Asymptotics of Bernoulli Gibbsian Line Ensembles

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July 29, 2020

The Gaussian universality class

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

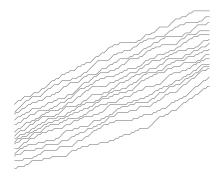
- Law of Large Numbers: $\frac{S_n}{n} \longrightarrow \mu$ as $n \to \infty$ almost surely.
- Central Limit Theorem: $\frac{S_n n\mu}{\sigma\sqrt{n}} \implies \mathcal{N}(0,1)$ as $n \to \infty$.
- **Donsker's Theorem:** For $t \in [0,1]$, let $W^{(n)}(t) = \frac{S_{nt} nt\mu}{\sigma\sqrt{n}}$ if $nt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

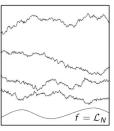
- If $S_{n+1} S_n \in \{0,1\}$, then $\{S_n\}_{n=1}^{\infty}$ is a Bernoulli random walk
- An avoiding Bernoulli line ensemble $\mathfrak{L} = (L_1, \ldots, L_k)$ consists of k avoiding Bernoulli random walks on an interval $[T_0, T_1]$, such that $L_1(s) \geq L_2(s) \geq \cdots \geq L_k(s)$ for $s \in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



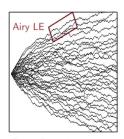
• Question: What does the limit look like as $k \to \infty$?

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 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

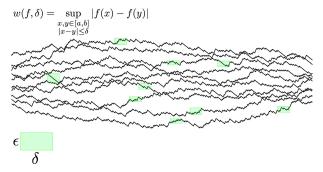


Figure: The Modulus of Continuity

Our Result

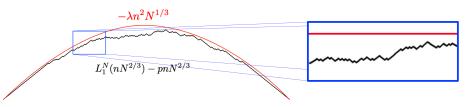
Theorem (DFFSTWZ)

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

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

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

- \bullet 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 '15], 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

Recall that to show tightness, we want to control:

- **1** Minimum of bottom curve L_k^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 (DFFSTWZ)

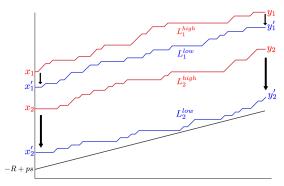
Fix $r, \epsilon > 0$. Then there exist constants M > 0 and $N_0 \in \mathbb{N}$ such that for all $N \geq N_0$,

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



Monotone coupling

Lowering entry and exit data \vec{x}, \vec{y} for the curves \implies curves shift down on whole interval [Corwin-Hammond '14]



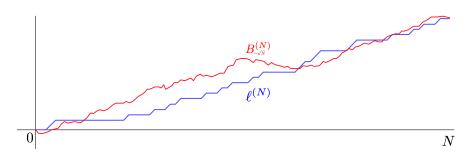
 \bullet \mathfrak{L}^{high} and \mathfrak{L}^{low} are "coupled":

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

• Use this to separate curves at endpoints – then treat as free Bernoulli bridges

Strong coupling

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



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



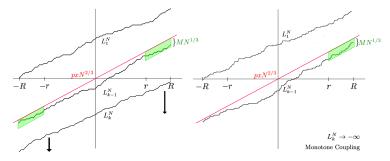
Controlling the minimum: pinning the bottom curve

Lemma (DFFSTWZ)

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

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

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



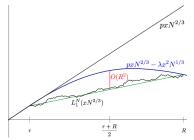
Couple L_k^N with a Brownian bridge: if "pinned" at two points in [r,R] and [-R,-r], it cannot be low on [-r,r] on scale $N^{1/3}$

Proving the pinning lemma

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

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

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