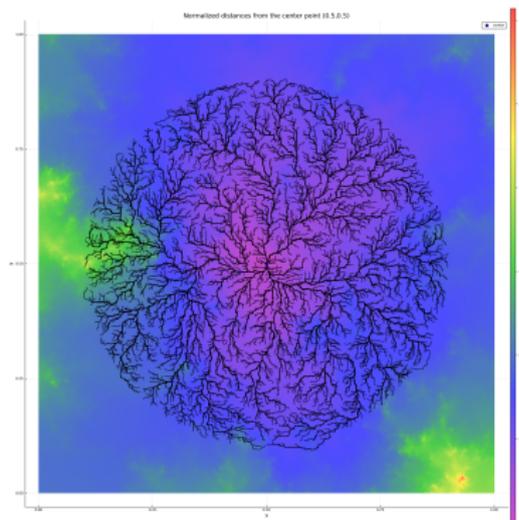
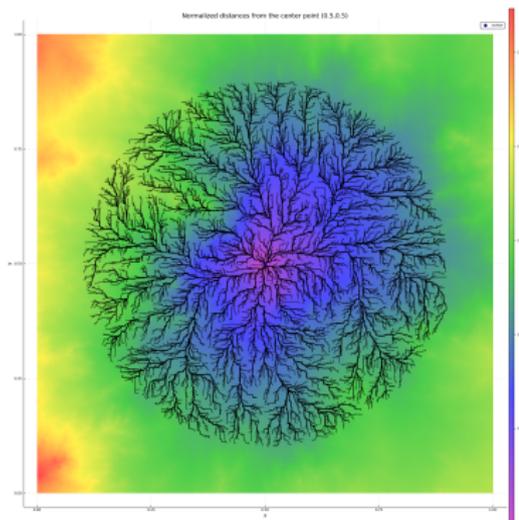


Subsequential scaling limits for Liouville graph distance

March 5, 2024

Liouville first-passage percolation

- ▶ “First passage percolation on the exponential of Gaussian free field”: Liouville first-passage percolation (LFPP) is first-passage percolation on $e^{\xi h}$, where h is a (discrete or circle-averaged continuous) Gaussian free field and ξ is an inverse-temperature parameter.



Distance exponent

- ▶ How does the metric (left–right crossing distance, point-to-point distance, diameter. . .) scale with the box size S (equivalently the mollification scale δ) and the temperature ξ ?
 - ▶ Y. Watabiki (1993): non-rigorous prediction for the distance exponent.
 - ▶ J. Ding–F. Zhang (PTRF, 2018): the exponent of the distance is *non-universal* among log-correlated Gaussian fields. Therefore, answering this question must use rather fine information about the Gaussian free field.
 - ▶ Ding–S. Goswami (CPAM, 2019): upper bounds on the exponent for high temperature $\xi \ll 1$ (seems to contradict Watabiki).
 - ▶ Ding–E. Gwynne (Preprint, 2018): bounds for general (subcritical) temperature.
 - ▶ Includes precise value only at one special value of ξ , which corresponds to expected equivalence between limiting metric and the Brownian map (J.P. Miller–S. Sheffield 2015–’16).

Limiting metric

- ▶ Does there exist a scaling limit of the metric as the box size goes to ∞ (equivalently, as the mollification scale goes to 0)?
 - ▶ Need to rescale the metric by the typical distance, which is so far unknown.
 - ▶ No non-universality expected here: a limiting metric should exist for a wide class of log-correlated Gaussian fields.
- ▶ Uniqueness of the limiting metric done via axiomatic characterization by Gwynne and Miller.

Liouville quantum gravity

- ▶ *Liouville quantum gravity* is supposed to be a random geometry given by reweighting the Lebesgue measure, or the Euclidean metric, by $e^{\gamma h}$, where h is a (continuum) Gaussian free field.
- ▶ Problem: h is not a function but rather a distribution, and so it cannot be exponentiated.
- ▶ Construction of LQG measure (B. Duplantier–S. Sheffield 2010):
 - ▶ Consider $\mu_\varepsilon(dx) = e^{\gamma h_\varepsilon(x) + \frac{1}{2}\gamma^2 \log \varepsilon} dx$, where h_ε is the circle average process of the Gaussian free field with circles of radius ε .
 - ▶ Note the *renormalization* by $\varepsilon^{\gamma^2/2}$.
 - ▶ If $\gamma < 2$ (critical temperature), then there is a random measure μ so that, with probability 1, we have

$$\lim_{k \rightarrow \infty} \mu_{2^{-k}} = \mu$$

weakly.

Liouville graph distance

- ▶ Liouville FPP is intuitive to define, but there is another notion of discretized metric related to Liouville quantum gravity.
- ▶ Let $\Omega \subset \mathbf{R}^2$. Consider the γ -LQG measure μ on some domain $\Omega' \supset \Omega$.
- ▶ For $x, y \in \Omega$, define $d_{\Omega', \Omega, \delta}^{LGD}(x, y)$ to be the minimum size of a set of *Euclidean* balls with centers in Ω , each having *LQG measure* at most δ , whose union forms a connected set that includes x and y .

Main result

Notation: if \mathfrak{R} is a box, let \mathfrak{R}^* be a box with three times the side lengths, centered around \mathfrak{R} .

Theorem

Suppose that $\gamma < 2$. Let \mathfrak{R} be a rectangular subset of \mathbf{R}^2 . For any sequence $\delta_n \downarrow 0$, there is a subsequence (δ_{n_k}) and a nontrivial random metric $d : \mathfrak{R} \times \mathfrak{R} \rightarrow \mathbf{R}$ so that

$$Q_{\delta_{n_k}}^{-1} d_{\mathfrak{R}^*, \mathfrak{R}, \delta_{n_k}}^{\text{LGD}} \rightarrow d$$

in distribution with respect to the uniform topology of functions $\mathfrak{R} \times \mathfrak{R} \rightarrow \mathbf{R}$. Here Q_δ is the median Liouville graph distance between and left- and right-hand sides of \mathfrak{R} .

- ▶ No assumption of high temperature beyond subcriticality.
- ▶ Don't need to know the scaling exponent (size of Q).

Multiscale analysis: the effect of the coarse field

- ▶ We would like to understand the LGD metric using multiscale analysis.
 - ▶ If \mathfrak{C} is a subbox of \mathfrak{R} , we would like to be able to relate $d_{\mathfrak{R}^*, \mathfrak{C}, \delta}^{LGD}$ and $d_{\mathfrak{C}^*, \mathfrak{C}, \delta}^{LGD}$.
- ▶ There is a relationship between $h_{\mathfrak{R}^*}$ and $h_{\mathfrak{C}^*}$: we can couple the GFFs so that $h_{\mathfrak{R}^*} \upharpoonright_{\mathfrak{C}^*} = h_{\mathfrak{C}^*} + h_{\mathfrak{R}^*: \mathfrak{C}^*}$, where $h_{\mathfrak{R}^*: \mathfrak{C}^*}$ (“coarse field”) is the harmonic interpolation of $h_{\mathfrak{R}^*}$ on $h_{\mathfrak{C}^*}$ (hence smooth). Moreover, $h_{\mathfrak{R}^*: \mathfrak{C}^*}$ and $h_{\mathfrak{C}^*}$ (“fine field”) are independent.
 - ▶ Compare this to the branching random walk picture.
 - ▶ This means that

$$d_{\mathfrak{R}^*, \mathfrak{C}, \delta''}^{LGD} \leq d_{\mathfrak{R}^*, \mathfrak{C}, \delta}^{LGD} \leq d_{\mathfrak{R}^*, \mathfrak{C}, \delta'}^{LGD}$$

where

$$\delta'' = \exp \left\{ -\gamma \min_{x \in \mathfrak{C}} h_{\mathfrak{R}^*: \mathfrak{C}^*}(x) \right\}, \quad \delta' = \exp \left\{ -\gamma \max_{x \in \mathfrak{C}} h_{\mathfrak{R}^*: \mathfrak{C}^*}(x) \right\}.$$

Multiscale analysis: the effect of the coarse field

- ▶ $d_{\mathfrak{R}^*, \mathfrak{G}, \delta''}^{LGD} \leq d_{\mathfrak{R}^*, \mathfrak{G}, \delta}^{LGD} \leq d_{\mathfrak{R}^*, \mathfrak{G}, \delta'}^{LGD}$, where
 $\delta'' = \exp\{-\gamma \min_{x \in \mathfrak{G}} h_{\mathfrak{R}^*: \mathfrak{G}^*}(x)\} \delta$,
 $\delta' = \exp\{-\gamma \max_{x \in \mathfrak{G}} h_{\mathfrak{R}^*: \mathfrak{G}^*}(x)\} \delta$.
- ▶ Also have *scaling property of LQG*:

$$d_{\alpha \mathfrak{R}^*, \alpha \mathfrak{G}, \alpha^{2+\gamma^2/2} \delta}^{LGD} \stackrel{\text{law}}{=} d_{\mathfrak{R}^*, \mathfrak{G}, \delta}^{LGD}.$$

- ▶ Maximum of the Gaussian free field: if $\mathfrak{G}_1, \dots, \mathfrak{G}_{K^2}$ is a set of translates of $\frac{1}{K} \mathfrak{R}$, all contained in \mathfrak{R} , then we have

$$\min_{i=1}^{K^2} \min_{x \in \mathfrak{G}} h_{\mathfrak{R}^*: \mathfrak{G}^*}(x) \gtrsim -2 \log K; \quad \max_{i=1}^{K^2} \max_{x \in \mathfrak{G}} h_{\mathfrak{R}^*: \mathfrak{G}^*}(x) \lesssim 2 \log K$$

with Gaussian concentration.

- ▶ Putting this all together: we have that

$$d_{\alpha_i \mathfrak{G}_i^*, \alpha_i \mathfrak{G}_i, \delta}^{LGD} \lesssim d_{\mathfrak{R}^*, \mathfrak{G}_i, \delta}^{LGD} \lesssim d_{\beta_i \mathfrak{G}^*, \beta_i \mathfrak{G}_i, \delta}^{LGD}$$

where (since $\gamma < 2$)

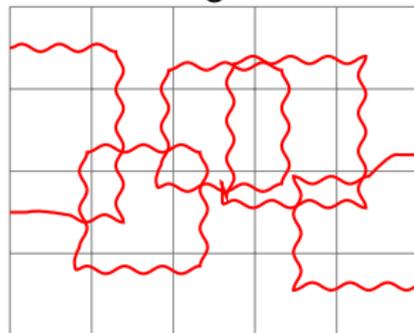
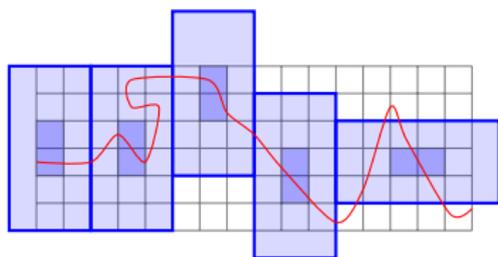
$$\max_i \beta_i \leq K^{\frac{2\gamma}{2+\gamma^2/2}} < K.$$

Subcriticality leads to averaging

- ▶ Recall that $\mathfrak{S}_1, \dots, \mathfrak{S}_{K^2}$ is a set of translates of $\frac{1}{K}\mathfrak{R}$.
- ▶ $d_{\alpha_i \mathfrak{S}_i^*, \alpha_i \mathfrak{S}_i, \delta}^{LGD} \leq d_{\mathfrak{R}^*, \mathfrak{S}_i, \delta}^{LGD} \lesssim d_{\beta_i \mathfrak{S}_i^*, \beta_i \mathfrak{S}_i, \delta}^{LGD}$, where
 $\max_i \beta_i \leq K^{\frac{\gamma}{2+\gamma^2/2}} < K$.
- ▶ “Subboxes look smaller than the overall box, even after considering the coarse field.”
- ▶ Idea: the crossing weight across a large box should feel the weight from many subboxes, and thus should be doing some kind of averaging, leading to concentration.

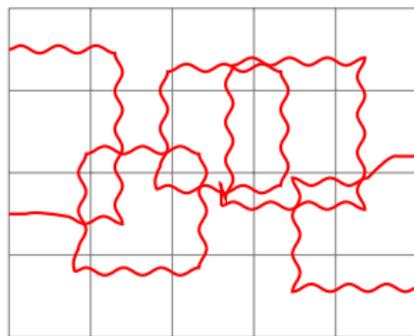
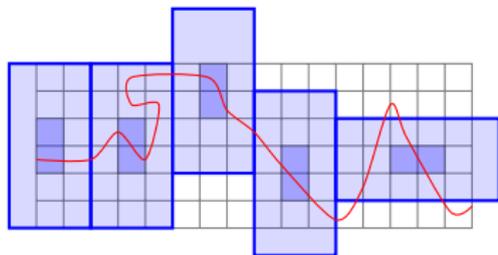
Percolation arguments

- ▶ Left-tails and right-tails of the crossings, as well as relationships between crossing distances at different scales, can be obtained through “percolation arguments”:



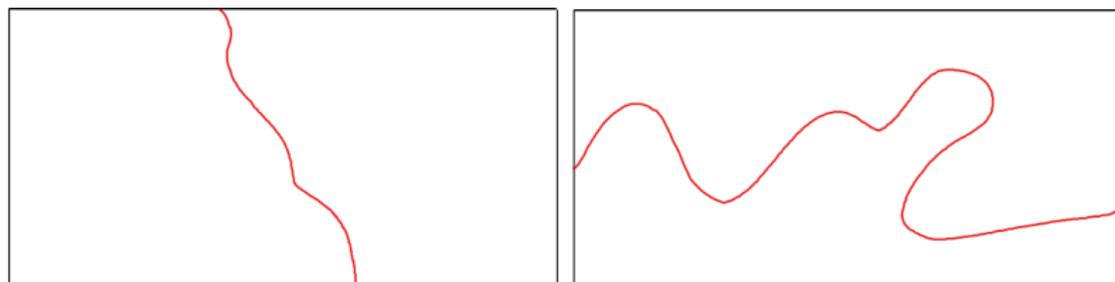
- ▶ A left-right crossing of the large box must include many left-right crossings of smaller boxes.
- ▶ A left-right crossing of the large box can be assembled from left-right circuits around smaller annuli.
- ▶ Similar to strategies used extensively before, most closely in Ding–D. high-temperature result and Dubédat–Falconet result for $*$ -scale invariant fields.

Concentration through percolation



- ▶
- ▶ Problem: the left and right tails aren't in terms of quite the same thing.
 - ▶ Left tail in terms of *small* quantiles of “easy” crossings at smaller scales.
 - ▶ Right tail in terms of *large* quantiles of “hard” crossings at smaller scales.
- ▶ So... need to relate easy and hard crossings, and need to relate small and large (but fixed) quantiles.

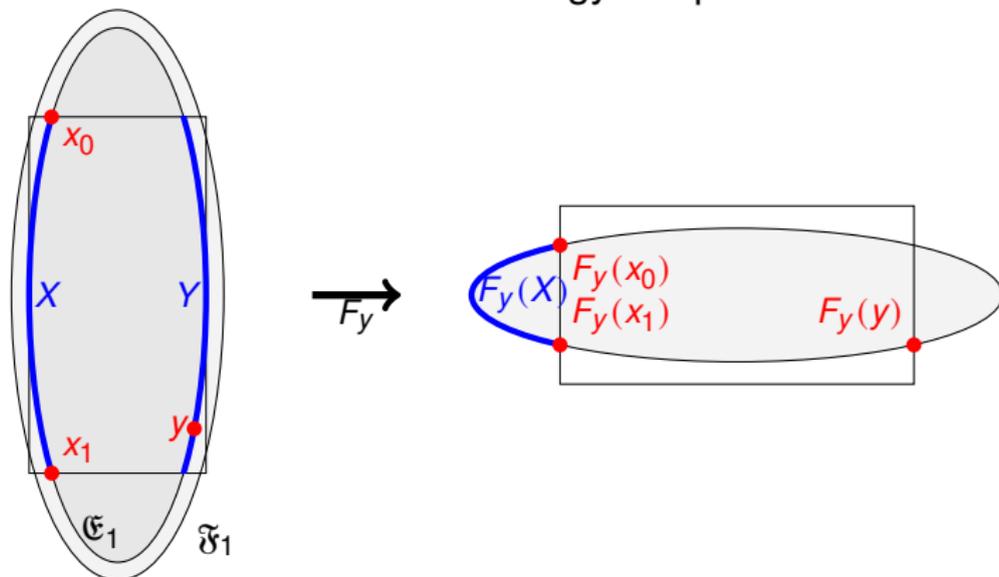
RSW result: relate easy and hard crossing quantiles



- ▶ First results: Russo, Seymour, Welsh for Bernoulli percolation in 1979–'80.
- ▶ Ding–D. high-temperature result for LFPP: adapted intricate inductive RSW result due to V. Tassion (2016).
- ▶ Dubédat–Falconet: introduced very simple RSW proof based on conformal invariance.
 - ▶ We adapt the Dubédat–Falconet proof to Liouville graph distance.

RSW strategy

- ▶ Dubédat–Falconet RSW strategy: map



- ▶ A positive fraction of easy–crossing geodesics can be mapped to hard–crossing geodesics if y is chosen correctly.
- ▶ Uses approximate conformal invariance of LGD (coming from conformal invariance of GFF).

Relating small and large quantiles

- ▶ In order to tie together the percolation arguments and the RSW arguments, we need to relate different quantiles of the crossing distances.
- ▶ To do this, we seek to bound $\text{Var}(\log d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right}))$.
- ▶ Then $p < q$ quantiles of $d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})$ are related by a multiplicative factor of at most

$$\exp \left\{ \left(q^{-1/2} + (1 - q)^{-1/2} \right) \sqrt{\text{Var}(\log d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right}))} \right\}.$$

Efron–Stein argument

- ▶ The field can be subdivided into a sum of “fine” and “coarse” pieces using the “white-noise decomposition.”
- ▶ “Fine” here means concentrated on boxes a factor of K smaller, where K is very large but fixed.
- ▶ Efron–Stein inequality:

$$\text{Var}(\log d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})) \leq \sum_i \mathbf{E} \left(\log \frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{R}, \delta}^{(i)}(\text{left-right})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right)^2,$$

where $\tilde{d}_{\mathfrak{R}^*, \mathfrak{R}, \delta}^{(i)}$ is the metric where piece i has been resampled.

- ▶ So the goal becomes to control the ratio between the resampled and original distances.

Resampling the coarse field

- ▶ The coarse field is smooth, and when it is resampled it has a Lipschitz effect on the crossing distance.
- ▶ Gaussian concentration and a variance bound on the coarse field implies that the variance coming from resampling the coarse field is bounded by $\log K$.

Resampling the fine field

- ▶ When we resample the fine field in a box, the weight coming from most of the geodesic (the part far away from where the field is resampled) doesn't change much.
- ▶ On the other hand, we can bound the change in the weight of the geodesic close to the resampled location by replacing the path with an annular crossing in that region.
- ▶ So the variance coming from resampling the small boxes is like

$$\sum_i \mathbf{E} \left(\frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{C}_i, \delta}^{(i)}(\text{circuit})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right)^2 \leq \left(\sum_i \frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{C}_i, \delta}^{(i)}(\text{circuit})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right) \left(\max_i \frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{C}_i, \delta}^{(i)}(\text{circuit})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right).$$

- ▶ Here, \mathfrak{C}_i is a box containing the area where the field is resampled, and $\tilde{d}_{\mathfrak{R}^*, \mathfrak{C}_i, \delta}^{(i)}(\text{circuit})$ is the weight of an annular circuit around \mathfrak{C}_i .

The effect of subcriticality

- ▶ In the inequality

$$\begin{aligned} \sum_i \mathbf{E} \left(\frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{S}_i, \delta}^{(i)}(\text{circuit})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right)^2 \\ \leq \left(\sum_i \frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{S}_i, \delta}^{(i)}(\text{circuit})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right) \left(\max_i \frac{\tilde{d}_{\mathfrak{R}^*, \mathfrak{S}_i, \delta}^{(i)}(\text{circuit})}{d_{\mathfrak{R}^*, \mathfrak{R}, \delta}(\text{left-right})} \right), \end{aligned}$$

the first factor is on the order of $\exp \{ C\sqrt{V} \}$, where V is the variance bound inductively attained at the smaller scale.

- ▶ That is, by the concentration at the smaller scale, size of first factor does not depend on K , but it does depend on the previously-attained variance bound which is necessary for the concentration at the smaller scale.
- ▶ The second factor is order $\exp \{ C\sqrt{V} \} K^{-c}$, because, as previously discussed, boxes at the smaller scale “look smaller,” even after considering the coarse field.

The bound on the variance of the logarithm

- ▶ Combining fine- and coarse-field contributions, obtain

$$V' := \text{Var}(\log d_{\mathbb{R}^*, \mathbb{R}, \delta}(\text{left-right})) \leq C \left(\log K + e^{C\sqrt{V}} K^{-c} \right).$$

- ▶ In order to close the induction, need a V so that

$$C \left(\log K + e^{C\sqrt{V}} K^{-c} \right) \leq V.$$

- ▶ V can be taken to be a small constant times $\log K$.
- ▶ Actually, there is an additional term in the variance bound, which goes to 0 as $S \rightarrow \infty$ but to ∞ as $K \rightarrow \infty$.
 - ▶ Comes from the inability of the path to be subdivided in the presence of LGD balls with large Euclidean diameter.
 - ▶ Thus the base case of the induction imposes additional challenges: use an *a priori* bound of J. Ding–O. Zeitouni–F. Zhang (preprint, 2018).

Further questions

- ▶ Can the result be extended to other similar discrete approximations? To other models?
 - ▶ What are good techniques for RSW results in the absence of conformal invariance?
- ▶ What are the scaling exponents?
- ▶ Any question you want to ask about the limiting metrics.