

# The $\alpha$ -Ford algebraic measure trees

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1 The  $\alpha$ -Ford model

2 Algebraic measure trees

3 Statistics on the limit trees

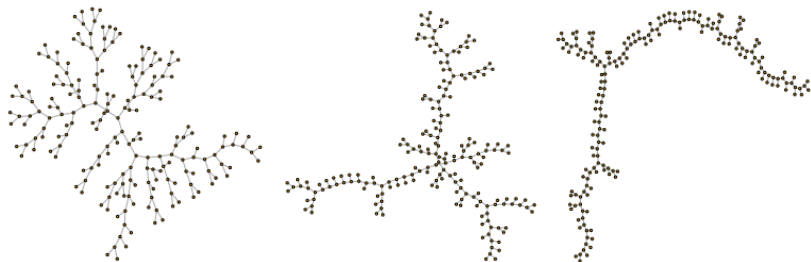


Figure:  $\alpha = 0$  on the left;  $\alpha = 0.5$  in the middle;  $\alpha = 0.9$  on the right.

Fix  $\alpha \in [0, 1]$ . The  $\alpha$ -Ford tree is a random **binary unrooted** tree with  $n$  leaves:

- Start with one edge (yielding 2 leaves).
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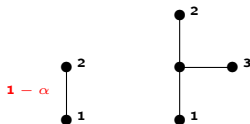
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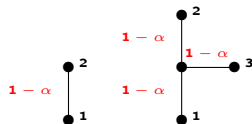
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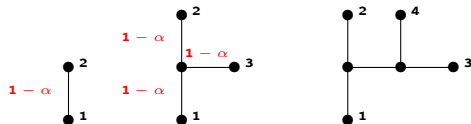




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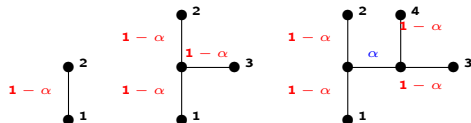
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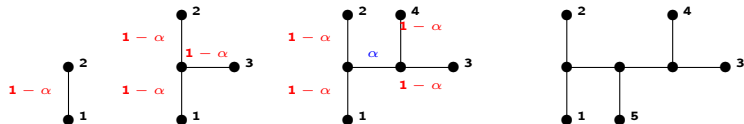
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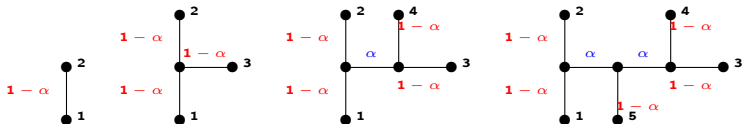
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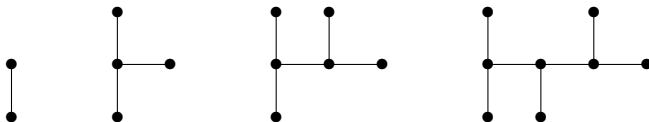
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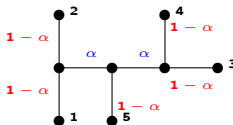


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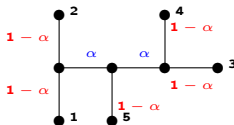
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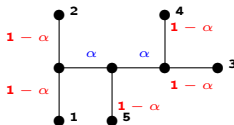
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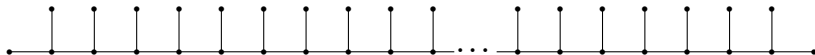
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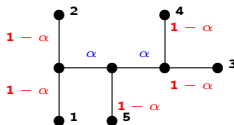
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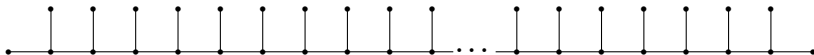


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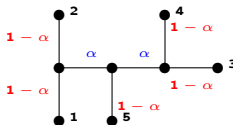
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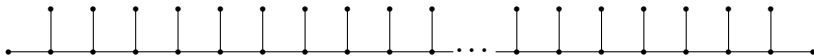
Goal

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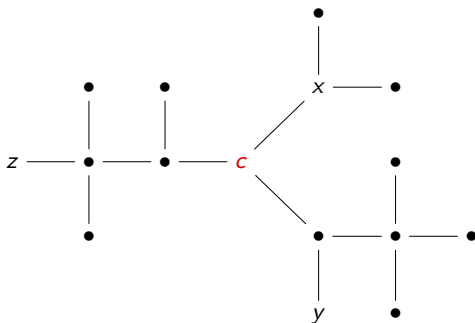
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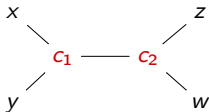
## Goal

What if the tree gets large? Need a *state space* of (*infinite*) trees with a “nice” *topology*.



For  $x, y, z \in T$  there exists a unique *branch point*  $c = c(x, y, z) \in T$  with

$$[x, y] \cap [y, z] \cap [z, x] = \{c\}.$$

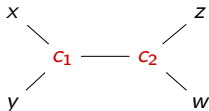


## Definition: algebraic tree

An **algebraic tree** is a set  $T$  together with a symmetric map  $c: T^3 \rightarrow T$  such that:

- (2-point condition) For  $x, y \in T$ ,  $c(x, y, y) = y$ .
- (3-point condition) For  $x, y, z \in T$ ,  $c(x, y, c(x, y, z)) = c(x, y, z)$ .
- (4-point condition) For  $x, y, z, w \in T$ ,

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## Definition

An **algebraic measure tree**  $(T, c, \mu)$  consists of an order separable algebraic tree  $(T, c)$  together with a probability measure  $\mu$  on  $(T, \mathcal{B}(T, c))$ .

We consider the subset of *binary algebraic measure trees*:

$$\mathbb{T}_2 := \{(T, c, \mu) : \deg(v) \leq 3 \forall v \in T, \text{atoms}(\mu) \subseteq \text{leaves}(T)\}.$$

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Idea: “Gromov-weak” topology

A sequence of trees converges to a limit tree if and only if all randomly sampled finite subtrees converge to the corresponding limit subtrees.

# “Sample shape convergence = convergence of subtrees”

When we sample  $n$  points according to  $\mu$ , we can look at the **tree shape** they span.

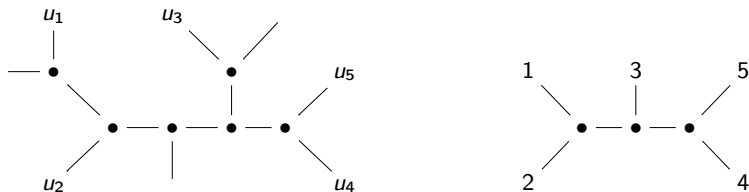


Figure: A tree  $T$  and the shape spanned by 5 points.



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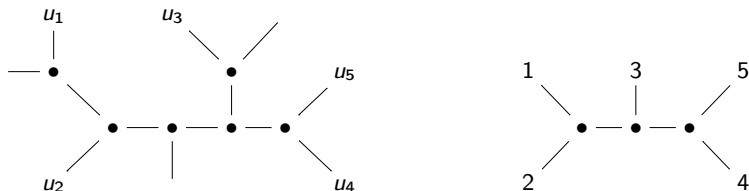


Figure: A tree  $T$  and the shape spanned by 5 points.

A sequence of algebraic measure trees is **sample-shape convergent** if for all  $n$ , the distributions of the  $n$ -tree shapes converge.

Let  $\alpha \in [0, 1]$ . For all  $n \in \mathbb{N}$ , the  $\alpha$ -Ford tree with  $n$  leaves defines a random tree in  $\mathbb{T}_2$  (we put the uniform distribution on the leaves).

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## Proposition [N., Winter '20]

As the number of leaves goes to infinity, the  $\alpha$ -Ford model converges in distribution in  $\mathbb{T}_2$  to a random *continuum* algebraic measure tree.

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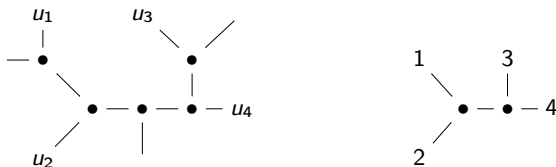
We call the limit the  $\alpha$ -Ford algebraic measure tree.

- $\alpha = 0$ : the Kingman algebraic measure tree.
- $\alpha = \frac{1}{2}$ : the algebraic measure Brownian Continuum Random Tree [Löhr, Mytnik, Winter '18].

We ignore the branch lengths, so we can not look at total length.

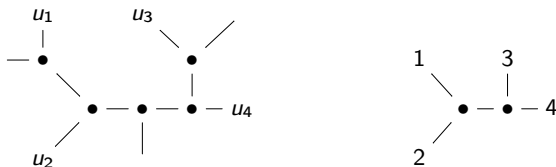
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- **Subtree masses:** sample 3 points according to  $\mu$  and look at how the mass is distributed around their branch point.

For  $\underline{u} = u_1, u_2, u_3$  three leaves of  $T$ , we consider the vector of the three masses of the components connected to  $c(\underline{u})$ :

$$\underline{\eta}(u_1, u_2, u_3) = (\eta_1(\underline{u}), \eta_2(\underline{u}), \eta_3(\underline{u})).$$

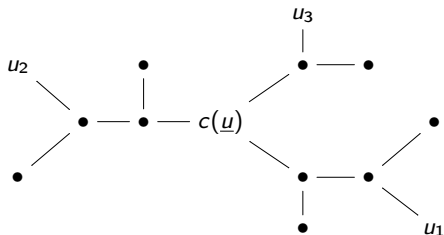


Figure: A tree with 8 leaves;  $\underline{\eta}(\underline{u}) = (\frac{3}{8}, \frac{3}{8}, \frac{2}{8})$ .

We look at polynomials of the form

$$\Phi^f(\chi) := \int_{T^3} \mu^{\otimes 3}(\underline{d}\underline{u}) f(\underline{\eta}(\underline{u})).$$



## Proposition [Aldous '94]

Let  $\mathbb{P}_{\text{CRT}}$  be the law of the **Brownian algebraic CRT**. Then for all  $f: \Delta_2 \rightarrow \mathbb{R}$  continuous bounded,

$$\mathbb{E}_{\text{CRT}} \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) f(\underline{\eta}(\underline{u})) \right] = \int_{\Delta_2} f(\underline{x}) \text{Dir}\left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)(d\underline{x}),$$

where  $\text{Dir}\left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$  is the Dirichlet distribution.

## Proposition [N., Winter '20]

Let  $\mathbb{P}_{\text{Kin}}$  be the law of the **Kingman algebraic measure tree**. Let  $B_{1,2}$  and  $B_{2,2}$  be two independent beta random variables, such that  $B_{1,2}$  has law  $\text{Beta}(1, 2)$  and  $B_{2,2}$  has law  $\text{Beta}(2, 2)$ . Then for all  $f: \Delta_2 \rightarrow \mathbb{R}$  continuous bounded,

$$\begin{aligned} \mathbb{E}_{\text{Kin}} \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) f(\underline{\eta}(\underline{u})) \right] \\ = \frac{1}{6} \sum_{\pi \in \mathcal{S}_3} \mathbb{E} [f \circ \pi^*(B_{1,2} B_{2,2}, B_{1,2}(1 - B_{2,2}), 1 - B_{1,2})], \end{aligned}$$

where  $\mathcal{S}_3$  is the set of permutations  $\{1, 2, 3\}$ , and for  $\pi \in \mathcal{S}_3$ ,  $\pi^*: \Delta_2 \rightarrow \Delta_2$  is the induced map  $\pi^*(\underline{x}) = (x_{\pi(1)}, x_{\pi(2)}, x_{\pi(3)})$ .

→ What about the other  $\alpha \in [0, 1]$ ?

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We can calculate all the moments through a recurrence relation.

$$\begin{aligned} \mathbb{E}_\alpha \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) \eta_1(\underline{u}) \right] &= \frac{1}{3}, \\ \mathbb{E}_\alpha \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) (\eta_1(\underline{u}))^2 \right] &= \frac{1}{5}, \\ \mathbb{E}_\alpha \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) (\eta_1(\underline{u}))^3 \right] &= \frac{11 - 7\alpha}{15(5 - 3\alpha)}, \\ \mathbb{E}_\alpha \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) (\eta_1(\underline{u}))^4 \right] &= \frac{37 - 25\alpha}{63(5 - 3\alpha)}, \\ \mathbb{E}_\alpha \left[ \int_{T^3} \mu^{\otimes 3}(d\underline{u}) (\eta_1(\underline{u}))^5 \right] &= \frac{145 - 165\alpha + 44\alpha^2}{42(5 - 3\alpha)(7 - 3\alpha)}. \end{aligned}$$



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The Annals of Probability, Vol. 21, No. 1, 248-289, 1993.



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