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 $S_n = X_1 + \cdots + X_n$.

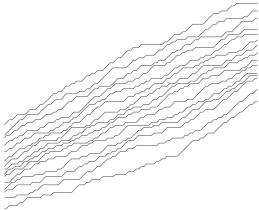
- Law of Large Numbers: $\frac{S_n}{n} \longrightarrow \mu$ as $n \to \infty$ almost surely.
- Central Limit Theorem: $\frac{S_n n\mu}{\sigma\sqrt{n}} \implies \mathcal{N}(0,1)$ as $n \to \infty$.
- **Donsker's Theorem:** For $t \in [0,1]$, let $W^{(n)}(t) = \frac{S_{nt} nt\mu}{\sigma\sqrt{n}}$ if $nt \in \mathbb{N}$, and linearly interpolate. Then $W^{(n)} \in C([0,1])$ and $W^{(n)} \Longrightarrow W$ 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 $s \in [T_0, T_1]$



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

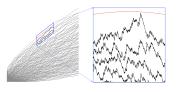


Figure: Trajectories of Dyson Brownian motion

- 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

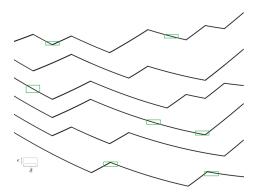


Figure: The Modulus of Continuity

Theorem (FFSTWZD)

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} denotes the Tracy-Widom distribution the one-point marginal for the Airy process
- [Dauvergne-Nica-Virág '19] showed that finite dimensional convergence of all curves implies tightness
- Our result shows that it suffices for the top curve to converge in the f.d. sense



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History of the line ensembles

Arguments in this paper are inspired by

- Brownian Gibbs property for Airy line ensembles [Corwin-Hammond '11] and KPZ line ensemble [Corwin-Hammond '13], 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 consider 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 (FFSTWZD)

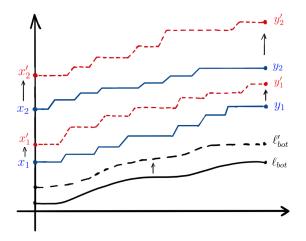
Fix $r, \epsilon > 0$. Then there exist constants R > 0 and $N_0 \in \mathbb{N}$ such that for all $N \ge N_0$,

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

Monotone coupling

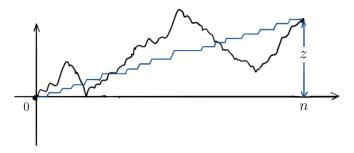
Shifting the endpoints or the bottom curve up(resp., down)

 \Rightarrow The whole line ensemble uniformly goes up(resp., down)



Strong coupling with Brownian bridges

[Dimitrov-Wu 19] A Bernoulli random walk and a Brownian bridge with the same endpoints are "strongly coupled", meaning that they stay close to each other.



Denote L(s) as a Bernoulli random walk from point (0,0) to (n,z), and B(s) is a Brownian bridge with the same endpoints, where $s \in [0,n]$. Then we have

$$\mathbb{P}(\sup_{s\in[0,n]}|L(s)-B(s)|\geq M\log n+x)<\epsilon$$

meaning that the maximal distance between them grows at the speed of $(\log n)^2$ with high probability.

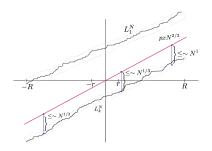
Controlling the minimum: pinning the bottom curve

Lemma (FFSTWZD)

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

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

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



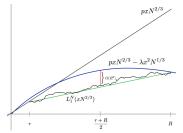
• Couple L_k^N with a Brownian bridge: if "pinned" at two points in [r, R] and [-R, -r], it cannot be low on scale $N^{1/3}$ on [-r, r]

Proving the pinning lemma

Recall our assumption:

$$\mathbb{P}\Big(\textcolor{red}{L_1^N(nN^{2/3})} - pnN^{2/3} + \textcolor{red}{\lambda n^2} N^{1/3} \leq xN^{1/3} \Big) \underset{N \to \infty}{\longrightarrow} F_{\mathcal{T}_W}(x)$$

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

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

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