

Multifractal analysis of Galton-Watson trees

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joint work with [Adam Kinnison](#) (Bath)

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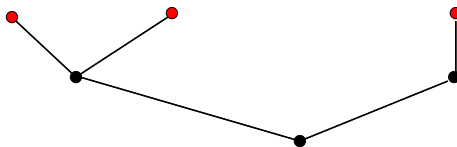
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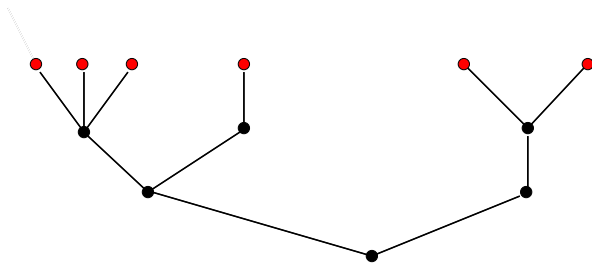
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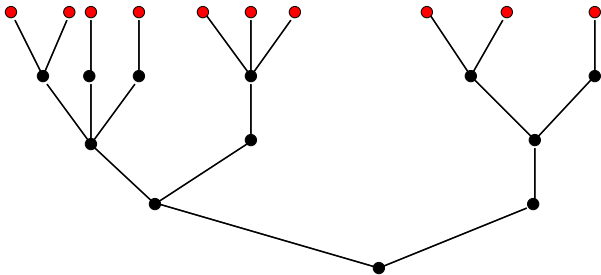
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Galton-Watson trees: the boundary

The **boundary ∂T of the tree** is the collection of all **rays**, i.e. the infinite paths

$$x = (v_0, v_1, v_2, \dots)$$

such that $v_0 = \rho$ and v_i is a child of v_{i-1} , for all $i \geq 1$.

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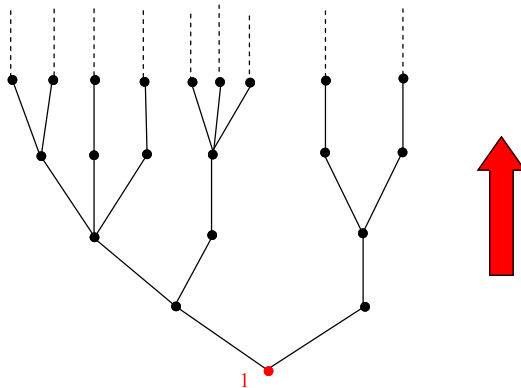
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Probability measures on ∂T correspond to (transient) **random walks** on the tree.

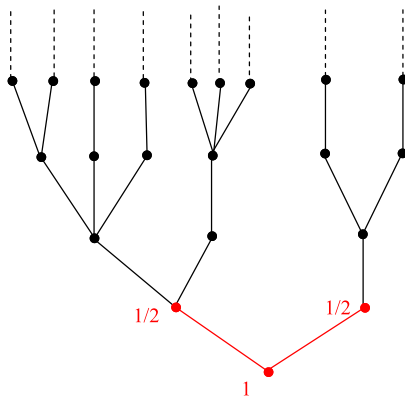
Two interesting measures on the boundary

Visibility or harmonic measure ν



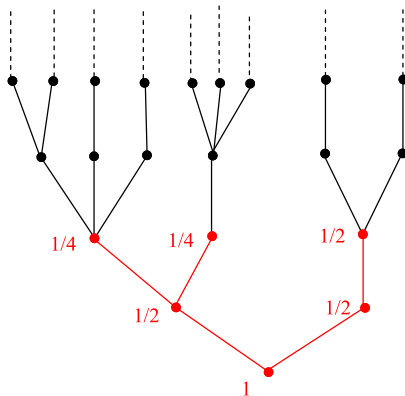
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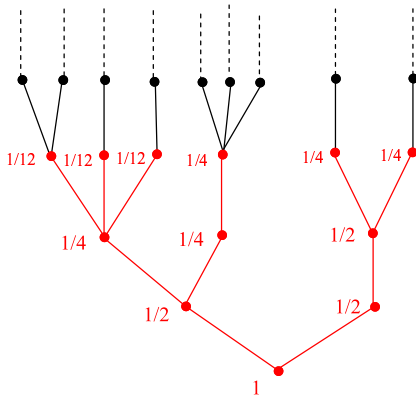
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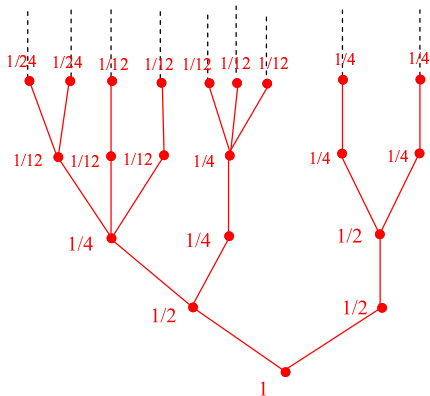
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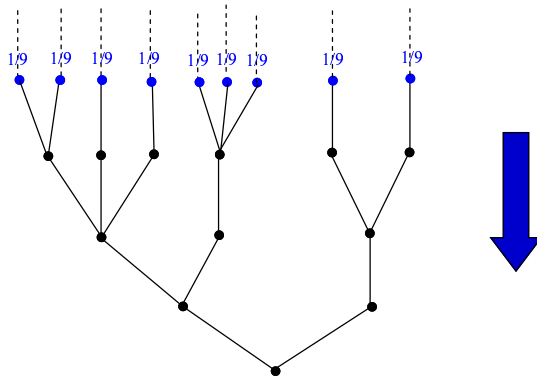
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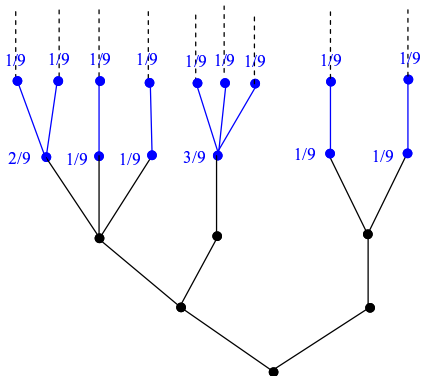
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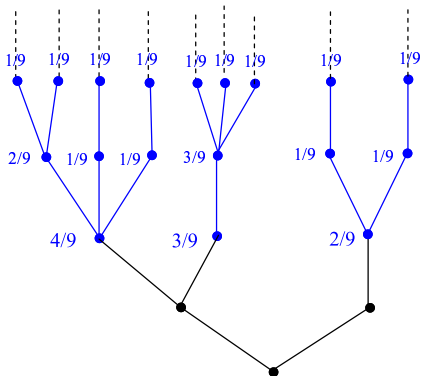
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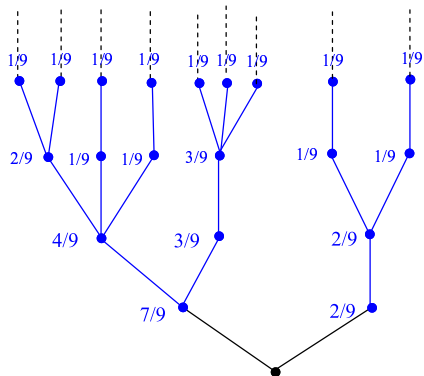
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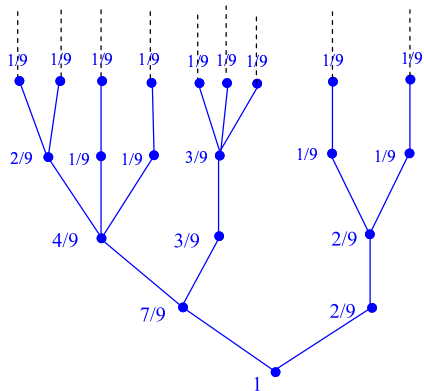
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The **upper local dimension of μ at $x \in \mathcal{X}$** is defined as

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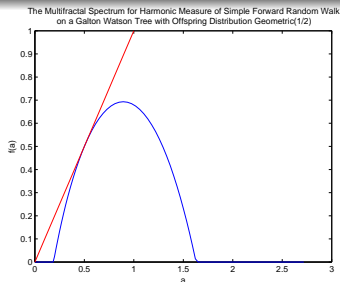
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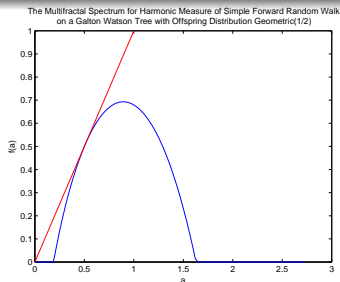
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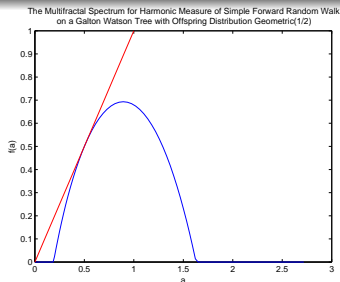
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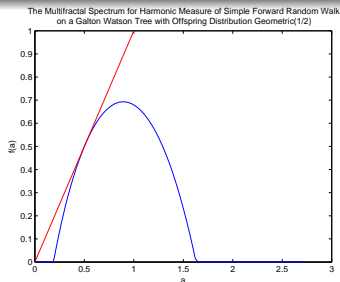
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The branching measure μ

We assume that $\mathbb{P}\{N = 1\} > 0$.

Theorem 2 (Shieh + Taylor 2002, M + Shieh 2004)

Let $a = \log \mathbb{E}N$ and $\tau = -\log \mathbb{P}\{N = 1\}/a$. Then, almost surely,

$$\underline{\dim}_\mu(x) = a \quad \text{for all } x \in \partial T,$$

and, for all $a \leq \theta \leq a(1 + \frac{1}{\tau})$,

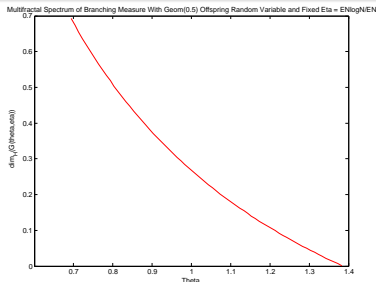
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Other references for simultaneous (or mixed) multifractal spectra:

Barreira and Schmeling 2000, Barreira and Saussol 2001, Olsen 2003.

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Theorem 3 (Kinnison + M 2009)

Let $a < \theta \leq a(1 + \frac{1}{r})$. Then, almost surely, for all $x \in \partial\mathcal{T}$ we have that

$$\overline{\dim}_\mu(x) \geq \theta \quad \text{implies} \quad \overline{\dim}_\nu(x) \geq \frac{\theta}{a} \underline{\dim}_\nu(x),$$

and therefore the local dimension $\dim_\nu(x)$ **never exists**.

Branching and visibility measure

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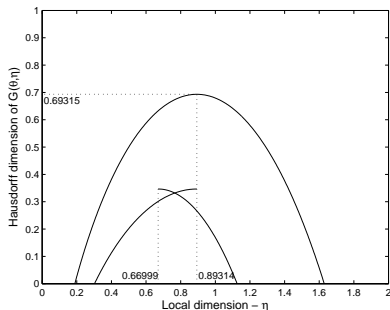
Slices through the spectrum

Left branch: If $\eta \leq \eta_{\text{typ}}$, then

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Right branch: If $\eta \geq \frac{a}{\theta} \eta_{\text{typ}}$, then

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The geometry behind Theorem 2

We first look at the multifractal spectrum for the **branching measure** μ ,

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Therefore the rays $x \in \partial T$ with

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- of order e^{am} vertices in the m th generation,
- and each independently has a chance of

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A refinement of this argument (based on percolation) yields the Hausdorff dimension of the set of rays with $\overline{\dim}_\mu(x) \geq \theta$.

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Hence for a ray with $\overline{\dim}_\mu(x) \geq \theta$ necessarily

$\overline{\dim}_\nu(x)$ and $\underline{\dim}_\nu(x)$ differ by a factor of at least $\frac{\theta}{a}$

and therefore the local dimension cannot exist for such a ray.

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The probability that a fixed vertex v is the first one in such a string is

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For m large, by the dimension spectrum for the visibility measure there are about

$$\exp\left\{m \inf_{\beta} \{-\eta\beta - \gamma(\beta)\}\right\}$$

vertices v which satisfy

$$\nu\{x \ni v_i\} \approx e^{-i\eta} \text{ for all } 1 \leq i \leq m.$$

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By the law of large numbers we therefore expect that we have about

$$\exp\{m(-\tau(\theta - a) + \inf_{\beta} \{-\eta\beta - \gamma(\beta)\})\}$$

vertices in generation m satisfying these two constraints.

The geometry behind Theorem 4

First look at $\eta \leq \eta_{\text{typ}}$. If a vertex v in generation m satisfies the two constraints, then its descendants satisfy the required lower bound

$$\nu\{x \ni v_i\} \approx e^{-m\eta} \geq e^{-i\eta} \text{ for } i = m+1, \dots, \frac{\theta}{a}m.$$

Therefore in generation $\frac{\theta}{a}m$ we expect about

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vertices satisfying the constraint on the branching measure for the radius e^{-m} and the constraint on the visibility measure for all radii $1 \geq r \geq e^{-m\frac{\theta}{a}}$.

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Next look at $\eta \geq \eta_{\text{typ}}^-$. If a vertex v in generation m satisfies the two constraints, then

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It is therefore necessary that the vertex v satisfies the stronger condition

$$\nu\{x \ni v_i\} \approx e^{-i\frac{\theta}{a}\eta} \text{ for } 1 \leq i \leq m.$$

to ensure that all descendants v_i in generations $i = m+1, \dots, \frac{\theta}{a}m$ still satisfy the required upper bound.

A.L. Kinnison and P. Mörters.

Simultaneous multifractal analysis of branching and visibility measure on a Galton-Watson tree. *Submitted for publication.*

Available at <http://people.bath.ac.uk/maspm>