Nonintersecting Brownian Excursions

Craig A. Tracy<br>Harold Widom

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## Outline of Talk

1. Brownian Excursion and the Airy Distribution
2. Nonintersecting Path Models and Extended Kernels
3. Extended Kernel for $n$-Brownian Excursion
4. Distribution of Bottom Curve at One Fixed Time: $P_{V}$
5. Scaling of Bottom Curve to Bessel Process, $\alpha=1 / 2$
6. Expected Areas under Bottom and Top Curves

## Brownian Excursion

A Brownian excursion (BE) $X(\tau)$ is a Brownian path conditioned to remain positive for $0<\tau<1$ with boundary conditions $X(0)=X(1)=0$. BE can be defined by scaling one-dimensional simple random walk conditioned to stay positive and conditioned to start and to end at the origin - a process known as Bernoulli excursion.

Let $B_{\tau}$ denote standard Brownian motion (BM), so that
$\mathbb{P}_{x}\left(B_{\tau} \in d y\right)=P(x, y, \tau) d y=\frac{1}{\sqrt{2 \pi \tau}} e^{-(x-y)^{2} /(2 \tau)} d y, \tau>0, x, y \in \mathbb{R}$,
Let

$$
H_{a}=\inf \left\{s: s>0, B_{s}=a\right\}=\text { hitting time of } a .
$$

Then
$\mathbb{P}_{x}\left(B_{\tau} \in d y, H_{0}>\tau\right)=(P(x, y, \tau)-P(x,-y, \tau)) d y:=P_{-}(x, y, \tau) d y$.

## Airy Distribution

Let $\mathcal{A}$ denote the area under a BE path.
Then

$$
F_{\mathrm{Ai}}(a):=\mathbb{P}(\mathcal{A}<a)
$$

is called the Airy distribution. (Not to be confused with the Airy process!) For example,

$$
\mathbb{E}(\mathcal{A})=\sqrt{\frac{\pi}{8}}, \quad \mathbb{E}\left(\mathcal{A}^{2}\right)=\frac{5}{12}, \ldots
$$

The Airy distribution appears as the limit law for a number of combinatorial problems: path length in trees, total displacement of a random parking sequence, ... (Flajolet \& Louchard)

## Extended Kernels

Given: stationary Markov process with continuous paths, transition probability density $P(x, y, \tau)$, and a family of $n$ nonintersecting paths, $\left\{X_{i}(\tau)\right\}_{i=1}^{n}$, beginning at $a_{1}, \ldots, a_{n}$ at time $\tau=0$ and ending at $b_{1}, \ldots, b_{n}$ at time $\tau=1$.

An extended kernel is a matrix kernel $K(x, y)=\left(K_{k \ell}(x, y)\right)_{k l=1, \ldots, m}$ depending on $0<\tau_{1}<\cdots<\tau_{m}<1$ with the following property:
Given functions $f_{k}$ the expected value of

$$
\prod_{k=1}^{m} \prod_{i=1}^{n}\left(1+f_{k}\left(X_{i}\left(\tau_{k}\right)\right)\right)
$$

is equal to $\operatorname{det}(I+K f)$, where $f$ denotes multiplication by $\operatorname{diag}\left(f_{k}\right)$. In the special case where $f_{k}=-\chi_{J_{k}}$ this is the probability that for $k=1, \ldots, m$ no path passes through the set $J_{k}$ at time $\tau_{k}$.

## Nonintersecting Brownian Bridges and Excursions

Define Extended Hermite kernel:

$$
K_{n}^{\mathrm{GUE}}(x, y ; \hat{\tau})=\left\{\begin{array}{rr}
\sum_{j=0}^{n-1} e^{j \hat{\jmath}} \varphi_{j}(x) \varphi_{j}(y) & \hat{\tau} \geq 0 \\
-\sum_{j=n}^{\infty} e^{j \hat{\jmath}} \varphi_{j}(x) \varphi_{j}(y) & \hat{\tau}<0
\end{array}\right.
$$

- Brownian Bridge Extended Kernel:

$$
\begin{gathered}
\frac{1}{\sqrt{2\left(1-\tau_{k}\right) \tau_{\ell}}} K_{n}^{\mathrm{GUE}}\left(X_{k}, Y_{\ell} ; \hat{\tau}_{k}-\hat{\tau}_{\ell}\right), \\
X_{k}=\frac{x}{\sqrt{2 \tau_{k}\left(1-\tau_{k}\right)}}, \quad Y_{\ell}=\frac{y}{\sqrt{\tau_{\ell}\left(1-\tau_{\ell}\right)}}, \frac{\tau_{k}}{1-\tau_{k}}=\hat{\tau}_{k}
\end{gathered}
$$

- Brownian Excursion Extended Kernel:
$\frac{1}{\sqrt{2\left(1-\tau_{k}\right) \tau_{\ell}}}\left\{K_{2 n}^{\mathrm{GUE}}\left(X_{k}, Y_{\ell} ; \hat{\tau}_{k}-\hat{\tau}_{\ell}\right)-K_{2 n}^{\mathrm{GUE}}\left(X_{k},-Y_{\ell} ; \hat{\tau}_{k}-\hat{\tau}_{\ell}\right)\right\}$


## Case of a Single Time: $m=1$

Order paths: $X_{n}(\tau)>\cdots>X_{1}(\tau)>0$

$$
\begin{aligned}
& \mathbb{P}\left(X_{1}(\tau) \geq x\right)=\operatorname{det}\left(I-K \chi_{J_{1}}\right), \quad J_{1}=(0, s) \\
& \mathbb{P}\left(X_{n}(\tau)<x\right)=\operatorname{det}\left(I-K \chi_{J_{2}}\right) \quad J_{2}=(s, \infty)
\end{aligned}
$$

where

$$
K=K_{2 n}^{\mathrm{GUE}}(x, y)-K_{2 n}^{\mathrm{GUE}}(x,-y), s=\frac{x}{\sqrt{\tau(1-\tau)}}
$$

- Operator $K$ is finite rank $\Longrightarrow$ distributions are $n \times n$ dets.
- Operator $K$ has kernel of the form

$$
\begin{gathered}
\frac{\varphi(x) \psi(y)-\psi(x) \varphi(y)}{x-y} \text { where } \\
\varphi(x)=n^{1 / 4} x^{1 / 4} \varphi_{2 n}(\sqrt{x}), \quad \psi(x)=n^{1 / 4} x^{-1 / 4} \varphi_{2 n-1}(\sqrt{x})
\end{gathered}
$$

## Painlevé V Representation

$$
\mathbb{P}\left(X_{1}(\tau) \geq x\right)=\exp \left(-\int_{0}^{s^{2}} \frac{r(t)}{t} d t\right)
$$

where $r$ satisfies the Jimbo-Miwa $\sigma$ form of Painlevé V , ${ }^{\text {a }}$ $s=x / \sqrt{\tau(1-\tau)}$ with boundary condition as $s \rightarrow 0^{+}$

$$
r(s)=r_{0} s^{3 / 2}+\mathrm{O}\left(s^{5 / 2}\right), r_{0}=\frac{1}{\sqrt{\pi}} \frac{1}{2^{2 n}}\binom{2 n}{n} \frac{4 n(2 n+1)}{3}
$$

- The number of BE curves, $n$, appears only as a parameter in the Painlevé V DE.
- Similar Painlevé V representation for $\mathbb{P}\left(X_{n}(\tau)<x\right)$.

$$
{ }^{\mathrm{a}} r=-\sigma, \nu_{0}=n, \nu_{1}=n+1 / 2, \nu_{2}=\nu_{3}=0
$$

## Limit Theorems as $\mathrm{n} \rightarrow \infty$

- Look at scaling of bottom curve as $n \rightarrow \infty$ :

$$
\begin{gathered}
x, y \rightarrow \sqrt{\frac{2 n(1-\tau)}{\tau}} x, y, \quad J_{k} \rightarrow \sqrt{\frac{2 n(1-\tau)}{\tau}} J_{k} \\
\tau_{k} \rightarrow \tau+\frac{\tau(1-\tau)}{2 n} \tau_{k}
\end{gathered}
$$

Then extended BE kernel approaches (with trace norm convergence) the extended Bessel kernel with $\alpha=1 / 2$ :

$$
\begin{aligned}
& \frac{2}{\pi} \int_{0}^{1} e^{\left(\tau_{k}-\tau_{\ell}\right) t^{2} / 2} \sin x t \sin y t d t, \quad k \geq \ell \\
&-\frac{2}{\pi} \int_{1}^{\infty} e^{\left(\tau_{k}-\tau_{\ell}\right) t^{2} / 2} \sin x t \sin y t d t, \quad k<\ell
\end{aligned}
$$

- Expect top curve to scale to Airy process. Top curve does not "feel presence of the wall" as $n \rightarrow \infty$. Checked that diagonal matrix elements approach (with trace norm convergence) the Airy kernel.


## Area under Bottom and Top Curves

Define

$$
\mathcal{A}_{n, L}=\int_{0}^{1} X_{1}(\tau) d \tau, \quad \mathcal{A}_{n, H}=\int_{0}^{1} X_{n}(\tau) d \tau
$$

- Expected area

$$
\begin{aligned}
\mathbb{E}\left(\mathcal{A}_{n, L}\right) & =\int_{0}^{1} \mathbb{E}\left(X_{1}(\tau)\right) d \tau \\
& =\int_{0}^{1} \int_{0}^{\infty} \mathbb{P}\left(X_{1}(\tau) \geq x\right) d x d \tau \\
& =\int_{0}^{1} \sqrt{2 \tau(1-\tau)} d \tau \cdot \int_{0}^{\infty} \operatorname{det}\left(I-K \chi_{(0, s)}\right) d s \\
& =\frac{\pi}{4 \sqrt{2}} \int_{0}^{\infty} \operatorname{det}\left(\delta_{j, k}-\left(\Psi_{j}, \Psi_{k}\right)\right)_{j, k=0}^{n-1} d s
\end{aligned}
$$

- Last expression good for numerical evaluation for small $n$.
- Higher moments are expressible in terms of integrals over Fredholm dets with extended kernel.


## Expected Area Asymptotics

Using convergence of BE kernel to Bessel kernel we derive (with some additional estimates to get convergence of the moments)

$$
\begin{gathered}
\mathbb{E}\left(\mathcal{A}_{n, L}\right) \sim \frac{c_{L}}{\sqrt{n}}, n \rightarrow \infty \text { where } \\
c_{L}=\frac{\pi}{8 \sqrt{2}} \int_{0}^{\infty} \operatorname{det}\left(I-K_{\text {Bessel }} \chi_{\left(0, s^{2}\right)}\right) d s \simeq 0.682808 .
\end{gathered}
$$

Compare with $\sqrt{n} \mathbb{E}\left(\mathcal{A}_{n, L}\right)$ for $n=5, \ldots, 11$
$0.667334,0.669708,0.671449,0.672784,0.673838,0.674691,0.675396$

- Top curve

$$
\begin{gathered}
\mathbb{E}\left(\mathcal{A}_{n, H}\right)=\frac{\pi}{2^{3 / 2}} \sqrt{n}+\frac{c_{H}}{n^{1 / 6}}+\mathrm{o}\left(n^{-1 / 6}\right) \\
c_{H}=\frac{\pi}{8 \cdot 2^{1 / 6}} \mu_{2} \simeq-0.619623767170
\end{gathered}
$$

Here $\mu_{2}$ is first moment of $F_{\mathrm{GUE}}$ largest eigenvalue distribution.

