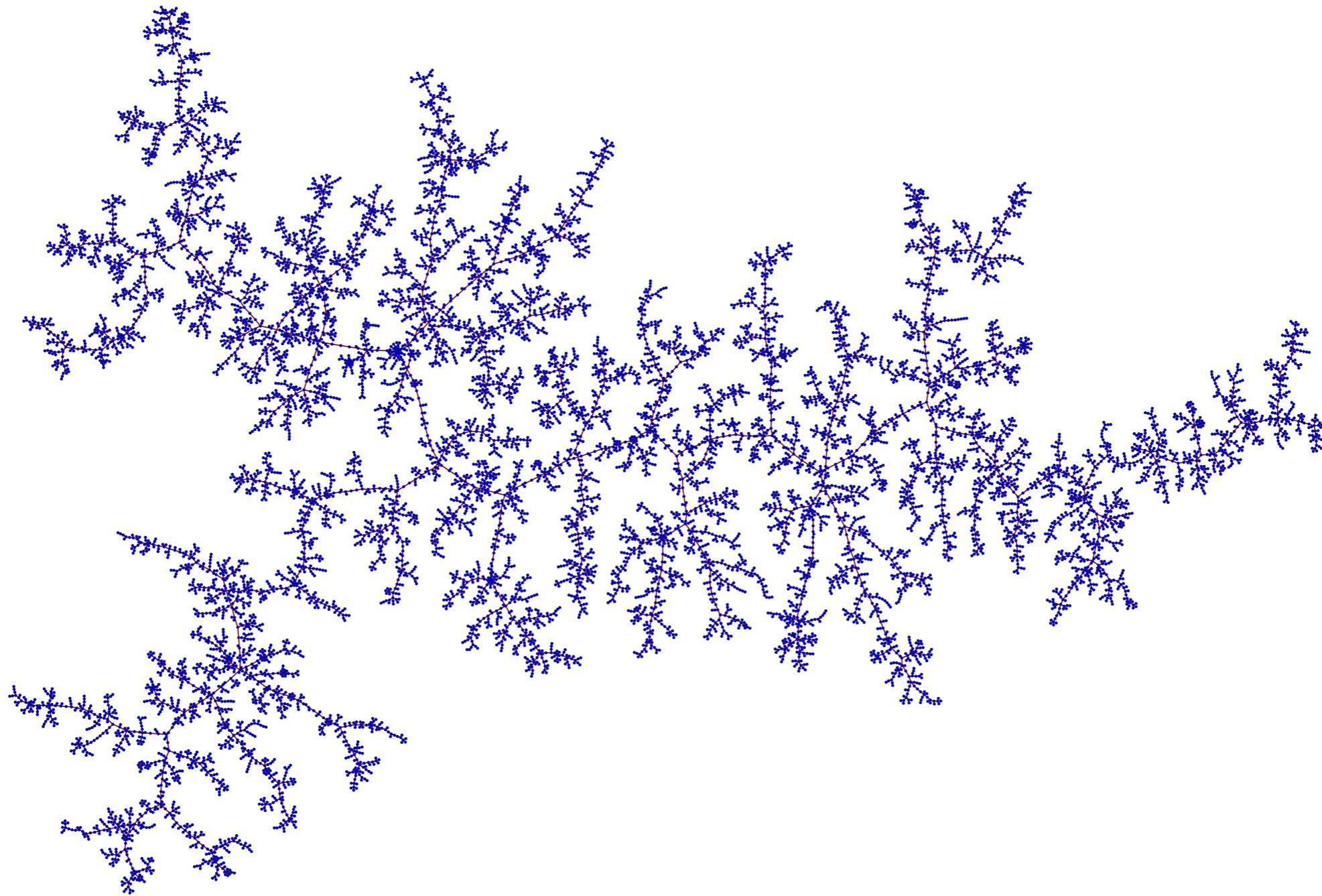
II. LOCAL LIMITS OF BGW TREES

III. SCALING LIMITS OF BGW TREES



What does a large BGW tree look like, globally?

A simulation of a large random critical GW tree

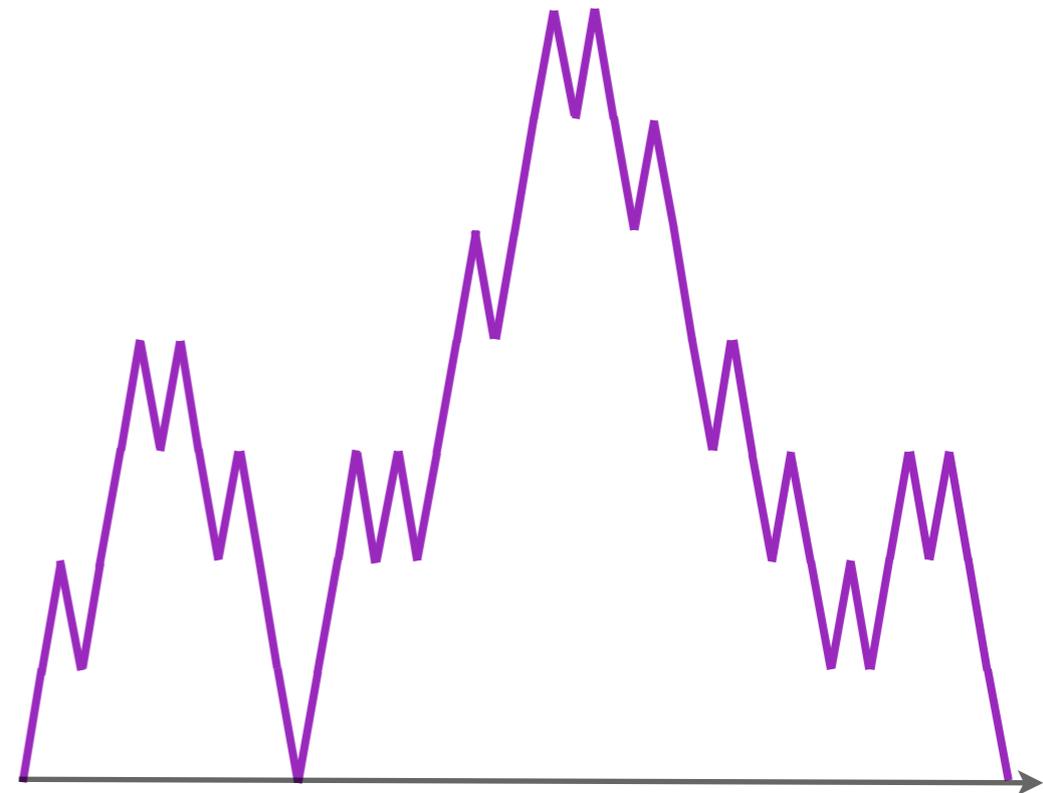
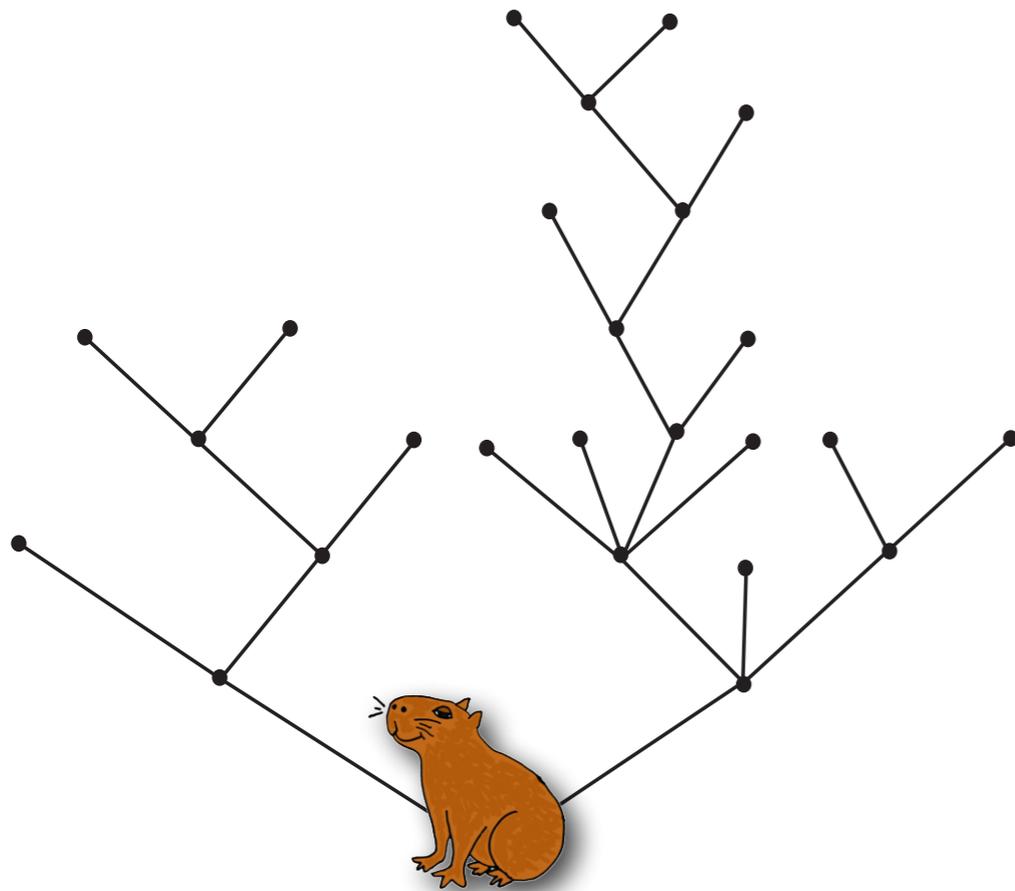


CODING TREES BY FUNCTIONS



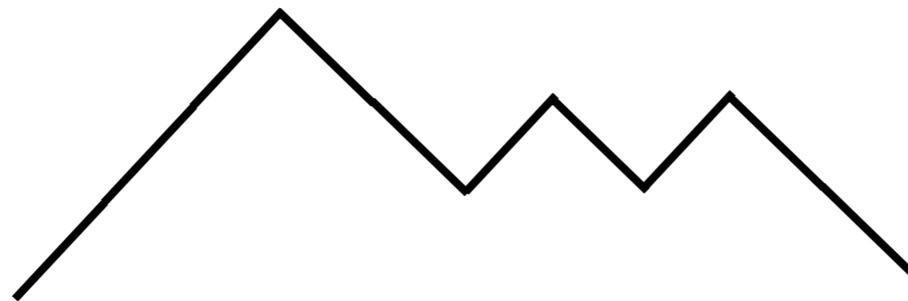
Contour function of a tree

Define the **contour function** of a tree:



Coding trees by contour functions

Knowing the contour function, it is easy to recover the tree.



SCALING LIMITS: FINITE VARIANCE



Scaling limits : finite variance

Let μ be an offspring distribution with **finite** positive variance such that $\sum_{i \geq 0} i\mu(i) = 1$. Let \mathcal{T}_n be a BGW tree conditioned on having n vertices.

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$$\left(\frac{1}{\sqrt{n}} C_{2nt}(\mathcal{T}_n) \right)_{0 \leq t \leq 1} \xrightarrow[n \rightarrow \infty]{(d)}$$

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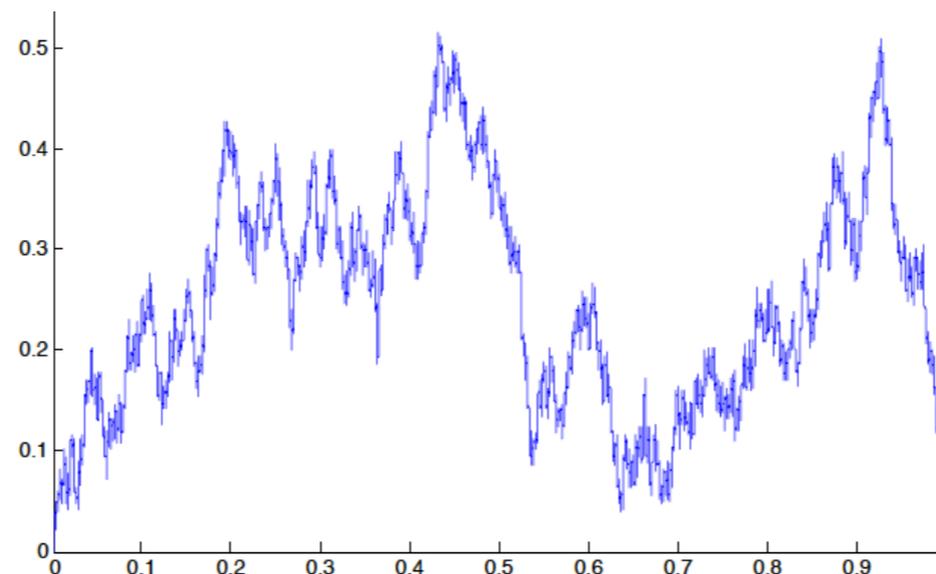
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DO THE DISCRETE TREES CONVERGE TO A CONTINUOUS TREE?



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Yes, if we view trees as compact metric spaces by equipping the vertices with the graph distance!

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be the r -neighborhoods of X and Y .

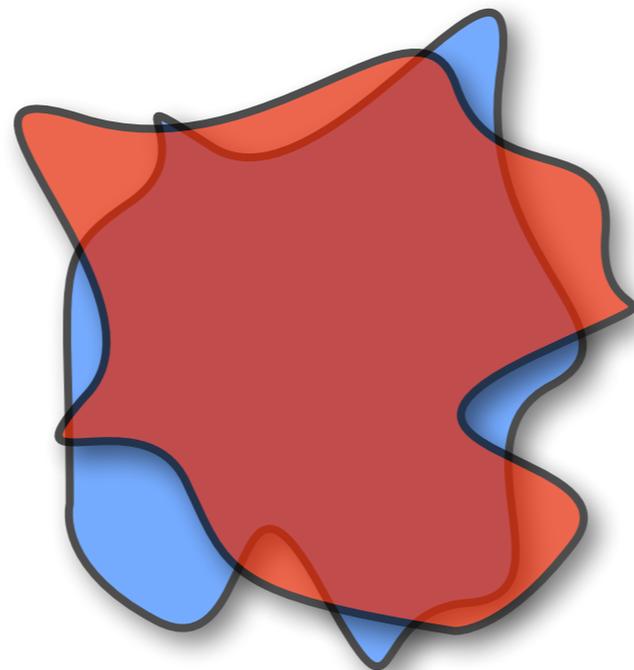
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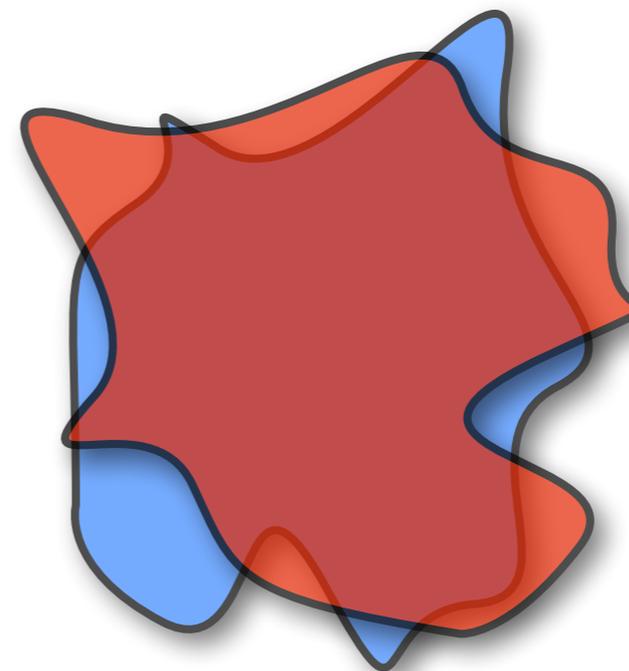
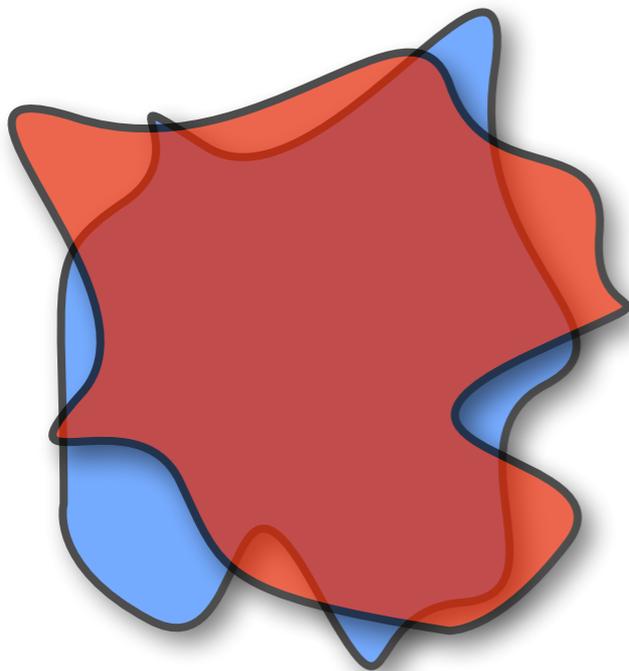
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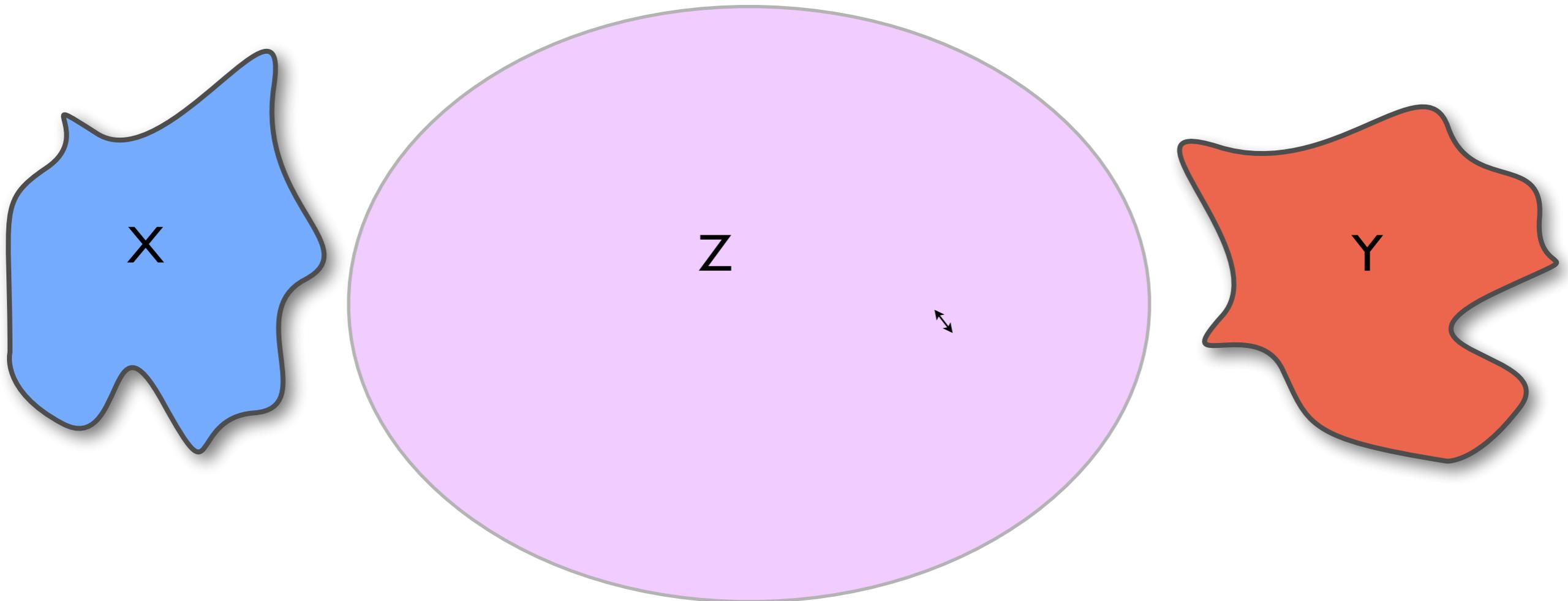


The Gromov–Hausdorff distance

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The Gromov–Hausdorff distance between X and Y is the smallest Hausdorff distance between all possible isometric embeddings of X and Y in a *same* metric space Z .

The Brownian tree

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Notation: for a metric space (Z, d) and $\alpha > 0$, $\alpha \cdot Z$ is the metric space $(Z, \alpha \cdot d)$.

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Notation: for a metric space (Z, d) and $\alpha > 0$, $\alpha \cdot Z$ is the metric space $(Z, \alpha \cdot d)$.

The metric space \mathcal{T}_e is called the *Brownian continuum random tree (CRT)*, and is coded by a Brownian excursion.

SCALING LIMITS: INFINITE VARIANCE CASE



Scaling limits: domain of attraction of a stable law

Fix $\alpha \in (1, 2)$. Let μ be an offspring distribution such that

$$\sum_{i \geq 0} i \mu_i = 1 \quad (\mu \text{ is critical})$$

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What does \mathcal{T}_n look like for large n ?

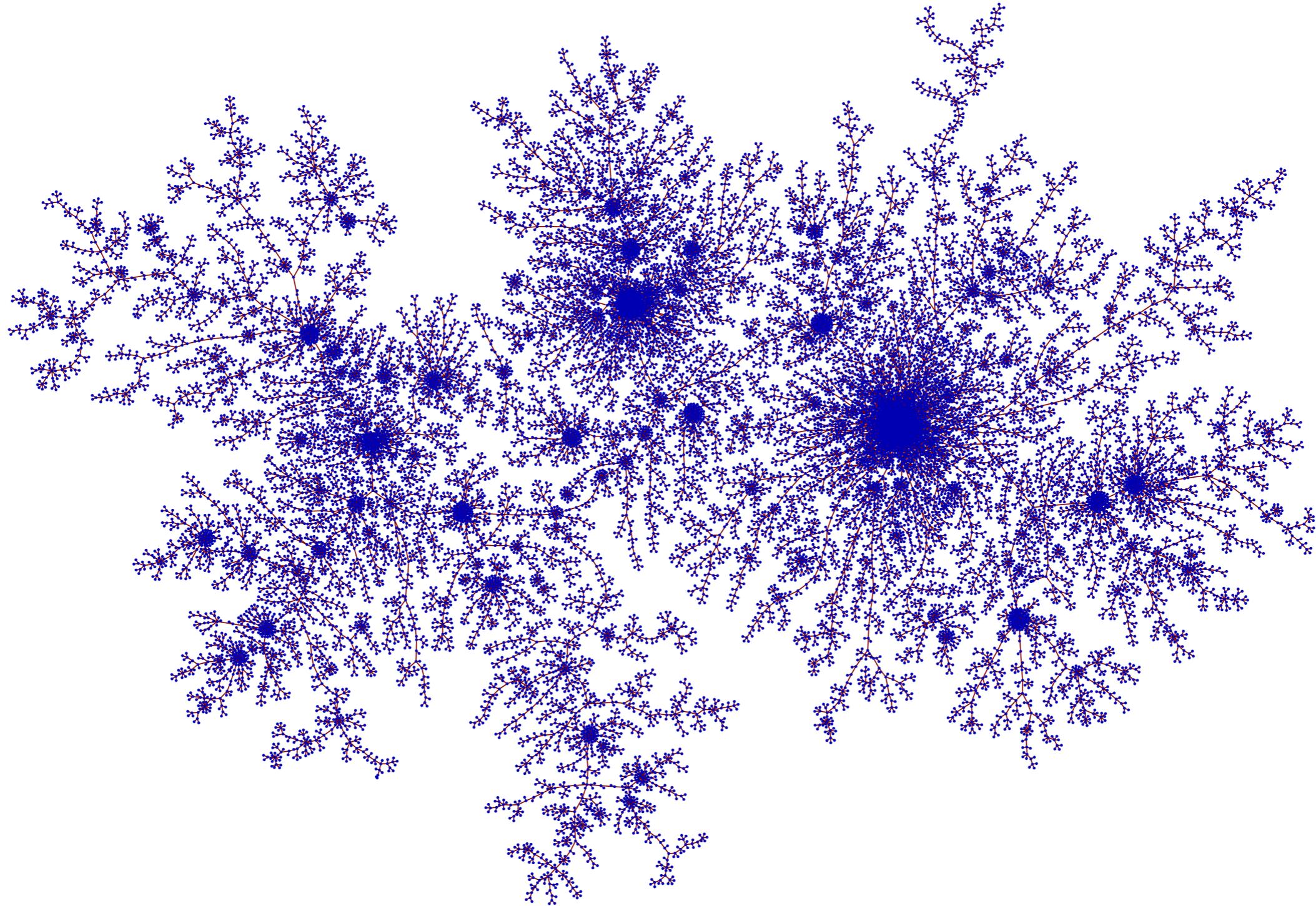


Figure: A large $\alpha = 1.1$ – stable tree

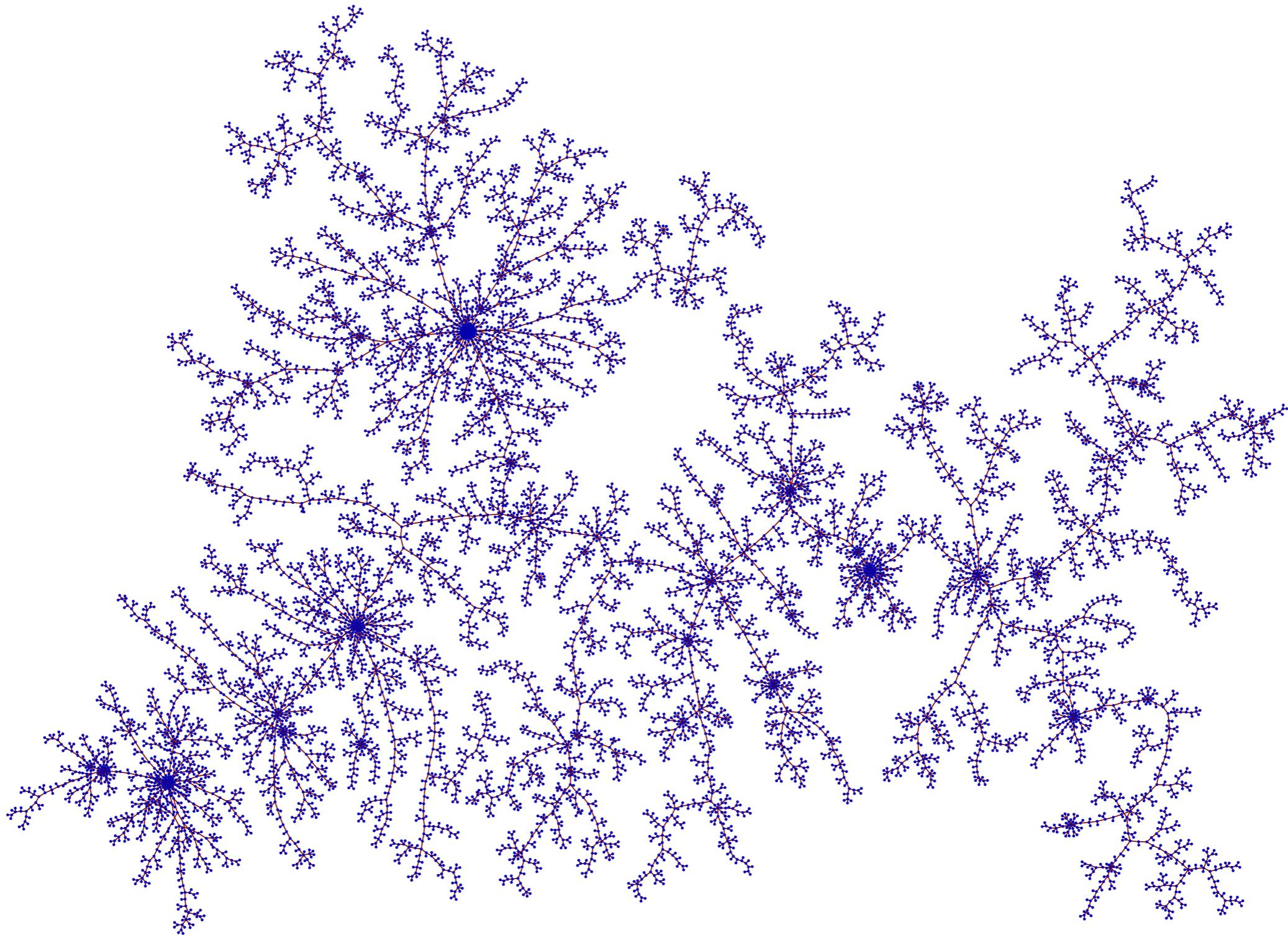


Figure: A large $\alpha = 1.5$ – stable tree

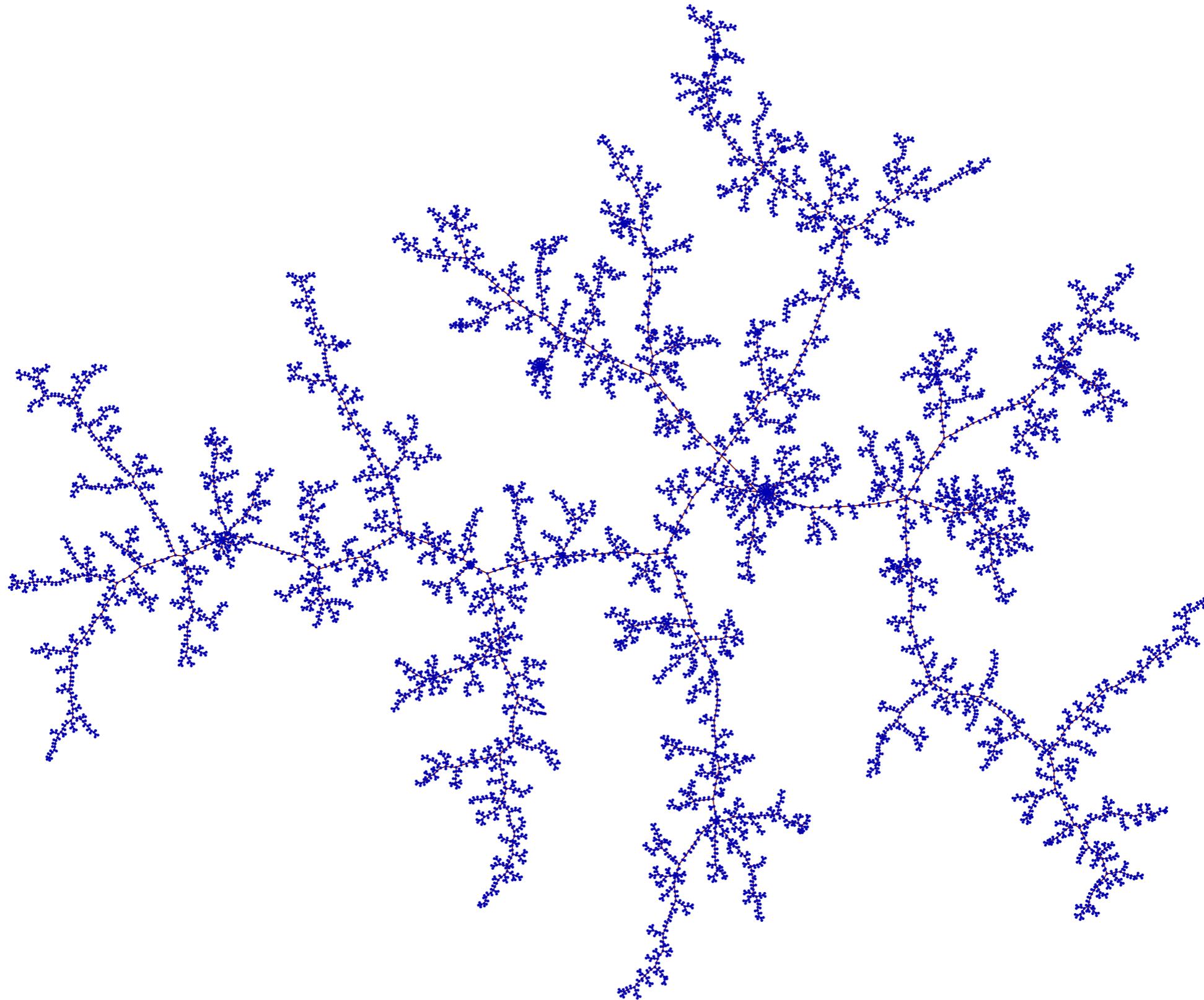


Figure: A large $\alpha = 1.9$ – stable tree

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Fix $\alpha \in (1, 2)$. Let μ be a **critical** offspring distribution such that $\mu_i \sim c/i^{1+\alpha}$.
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↗ The maximal degree of \mathcal{T}_n is of order $n^{1/\alpha}$.

CONDENSATION : SUBCRITICAL CASE

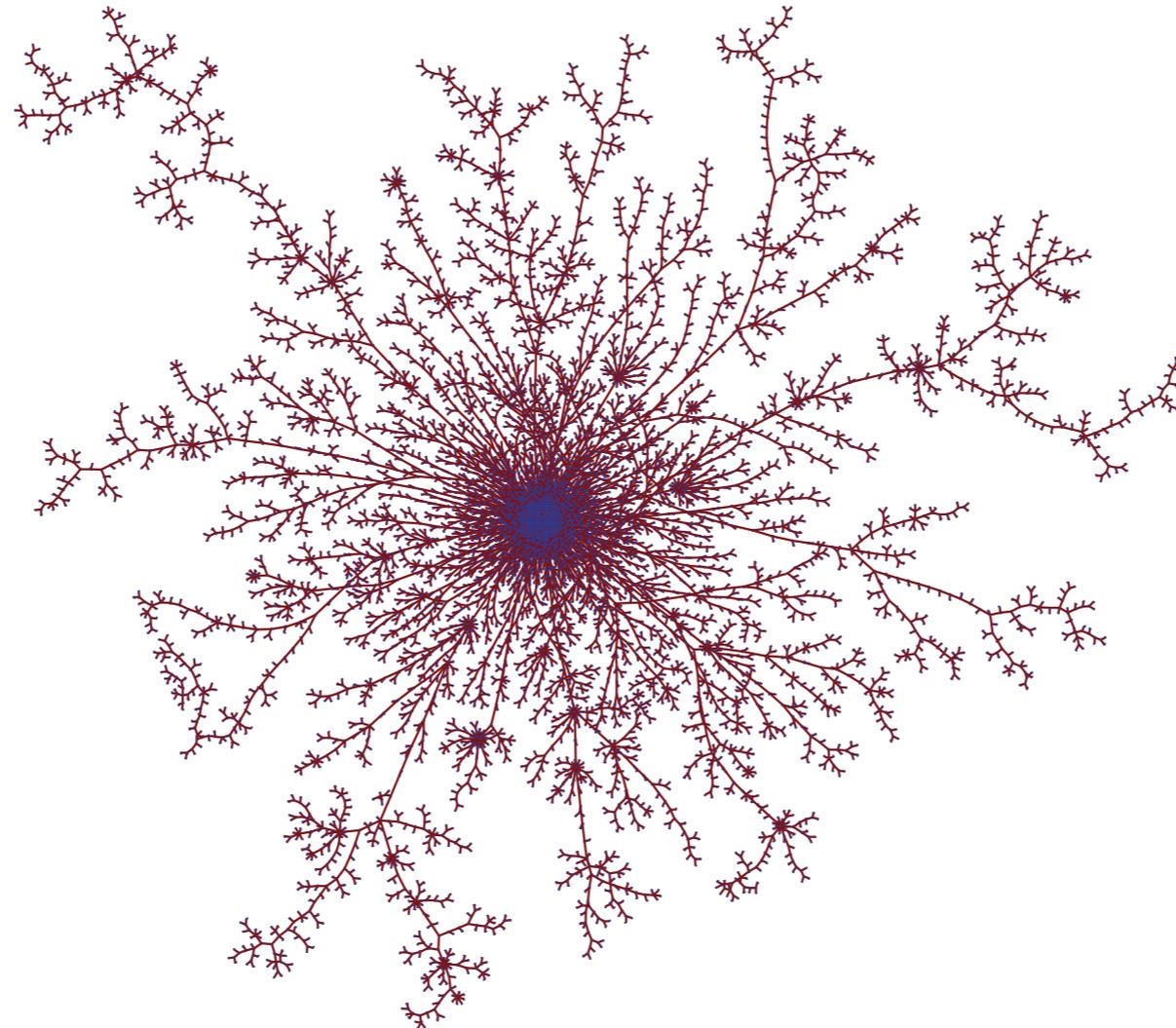


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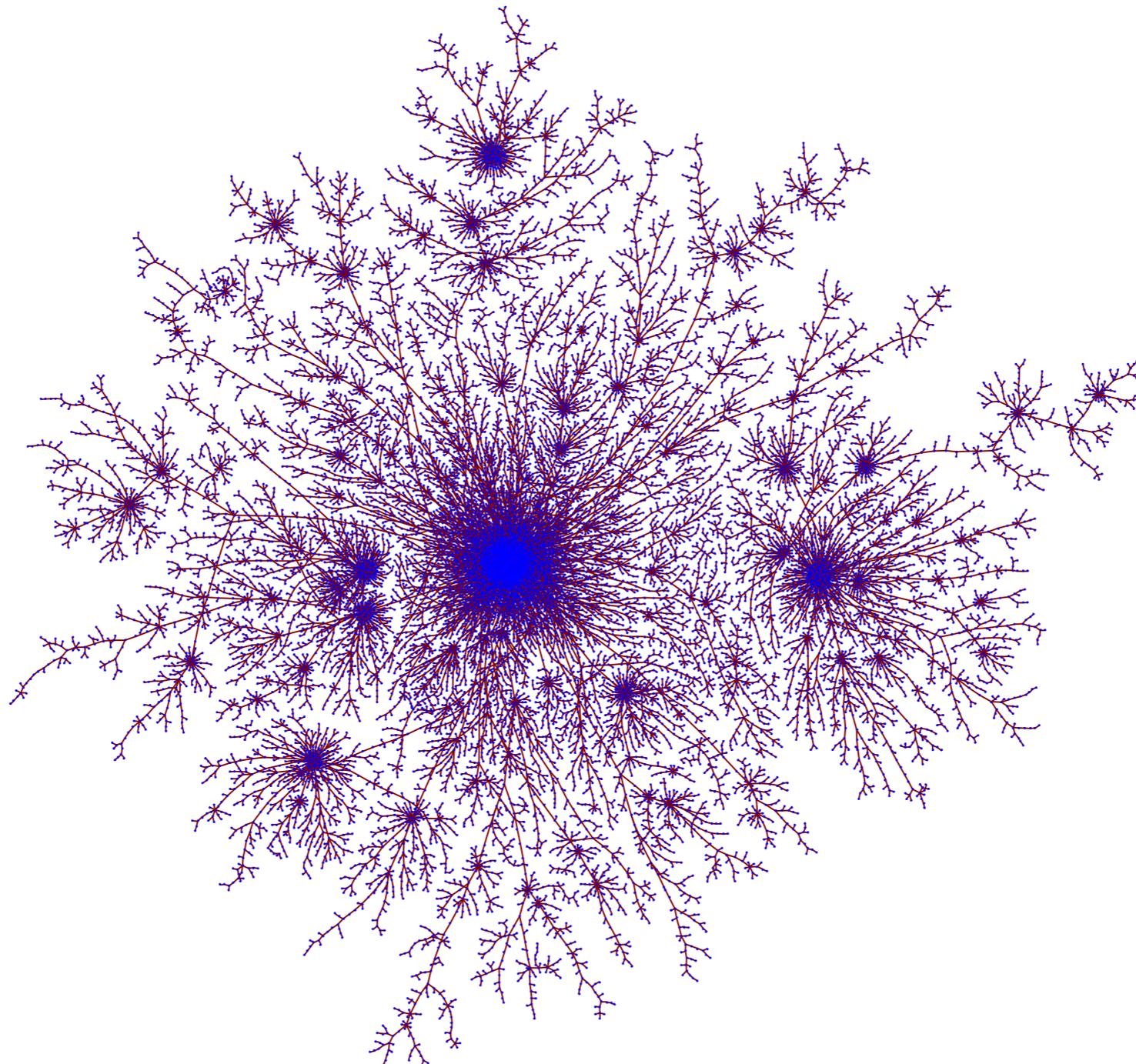


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For example, if

$$\mu_i \sim \frac{1}{\ln(i)^2 i^2},$$

the maximal degree is of order $n/\ln(n)$, the maximum of the other degrees is of order $n/\ln(n)^2$, and the height of the vertex with maximal degree is of order $\ln(n)$.

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Then distances in \mathcal{T}_n are of order \sqrt{n} (up to a constant), and the scaling limit is the **Brownian CRT**.

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Then distances in \mathcal{T}_n are of order $n^{1/\alpha}$ (up to a slowly varying function), and the scaling limit is the **α -stable tree**.
- μ is subcritical and $\mu(n) = L(n)/n^{1+\beta}$ with $\beta > 1$ and L slowly varying.
Then condensation occurs: there is a unique vertex of degree of order n (up to a constant), the other degrees are of order $n^{1/\min(2,\beta)}$ (up to a slowly varying constant), the height of the vertex with maximal degree converges in distribution, the height of the tree is of order $\ln(n)$ and there are no nontrivial scaling limits.

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- μ is critical and $\mu(n) = L(n)/n^2$ with L slowly varying.
Condensation occurs, but at a smaller scale, that is $n/L_1(n)$ (where L_1 is slowly varying), the other degrees are of order $n/L_2(n)$ (where L_2 is slowly varying, with $L_2 = o(L_1)$), and the height of the vertex with maximal degree converges in probability to ∞ .