Asymptotics of Bernoulli Gibbsian Line Ensembles

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The Gaussian universality class

Let $\{X_i\}$ be a sequence of independent identically distributed random variables with mean μ and variance σ^2 . Let $S_n = X_1 + \cdots + X_n$.

- Law of Large Numbers: $\frac{S_n}{n} \longrightarrow \mu$ as $n \to \infty$ almost surely.
- Central Limit Theorem: $\frac{S_n n\mu}{\sigma\sqrt{n}} \implies \mathcal{N}(0,1)$ as $n \to \infty$.
- **Donsker's Theorem:** For $t \in [0,1]$, let $W^{(n)}(t) = \frac{S_{nt} nt\mu}{\sigma\sqrt{n}}$ if $nt \in \mathbb{N}$, and linearly interpolate. Then $W^{(n)} \in C([0,1])$ and $W^{(n)} \implies W$ as $n \to \infty$, a standard Brownian motion on [0,1].

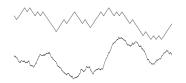
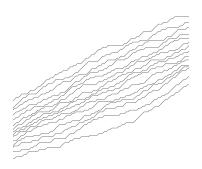


Figure: An example of a random walk and a Brownian motion.

Multiple random walks

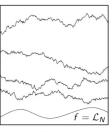
- If $S_{n+1} S_n \in \{0,1\}$, then $\{S_n\}_{n=1}^{\infty}$ is a Bernoulli random walk
- An avoiding Bernoulli line ensemble $\mathfrak{L}=(L_1,\ldots,L_k)$ consists of k avoiding Bernoulli random walks on an interval $[T_0,T_1]$ with random initial and ending points $\mathfrak{L}(T_0),\mathfrak{L}(T_1)$, such that $L_1(s)\geq L_2(s)\geq \cdots \geq L_k(s)$ for $s\in [T_0,T_1]$
- Special case of Bernoulli Gibbsian line ensembles



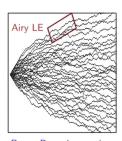
• Question: What does the limit look like as $k \to \infty$?

Airy Line Ensemble

As $k \to \infty$, k avoiding random walks are conjectured to converge to the *Airy line ensemble* \mathcal{A} , and the top curve to the *Airy process* \mathcal{A}_1



Avoiding Brownian bridges



Dyson Brownian motion

- Increasing the number of paths pushes us outside of the Gaussian universality class and into the Kardar-Parisi-Zhang (KPZ) universality class
- Open problem: Show that "generic" random walks with "generic" initial and terminal conditions converge to the Airy line ensemble
- We consider this problem for Bernoulli random walks; the proof is only known if all walks start from 0

Convergence to the Airy Line Ensemble

Two sufficient conditions for convergence in distribution:

- Finite dimensional convergence difficult, requires exact algebraic formulas
- *Tightness* (existence of weak subsequential limits) easier, more qualitative/analytic We focused on tightness, which we prove by controlling the maximum, the minimum, and the modulus of continuity

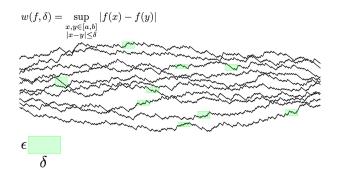


Figure: The modulus of continuity

Our Result

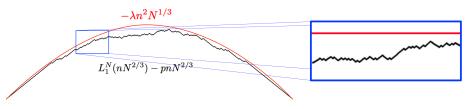
Theorem (DFFSTWZ)

Let $\{\mathfrak{L}^N=(L_1^N,\dots,L_k^N)\}_{N=1}^\infty$ be a sequence of k avoiding Bernoulli random walks. Fix $p\in(0,1)$ and $\lambda>0$, and suppose that for all $n\in\mathbb{Z}$ we have

$$\lim_{N \to \infty} \mathbb{P} \big(L_1^N(nN^{2/3}) - pnN^{2/3} + \lambda n^2 N^{1/3} \le N^{1/3} x \big) = F_{TW}(x).$$

Then $\{\mathfrak{L}^N\}$ is a tight sequence.

- F_{TW} is the *Tracy-Widom distribution* the one-point marginal for the Airy process
- [Dauvergne-Nica-Virág '19] Finite dimensional convergence of all curves implies tightness
- Our result shows that it suffices for the integer time one-point marginals of the top curve to converge to F_{TW}



History of the line ensembles

Arguments in this paper are inspired by

- Brownian Gibbs property for Airy line ensembles [Corwin-Hammond '14] and KPZ line ensemble [Corwin-Hammond '16], which address similar issues for continuous line ensembles
- Transversal fluctuations of the ASEP, stochastic six vertex model, and Hall-Littlewood line ensembles [Corwin-Dimitrov '17], which considers similar questions in a discrete setting

Proving tightness

To show tightness, we want to control:

- **1** Minimum of bottom curve L_{k-1}^N
- **2** Maximum of top curve L_1^N
- **1** Modulus of continuity of each curve L_i^N

We will focus on bounding the minimum:

Lemma 1 (DFFSTWZ)

Fix $r, \epsilon > 0$. Then there exist constants M > 0 and $N_0 \in \mathbb{N}$ such that for all $N \geq N_0$,

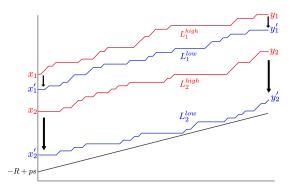
$$\mathbb{P}\Big(\inf_{x \in [-r,r]} \left(L_{k-1}^{N}(xN^{2/3}) - pxN^{2/3} \right) < -MN^{1/3} \Big) < \epsilon.$$



Monotone coupling

Lowering entry and exit data \vec{x}, \vec{y} for the curves \implies curves shift down on whole interval

We proved this by adapting arguments from [Corwin-Hammond '14]



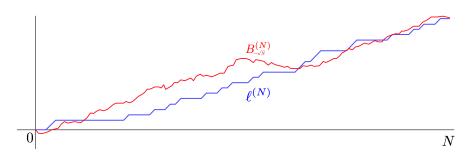
 \bullet \mathfrak{L}^{high} and \mathfrak{L}^{low} are coupled, in particular

$$\mathbb{P}(L_2^{high} < -R) \leq \mathbb{P}(L_2^{low} < -R)$$



Strong coupling

A Bernoulli random walk $\ell^{(N)}$ on [0,N] can be coupled with an "exponentially close" Brownian bridge $B^{(N)}$ with variance $O(\sqrt{N})$ [Dimitrov-Wu '19]



$$\mathbb{P}\Big(\sup_{s\in[0,N]}\left|\ell^{(N)}(s)-B^{(N)}(s)\right|\geq M(\log N)^2+x\Big)< Ke^{-Ax}$$



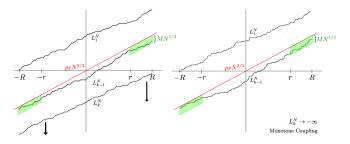
Proving Lemma 1: pinning the bottom curve

Lemma 2 (DFFSTWZ)

For any $r, \epsilon > 0$, there exists R > r and a constant M > 0 so that for large N,

$$\mathbb{P}\Big(\max_{\boldsymbol{x}\in[r,R]}\big(L_{k-1}^N(\boldsymbol{x}N^{2/3})-\Pr^{\boldsymbol{x}N^{2/3}}\big)<-MN^{1/3}\Big)<\epsilon.$$

The same is true of the maximum on [-R, -r].



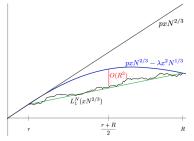
- Use monotone coupling to push L_{k}^{N} to $-\infty$
- Strongly couple L_{k-1}^N with a Brownian bridge: if "pinned" at two points in [r,R] and [-R,-r], it cannot be low on [-r,r] on scale $N^{1/3}$

Proving Lemma 2

Recall our assumption:

$$\mathbb{P}\Big(L_1^N(nN^{2/3}) - pnN^{2/3} + \lambda n^2N^{1/3} \le xN^{1/3}\Big) \underset{N \to \infty}{\longrightarrow} F_{TW}(x)$$

The top curve looks like a parabola with an affine shift on large scales



• Two curves: if L_2^N is low on [r, R], L_1^N looks like a free Brownian bridge

$$\left[-\lambda \left(\frac{R+r}{2}\right)^2\right] - \left[-\lambda \left(\frac{R^2+r^2}{2}\right)\right] = \lambda \frac{R^2+r^2-2rR}{4} = O(R^2)$$

ullet For large R, the top curve would be far from the parabola at the midpoint!

