Ergodicity of the Airy line ensemble

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Abstract

In this paper, we establish the ergodicity of the Airy line ensemble with respect to horizontal shifts. This shows that it is the only candidate for Conjecture 3.2 in [3], regarding the classification of ergodic line ensembles satisfying a certain Brownian Gibbs property after a parabolic shift.

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1 Introduction

The Airy2 process was introduced in [8] and describes fluctuations in random matrices, random surfaces and KPZ growth models. For instance, the Airy2 process describes the scaling limit of the largest eigenvalue in GUE Dyson’s Brownian motion as the number of eigenvalues goes to infinity. For more information and background, see [8, 7, 3, 4, 9] and reference therein.

One illuminating way to view the Airy2 process is as the top line of the Airy line ensemble whose finite dimensional distributions are described by a determinantal process with the extended Airy2 kernel as its correlation kernel. This determinantal process was introduced in [8] and the existence of a continuous version was established in [3]. In the Dyson’s Brownian motion context, the Airy line ensemble describes the limiting measure focusing on the top collection of evolving eigenvalues.

The Airy2 process is stationary with respect to horizontal shifts and Equation (5.15) in [8] showed that it satisfies the strong mixing condition and hence is ergodic. In this paper, we extend this result to the Airy line ensemble. This shows that it is the only candidate for Conjecture 3.2 in [3], regarding the classification of ergodic line ensembles satisfying a certain Brownian Gibbs property after a parabolic shift.

1.1 The Airy line ensemble and the Brownian Gibbs property

In order to define the Airy line ensemble we first introduce the concept of a line ensemble and the Brownian Gibbs line ensembles. We follow the notations of [3].

Definition 1.1. Let Σ be a (possibly infinite) interval of Z, and let Λ be an interval of R. Consider the set X of continuous functions f : Σ × Λ → R endowed with the topology
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of uniform convergence on compact subsets of $\Sigma \times \Lambda$. Let $\mathcal{C}$ denote the sigma-field generated by Borel sets in $X$.

A $\Sigma$-indexed line ensemble $\mathcal{L}$ is a random variable defined on a probability space $(\Omega, B, P)$, taking values in $X$ such that $\mathcal{L}$ is a $(B, \mathcal{C})$-measurable function. Intuitively, $\mathcal{L}$ is a collection of random continuous curves (even though we use the word “line” we are referring to continuous curves), indexed by $\Sigma$, each of which maps $\Lambda$ into $R$. We will often slightly abuse notation and write $\mathcal{L} : \Sigma \times \Lambda \rightarrow R$, even though it is not $\mathcal{L}$ which is such a function, but rather $\mathcal{L}(\omega)$ for each $\omega \in \Omega$. Furthermore, we write $\mathcal{L}_i := (\mathcal{L}(\omega))(i, \cdot)$ for the line indexed by $i \in \Sigma$.

We turn now to formulating the Brownian Gibbs property.

**Definition 1.2.** Let $\{x_1 > \cdots > x_k\}$ and $\{y_1 > \cdots > y_k\}$ be two sets of real numbers. Let $a, b \in \mathbb{R}$ satisfy $a < b$, and let $f, g : [a, b] \rightarrow \mathbb{R}^*$ (where $\mathbb{R}^* = \mathbb{R} \cup \{-\infty, +\infty\}$) be two given continuous functions that satisfy $f(r) > g(r)$ for all $r \in [a, b]$ as well as the boundary conditions $f(a) > x_1$, $f(b) > y_1$, and $g(a) < x_k$, $g(b) < y_k$.

The $(f, g)$-avoiding Brownian line ensemble on the interval $[a, b]$ with entrance data $(x_1, \ldots, x_k)$ and exit data $(y_1, \ldots, y_k)$ is a line ensemble $Q$ with $\Sigma = \{1, \ldots, k\}$, $\Lambda = [a, b]$ and with the law of $Q$ equal to the law of $k$ independent Brownian bridges with diffusion coefficient $1$ ($B_i : [a, b] \rightarrow \mathbb{R}$) $i = 1, \ldots, k$ from $B_i(a) = x_i$ to $B_i(b) = y_i$, conditioned on the event that $f(r) > B_i(r) > B_j(r) > \cdots > B_k(r) > g(r)$ for all $r \in [a, b]$. Note that any such line ensemble $Q$ is necessarily non-intersecting.

Now fix an interval $\Sigma \subseteq \mathbb{Z}$ and $\Lambda \subseteq \mathbb{R}$ and let $K = \{k_1, k_1 + 1, \ldots, k_2 - 1, k_2\} \subseteq \Sigma$ and $a, b \in \Lambda$, with $a < b$. Set $f = \mathcal{L}_{k_1-1}$ and $g = \mathcal{L}_{k_2+1}$ with the convention that if $k_1 - 1 \notin \Sigma$ then $f \equiv +\infty$ and likewise if $k_2 + 1 \notin \Sigma$ then $g \equiv -\infty$. Write $D_{K,a,b} = K \times (a, b)$ and $D_{K,a,b}^\text{ext} = (\Sigma \times \Lambda) \setminus D_{K,a,b}$. A $\Sigma$- indexed line ensemble $\mathcal{L} : \Sigma \times \Lambda \rightarrow \mathbb{R}$ is said to have the Brownian Gibbs property if

\[
\text{Law}\left(\mathcal{L}\big|_{D_{K,a,b}^\text{ext}} \text{ conditional on } \mathcal{L}\big|_{D_{K,a,b}}\right) = \text{Law}(Q),
\]

where $Q_i = \tilde{Q}_{i-k_1+1}$ and $\tilde{Q}$ is the $(f, g)$-avoiding Brownian line ensemble on $[a, b]$ with entrance data $(\mathcal{L}_{k_1}(s), \ldots, \mathcal{L}_{k_2}(s))$ and exit data $(\mathcal{L}_{k_2}(t), \ldots, \mathcal{L}_{k_1}(t))$. Note that $\tilde{Q}$ is introduced because, by definition, any such $(f, g)$-avoiding Brownian line ensemble is indexed from $1$ to $k_2 - k_1 + 1$, but we want $Q$ to be indexed from $k_1$ to $k_2$.

**Definition 1.3.** The Airy line ensemble is a $\mathbb{N} \times \mathbb{R}$ indexed line ensemble which we denote by $\mathcal{A}$. Given $I \subset \mathbb{R}$, let $\mathcal{A}(I) = \{\mathcal{A}(i, t) | i \in \mathbb{N}, t \in I\}$. The defining property of $\mathcal{A}$ is the following: for all $I = \{l_1, \ldots, l_n\}$, and $n \geq 1$, as a point process on $I \times \mathbb{R}$, $\mathcal{A}(I)$ is a determinantal process whose kernel is the extended Airy$_2$ kernel $K_2^\text{ext}$ such that

\[
K_2^\text{ext}(s, x; t, y) = \begin{cases} 
\int_0^\infty d\lambda e^{-\lambda(s-t)} \text{Ai}(x + \lambda) \text{Ai}(y + \lambda) & \text{if } s \geq t, \\
- \int_{-\infty}^0 d\lambda e^{-\lambda(s-t)} \text{Ai}(x + \lambda) \text{Ai}(y + \lambda) & \text{if } s < t,
\end{cases}
\]

(1.1)

where $\text{Ai}(\cdot)$ is the Airy function.

It is not a priori clear that there exists a continuous line ensemble satisfying the correlation functions in Definition 1.3. [3] showed that the Airy line ensemble exists and satisfies the Brownian Gibbs property after a parabolic shift. To be precise, we have

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**Theorem 1.4** (Theorem 3.1 in [3]). There is a unique $\mathbb{N} \times \mathbb{R}$ indexed line ensemble $A$ satisfying Definition 1.3. Moreover, the line ensemble $L$ given by

$$L_i(x) = \frac{1}{\sqrt{2}}(A_i(x) - x^2)$$

(1.2)

satisfies the Brownian Gibbs property.

**Definition 1.5.** The top line of the Airy line ensemble is called Airy$_2$ process.

### 1.2 Main result and motivation

From the definition of the extended Airy$_2$ kernel (1.1), the Airy line ensemble and its marginal, the Airy$_2$ process (i.e. $A_1$), are invariant under horizontal shift in the $x$-coordinate (i.e. are stationary). It is proved in [8] that the Airy$_2$ process is ergodic, and in fact, satisfies the strong mixing condition. (Technical details regarding ergodicity are provided in Section 1.3). In this paper, we show that:

**Theorem 1.6.** The Airy line ensemble is ergodic with respect to horizontal shifts.

Theorem 1.6 can be seen as a multi-line extension of the ergodicity of the Airy$_2$ process. Besides its independent interest, one motivation for this result is our desire to classify stationary ergodic line ensembles which display the Brownian Gibbs property after the parabolic shift given by (1.2).

According to Theorem 1.4, $L_i(x) = \frac{1}{\sqrt{2}}(A_i(x) - x^2)$ satisfies the Brownian Gibbs property. In [3], the authors formulated a conjecture which was originally suggested by Scott Sheffield:

**Conjecture 1.7** (Conjecture 3.2 of [3]). Let $\Theta = \{\theta|s \in \mathbb{R}\}$ denote the horizontal shift group of $\mathbb{N} \times \mathbb{R}$-indexed line ensembles. More precisely, given a line ensemble $L$, let

$$\theta_s L_i(x) = L_i(x + s) \forall i \in \mathbb{N}, x \in \mathbb{R}.$$  

We say that an $\mathbb{N} \times \mathbb{R}$-indexed line ensemble $L$ is horizontal shift-invariant if $\theta_s L$ is equal in distribution to $L$ for each $s \in \mathbb{R}$. Let $\%_\Theta$ be the set of Brownian Gibbs line ensemble measure $L$’s such that $2^{1/2}L_i(x) + x^2$ is horizontal shift-invariant. Then as a convex set, the extremal points of $\%_\Theta$ consists of $\{L^c|c \in \mathbb{R}\}$ where

$$L^c_i(x) = \frac{1}{\sqrt{2}}(A_i(x) - x^2) + c$$

(1.3)

and $A$ is the Airy line ensemble.

Beyond its intrinsic interest, this conjecture is worth investigating in light of its possible use as an invariance principle for deriving convergence of systems to the Airy line ensemble. As such, the characterization could serve as a route to universality results. For example, Section 2.3.3 in [2] suggests a possible route to prove that the KPZ line ensemble converges to the Airy line ensemble minus a parabola (and hence the narrow-wedge initial data KPZ equation converges to the Airy$_2$ process minus a parabola) based on the above conjecture.

In Section 3 we will prove that Theorem 1.6 implies:

**Theorem 1.8.** $L^c$ defined in Conjecture 1.7 are extremal Brownian Gibbs line ensembles.

Theorem 1.8 reduces Conjecture 1.7 to the following:
Conjecture 1.9. Given $c \in \mathbb{R}$, there is a unique Brownian Gibbs line ensemble $L \in \mathcal{G}_\Theta$ such that
\[ E[L_1(x) + 2^{-1/2}x^2] = c \]
for all $x \in \mathbb{R}$.

When Conjecture 1.7 was formulated in [3], it was not shown that $L^c$ defined in (1.3) are extremal. Therefore even if the uniqueness in Conjecture 1.9 was established, it would not necessarily follow that extremal points in $\mathcal{G}_\Theta$ were related to the Airy line ensemble. Theorem 1.8 rules out the possibility that the Airy line ensemble is a non-trivial convex combination of extremal points in $\mathcal{G}_\Theta$, thus reducing Conjecture 1.7 to Conjecture 1.9.

1.3 Ergodicity of the Airy line ensemble

We recall some basic facts in ergodic theory in the context of $\mathbb{N} \times \mathbb{R}$-indexed line ensemble. As a matter of convention, all line ensembles are assumed to be indexed by $\mathbb{N} \times \mathbb{R}$ unless otherwise noted.

Recall $\Theta = \{\theta_s | s \in \mathbb{R}\}$ is the horizontal shift group of $\mathbb{N} \times \mathbb{R}$-indexed line ensembles.

Definition 1.10. Suppose $L$ is a horizontal shift-invariant line ensemble on the probability space $(\Omega, \mathcal{C}, P)$. We say $A \in \mathcal{C}$ is shift-invariant if $\theta_s A := \{\theta_s \omega | \omega \in A\} = A$ for all $s \in \mathbb{R}$. Then $L$ is ergodic if for all shift-invariant $A, P[A] = 0$ or 1.

It is well known that ergodicity follows from the strong mixing condition.

Definition 1.11. Suppose $L$ is a horizontal shift-invariant line ensemble on the probability space $(\Omega, \mathcal{C}, P)$. $L$ is said to satisfy the strong mixing condition if for all $A, B \in \mathcal{C}$,
\[ \lim_{T \to \infty} P[\theta_T A, B] = P[A]P[B]. \]

Proposition 1.12. If a horizontal shift-invariant line ensemble $L$ satisfies the strong mixing condition, it is ergodic.

Proof. Suppose $A$ is shift-invariant. Then $P[\theta_T A, A] = P[A]$. Let $T$ tend to infinity, by the strong mixing condition, $P[A] = P[A]P[A]$, which means $P[A] = 0$ or 1.

To prove Theorem 1.6, we actually prove a stronger result:

Proposition 1.13. The Airy line ensemble satisfies the strong mixing condition given in Definition 1.11.

Now we consider the Airy line ensemble $\mathcal{A}$. Fix $m \in \mathbb{N}$ and $t_1 < t_2 < \cdots < t_m$. The Airy line ensemble restricted to $\mathbb{N} \times \{t_1, t_2, \cdots, t_m\}$ is a point process on $\{t_1, t_2, \cdots, t_m\} \times \mathbb{R}$. For $1 \leq i \leq m$ and $k_i \in \mathbb{N}$, suppose $I = \{I_j^i | 1 \leq j \leq k_i\}$ is a collection of intervals on $\mathbb{N} \times \mathbb{R}$ satisfying
\[ \{I_j^i\}_{1 \leq j \leq k_i} \text{ are disjoint intervals on } \{t_i\} \times \mathbb{R} \text{ for all } i, \text{ and} \]
\[ \inf \left\{ x : \exists 1 \leq i \leq m, 1 \leq j \leq k_i \text{ such that } (t_i, x) \in I_j^i \right\} > -\infty. \] (1.4)

Let $N_{I_i^i}$ denote the number of particles in $I_j^i$ and let
\[ G(t_1, t_2, \cdots, t_m, I) = \left\{ A \in \mathcal{C} | A \in \sigma \left( \{N_{I_i^i} | 1 \leq i \leq m, 1 \leq j \leq k_i\} \right) \right\}. \]

We prove Proposition 1.13 by showing that
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**Lemma 1.14.** Fix $m \in \mathbb{N}$, $t_1 < t_2 < \cdots < t_m$ and $\mathcal{I}$, for $A, B \in \mathcal{G}(t_1, t_2, \cdots, t_m, \mathcal{I})$,

$$\lim_{T \to \infty} \mathbb{P}[\theta_T(A), B] = \mathbb{P}[A|\mathbb{P}[B]. \tag{1.5}$$

**Proof of Proposition 1.13 based on Lemma 1.14.** Recall a result in measure theory (see, for example,[6]).

**Theorem 1.15.** Let $(X, \mathcal{C}, \mathbb{P})$ be a probability space. Let $B \subset \mathcal{C}$ an algebra generating $C$. Then for all $A \in \text{Cand } \varepsilon > 0$, we can find $A' \in B$ such that $\mu(A\Delta A') \leq \varepsilon$. Here $A\Delta A' = A'\setminus A \setminus A'$ is the symmetric difference of $A$ and $A'$.

Let $\mathcal{F}$ be the $\sigma$-algebra generated by the Airy line ensemble and $\mathcal{F}'$ be the union of all $\mathcal{G}(t_1, t_2, \cdots, t_m, \mathcal{I})$ where $(t_1, t_2, \cdots, t_m)$ varies over all finite collections of real numbers and $\mathcal{I}$ varies over all finite collections of intervals satisfying (1.4). Since

$$\mathcal{G}(t_1, t_2, \cdots, t_m, \mathcal{I}) \cup \mathcal{G}(t_1', t_2', \cdots, t_n', \mathcal{I}') \subset \mathcal{G}(t_1, t_2, \cdots, t_m, t_1', t_2', \cdots, t_n', \mathcal{I} \cup \mathcal{I}')$$

for all choices of $(t_1, t_2, \cdots, t_m, \mathcal{I})$ and $(t_1', t_2', \cdots, t_n', \mathcal{I}')$, $\mathcal{F}'$ is an algebra. Furthermore, if we restrict the Airy line ensemble to $\mathbb{N} \times \mathbb{Q}$, then $A_i(r)$ is measurable with respect to the $\sigma$-algebra generated by $\mathcal{F}'$. By the continuity of the Airy line ensemble, $\mathcal{F}$ equals to the $\sigma$-algebra generated by $\mathcal{F}'$.

Therefore for $A, B \in \mathcal{F}$ and $\varepsilon > 0$, there exist $A', B' \in \mathcal{F}'$ such that $\mathbb{P}[A\Delta A'] < \varepsilon$, $\mathbb{P}[B\Delta B'] < \varepsilon$. By Lemma 1.14,

$$\lim_{T \to \infty} \mathbb{P}[\theta_T(A'), B'] = \mathbb{P}[A'|\mathbb{P}[B]].$$

Since $\varepsilon$ can be arbitrarily small,

$$\lim_{T \to \infty} \mathbb{P}[\theta_T(A), B] = \mathbb{P}[A|\mathbb{P}[B]. \tag{1.5}$$

\[\square\]

### 1.4 Strategy and outline

Now we sketch our strategy for proving our main results, Theorem 1.6 and 1.8. From the argument above, Lemma 1.14 implies Proposition 1.13, which further implies Theorem 1.6. Thus Theorem 1.6 boils down to proving Lemma 1.14, which is the content of Section 2.

Since $N(t)$ are discrete random variables, their joint distribution is governed by the moment generating function. By a standard fact in determinantal point process (Lemma 2.1), the generating function can be expressed as a Fredholm determinant of $K_{2}^{\text{ext}}$ on $L^2((t_1, t_2, \cdots, t_m) \times \mathbb{R})$.

Since there is a time shift $T$ in Lemma 1.14, we need to consider $2m$ time moments. The trace class operator involved in the left hand side of (1.5) can be regarded as a $2m$ by $2m$ operator valued matrix. Based on the known estimates of Airy$_2$ kernel, one can show that as a $2 \times 2$ block matrix of block size $m$, the “off diagonal” terms will vanish as $T \to \infty$. This shows the factorization of the generating function (Lemma 2.2). Then using complex analysis of several variables one can extract Lemma 1.14 from Lemma 2.2.

To prove Theorem 1.8, we follow the standard method in [5]. As shown in Chapter 14 of [5], given a translation group and a Gibbs specification, for the set of translation invariant Gibbs measures defined on $S = \mathbb{Z}^d$, extremal points coincide with ergodic ones with respect to the translation group. In our case, the underlying space is $S = \mathbb{R} \times \mathbb{N}$. Thanks to the fact that the Airy line ensemble is a collection of continuous curves [3], the same argument in [5] works for the Airy line ensemble. The details are provided in Section 3.
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2 Strong mixing for $N_{I_i}^j$

2.1 Express the generating function by Fredholm determinant

We first recall a useful formula from determinantal point process. It appears in [1] as formula (2.4).

Lemma 2.1. Suppose $X$ is a determinantal point process on a locally compact space of $X$ with kernel $K$. Let $\phi$ be a function on $X$ such that the kernel $(1 - \phi(x))K(x, y)$ defines a trace class operator $(1 - \phi)K$ in $L^2(X)$. Then

$$E \left[ \prod_{x_i \in X} \phi(x_i) \right] = \det \left( 1 - (1 - \phi)K \right). \tag{2.1}$$

Fix $z_1, z_2, \ldots, z_m \in \mathbb{C}$ such that $|z_i| \leq 1$ for all $i$. Let $I_1, I_2, \ldots, I_m$ be a family of pairwise disjoint subsets of $X$ and $Q$ is a multiplication operator defined by

$$Qf(x) = \sum_{i=1}^m (1 - z_i)1_{x \in I_i}f(x). \tag{2.2}$$

Denote $N_{I_i}(1 \leq i \leq m)$ to be the number of particles in $I_i$ of the random configuration $X$. Specifying $\phi(x) = \sum_{i=1}^m z_i 1_{x \in I_i} + 1_{x \in \cap_{i=1}^m I_i^c}$ in Lemma 2.1 leads to

$$E \left[ \prod_{i=1}^m z_i^{N_{I_i}} \right] = \det(I - QK)_{L^2(X)}. \tag{2.3}$$

2.2 Proof of Lemma 1.14

Let the Airy Hamiltonian be defined as

$$H = -\Delta + x.$$ $H$ has the shifted Airy functions $\Lambda_\lambda(x) = \Lambda(x - \lambda)$ as its generalized eigenfunctions: $H\Lambda_\lambda(x) = \lambda\Lambda_\lambda(x)$. Define the Airy kernel $K_2$ as the projection of $H$ onto its negative generalized eigenspace:

$$K_2(x, y) = \int_0^\infty d\lambda \Lambda(x + \lambda) \Lambda(y + \lambda).$$

Consider $t_1 < t_2 < \cdots < t_n$, $t_i \in \mathbb{R}$. For $1 \leq i \leq n$, suppose $\{I_i^j\}_{1 \leq j \leq k_i}$ are intervals on $\{t_i\} \times \mathbb{R}$ satisfying the condition in (1.4) and

$$M_0 = -\inf \{x | (t_i, x) \in I_i^j \text{ for some } i, 1 \leq j \leq k_i \}. \tag{2.4}$$

Let $N_{I_i^j}$ be the number of particles in the interval $I_i^j$. From Lemma 2.1 and (2.3), for $\{z_i^j | 1 \leq i \leq n, 1 \leq j \leq k_i, |z_i^j| \leq 1 \}$

$$E \left[ \prod_{i=1}^n \prod_{j=1}^{k_i} (z_i^j)^{N_{I_i^j}} \right] = \det(I - QK_2^{\text{ext}})_{L^2((t_1, t_2, \ldots, t_n) \times \mathbb{R})}, \tag{2.5}$$

where $Q$ is a multiplication operator defined as

$$Qf(t, x) = \sum_{1 \leq i \leq n, 1 \leq j \leq k_i} (1 - z_i^j)1_{(t_i, x) \in I_i^j}f(t, x). \tag{2.6}$$
Lemma 2.2. Let
\[ \det \left( I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq n} \right) \]
We write \( \det \left( I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq n} \right) \) to be the Fredholm determinant of \( Q^{\text{ext}}_{K^2} \). For \( 1 \leq m \leq n \), we write \( [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq m} \) as the operator corresponding to the submatrix of \( Q^{\text{ext}}_{K^2} \) which consists of entries indexed by \( \{1, \cdots, m\} \times \{1, \cdots, m\} \) and \( \det(I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq m}) \) as its Fredholm determinant. Let \( \det(I - [Q^{\text{ext}}_{K^2}]_{m+1 \leq i,j \leq n}) \) is defined in the same manner.

According to the convention above,
\[ E \left[ \prod_{i=1}^{n} k_i \prod_{j=1}^{n} (z_{i,j}^{ext})^{N_i^j} \right] = \det \left( I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq n} \right). \tag{2.8} \]

Now we assume \( n = 2m \) is an even number and \( \{t_1 < t_2 < \cdots < t_n\} \) is a fixed set of real numbers. \( \{t_{m+i}, t_{m+2} \cdots, t_n\} \) is a shift of \( \{t_1, t_2, \cdots, t_m\} \) by \( T \). To be precise, \( t_{m+i} = t_i + T \) for all \( 1 \leq i \leq m \).

Similar to (2.8), we have
\[ E \left[ \prod_{i=1}^{n} k_i \prod_{j=1}^{n} (z_{i,j}^{ext})^{N_i^j} \right] = \det \left( I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq m} \right), \]
\[ E \left[ \prod_{i=m+1}^{n} \prod_{j=1}^{N_i^j} (z_{i,j}^{ext})^{N_i^j} \right] = \det \left( I - [Q^{\text{ext}}_{K^2}]_{m+1 \leq i,j \leq n} \right). \tag{2.9} \]

Lemma 2.2. Let
\[ R(z,T) = E \left[ \prod_{i=1}^{n} k_i \prod_{j=1}^{n} (z_{i,j}^{ext})^{N_i^j} \right] - E \left[ \prod_{i=1}^{m} k_i \prod_{j=1}^{m} (z_{i,j}^{ext})^{N_i^j} \right] E \left[ \prod_{i=m+1}^{n} \prod_{j=1}^{N_i^j} (z_{i,j}^{ext})^{N_i^j} \right] \]
\[ = \det \left( I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq n} \right) - \det \left( I - [Q^{\text{ext}}_{K^2}]_{1 \leq i,j \leq m} \right) \det \left( I - [Q^{\text{ext}}_{K^2}]_{m+1 \leq i,j \leq n} \right). \tag{2.10} \]
then \( \lim_{T \to \infty} R(z,T) = 0 \) for all \( z \).

We postpone the proof of Lemma 2.2 to Section 2.3. Now we prove Lemma 1.14 based on it. Let \( N_i^j \in \mathbb{N} \) for all \( 1 \leq i \leq n, 1 \leq j \leq k_i \). Define the partial differential operators
\[ \partial_{i,m+1, n} = \frac{\partial^{\sum_{j=1}^{n} N_i^j}}{\prod_{j=1}^{n} \prod_{i=1}^{n} \partial (z_{i,j}^{ext})^{N_i^j}}. \tag{2.11} \]

Then by Lemma 2.2
\[ \partial_{i,m+1, n} E \left[ \prod_{i=1}^{n} \prod_{j=1}^{n} (z_{i,j}^{ext})^{N_i^j} \right] = \partial_{i,m+1, n} E \left[ \prod_{i=m+1}^{n} \prod_{j=1}^{N_i^j} (z_{i,j}^{ext})^{N_i^j} \right] + \partial_{i,n} R(z,T). \tag{2.12} \]
R(z,T) is analytic function bounded by 2 when |z| < 1. By Montel’s theorem [10, Theorem 1.4.31], every subsequence of R(z,T) has a further subsequence which converges locally uniformly. This implies that \( \lim_{T \to \infty} R(z,T) = 0 \) locally uniformly in |z| < 1.

By the Cauchy inequalities [10, Theorem 1.3.3], the magnitude of partial derivatives of an analytic function at a certain point is controlled by the magnitude of the function around that point. Therefore \( \partial_{1,n} R(z,T) \) converges uniformly for |z| < 1. In particular, \( \lim_{T \to \infty} \partial_{1,n} R(0, T) = 0 \).

Let
\[
A' = \{ N_{i}^{j} : \text{all } 1 \leq i \leq m, 1 \leq j \leq k_i \}, \\
B' = \{ N_{i+m}^{j} : \text{all } 1 \leq i \leq m, 1 \leq j \leq k_i \}.
\]

Taking \( z = 0 \) in (2.12) we have
\[
\mathcal{P}[A', \theta_{T} B'] = \mathcal{P}[A'] \mathcal{P}[\theta_{T} B'] + o(1) = \mathcal{P}[A'] \mathcal{P}[B'] + o(1) \quad \text{as } T \to \infty,
\]
where the second equality is due to the horizontal shift invariance of the Airy line ensemble.

Therefore for all \( A, B \in \mathcal{G}(t_1, \ldots, t_m, T) \),
\[
\lim_{T \to \infty} \mathcal{P}[A, \theta_{T} B] = \mathcal{P}[A] \mathcal{P}[B].
\]

This finishes the proof of Lemma 1.14.

2.3 Proof of Lemma 2.2

The proof of Lemma 2.2 uses the following fact.

**Lemma 2.3.** Let \( P_a \) be the multiplication operator of \( 1_{x \geq a} \) and \( y > 0 \). Then \( P_a K_2, P_a e^{-yH}(I - K_2) \) and \( P_a e^{yH} K_2 \) are trace class operators. Moreover,
\[
\lim_{y \to \infty} \| P_a e^{-yH}(I - K_2) \|_1 = \lim_{y \to \infty} \| P_a e^{yH} K_2 \|_1 = 0.
\]

**Proof.** Let \( \varphi(x) = (1 + x^2)^{1/2} \) and define the multiplication operator \( Mf(x) = \varphi(x)f(x) \).

In the proof of Proposition 3.2 in [4], it was shown that
\[
\| P_a e^{-yH} M \|_2 \leq C.
\]
where \( \| \cdot \|_2 \) is the Hilbert-Schmidt norm. Actually by taking \( n = 2 \) in the proof of Proposition 3.2 in [4], we have
\[
\| P_a e^{-yH} M \|_2 = \left\| e^{-yH} - (I - P_a)e^{-yH} \right\|_2 \leq C.
\]

On the other hand
\[
\| M^{-1} e^{yH} K_2 \|_2 = \int_{R^2} dx \, dz \int_{(-\infty,0]^2} d\lambda \, d\bar{\lambda} \varphi(x)^{-2} e^{i(\lambda + \bar{\lambda})y} \operatorname{Ai}(x - \lambda) \operatorname{Ai}(z - \bar{\lambda})
\]
\[
= \int_{-\infty}^{\infty} dx \int_{-\infty}^{0} d\lambda \varphi(x)^{-2} e^{2\lambda y} \operatorname{Ai}(x - \lambda)^2
\]
\[
\leq \frac{c}{2y} \| \varphi^{-1} \|_2^2.
\]
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where \( c = \max_{x \in \mathbb{R}} \text{Ai}(x)^2 < \infty \).

Similarly,

\[
\| M^{-1} e^{-yH} (I - K) \|^2_2 \leq \frac{c}{2y} \| \varphi^{-1} \|^2_2.
\]

(2.18)

Since \( \|AB\|_1 \leq \|A\|_2 \|B\|_2 \),

\[
\| P_{a}K \|_1 \leq \| P_{a} e^{yH} M \|_2 \| M^{-1} e^{-yH} K \|_2 < \infty.
\]

Fix \( b > 0 \), there exists a constant \( C_b \) such that

\[
\| P_{a} e^{yH} K \|_1 \leq \| P_{a} e^{-bH} M \|_2 \| M^{-1} e^{-(y-b)H} K \|_2 \leq \frac{C_b}{y},
\]

\[
\| P_{a} e^{-yH} (I - K) \|_1 \leq \| P_{a} e^{-bH} M \|_2 \| M^{-1} e^{-(y-b)H} (I - K) \|_2 \leq \frac{C_b}{y}.
\]

This finishes the proof.

To prove Lemma 2.2, we come back to (2.7).

\[
R(z, T) = \det \left( I - [Q_{K_2}^{\text{ext}}]_{i \leq j \leq n} \right) - \det \left( I - [Q_{K_2}^{\text{ext}}]_{i \leq j \leq m} \right) \det \left( I - [Q_{K_1}^{\text{ext}}]_{m+1 \leq i,j \leq n} \right).
\]

(2.19)

By condition (1.4), we can always replace \( Q_{i_2} \) by \( Q_{i_1} P_{a} \) for \( a < -M_0 \). Therefore from Lemma 2.3,

\[
\lim_{T \to \infty} \| Q_{i_1} e^{t_1-t_2} H K_2 \|_1 = 0, \quad 1 \leq j \leq m < i \leq n,
\]

\[
\lim_{T \to \infty} \| Q_{i_1} e^{t_1-t_2} H (K_2 - I) \|_1 = 0, \quad 1 \leq i \leq m < j \leq n.
\]

Since Fredholm determinant is continuous respect to trace norm, \( \lim_{T \to \infty} R(z, T) = 0. \)

3 Extremal Gibbs measure and ergodicity

3.1 The Gibbs property of the Airy line ensemble

We first adjust the notations to make it consistent with those in [5].

Let \( S = \mathbb{N} \times \mathbb{R} \) denote the parameter set and \( (\Omega, \mathcal{F}, P) \) denote the probability space of the Airy line ensemble. Canonically, we choose \( \Omega \) to be the set of continuous functions from \( S \) to \( \mathbb{R} \) and \( \mathcal{F} \) be the Borel \( \sigma \)-algebra with respect to the locally uniformly convergence topology. Let \( \mathcal{L} \) be the set of all allowable finite non-empty subsets of \( S \). Here allowable sets are those of the form \( \{k_1, \ldots, k_2\} \times (a, b) \), where \( k_1 \) runs through \( \mathcal{L} \) and \( \mathcal{F}_{S \setminus \Lambda} \) be the \( \sigma \)-algebra of the Airy line ensemble restricted to \( S \setminus \Lambda \).

From Theorem 1.4, the Airy line ensemble satisfies certain Gibbs property, which we formulate now.

Let \( \{x_1 > \cdots > x_k\} \) and \( \{y_1 > \cdots > y_k\} \) be two sets of real numbers. Let \( a, b \in \mathbb{R} \) satisfy \( a < b \), and let \( f, g : [a, b] \to \mathbb{R}^* \) (where \( \mathbb{R}^* = \mathbb{R} \cup \{ -\infty, +\infty \} \) ) be two given continuous functions that satisfy \( f(r) > g(r) \) for all \( r \in [a, b] \) as well as the boundary conditions \( f(a) > x_1, f(b) > y_1 \) and \( g(a) < x_k, g(b) < y_k \). The shifted \( (f, g) \)-avoiding Brownian line ensemble on the interval \( [a, b] \) with entrance data \( \{x_1, \ldots, x_k\} \) and exit data \( \{y_1, \ldots, y_k\} \) is a line ensemble \( \mathcal{L} \) such that \( 2^{-1/2}(\mathcal{L} - x^2) \) is a \( (2^{-1/2}(f(x^2)), 2^{-1/2}(g(x^2))) \)-avoiding Brownian line ensemble on the interval \( [a, b] \) with entrance data \( \{2^{-1/2}(x_1 - a^2), \ldots, 2^{-1/2}(x_k - a^2)\} \) and exit data \( \{2^{-1/2}(y_1 - b^2), \ldots, 2^{-1/2}(y_k - b^2)\} \).
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**Definition 3.1.** Suppose $\Lambda = \{k_1, \cdots, k_2\} \times (a, b)$ and $\gamma_{\Lambda}$ is a probability kernel from $(\Omega, \mathcal{F})$ to $(\Omega, \mathcal{F})$ defined as follows:

For $\omega \in \Omega$, $\gamma_{\Lambda}(\cdot | \omega)$ coincides with $\omega$ outside $\Lambda$; in $\Lambda$, $\gamma_{\Lambda}(\cdot | \omega)$ is the law of the shifted $(\omega(k_2, a), \omega(k_2, b))$-avoiding Brownian line ensemble on $[a, b]$ with entrance data $(\omega(k_1, a), \omega(k_2, a))$ and exit data $(\omega(k_1, b), \omega(k_2, b))$. Here we make the convention that $\omega(0, x) = \infty$.

$\gamma = (\gamma_{\Lambda})_{\Lambda \in \mathcal{L}}$ is the family of Gibbs specifications that describes the Gibbs property of the Airy line ensemble. Since the Airy line ensemble is stationary, $\gamma$ is also horizontal shift-invariant, which means

$$\gamma_{\Lambda + T}(\theta_T A | \theta_T \omega) = \gamma_{\Lambda}(A | \omega), \quad (\Lambda \in \mathcal{L}, T \in \mathbb{R}, \omega \in \Omega). \quad (3.1)$$

With these notations, Theorem 1.8 can be formulated in this way.

**Theorem 3.2.** Let $\mathcal{G}_0(\gamma)$ be the simplex of all probability measures $\mu$ on $(\Omega, \mathcal{F})$ such that

$$\mu(\theta_T A) = \mu(A) \quad \text{and} \quad \mu(A | \mathcal{F}_{\Lambda}) = \gamma(A | \cdot) \quad \mu \text{ a.s. for all } A \in \mathcal{F}, \Lambda \in \mathcal{L} \text{ and } T \in \mathbb{R}. \quad (3.2)$$

Then the Airy line ensemble is an extreme point of $\mathcal{G}_0(\gamma)$.

**3.2 Proof of Theorem 3.2**

Theorem 3.2 follows from a standard result in ergodic theory which is proved as Corollary 7.4 in [5].

**Lemma 3.3.** Let $(\Omega, \mathcal{F})$ be a measurable space, $\Pi$ a non-empty set of probability kernels from $\mathcal{F}$ to $\mathcal{F}$, and

$$\mathcal{P}_\Pi = \{ \mu \in \mathcal{P}(\Omega, \mathcal{F}) : \mu \pi = \mu \text{ for all } \pi \in \Pi \}, \quad (3.3)$$

be the convex set of all $\Pi$-invariant probability measures on $(\Omega, \mathcal{F})$. Let $\mu \in \mathcal{P}_\Pi$ be given and $\mathcal{I}_\Pi(\mu) = \cap_{\pi \in \Pi} \mathcal{I}_\pi(\mu)$ where $\mathcal{I}_\pi(\mu) = \{ A \in \mathcal{F} : \pi(A | \cdot) = 1_A \mu - \text{a.s.} \}$. Then $\mu$ is extreme if and only if $\mu$ is trivial on $\mathcal{I}_\Pi(\mu)$.

Define a family $\hat{\Theta} = \{ \hat{\theta}_t : t \in \mathbb{R} \}$ of probability kernels $\hat{\theta}_t$ from $\mathcal{F}$ to $\mathcal{F}$ by

$$\hat{\theta}_t(A | \omega) = 1_A(\theta_t \omega) \quad (t \in \mathbb{R}, A \in \mathcal{F}, \omega \in \Omega).$$

Using the notation of Lemma 3.3, $\mathcal{P}_\hat{\Theta}$ is the set of all horizontal shift-invariant measures.

By definition, a probability measure $\mu$ belongs to $\mathcal{G}_0(\gamma)$ if and only if $\mu$ is preserved by all probability kernels in

$$\Pi = \{ \gamma_{\Lambda} : \gamma \in \mathcal{L} \} \cup \mathcal{P}_\hat{\Theta}.$$

Therefore $\mu$ is an extreme in $\mathcal{G}_0(\gamma)$ if and only if $\mu$ is trivial on $\mathcal{I}_\Pi(\mu)$.

We claim that $\mathbb{P}$ (the probability measure for the Airy line ensemble) is trivial on $\cap_{\pi \in \Pi} \mathcal{I}_\pi(\mu)$, which is sufficient to prove Theorem 3.2.

To show the triviality, for given $A \in \cap_{\pi \in \Pi} \mathcal{I}_\pi(\mu)$, let $B = \cup_{t \in Q} \theta_t A$. Then $\theta_t B = B$ for all $t \in Q$ and $\mathbb{P}[A \Delta B] \leq \sum_{t \in Q} \mathbb{P}[A \Delta \theta_t A] = 0$. Therefore $\mathbb{P}[A] = \mathbb{P}[B]$.

As in the proof of Proposition 1.12, $\mathbb{P}[B] = \lim_{n \to \infty} \mathbb{P}[\theta_n B, B] = \mathbb{P}[B]^2$, which means $\mathbb{P}[A] = \mathbb{P}[B] = 0$ or 1.
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References

[1] Alexei Borodin. Determinantal point processes. *The Oxford handbook of random matrix theory*, pages 231–249, 2011. MR-2932631

[2] Ivan Corwin and Alan Hammond. KPZ line ensemble. *arXiv preprint arXiv:1312.2600*, 2013.

[3] Ivan Corwin and Alan Hammond. Brownian Gibbs property for Airy line ensembles. *Inventiones mathematicae*, 195(2):441–508, 2014. MR-3152753

[4] Ivan Corwin, Jeremy Quastel, and Daniel Remenik. Continuum statistics of the Airy\textsubscript{2} process. *Communications in Mathematical Physics*, 317(2):347–362, 2013. MR-3010187

[5] Hans-Otto Georgii. *Gibbs measures and phase transitions*, volume 9. Walter de Gruyter, 2011. MR-2807681

[6] Paul Richard Halmos. *Measure theory*, volume 2. van Nostrand New York, 1950. MR-0033869

[7] Kurt Johansson. Discrete polynuclear growth and determinantal processes. *Communications in Mathematical Physics*, 242(1-2):277–329, 2003. MR-2018275

[8] Michael Prähofer and Herbert Spohn. Scale invariance of the PNG droplet and the Airy process. *Journal of Statistical Physics*, 108(5-6):1071–1106, 2002. MR-1933446

[9] Jeremy Quastel and Daniel Remenik. Airy processes and variational problems. *arXiv preprint arXiv:1301.0750*, 2013.

[10] Volker Scheidemann. *Introduction to complex analysis in several variables*. Springer, 2005. MR-2176976

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