

# Quantum Gravity, Random Tensors and Renormalization I

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Regards sur la gravité quantique  
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## Plan of the Mini-Course

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A Recent Reference: Riv., "The Tensor Track, III," [arXiv:1311.1461](https://arxiv.org/abs/1311.1461)

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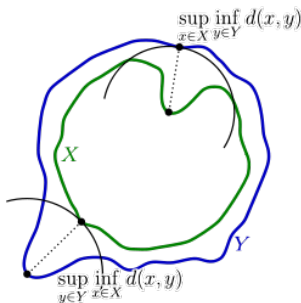
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Remark: Feynman graphs are **metric** spaces (for the graph distance) hence sum over Feynman graphs can in principle lead to random metric spaces as desired for quantizing gravity.

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Large  $N$  behavior studied with  $1/N$  expansion

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Invariant interactions:  $(\bar{\phi} \cdot \phi)^p$  factorized, hence **not connected**. For  $p = 2$ , interacting  $\phi^4$ -type vector model

$$Z_{\text{int}}(\lambda, N) = \frac{1}{(2i\pi)^N} \int e^{-(\bar{\phi} \cdot \phi) - \lambda(\bar{\phi} \cdot \phi)^2} d\bar{\phi} d\phi$$

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**Propagator:** from models to QFT with renormalization group flow

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(Single) Matrix Models have exactly **one** connected invariant interaction at every (even) degree, namely  $\text{Tr}(MM^\dagger)^p$ .

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Many connected invariant interaction at every (even) degree.

$$Z_1^c(n) = 1, 0, 0, 0, 0, \dots$$

$$Z_2^c(n) = 1, 1, 1, 1, 1, 1, \dots$$

$$Z_3^c(n) = 1, 3, 7, 26, 97, 624, 4163, \dots$$

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Remark: symmetry of interactions could be interpreted as pre-geometric locality.

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If we apply this idea to **matrix models** we obtain **non-commutative quantum field theory**. If we apply this idea to **tensor models** we obtain a higher category of quantum field theories which we call **tensorial group field theories**.

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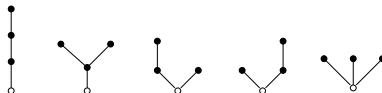
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The second space can be seen as a set of random labels living on the first.

# The Continuous Random Tree

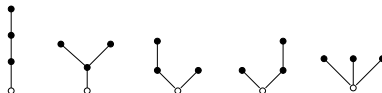
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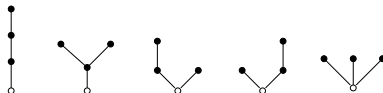
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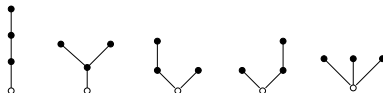
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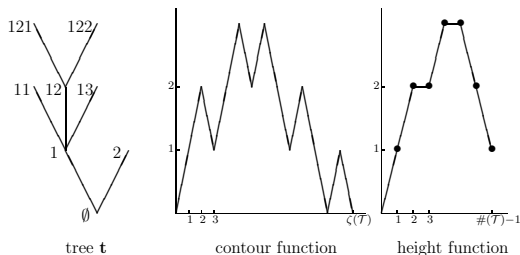
The equidistributed measure on plane trees converges (in Gromov-Hausdorff sense) to a universal object as  $n \rightarrow \infty$ , namely the Continuous Random Tree (CRT).

## Main Properties of the Continuous Random Tree

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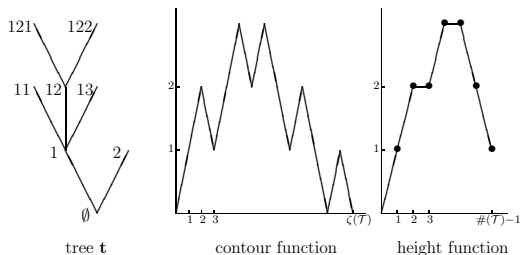
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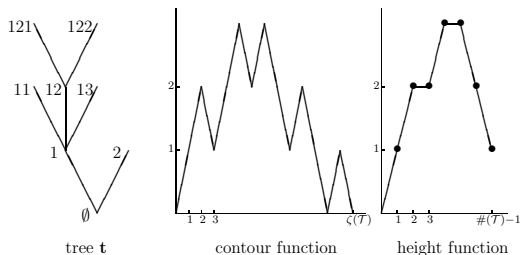
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$$d_{\text{Hausdorff}} = 2, d_{\text{spectral}} = 4/3.$$

# CRT and Vector Models

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Simplest interacting complex  $\phi^4$  vector model: conjugate vector fields  
( $\phi = \{\phi_i\}, \bar{\phi} = \{\bar{\phi}_i\}, i = 1, \dots, N, \lambda(\bar{\phi} \cdot \phi)^2$  interaction).

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$$\begin{aligned} Z(\lambda, N) &= \frac{1}{(2i\pi)^N} \int d\sigma \frac{e^{-\sigma^2/2}}{\sqrt{2\pi}} \int e^{-(\bar{\phi} \cdot \phi) + i\sqrt{2\lambda}(\bar{\phi} \cdot \phi)\sigma} d\bar{\phi} d\phi \\ &= \int \frac{d\sigma}{\sqrt{2\pi}} e^{-\sigma^2/2 - N \log(1 - i\sqrt{2\lambda}\sigma)} \end{aligned}$$

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We recognize  $\tau_c(z) = \sqrt{z}T(z)$  where  $T(z) = \sum_n C_n z^n = \frac{1}{2z}[1 - \sqrt{1 - 4z}]$ .

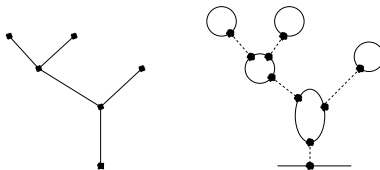
$$\begin{aligned} \lim_{N \rightarrow \infty} \frac{\log Z(\lambda, N)}{N} &= -f(\tau_c) = -\tau_c^2/2 - \log(1 - \sqrt{z}\tau_c) \\ &= \frac{-1}{4z}[1 - \sqrt{1 - 4z} - 2z] - \log\left[\frac{1}{2}(1 + \sqrt{1 - 4z})\right]. \end{aligned}$$

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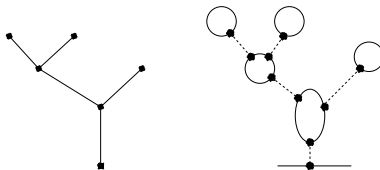
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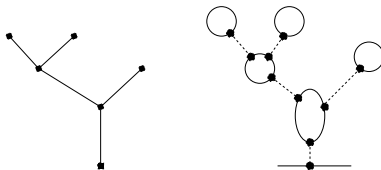


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- $(1/N)^p$ -sub leading term correspond to add exactly  $p$  loops. Double scaling would include all graphs hence unstable at  $z > 0$ .

# The Dyck Map

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One can understand the **metric properties** of the CRT via a nice **one-to-one map**.

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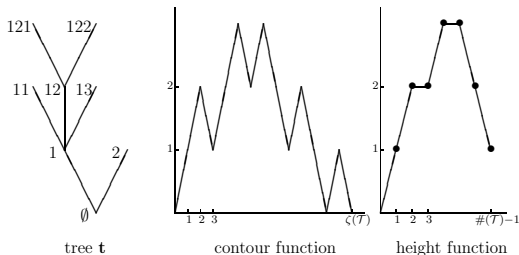
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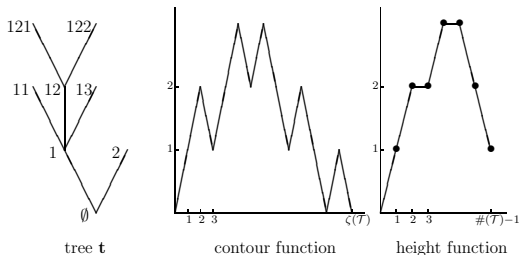
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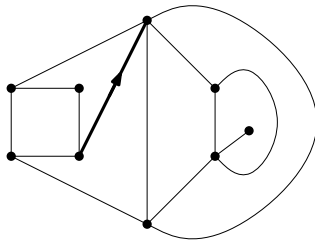
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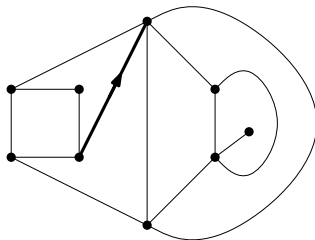


This Dyck map easily proves  $d_{Hausdorff} = 2$ ; with further work  $d_{spectral}=4/3$ .

## Planar Rooted Quadrangulations

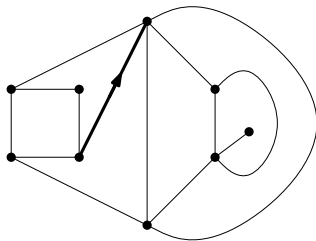


## Planar Rooted Quadrangulations

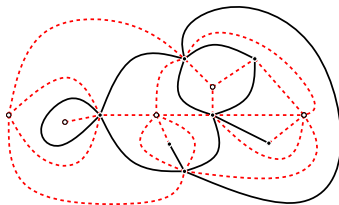


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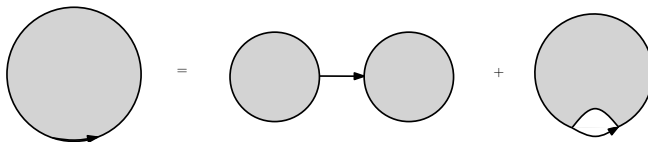
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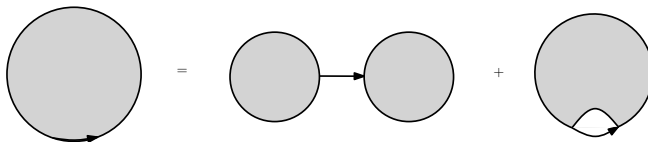
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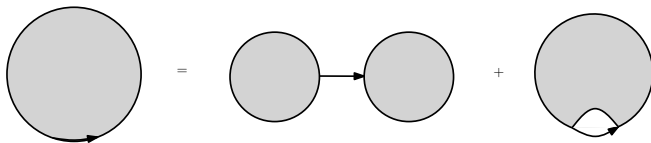


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$$Q_n = 3^n \frac{2}{n+2} \frac{1}{n+1} \binom{2n}{n}.$$

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$$2 - 2g = V - L + F = -V + F, \text{ since } L = 2V.$$

hence planar graphs lead at  $N \rightarrow \infty$  ('t Hooft, 1974).

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- Double scaling:  $N \rightarrow \infty$  and  $\lambda \rightarrow \lambda_c = 1/24$ ,  $\kappa^{-1} = N^5/4(\lambda - \lambda_c)$  fixed

$$G_{2,double\ scaling}(\lambda) = \sum_h a_h \kappa^{2h}$$

includes all graphs, at  $\lambda > 0$ , hence unstable.

# The Schaeffer Map

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One can understand the **metric properties** of the Brownian sphere via a nice **one-to-one map**.

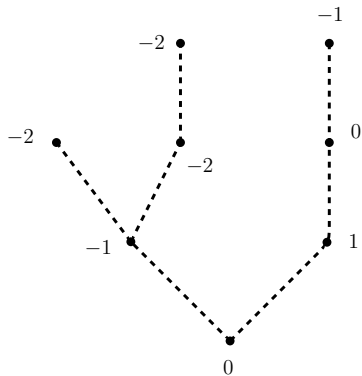
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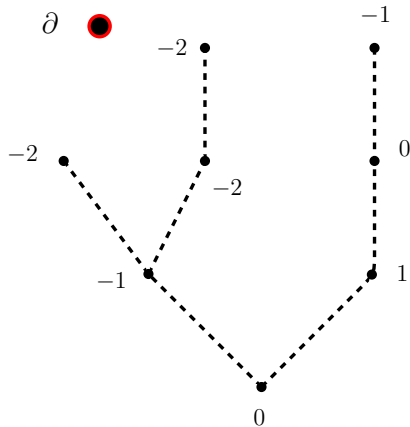
$$(n+2)Q_n = 2 \cdot 3^n C_n, \quad .$$

The Schaeffer map identifies **rooted, pointed** planar quadrangulations with **well-labeled, oriented** rooted plane trees.

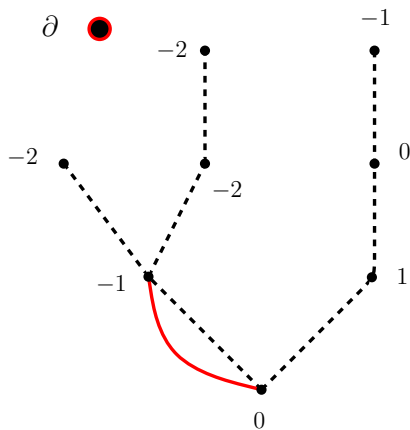
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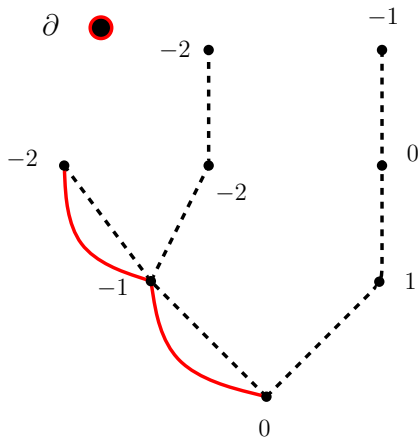
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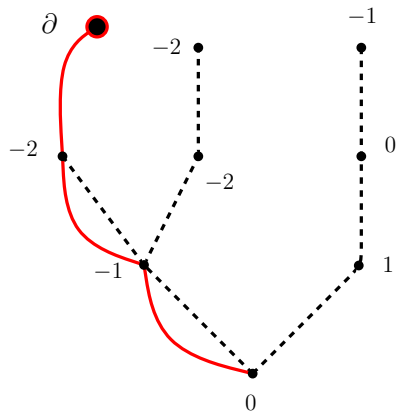
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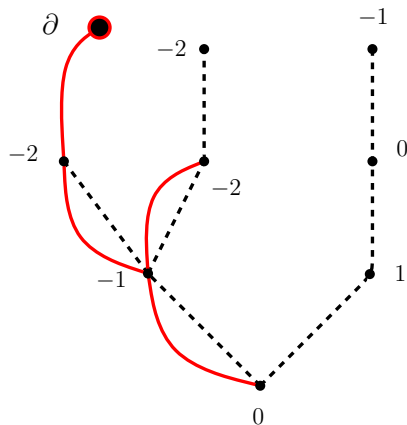
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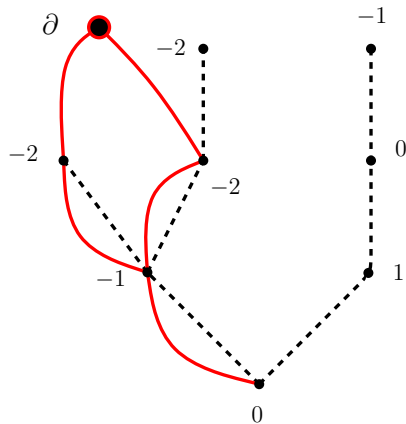
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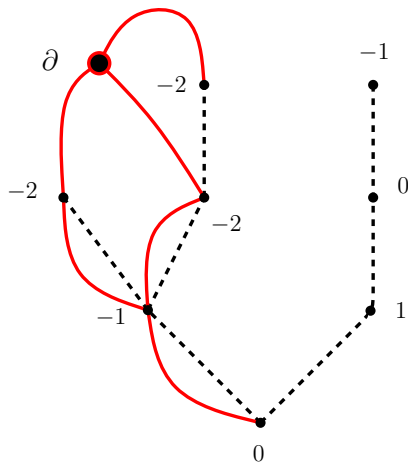
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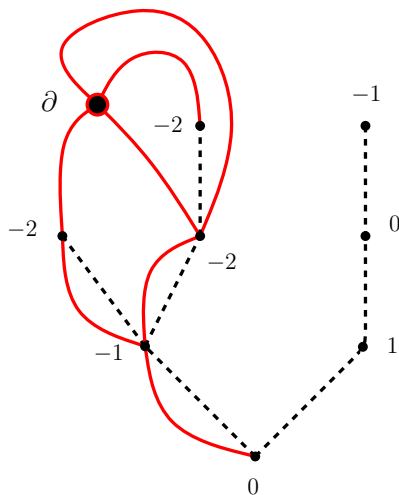
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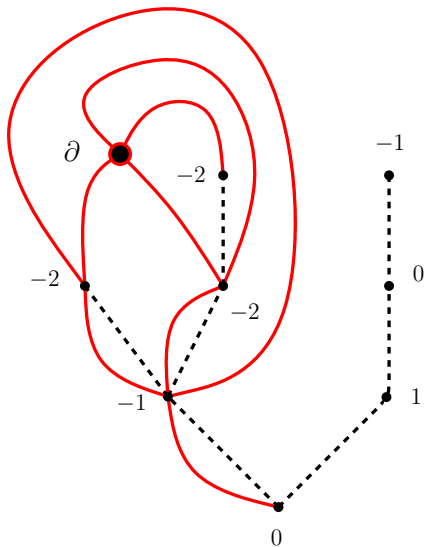
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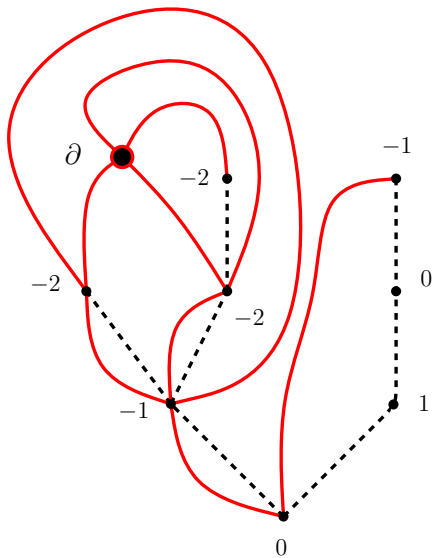
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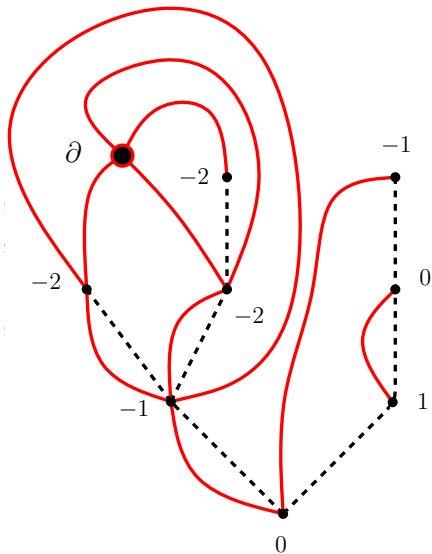
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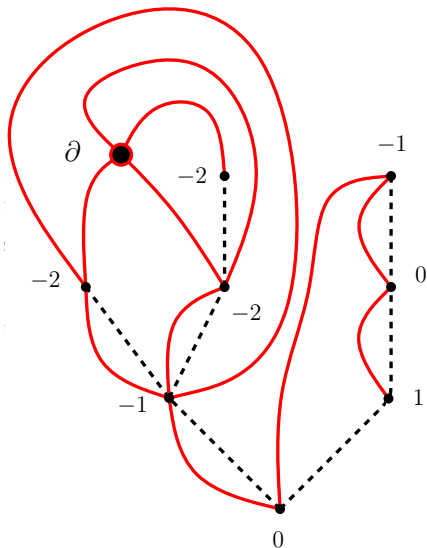
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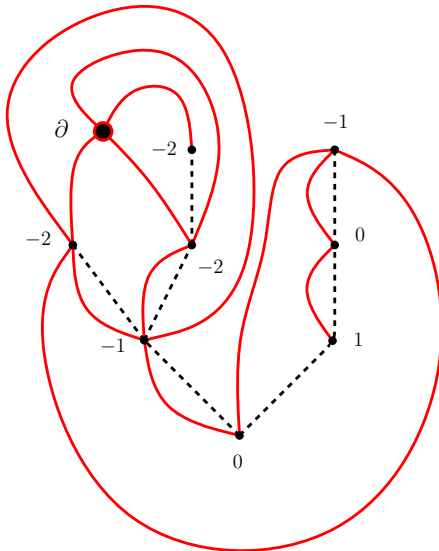
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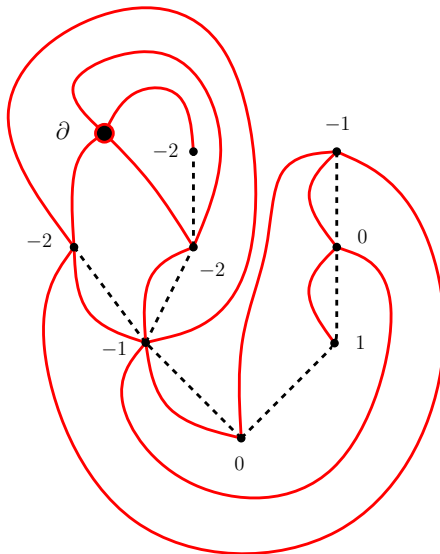
# Large Quadrangulations



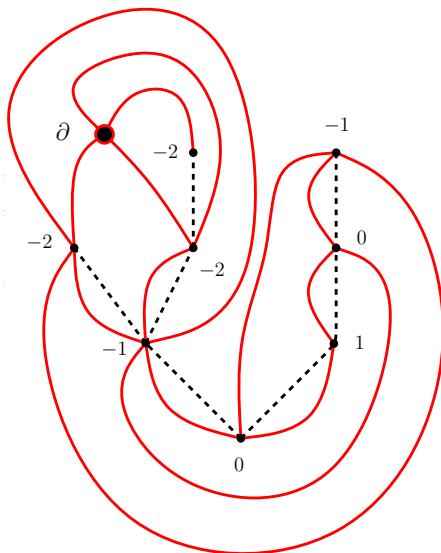
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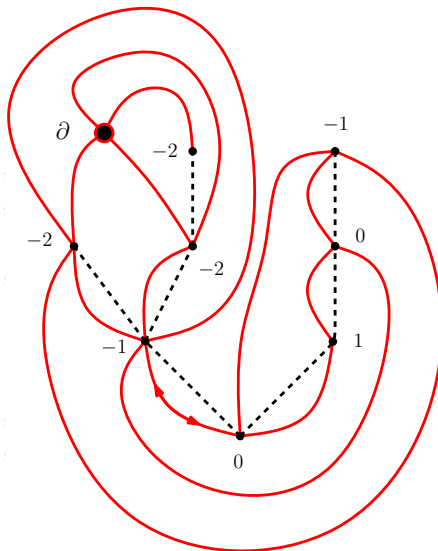
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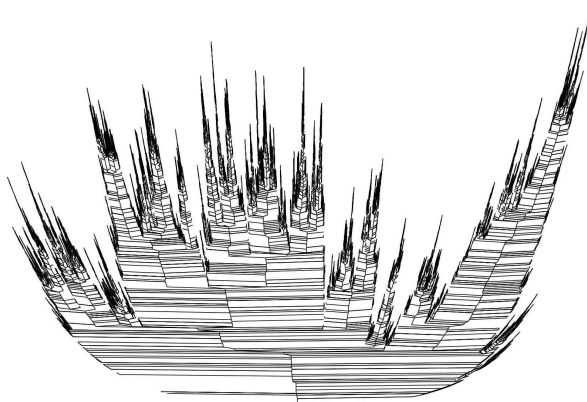
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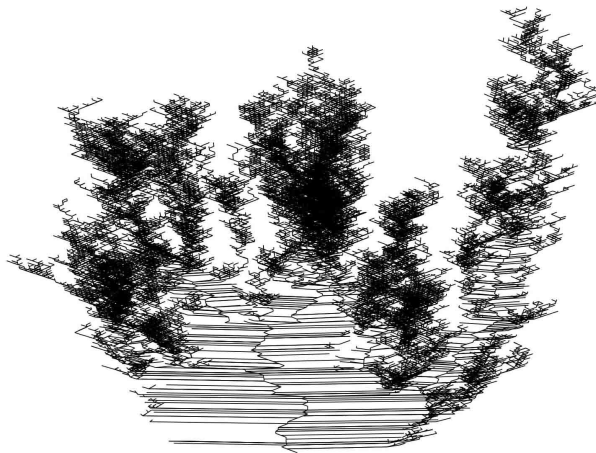
This space has Hausdorff dimension 4 and is **almost surely homeomorphic to the two-dimensional sphere**. It is expected to have spectral dimension 2.

## A Look at Large Random Quadrangulations

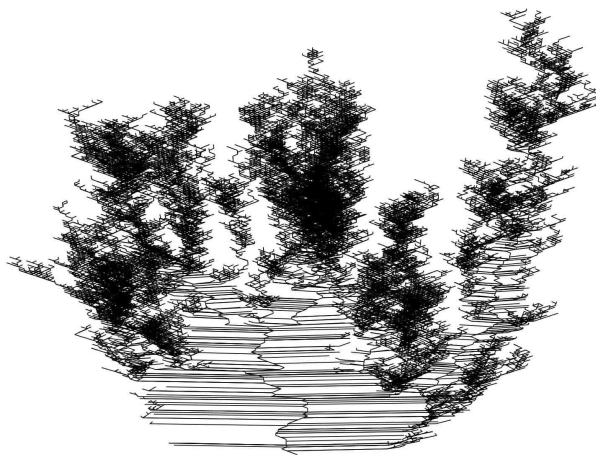


The Probabilist's View: The Brownian Snake, Head on

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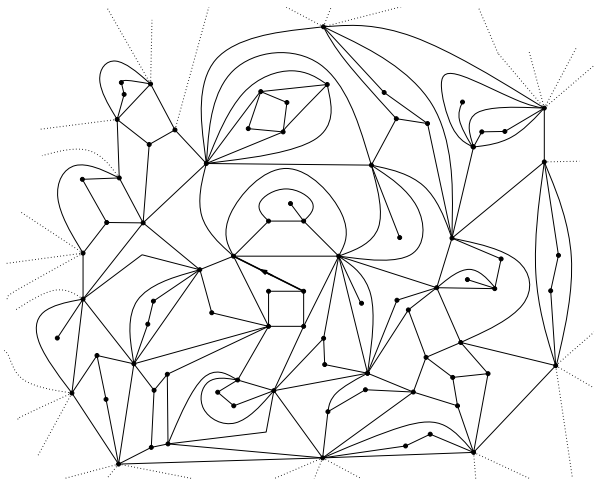


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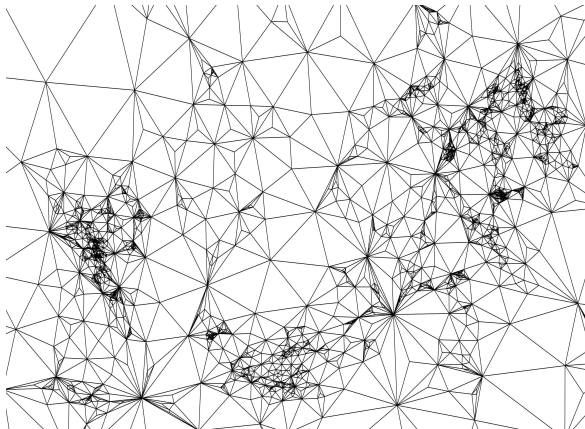
The Probabilist's View: The Brownian Snake, Profile

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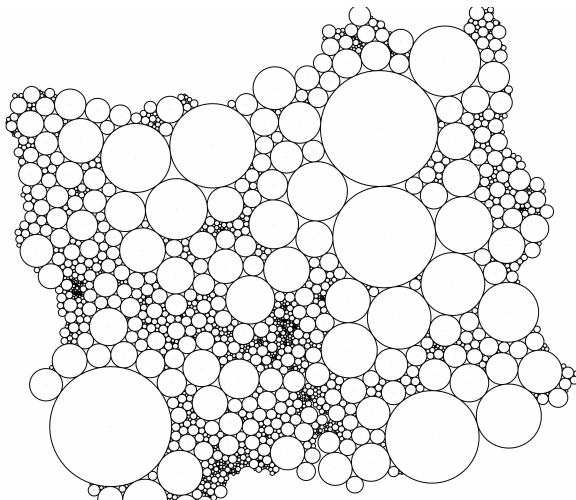
The Topological View

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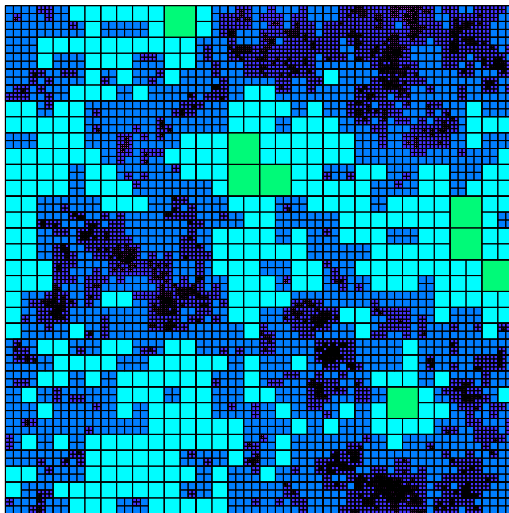
Uniformized Through Riemann Mapping Theorem

## A Look at Large Random Quadrangulations



Using the Circle Packing Theorem (Courtesy: Krikun)

## A Look at Large Random Quadrangulations



The Liouville Theory (Courtesy: Duplantier)

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$\Delta$  and  $x$  are scaling Euclidean dimensions **coupled** or **not** to quantum gravity plus matter with conformal central charge  $c$  (for pure gravity  $c = 0, \gamma = \sqrt{8/3}$ ; for Ising,  $c = 1/2, \gamma = \sqrt{3}$ ).

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- This Brownian sphere is best thought of as a CRT equipped with fluctuation fields, the Schaeffer labels, which add loops, hence space-time shortcuts.
- What about higher dimensions QG3, QG4... ?