

The hard-edge tacnode process for Brownian motion

Patrik L. Ferrari* Bálint Vető†

Abstract

We consider N non-intersecting Brownian bridges conditioned to stay below a fixed threshold. We consider a scaling limit where the limit shape is tangential to the threshold. In the large N limit, we determine the limiting distribution of the top Brownian bridge conditioned to stay below a function as well as the limiting correlation kernel of the system. It is a one-parameter family of processes which depends on the tuning of the threshold position on the natural fluctuation scale. We also discuss the relation to the six-vertex model and the Aztec diamond on restricted domains.

1 Introduction

Non-intersecting walks have appeared naturally in the descriptions of many physical systems as well as in mathematics. To mention just a few examples, the polynuclear growth model (describing the growth of an interface) is based on the representation as non-intersecting random walks [34, 50], the Aztec diamond (and similar combinatorial models of random tiling) has a similar mathematical description [9, 36], Markov chains on Young diagrams related to the Plancherel measure [7, 12], and the evolution of eigenvalues of random matrices as the GUE Dyson's Brownian motion [26] can be expressed and analyzed as non-intersecting Brownian motions [27, 47]. The analysis was possible because of the determinantal structure of correlation functions [6, 13, 27].

In this paper, we study non-intersecting Brownian motions starting and ending at a fixed position with the extra constraint that they stay below a given threshold as illustrated in Figure 1. The motivation for these investigations is twofold:

(a) The six-vertex model with domain wall boundary conditions (DWBC) can be expressed as a system of non-intersecting line ensembles (in discrete space and time) [30] with fixed starting and ending points. In particular, at the free-fermion line, there is a mapping to the Aztec diamond [57] and thus by [36] we know that the border of the lines are described in the limit of large system by the Airy₂ process [50]. Recent studies of limit shapes (not only for the free-fermion case) consider also geometries beyond the classical DWBC [15–18]. This raised the natural question on the description of the limit process for the border of the line ensemble for L -shaped domains or for pentagonal domains obtained from a square by removing a triangle at the corner. Although we do not do the analysis for this discrete case, if the removed triangular piece is tangential to the the limit

*Institute for Applied Mathematics, Bonn University, Endenicher Allee 60, 53115 Bonn, Germany. E-mail: ferrari@uni-bonn.de

†MTA–BME Stochastics Research Group, Egrý J. u. 1, 1111 Budapest, Hungary. E-mail: vetob@math.bme.hu

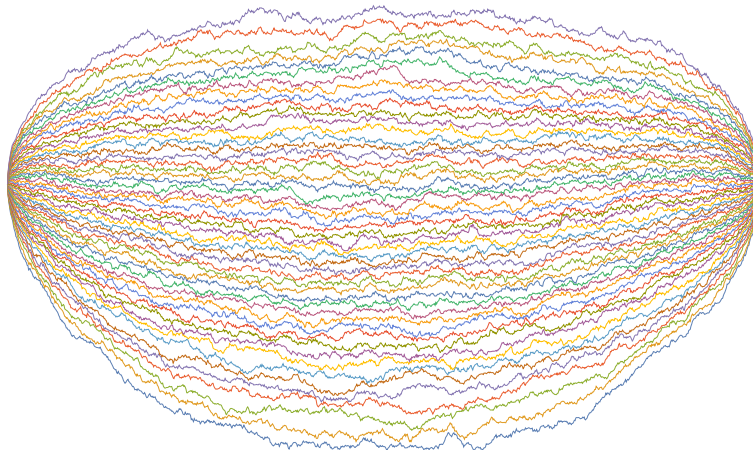


Figure 1: Illustration of $N = 50$ non-intersecting Brownian motions conditioned to stay below the black threshold.

shape of the lines for the DWBC, then under appropriate scaling, the limiting process should be exactly the one we study in this paper. See Section 3 for further discussions.

(b) Non-intersecting Brownian motions have attracted a lot of interest also because of their relations to the eigenvalues of Hermitian random matrices subjected to Dyson's Brownian motion [2, 5, 10, 40–42, 47, 54, 55]. Discrete versions have been studied as well [29, 35, 46, 48, 50]. More recently, the situation where the limit shape of two sets of non-intersecting Brownian motions just touch in a tacnode geometry has been studied, first in a random walk setting in [3], then via a 4×4 Riemann-Hilbert problem [23] and with a more direct approach in [32, 38]. The equality of the formulas for the correlation kernel of the tacnode process obtained in [23] and in [32] was verified directly in [21]. The tacnode was observed also in random tiling models [1, 4]. The tacnode geometry occurs also if the Brownian motions are conditioned to stay positive and to start and end away from 0 at a distance so that the limit shape becomes tangential to 0. This has been studied in [22, 24] for the case of non-intersecting Bessel processes. However, finding explicit formulas for the tacnode limit process for Brownian motions conditioned to stay positive remained open.

We mention that if the starting points of Brownian bridges (or more generally Bessel processes) are set to 0, the ending points are the same for all paths and it is scaled with the number of paths, then the limit shape of the non-intersecting paths conditioned to stay positive separates from 0 at some time in $(0, 1)$. In the neighbourhood of the point of separation, the hard-edge Pearcey process appears [24, 25, 45].

In this paper, we consider N Brownian bridges starting from 0 at time 0 and ending at 0 at time 1. We condition the Brownian bridges not to intersect for times $t \in (0, 1)$ and denote by $B_N(t)$ the position of the top bridge at time t . This is also known as Brownian watermelon and it is well-known that under appropriate scaling, B_N converges to the \mathcal{A}_2 process \mathcal{A}_2 :

$$2N^{1/6} \left(B_N \left(\frac{1}{2}(1 + uN^{-1/3}) \right) - \sqrt{N} \right) \rightarrow \mathcal{A}_2(u) - u^2 \quad (1.1)$$

as $N \rightarrow \infty$. Therefore, if we consider the Brownian watermelon conditioned to stay below a threshold of height $\sqrt{N} + \frac{1}{2}RN^{-1/6}$, then the probability that the conditioning is effective is in $(0, 1)$ also in the $N \rightarrow \infty$ limit. Thus we will see a new non-trivial limit

process, which we call *hard-edge tacnode process for Brownian motions*. This process is characterized by its finite dimensional distributions as given in Theorem 2.6. When $R \rightarrow \infty$, the constraint becomes irrelevant and the top path will be the Airy_2 process (see Remark 2.11). When $R \rightarrow -\infty$, after appropriate rescaling, the limit process should be the one with extended Bessel kernel [54] which was also derived for non-intersecting Brownian excursions studied in [56].

The derivation of our result does not use the standard determinantal point process approach [37], rather we start with a Fredholm determinant expression with path integral kernel obtained in [8, 49] which gives the probability that the top path of N Brownian bridges stays below a given function over an open time subinterval of $[0, 1]$, see Proposition 2.1. First we extend the conditioning to the full time interval (see Theorem 2.3). The finite dimensional distributions are then written as ratios of probabilities for two threshold functions leading to Theorem 2.4. Using [8], we can rewrite the Fredholm determinant of a path integral kernel to a Fredholm determinant of an extended kernel which is indeed the correlation kernel as shown in Theorem 2.5. ([8] is a generalization of what was present in [50]. The importance of [50] was rediscovered and extended in [20] in the setting of the Airy processes.) Notice that with the present method, we directly get formulas for quantities such as distribution of the maximum of B_N (conditioned to stay below the threshold). This quantity is not directly accessible by the standard method leading to the finite dimensional distributions. Finally we perform the asymptotic analysis both for the correlation kernel (see Theorem 2.6) and for the probability that the hard-edge tacnode process stays below a given function (see Theorem 2.9).

Outline: In Section 2, we define the model and present the results of this paper. Section 3 contains a short discussion on the relation with the six-vertex model. In Section 4, we determine the multipoint distribution of B_N conditioned to stay below a constant threshold. Section 5 contains the extension of the formula of [49] to the full time interval. In Section 6, we prove the formula for the correlation kernel. The large N asymptotic analysis is performed in Section 7. Finally, Section 8 contains the proof of several technical lemmas.

Acknowledgements: The work of P.L. Ferrari is supported by the German Research Foundation via the SFB 1060–B04 project. The work of B. Vetó is supported by OTKA (Hungarian National Research Fund) grant K100473. He is grateful for the Postdoctoral Fellowship of the Hungarian Academy of Sciences and for the Bolyai Research Scholarship. The authors are grateful for discussions with F. Colomo and A. Sportiello about their work and to both ICERM and the Galileo Galilei Institute which provided the platform to make such discussions possible.

2 Model and main results

The model

The model considered in this paper is the following system of N non-intersecting Brownian bridges. Consider N standard Brownian bridges $B_1(t), \dots, B_N(t)$ which start from zero at time $t = 0$ and end at zero at time $t = 1$, and condition them on having no intersection in $t \in (0, 1)$ in Doob's sense. To denote the paths, we use the convention $B_1(t) \leq \dots \leq B_N(t)$ with strict inequality for $t \in (0, 1)$.

The starting point of the work is a formula for the distribution of the top path $B_N(t)$ conditioned to stay below a given function, based on [8] and [49]. To state it, we need some notations. Let $H_n(x)$ denote the n th Hermite polynomial defined by

$$H_n(x) = (-1)^n e^{x^2} \frac{d^n}{dx^n} e^{-x^2} \quad (2.1)$$

which form an orthogonal system with respect to the weight $e^{-x^2} dx$ on \mathbb{R} , i.e.

$$\int_{\mathbb{R}} H_n(x) H_m(x) e^{-x^2} dx = \sqrt{\pi} 2^n n! \delta_{n,m}. \quad (2.2)$$

Define the harmonic oscillator functions

$$\varphi_n(x) = \pi^{-1/4} 2^{-n/2} (n!)^{-1/2} e^{-x^2/2} H_n(x) \quad (2.3)$$

and the Hermite kernel

$$K_{\text{Herm},N}(x, y) = \sum_{n=0}^{N-1} \varphi_n(x) \varphi_n(y). \quad (2.4)$$

With the Laplacian Δ on \mathbb{R} , let

$$D = -\frac{1}{2}(\Delta - x^2 + 1) \quad (2.5)$$

be the differential operator for which the eigenfunctions are the harmonic oscillator functions, that is, $D\varphi_n = n\varphi_n$. Then $K_{\text{Herm},N}$ is a projection to the space spanned by the eigenfunctions $\varphi_0, \dots, \varphi_{N-1}$.

For some $0 < a < b < 1$, let $H^1([a, b])$ be the set of square integrable functions with square integrable derivative. The following statement is a consequence of Propositions 2.1 (which goes back to Proposition 4.3 of [8]) and Proposition 2.2 in [49].

Proposition 2.1 (Nguyen-Remenik [49]). *Let $0 < a < b < 1$ and $h \in H^1([a, b])$ and denote by $B_N(t)$ the top path of N non-intersecting Brownian bridges. Then*

$$\mathbf{P}(B_N(t) < h(t) \text{ for } t \in [a, b]) = \det(\mathbb{1} - K_{\text{Herm},N} + \Theta_{A,B} e^{(B-A)D} K_{\text{Herm},N})_{L^2(\mathbb{R})} \quad (2.6)$$

where $A = \frac{1}{2} \ln \frac{a}{1-a}$, $B = \frac{1}{2} \ln \frac{b}{1-b}$, and D is the differential operator defined in (2.5). Further,

$$\begin{aligned} \Theta_{A,B}(x, y) &= e^{(y^2 - x^2)/2 + B} \frac{\exp\left(-\frac{(e^B y - e^A x)^2}{4(\beta - \alpha)}\right)}{\sqrt{4\pi(\beta - \alpha)}} \\ &\quad \times \mathbf{P}_{\widehat{b}(\alpha) = e^A x, \widehat{b}(\beta) = e^B y} \left(\widehat{b}(\tau) \leq \frac{1 + 4\tau}{\sqrt{2}} h\left(\frac{4\tau}{1 + 4\tau}\right) \text{ for } \tau \in [\alpha, \beta] \right) \end{aligned} \quad (2.7)$$

where $\alpha = \frac{1}{4} e^{2A} = \frac{1}{4} \frac{a}{1-a}$ and $\beta = \frac{1}{4} e^{2B} = \frac{1}{4} \frac{b}{1-b}$. In (2.7), $\widehat{b}(\tau)$ denotes a Brownian bridge with diffusion coefficient 2 starting at $\widehat{b}(\alpha) = e^A x$ and ending at $\widehat{b}(\beta) = e^B y$.

Finite N result

First of all, we extend Proposition 2.1 so that the condition for the N non-intersecting Brownian bridges to stay below a function can be imposed for the whole $[0, 1]$. Since we are ultimately interested in the distribution of N non-intersecting Brownian bridges conditioned to stay below a constant, we consider functions h is such that for some $0 < t_1 < t_2 < 1$ and $r > 0$,

$$h(t) \leq r \quad \text{for } t \in [0, 1] \quad \text{and} \quad h(t) = r \quad \text{for } t \in [0, 1] \setminus (t_1, t_2). \quad (2.8)$$

Motivated by the definition (2.7), let

$$\tau_i = \frac{1}{4} \frac{t_i}{1 - t_i} \quad \text{for } i = 1, 2 \quad \text{and} \quad \tilde{h}(\tau) = \frac{1 + 4\tau}{\sqrt{2}} \left[h \left(\frac{4\tau}{1 + 4\tau} \right) - r \right]. \quad (2.9)$$

Further, for such a function h , define

$$\begin{aligned} & T_{\alpha_1, \alpha_2}^h(u, v) \\ &= \frac{d}{dv} \mathbf{P}_{\tilde{b}(\alpha_1)=u} \left(\tilde{b}(\tau) \leq 0 \text{ for } \tau \in [\alpha_1, \alpha_2], \tilde{b}(\tau) \leq \tilde{h}(\tau) \text{ for } \tau \in (\tau_1, \tau_2), \tilde{b}(\alpha_2) \leq v \right), \end{aligned} \quad (2.10)$$

where $\tau_1, \tau_2 \in [\alpha_1, \alpha_2]$ and \tilde{h} are as in (2.9). The Brownian motion \tilde{b} above has diffusion coefficient 2.

For any $u, v \in \mathbb{R}$ and n, m integers, introduce the functions

$$\Phi_\tau^n(u) = \frac{1}{\pi i} \int_{i\mathbb{R}} dW W^n e^{\tau(\sqrt{2}r - 2W)^2 - \sqrt{2}rW} (f_W(u) - f_W(-u)), \quad (2.11)$$

$$\Psi_\tau^m(v) = \frac{1}{2\pi i} \oint_{\Gamma_0} dZ Z^{-(m+1)} e^{-\tau(\sqrt{2}r - 2Z)^2 + \sqrt{2}rZ} (g_Z(v) - g_Z(-v)) \quad (2.12)$$

with

$$f_W(u) = e^{(\sqrt{2}r - 2W)u} \quad \text{and} \quad g_Z(v) = e^{-(\sqrt{2}r - 2Z)v} \quad (2.13)$$

and define the kernel

$$K_0(n, m) = \frac{1}{2\pi i} \oint_{\Gamma_0} dZ \frac{(\sqrt{2}r - Z)^n}{Z^{m+1}} e^{-2r^2 + 2\sqrt{2}rZ}. \quad (2.14)$$

They satisfy the following compatibility conditions (see Section 8 for the proof).

Proposition 2.2. *Let $\phi_t(x, y) = \frac{1}{\sqrt{2\pi t}} \exp(-(x - y)^2/2t)$ and set*

$$T_{\tau_1, \tau_2}(x, y) = \phi_{2(\tau_2 - \tau_1)}(y - x) - \phi_{2(\tau_2 - \tau_1)}(y + x) \quad (2.15)$$

for any $x, y \in \mathbb{R}$. Then, for any $0 < \tau_1 < \tau_2$, $0 < \tau$ and $u, v \in \mathbb{R}$, the following compatibility relations are satisfied:

$$\int_{\mathbb{R}_-} du \Phi_{\tau_1}^n(u) T_{\tau_1, \tau_2}(u, v) = \Phi_{\tau_2}^n(v), \quad (2.16)$$

$$\int_{\mathbb{R}_-} dv T_{\tau_1, \tau_2}(u, v) \Psi_{\tau_2}^m(v) = \Psi_{\tau_1}^m(u), \quad (2.17)$$

$$\int_{\mathbb{R}_-} du \Phi_\tau^n(u) \Psi_\tau^m(u) = (\mathbb{1} - K_0)(n, m). \quad (2.18)$$

We can now state the extension of Proposition 2.1 to the conditioning on the full time interval.

Theorem 2.3 (Full time span conditioning). *Let the function $h \in H^1([0, 1])$ satisfy (2.8) for some $0 < t_1 < t_2 < 1$. Then*

$$\mathbf{P}(B_N(t) < h(t) \text{ for } t \in [0, 1]) = \det(\mathbb{1} - K_N^h)_{L^2(\{0,1,\dots,N-1\})} \quad (2.19)$$

where the kernel K_N^h is given by

$$K_N^h(n, m) = \mathbb{1}(n, m) - \int_{\mathbb{R}} du \int_{\mathbb{R}} dv \Phi_{\tau_1}^n(u) T_{\tau_1, \tau_2}^h(u, v) \Psi_{\tau_2}^m(v). \quad (2.20)$$

A consequence of Theorem 2.3 we get the probability of the conditioned process.

Theorem 2.4. *Under the assumptions of Theorem 2.3, we have*

$$\begin{aligned} \mathbf{P}(B_N(t) < h(t) \text{ for } t \in [0, 1] \mid B_N(t) < r \text{ for } t \in [0, 1]) \\ = \det(\mathbb{1} - K_{\tau_1} + T_{\tau_1, \tau_2}^h K_{\tau_2, \tau_1})_{L^2(\mathbb{R}_-)}, \end{aligned} \quad (2.21)$$

where $K_{\tau_1} = K_{\tau_1, \tau_1}$ and K_{τ_2, τ_1} is given by

$$K_{\tau_2, \tau_1}(u, v) = \sum_{n, m=0}^{N-1} \Psi_{\tau_2}^n(u) (\mathbb{1} - K_0)^{-1}(n, m) \Phi_{\tau_1}^m(v). \quad (2.22)$$

For N non-intersecting Brownian bridges conditioned to stay below a constant level r for $[0, 1]$, we know by the Karlin–McGregor type formulas and Eynard–Mehta theorem [27, 39] that it forms a determinantal process. We compute its correlation kernel which characterizes the finite dimensional distributions of the process. Conditioning N non-intersecting Brownian bridges to stay below r corresponds to the $h \equiv r$ constant choice in (2.8). In this case, (2.10) becomes $\mathbb{1}_{u < 0} T_{\alpha_1, \alpha_2}(u, v) \mathbb{1}_{v < 0}$ by the reflection principle. The correlation kernel of N non-intersecting Brownian bridges conditioned to be below a constant level is given as follows.

Theorem 2.5 (Correlation kernel). *The system of N non-intersecting Brownian bridges conditioned to stay below the constant level r for time $[0, 1]$ forms a determinantal process with extended correlation kernel defined for $t_1, t_2 \in [0, 1]$ and $x_1, x_2 \leq r$ by*

$$K_{\text{ext}}(t_1, x_1; t_2, x_2) = \frac{1}{\sqrt{2(1-t_1)(1-t_2)}} K^{\text{ext}}(\tau_1, u_1; \tau_2, u_2) \quad (2.23)$$

where we used the variables

$$\tau_i = \frac{1}{4} \frac{t_i}{1-t_i}, \quad u_i = \frac{x_i - r}{\sqrt{2}(1-t_i)} \quad (2.24)$$

due to (2.9) and the kernel

$$K^{\text{ext}}(\tau_1, u_1; \tau_2, u_2) = -\mathbb{1}_{\tau_1 < \tau_2} T_{\tau_1, \tau_2}(u_1, u_2) + \sum_{n, m=0}^{N-1} \Psi_{\tau_1}^n(u_1) (\mathbb{1} - K_0)^{-1}(n, m) \Phi_{\tau_2}^m(u_2). \quad (2.25)$$

In particular, the gap probabilities of N non-intersecting Brownian bridges conditioned to stay below level r can be expressed for any $t_1, \dots, t_k \in [0, 1]$ and $h_1, \dots, h_k \leq r$ as

$$\mathbf{P}(B_N(t_1) < h_1, \dots, B_N(t_k) < h_k \mid B_N(t) < r, t \in [0, 1]) = \det(\mathbb{1} - QK^{\text{ext}})_{L^2(\{\tau_1, \dots, \tau_k\} \times \mathbb{R}_-)} \quad (2.26)$$

with

$$Qf(\tau_i, u) = \mathbb{1}_{u \geq \eta_i} f(\tau_i, u), \quad \tau_i = \frac{1}{4} \frac{t_i}{1-t_i}, \quad \eta_i = \frac{1 + 4\tau_i}{\sqrt{2}}(h_i - r). \quad (2.27)$$

Large N asymptotic result

As mentioned in the introduction, we want to analyze the $N \rightarrow \infty$ limit under the scaling limit where the probability that the top Brownian motion without conditioning crosses the threshold h stays asymptotically away from 0 and 1. This means that we need to scale the threshold r as well as time and space as follows:

$$t = \frac{1 + TN^{-1/3}}{2}, \quad r = \sqrt{N} + \frac{RN^{-1/6}}{2}, \quad h = \sqrt{N} + \frac{(R + H)N^{-1/6}}{2} \quad (2.28)$$

with $H \leq 0$. Let us first describe ingredients of the limiting correlation kernel. For any parameter s , let

$$\text{Ai}^{(s)}(x) = e^{2s^3/3+xs} \text{Ai}(s^2 + x). \quad (2.29)$$

Then we introduce the functions

$$\begin{aligned} \widehat{\Phi}_T^\xi(U) &= \text{Ai}^{(T)}(R + \xi + U) - \text{Ai}^{(T)}(R + \xi - U), \\ \widehat{\Psi}_T^\zeta(U) &= \text{Ai}^{(-T)}(R + \zeta + U) - \text{Ai}^{(-T)}(R + \zeta - U), \end{aligned} \quad (2.30)$$

and the shifted GOE kernel

$$\widehat{K}_0(\xi, \zeta) = 2^{-1/3} \text{Ai}(2^{-1/3}(2R + \xi + \zeta)). \quad (2.31)$$

The next theorem establishes the convergence of the rescaled kernel and the existence of the hard-edge tacnode process which is the limiting determinantal point process.

Theorem 2.6 (The hard-edge tacnode process). *Consider the scaling*

$$t_i = \frac{1 + T_i N^{-1/3}}{2}, \quad r = \sqrt{N} + \frac{RN^{-1/6}}{2}, \quad x_i = \sqrt{N} + \frac{(R + U_i)N^{-1/6}}{2}. \quad (2.32)$$

Then the extended correlation kernel of N non-intersecting Brownian bridges conditioned to stay below a constant level converges uniformly on compact sets, i.e.,

$$\lim_{N \rightarrow \infty} \frac{N^{-1/6}}{2} K_{\text{ext}}(t_1, x_1; t_2, x_2) = \widehat{K}^{\text{ext}}(T_1, U_1; T_2, U_2) \quad (2.33)$$

where the limiting kernel \widehat{K}^{ext} is given by

$$\widehat{K}^{\text{ext}}(T_1, U_1; T_2, U_2) = -\mathbb{1}_{T_1 < T_2} T_{T_1, T_2}(U_1, U_2) + \int_{\mathbb{R}_+} d\xi \int_{\mathbb{R}_+} d\zeta \widehat{\Psi}_{T_1}^\xi(U_1) (\mathbb{1} - \widehat{K}_0)^{-1}(\xi, \zeta) \widehat{\Phi}_{T_2}^\zeta(U_2) \quad (2.34)$$

where $T_1, T_2 \in \mathbb{R}$ and $U_1, U_2 \leq 0$.

As a consequence, the hard-edge tacnode process \mathcal{T} exists as the limit of N non-intersecting Brownian bridges conditioned to stay below a constant level under the given scaling. It is characterized by the following gap probabilities. For any fixed integer k and $T_1, \dots, T_k \in \mathbb{R}$ and for any compact set $E \subseteq \{T_1, \dots, T_k\} \times \mathbb{R}_-$,

$$\mathbf{P}(\mathcal{T} \cap E = \emptyset) = \det(\mathbb{1} - \widehat{K}^{\text{ext}})_{L^2(E)}. \quad (2.35)$$

Remark 2.7. It is easy to verify that

$$\begin{aligned}
& \lim_{R \rightarrow \infty} \widehat{K}^{\text{ext}}(T_1, U_1 - R; T_2, U_2 - R) e^{2(T_1^3 - T_2^3)/3 + T_1 U_1 - T_2 U_2} \\
&= K_{\text{Ai}}(T_1, U_1 + T_1^2; T_2, U_2 + T_2^2) \\
&= -\frac{e^{-\frac{(U_1 - U_2)^2}{4(T_2 - T_1)}}}{\sqrt{4\pi(T_2 - T_1)}} \mathbb{1}_{T_1 < T_2} + \int_{\mathbb{R}_+} d\xi e^{-\xi(T_2 - T_1)} \text{Ai}(\xi + U_1 + T_1^2) \text{Ai}(\xi + U_2 + T_2^2).
\end{aligned} \tag{2.36}$$

K_{Ai} is known as the extended Airy kernel [36, 50]. Indeed, as $R \rightarrow \infty$, $(\mathbb{1} - \widehat{K}_0)^{-1} \rightarrow \mathbb{1}$, but also $\widehat{\Phi}_T^\xi(U - R) e^{-2T^3/3 - TU} \rightarrow e^{T\xi} \text{Ai}(\xi + U + T^2)$, and $\widehat{\Psi}_T^\zeta(U - R) e^{2T^3/3 + TU} \rightarrow e^{-T\zeta} \text{Ai}(\zeta + U + T^2)$. This is not surprising since in the $R \rightarrow \infty$ limit the constraint by the threshold disappears and we are back to the standard watermelon case. The top paths in the system of non-intersecting Brownian bridges converge to the Airy point field, which have correlation kernel given by K_{Ai} .

As in [21] and in [24], the soft edge or hard-edge tacnode process usually has a natural temperature parameter (here is the threshold R), and the derivative of the correlation kernel with respect to the temperature parameter has a low rank structure. In particular, the temperature derivative of the correlation kernel of the soft edge tacnode process is rank two, which was proved in [21] to hold for the formulas obtained in [23] and in [32] yielding a direct proof for the equivalence of the two formulations. In [24], the rank one structure of the temperature derivative of the hard-edge tacnode kernel was shown obtained from non-intersecting squared Bessel processes with even dimension. This gives the importance of the next proposition about the derivative with respect to the microscopic position parameter of the threshold, which is proved in Section 8.

Proposition 2.8. *The derivative of the extended correlation kernel of the hard-edge tacnode process with respect to parameter R has rank one, that is,*

$$\frac{\partial}{\partial R} \widehat{K}^{\text{ext}}(T_1, U_1; T_2, U_2) = -f(T_1, U_1)g(T_2, U_2) \tag{2.37}$$

where

$$f(T_1, U_1) = \int_{\mathbb{R}_+} d\xi \widehat{\Psi}_{T_1}^\xi(U_1) (\mathbb{1} - \widehat{K}_0)^{-1}(\xi, 0), \tag{2.38}$$

$$g(T_2, U_2) = \int_{\mathbb{R}_+} d\zeta (\mathbb{1} - \widehat{K}_0)^{-1}(0, \zeta) \widehat{\Phi}_{T_2}^\zeta(U_2). \tag{2.39}$$

From Theorem 2.6 we have the finite dimensional distributions for the limiting process, which does not cover properties as the probability that the limiting process stay below a given function. This can be obtained by performing the large N asymptotics of Theorem 2.4.

Theorem 2.9. *Consider the top path of N non-intersecting Brownian motions conditioned to stay below $r = \sqrt{N} + \frac{1}{2}RN^{-1/6}$ rescaled as*

$$\mathcal{B}_N^R(T) = 2N^{1/6} \left(B_N \left(\frac{1}{2}(1 + TN^{-1/3}) \right) - \sqrt{N} \right). \tag{2.40}$$

Let $T_1 < T_2$ be given as well as a function $H \in H^1([T_1, T_2])$ with $H \leq R$. Then

$$\lim_{N \rightarrow \infty} \mathbf{P} \left(\mathcal{B}_N^R(T) \leq H(T) \text{ for } T \in [T_1, T_2] \right) = \det(\mathbb{1} - K_{T_1} + T_{T_1, T_2}^{H-R} K_{T_2, T_1})_{L^2(\mathbb{R}_-)} \tag{2.41}$$

where $K_{T_1} = K_{T_1, T_1}$ and $K_{T_1, T_2}(U_1, U_2) := \widehat{K}^{\text{ext}}(T_1, U_1; T_2, U_2)$ defined in (2.34).

Remark 2.10. We have $K_{T_1, T_2}(U_1 - R, U_2 - R) \rightarrow K_{\text{Ai}}(T_1, U_1 + T_1^2; T_2, U_2 + T_2^2)$ (up to a conjugation) as $R \rightarrow \infty$, see (2.36). Also $T_{T_1, T_2}^{H-R}(U_1 - R, U_2 - R) = T_{T_1, T_2}^H(U_1, U_2)$. The constraint $H \leq R$ becomes trivially satisfied in the $R \rightarrow \infty$ limit. Thus, as expected, we recover the probability that the Airy₂ process stay below H :

$$\lim_{R \rightarrow \infty} \det(\mathbb{1} - K_{T_1} + T_{T_1, T_2}^{H-R} K_{T_2, T_1})_{L^2(\mathbb{R}_-)} = \mathbf{P}(\mathcal{A}_2(T) - T^2 \leq H(T), T \in [T_1, T_2]) \quad (2.42)$$

where \mathcal{A}_2 is the Airy₂ process, compare with Theorem 2 and 3 of [20].

Remark 2.11. There are two natural ways to obtain the hard-edge tacnode process as the limit of non-intersecting Brownian bridges conditioned to stay below a constant level. The first option is what we follow in the present paper: we keep the number of paths fixed first and we characterize the distribution of the paths conditioned to stay below a constant level for $[0, 1]$, see Theorem 2.5. Then we let the number of paths $N \rightarrow \infty$ in Theorem 2.6.

An alternative approach is also possible. Imposing the condition that the Brownian bridges stay under a constant level on a fixed interval $[a, b]$ with $0 < a < b < 1$ and letting the number of paths $N \rightarrow \infty$ first, then the limit is an Airy₂ process conditioned to stay below a parabola for a fixed finite interval (compare with (2.42) for constant H). In the second step, by letting this interval grow to \mathbb{R} , the same hard-edge tacnode process is obtained as in Theorem 2.6.

3 Relation to the six-vertex model and the Aztec diamond

The six-vertex model is a statistical mechanics model with short range interaction which is however sensitive to the boundary conditions. For instance, imposing the so-called domain wall boundary conditions (DWBC), it was noticed in [43] that it has a macroscopic influence on the system. In this setting, the model has two free parameters. When these parameters satisfy a given equation, the system becomes “free-fermion” and there is a (many-to-one) mapping to the Aztec diamond [57]. For the free-fermion case, one can associate a set of non-intersecting lines to the six-vertex configurations, from which the Aztec diamond configurations can be recovered [30]. These are illustrated in Figure 2.

In the recent papers on the six-vertex model [15–18], questions concerning the limit shape and correlation functions have been addressed for the six-vertex model also for other domains. In particular, domains obtained from a square by cutting off a triangle or a rectangle from the corner were considered with DWBC. In terms of the Aztec diamond, this corresponds to conditioning the dominoes in the top corner to be all fixed and horizontal. The fixed dominoes form the region which has been cut out.

The Aztec diamond has been studied very well. In particular, denote the size of the Aztec diamond by N . One can think of lines in discrete time $t \in [-N, N]$.

Theorem 3.1 (Theorem 1.1 of [36]). *Denote by $X_N(t)$ the top line of the Aztec diamond at time t . Then*

$$\frac{X_N(2^{-1/6} N^{2/3} T) - N/\sqrt{2}}{2^{-5/6} N^{1/3}} \rightarrow \mathcal{A}_2(T) - T^2 \quad (3.1)$$

in the sense of finite dimensional distributions. Here \mathcal{A}_2 is the Airy₂ process.

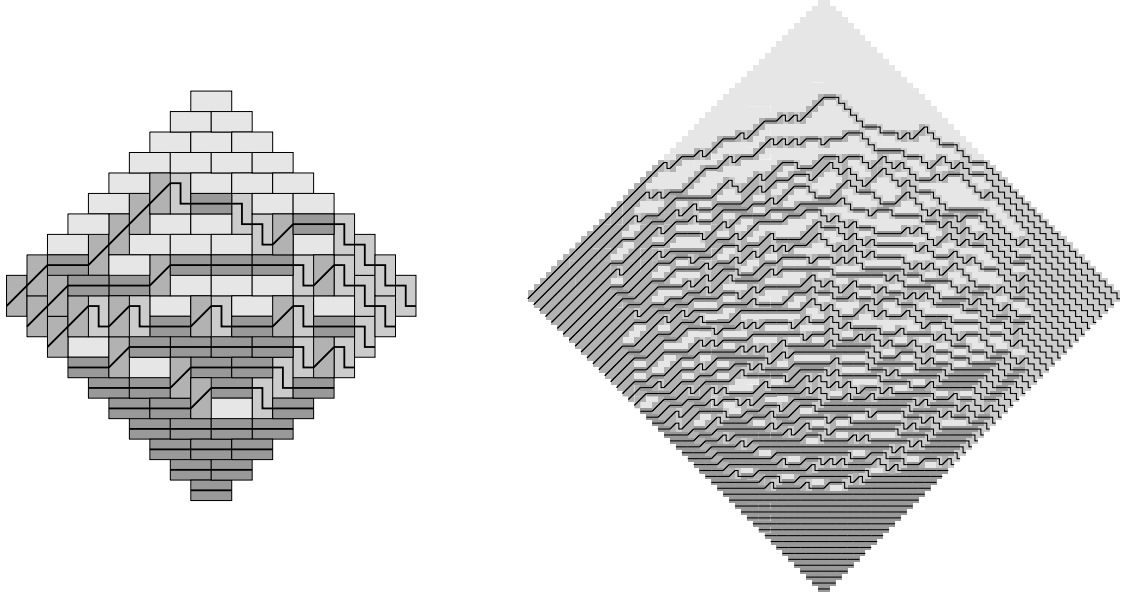


Figure 2: Illustration of the non-intersecting line ensemble for an Aztec diamond/six-vertex model with DWBC of size $N = 10$ (left) and size $N = 50$ (right).

The result is derived by analyzing the point process of the lines. Consider the $(N^{2/3}, N^{1/3})$ windows around the top line of Figure 2, i.e. if (t, x) denotes the coordinates of the lines in Figure 2, one considers

$$(t, x) = (2^{-1/6} N^{2/3} T, N/\sqrt{2} + 2^{-5/6} N^{1/3} U). \quad (3.2)$$

Then under this scaling, the lines converge to a determinantal point process with correlation kernel given by the extended Airy kernel, see (2.36). This is the same limit as the appropriate scaling limit obtained from N non-intersecting Brownian bridges as $N \rightarrow \infty$. Notice that the scaling of the horizontal and vertical directions is compatible with the Brownian scaling (as it is the case for the limit process since the Airy_2 process is locally Brownian [14, 19, 33]).

L-shaped case: Under the scaling (3.2), cutting out a square from the top of the Aztec diamond such that its lower tip is at height $N/\sqrt{2} + 2^{-5/6} N^{1/3} R$ is asymptotically equivalent to forbidding only a vertical line segment down to the tip of the square. Denote by X_N^R the top line in this case. Then, from the above discussion, we expect the following:

Conjecture 3.2. *Define*

$$X_N^{R, \text{resc}}(T) = \frac{X_N^R(2^{-1/6} N^{2/3} T) - N/\sqrt{2}}{2^{-5/6} N^{1/3}}. \quad (3.3)$$

Then for any given $T_1 < T_2 < \dots < T_k$ and $U_1, \dots, U_k \leq R$,

$$\lim_{N \rightarrow \infty} \mathbf{P} \left(\bigcap_{\ell=1}^k \{X_N^{R, \text{resc}}(T_\ell) \leq U_\ell\} \right) = \frac{\mathbf{P} \left(\bigcap_{\ell=1}^k \{\mathcal{A}_2(T_\ell) - T_\ell^2 \leq U_\ell\} \cap \{\mathcal{A}_2(0) \leq R\} \right)}{\mathbf{P}(\mathcal{A}_2(0) \leq R)} \quad (3.4)$$

where \mathcal{A}_2 is the Airy_2 process [36, 50]. As a consequence

$$\lim_{N \rightarrow \infty} \mathbf{P}(X_N^{R, \text{resc}} \leq U) = \frac{F_{\text{GUE}}(\min\{U, R\})}{F_{\text{GUE}}(R)}, \quad (3.5)$$

where F_{GUE} is the GUE Tracy–Widom distribution function [53].

Pentagonal case: Under the scaling (3.2), cutting out a triangle on the top corner at height $N/\sqrt{2} + 2^{-5/6}N^{1/3}R$ becomes asymptotically a conditioning to stay below a fixed height R . Denote by X_N^R be the top line in this case. Then, we expect to have the following:

Conjecture 3.3. *Define*

$$X_N^{R,\text{resc}}(T) = \frac{X_N^R(2^{-1/6}N^{2/3}T) - N/\sqrt{2}}{2^{-5/6}N^{1/3}}. \quad (3.6)$$

Then

$$\lim_{N \rightarrow \infty} \mathbf{P} \left(\bigcap_{\ell=1}^k \{X_N^{R,\text{resc}}(T_\ell) \leq U_\ell\} \right) = \det \left(\mathbb{1} - \widehat{K}^{\text{ext}} \right)_{L^2(E)} \quad (3.7)$$

with the set $E = \{(T_1, [U_1 - R, 0]) \times \dots \times (T_k, [U_k - R, 0])\}$.

4 Multipoint distribution and heuristics for the correlation kernel

In this section, we consider the process of N non-interaction Brownian bridges conditioned to stay below a constant level. First we prove Theorem 2.4, that is, the probability that this conditional process stays below a function of the form (2.8) can be written as a Fredholm determinant of the kernel K_{ext} . As a consequence, we show that the multipoint distribution of the conditional process also has a Fredholm determinantal form, which is part of the statement of Theorem 2.5. This does not imply that K_{ext} is the correlation kernel for the point process of the non-intersecting Brownian bridges, but it gives a potential candidate for it. The proof that K_{ext} is actually the correlation kernel is performed directly in Section 6.

Our heuristic derivation of the correlation kernel for N non-intersecting Brownian bridges conditioned to stay below a constant is based on the formula given in Theorem 2.3 for the probability that the top path of N Brownian bridges is below a function. First we verify that the kernel which appears in Theorem 2.3 is trace class.

Lemma 4.1. *For any function $h : [0, 1] \rightarrow \mathbb{R}$ and for any fixed integer N , the operator with kernel K_N^h given in (2.20) is trace class on $L^2(\{0, 1, \dots, N-1\})$.*

Proof. Since N is fixed, it is enough to show that $K_N^h(n, m)$ is finite for any $n, m < N$, that is, the double integral in (2.20) is finite. By (2.12), one clearly has $|\Psi_{\tau_2}^m(v)| \leq Ce^{c|v|}$ for some finite constants C and c . By definition (2.10), $|T_{\tau_1, \tau_2}(u, v)| \leq \phi_{2(\tau_2 - \tau_1)}(y - x)$, hence

$$\int_{\mathbb{R}} dv |T_{\tau_1, \tau_2}^h(u, v) \Psi_{\tau_2}^m(v)| \leq Ce^{c|u|} \quad (4.1)$$

for some finite constants C and c . On the other hand, (6.15) shows that $\Phi_{\tau_1}^n(u)$ has a Gaussian decay in u , i.e.

$$|\Phi_{\tau_1}^n(u)| \leq Ce^{-\frac{u^2}{8\tau_1}} \quad (4.2)$$

for some finite C . This completes the proof. \square

For any function $f \in L^2(\mathbb{R})$, let

$$P_\eta f(x) = \mathbb{1}_{x \geq \eta} f(x), \quad \overline{P}_\eta f(x) = \mathbb{1}_{x < \eta} f(x) \quad (4.3)$$

be the projection operators.

Proof of Theorem 2.4. The strategy of the proof is to compare the kernel K_N^h for a general h of the form (4.9) to the one which corresponds to the constant $h \equiv r$. For $h \equiv r$, in the second term on the right-hand side of (2.20) one has to insert

$$T_{\tau_1, \tau_2}^r = \overline{P}_0 T_{\tau_1, \tau_2} \overline{P}_0 \quad (4.4)$$

by comparing (2.10), (2.15) and (2.9). Hence the kernel K_N^r for $h \equiv r$ simplifies to

$$K_N^r(n, m) = \mathbb{1}(n, m) - \int_{\mathbb{R}_-} du \int_{\mathbb{R}_-} dv \Phi_{\tau_1}^n(u) T_{\tau_1, \tau_2}(u, v) \Psi_{\tau_2}^m(v) = K_0(n, m) \quad (4.5)$$

as a consequence of Proposition 2.2.

Hence we can write the kernel K_N^h for a general h of the form (2.8) as

$$K_N^h(n, m) = K_0(n, m) + \int_{\mathbb{R}} du \int_{\mathbb{R}} dv \Phi_{\tau_1}^n(u) (\overline{P}_0 T_{\tau_1, \tau_2} \overline{P}_0 - T_{\tau_1, \tau_2}^h)(u, v) \Psi_{\tau_2}^m(v). \quad (4.6)$$

The conditional probability on the left-hand side of (2.21) is written as a ratio of two Fredholm determinants: using (4.5) we get

$$\begin{aligned} & \mathbf{P} (B_N(t) < h(t) \text{ for } t \in [0, 1] \mid B_N(t) < r \text{ for } t \in [0, 1]) \\ &= \frac{\det(\mathbb{1} - K_N^h)_{\ell^2(\{0, 1, \dots, N-1\})}}{\det(\mathbb{1} - K_0)_{\ell^2(\{0, 1, \dots, N-1\})}} \\ &= \det(\mathbb{1} - (K_N^h - K_0)(\mathbb{1} - K_0)^{-1})_{\ell^2(\{0, 1, \dots, N-1\})} \end{aligned} \quad (4.7)$$

where we used the multiplicative property of the determinant in the second equality. By the cyclic property of the determinant and by using (4.6), (2.22) and (2.17), one obtains

$$\begin{aligned} & \det(\mathbb{1} - (K_N^h - K_0)(\mathbb{1} - K_0)^{-1})_{\ell^2(\{0, 1, \dots, N-1\})} \\ &= \det(\mathbb{1} - (\overline{P}_0 T_{\tau_1, \tau_2} \overline{P}_0 - T_{\tau_1, \tau_2}^h) \Psi_{\tau_2} (\mathbb{1} - K_0)^{-1} \Phi_{\tau_1})_{L^2(\mathbb{R}_-)} \\ &= \det(\mathbb{1} - K_{\tau_1} + T_{\tau_1, \tau_2}^h K_{\tau_2, \tau_1})_{L^2(\mathbb{R}_-)} \end{aligned} \quad (4.8)$$

which completes the proof. \square

To obtain the multipoint distribution of N non-intersecting Brownian bridges conditioned to be under the constant level r in the time interval $[0, 1]$, we specialize the probability that the top path of N non-intersecting Brownian bridges stays below a function h given by (2.8). Namely, for $0 < t_1 < \dots < t_k < 1$, we consider the function

$$h(x) = \begin{cases} r & \text{if } x \neq t_i \text{ for } i = 1, \dots, k \\ h_i & \text{if } x = t_i \end{cases} \quad (4.9)$$

for some $h_i \leq r$ for $i = 1, \dots, k$. Since h given by (4.9) is not in $H^1([0, 1])$, one has to verify that Theorem 2.3 can be used. We prove the following lemma in Section 8.

Lemma 4.2. *Theorem 2.3 remains valid for a function h defined in (4.9).*

The multipoint distribution of N non-intersecting Brownian bridges conditioned to stay below a constant level can be expressed as follows.

Proposition 4.3. *Let h be a function given by (4.9). Then the following conditional probability for the top path of N non-intersecting Brownian bridges can be written in a Fredholm determinant form as*

$$\begin{aligned} \mathbf{P} (B_N(t) < h(t) \text{ for } t \in [0, 1] \mid B_N(t) < r \text{ for } t \in [0, 1]) \\ = \det \left(\mathbb{1} - K_{\tau_1} + \overline{P}_{\eta_1} T_{\tau_1, \tau_2} \overline{P}_{\eta_2} \cdots T_{\tau_{k-1}, \tau_k} \overline{P}_{\eta_k} (T_{\tau_1, \tau_k})^{-1} K_{\tau_1} \right)_{L^2(\mathbb{R}_-)}. \end{aligned} \quad (4.10)$$

Proof. By Lemma 4.2, Theorem 2.3 holds for this choice of h as well. The same steps used in the proof of Theorem 2.4 lead to the result. We just need to replace T_{τ_1, τ_2}^h with the corresponding expression for a general h of the form (4.9), namely with

$$T_{\tau_1, \tau_k}^h = \overline{P}_{\eta_1} T_{\tau_1, \tau_2} \overline{P}_{\eta_2} T_{\tau_2, \tau_3} \cdots T_{\tau_{k-1}, \tau_k} \overline{P}_{\eta_k} \quad (4.11)$$

from (2.10) and using (2.9) with (2.27). \square

Using the result of [8], the Fredholm determinant with the path integral kernel on the right-hand side of (4.10) can be rewritten as in Proposition 4.4 below. Hence the second part of Theorem 2.5 about the gap probabilities follows from Proposition 4.3 and 4.4. This is weaker than proving that K_{ext} is the correlation kernel for N non-intersecting Brownian bridges conditioned to stay below level r . We prove in Section 6 that K_{ext} is actually the correlation kernel.

Proposition 4.4. *For the Fredholm determinant on the right-hand side of (4.10), the following identity hold*

$$\begin{aligned} \det \left(\mathbb{1} - K_{\tau_1} + \overline{P}_{\eta_1} T_{\tau_1, \tau_2} \overline{P}_{\eta_2} \cdots T_{\tau_{k-1}, \tau_k} \overline{P}_{\eta_k} (T_{\tau_1, \tau_k})^{-1} K_{\tau_1} \right)_{L^2(\mathbb{R}_-)} \\ = \det(\mathbb{1} - QK^{\text{ext}})_{L^2(\{\tau_1, \dots, \tau_k\} \times \mathbb{R}_-)} \end{aligned} \quad (4.12)$$

where the extended kernel K^{ext} is given by (2.25) and Q is defined in (2.27).

Proof. Applying formally Theorem 3.3 of [8] with $\mathcal{W}_{\tau_i, \tau_j} = T_{\tau_i, \tau_j}$ and with K_{τ_i} defined by (2.22) would give (4.12). This is however not correct because the operator K_{τ} with kernel given in (2.22) is not a bounded operator and the assumptions of Theorem 3.3 of [8] are not satisfied.

Hence we introduce the following conjugation in order to circumvent this issue. Let

$$\begin{aligned} \overline{\Phi}_{\tau}^n(u) &= e^{\frac{u^2}{C\tau}} \Phi_{\tau}^n(u), \\ \overline{T}_{\tau_i, \tau_j}(u, v) &= e^{-\frac{u^2}{C\tau_i} + \frac{v^2}{C\tau_j}} T_{\tau_i, \tau_j}(u, v), \\ \overline{\Psi}_{\tau}^m(v) &= e^{-\frac{v^2}{C\tau}} \Psi_{\tau}^m(v) \end{aligned} \quad (4.13)$$

where C is a sufficiently large constant which depends on τ_1, \dots, τ_k in such a way the operators $\overline{T}_{\tau_i, \tau_{i+1}}$ are bounded. The condition of boundedness of $\overline{T}_{\tau_i, \tau_{i+1}}$ is

$C > 4(\tau_{i+1} - \tau_i)/\tau_{i+1}$, because then the v^2 term in the exponent has negative sign in $\overline{T}_{\tau_i, \tau_{i+1}}(u, v)$ in (4.13). Further in this case,

$$\int_{\mathbb{R}} dv \overline{T}_{\tau_i, \tau_{i+1}}(u, v) = \sqrt{\frac{C\tau_{i+1}}{C\tau_{i+1} - 4(\tau_{i+1} - \tau_i)}} e^{-\frac{4(\tau_{i+1} - \tau_i)u^2}{C\tau_{i+1}(C\tau_{i+1} - 4(\tau_{i+1} - \tau_i))}} \quad (4.14)$$

which has Gaussian decay in u .

Replacing Φ_τ^n and Ψ_τ^m by $\overline{\Phi}_\tau^n$ and $\overline{\Psi}_\tau^m$ in the definition (2.22) of K_τ , we get the kernel

$$\overline{K}_\tau(u, v) = e^{-\frac{u^2}{C\tau} + \frac{v^2}{C\tau}} K_\tau(u, v). \quad (4.15)$$

Note that the Fredholm determinant on the left-hand side of (4.12) does not change if the operators K and T are replaced by \overline{K} and \overline{T} since it is just a conjugation, i.e.

$$\begin{aligned} & \det(\mathbb{1} - K_{\tau_1} + \overline{P}_{\eta_1} T_{\tau_1, \tau_2} \overline{P}_{\eta_2} \cdots T_{\tau_{k-1}, \tau_k} \overline{P}_{\eta_k} (T_{\tau_1, \tau_k})^{-1} K_{\tau_1})_{L^2(\mathbb{R}_-)} \\ &= \det(\mathbb{1} - \overline{K}_{\tau_1} + \overline{P}_{\eta_1} \overline{T}_{\tau_1, \tau_2} \overline{P}_{\eta_2} \cdots \overline{T}_{\tau_{k-1}, \tau_k} \overline{P}_{\eta_k} (\overline{T}_{\tau_1, \tau_k})^{-1} \overline{K}_{\tau_1})_{L^2(\mathbb{R}_-)}. \end{aligned} \quad (4.16)$$

To apply Theorem 3.3 of [8] (with the minor modification that now the space is $L^2(\mathbb{R}_-)$) with $\mathcal{W}_{\tau_i, \tau_j} = \overline{T}_{\tau_i, \tau_j}$ and with \overline{K}_{τ_i} , we check the three assumptions of the theorem. For Assumption 1, all the operators which appear are bounded. In particular, the boundedness of $\overline{T}_{\tau_i, \tau_j}$ was checked above. The operator \overline{K}_τ is also bounded if $C > 4$ by comparing the Gaussian decay $\Phi_\tau^n(u) \sim e^{-\frac{u^2}{4\tau}}$ with the conjugation (4.13).

Assumption 2 about compatibility is rather clear using the interpretation of T_{τ_i, τ_j} as a Brownian bridge transition kernel and by Proposition 2.2. Since the kernels of all the conjugated operators $\mathcal{W}_{\tau_i, \tau_j}$ and \overline{K}_{τ_i} which appear have Gaussian decay, the trace class properties needed for Assumption 3 are straightforward to check. Hence Theorem 3.3 of [8] can be used which gives (4.12) with K_{ext} replaced by its conjugated version on the right-hand side, but the conjugation can be removed without changing the Fredholm determinant. \square

5 Extension of the Nguyen–Remenik formula

In this section, we extend Proposition 2.1, the Nguyen–Remenik formula for the probability that N non-intersecting Brownian bridges stay below a given function on $[a, b]$ for any fixed $0 < a < b < 1$ to the probability that the Brownian bridges stay below a function h of the form (2.8) on $[0, 1]$.

Proof of Theorem 2.3. First we express the Brownian bridge probability on the right-hand side of (2.7) for the special choice of the function h given in (2.8) in terms of T_{τ_1, τ_2}^h . By introducing the drifted and shifted Brownian bridge $\tilde{b}(\tau) = \widehat{b}(\tau) - (1 + 4\tau)r/\sqrt{2}$, one can write

$$\begin{aligned} & \mathbf{P}_{\widehat{b}(\alpha) = e^A x, \widehat{b}(\beta) = e^B y} \left(\widehat{b}(\tau) \leq \frac{1 + 4\tau}{\sqrt{2}} h \left(\frac{4\tau}{1 + 4\tau} \right) \text{ for } \tau \in [\alpha, \beta] \right) \\ &= \mathbf{P}_{\tilde{b}(\alpha) = e^A x - (1 + 4\alpha)r/\sqrt{2}, \tilde{b}(\beta) = e^B y - (1 + 4\beta)r/\sqrt{2}} \left(\tilde{b}(\tau) \leq 0 \text{ for } \tau \in [\alpha, \beta], \tilde{b}(\tau) \leq \tilde{h}(\tau) \text{ for } \tau \in (\tau_1, \tau_2) \right) \end{aligned} \quad (5.1)$$

where τ_1, τ_2 and \tilde{h} are defined by (2.9). Using the notation (2.15), we can condition on the values of the Brownian bridge $\tilde{b}(\tau)$ at times τ_1 and τ_2 and rewrite the right-hand side of (5.1) as

$$\begin{aligned} & \mathbf{P}_{\substack{\tilde{b}(\alpha)=e^A x - (1+4\alpha)r/\sqrt{2} \\ \tilde{b}(\beta)=e^B y - (1+4\beta)r/\sqrt{2}}} \left(\tilde{b}(\tau) \leq 0 \text{ for } \tau \in [\alpha, \beta], \tilde{b}(\tau) \leq \tilde{h}(\tau) \text{ for } \tau \in (\tau_1, \tau_2) \right) \\ &= \mathbb{1}_{\substack{x \leq \sqrt{2}r \cosh A \\ y \leq \sqrt{2}r \cosh B}} \frac{\int_{-\infty}^0 du \int_{-\infty}^0 dv T_{\alpha, \tau_1} \left(e^A x - \frac{(1+4\alpha)r}{\sqrt{2}}, u \right) T_{\tau_1, \tau_2}^h(u, v) T_{\tau_2, \beta} \left(v, e^B y - \frac{(1+4\beta)r}{\sqrt{2}} \right)}{\phi_{2(\beta-\alpha)}(e^B y - e^A x - 2\sqrt{2}(\beta-\alpha)r)} \end{aligned} \quad (5.2)$$

where the indicator on the right-hand side of (5.2) comes from the condition that the starting point and the endpoint of the Brownian bridge $\tilde{b}(\tau)$ should be below 0 to get a non-zero probability.

Next we compare the operator $\Theta_{A,B}$ to the case of N non-intersecting Brownian bridges not conditioned to stay below any function, that is, the free case. We express $\Theta_{A,B}$ as the operator for the free case minus a remainder. From the representation of $\Theta_{A,B}$ as the solution operator of a boundary value problem given in [49], one obtains that

$$e^{-(B-A)D}(x, y) = e^{(y^2-x^2)/2+B} \frac{\exp\left(-\frac{(e^B y - e^A x)^2}{4(\beta-\alpha)}\right)}{\sqrt{4\pi(\beta-\alpha)}} \quad (5.3)$$

which corresponds to $\Theta_{A,B}$ with the choice $h = \infty$. By defining

$$\begin{aligned} R_{A,B}(x, y) &= e^{(y^2-x^2)/2+B} \frac{\exp\left(-\frac{(e^B y - e^A x)^2}{4(\beta-\alpha)}\right)}{\sqrt{4\pi(\beta-\alpha)}} \\ &\times \left(1 - \frac{\int_{-\infty}^0 du \int_{-\infty}^0 dv T_{\alpha, \tau_1} \left(e^A x - \frac{(1+4\alpha)r}{\sqrt{2}}, u \right) T_{\tau_1, \tau_2}^h(u, v) T_{\tau_2, \beta} \left(v, e^B y - \frac{(1+4\beta)r}{\sqrt{2}} \right)}{\phi_{2(\beta-\alpha)}(e^B y - e^A x - 2\sqrt{2}(\beta-\alpha)r)} \right), \end{aligned} \quad (5.4)$$

we can write the operator identity

$$\Theta_{A,B} = \overline{P}_{\sqrt{2}r \cosh A} (e^{-(B-A)D} - R_{A,B}) \overline{P}_{\sqrt{2}r \cosh B} \quad (5.5)$$

using the notation (4.3).

For the proof of Theorem 2.3, we need to take the limit $a \rightarrow 0$ and $b \rightarrow 1$ in (2.6). Thus we set

$$A = -L, \quad B = L \quad \text{which means} \quad a = \frac{1}{1 + e^{2L}}, \quad b = \frac{e^{2L}}{1 + e^{2L}}. \quad (5.6)$$

We decompose the operator $\Theta_{-L,L}$ as a sum of the operator which corresponds to the free case, the remainder operator and an error term as

$$\Theta_{-L,L} = e^{-2LD} - R_{-L,L} - \Omega_L \quad (5.7)$$

where the error term is

$$\Omega_L = e^{-2LD} - R_{-L,L} - \overline{P}_{\sqrt{2}r \cosh L} (e^{-2LD} - R_{-L,L}) \overline{P}_{\sqrt{2}r \cosh L}. \quad (5.8)$$

Since $K_{\text{Herm},N}$ defined by (2.4) is a projector on a subspace of eigenvectors of D , it commutes with e^{LD} and thus one has $e^{2LD}K_{\text{Herm},N} = (e^{LD}K_{\text{Herm},N})^2$. Using the identity $\det(\mathbb{1} + AB) = \det(\mathbb{1} + BA)$, Proposition 2.1 can be written as

$$\begin{aligned} \mathbf{P}(B_N(t) < h(t) \text{ for } t \in [0, 1]) \\ = \lim_{L \rightarrow \infty} \det(\mathbb{1} - K_{\text{Herm},N} + e^{LD}K_{\text{Herm},N}\Theta_{-L,L}e^{LD}K_{\text{Herm},N})_{L^2(\mathbb{R})}. \end{aligned} \quad (5.9)$$

Next we use the decomposition (5.7) of $\Theta_{-L,L}$. We prove the following lemma in Section 8.

Lemma 5.1. *The error term $\tilde{\Omega}_L = e^{LD}K_{\text{Herm},N}\Omega_L e^{LD}K_{\text{Herm},N}$ goes to 0 in trace norm as $L \rightarrow \infty$.*

Thus, by Lemma 5.1, in the $L \rightarrow \infty$ limit, we can neglect the error term in the Fredholm determinant on the right-hand side of (5.9) (use for example Lemma 4 in Chap. XIII.17 of [51]). Consequently one obtains

$$\mathbf{P}(B_N(t) < h(t) \text{ for } t \in [0, 1]) = \lim_{L \rightarrow \infty} \det(\mathbb{1} - e^{LD}K_{\text{Herm},N}R_{-L,L}e^{LD}K_{\text{Herm},N})_{L^2(\mathbb{R})}. \quad (5.10)$$

By using the cyclic property of the Fredholm determinant again and by the definition (2.4) of $K_{\text{Herm},N}$, one can write

$$\begin{aligned} \det(\mathbb{1} - e^{LD}K_{\text{Herm},N}R_{-L,L}e^{LD}K_{\text{Herm},N})_{L^2(\mathbb{R})} &= \det(\mathbb{1} - R_{-L,L}e^{2LD}K_{\text{Herm},N})_{L^2(\mathbb{R})} \\ &= \det(\mathbb{1} - \varphi^*R_{-L,L}e^{2LD}\varphi)_{L^2(\{0,1,\dots,N-1\})} \end{aligned} \quad (5.11)$$

where $\varphi : L^2(\{0, 1, \dots, N-1\}) \rightarrow L^2(\mathbb{R})$ is an operator defined by

$$(\varphi f)(x) = \sum_{n=0}^{N-1} \varphi_n(x) f(n). \quad (5.12)$$

The rest of the proof of Theorem 2.3 now follows from the Proposition 5.2 below about the equality of kernels, since the prefactor in front of K_N^h on the right-hand side of (5.13) is just a conjugation which can be removed without changing the value of the corresponding Fredholm determinant. \square

Proposition 5.2. *Let $h \in H^1([a, b])$ be a function which satisfies (2.8). Then for any N and L , one has*

$$(\varphi^*R_{-L,L}e^{2LD}\varphi)_{n,m} = \sqrt{\frac{m!}{n!} \frac{2^n}{2^m} \frac{e^{Lm}}{e^{Ln}}} K_N^h(n, m) \quad (5.13)$$

for all $n, m = 0, 1, \dots, N-1$.

Remark 5.3. Notice that K_N^h on the right-hand side of (5.13) does not depend on L , hence up to the conjugation neither the left-hand side does, which is a priori not at all obvious. This fact shows that the $L \rightarrow \infty$ limit of the right-hand side of (2.6) with (5.6) is obtained up to conjugation by simply removing the projections from $\Theta_{-L,L}$ in (5.5).

Proof of Proposition 5.2. We use the following two integral representations of the harmonic oscillator functions:

$$\varphi_n(x) = \sqrt{\frac{2^n}{n!}} \pi^{1/4} e^{x^2/2} \frac{1}{\pi i} \int_{i\mathbb{R}} dw e^{w^2 - 2wx} w^n, \quad (5.14)$$

$$\varphi_n(x) = \sqrt{\frac{n!}{2^n}} \pi^{-1/4} e^{-x^2/2} \frac{1}{2\pi i} \oint_{\Gamma_0} dz \frac{e^{-z^2 + 2zx}}{z^{n+1}} \quad (5.15)$$

where the integration contour Γ_0 is a small circle around 0 with counterclockwise orientation. To compute the kernel on the left-hand side of (5.13), we substitute (2.15) in the double integral in the definition (5.4) of $R_{-L,L}$. In this way, we get the terms

$$\begin{aligned}
Q_0(u, v, X, Y) &= \phi_{2\tau_1 - e^{-2L}/2}(u - X) \phi_{2(\tau_2 - \tau_1)}(v - u) \phi_{e^{2L}/2 - 2\tau_2}(Y - v), \\
Q_1(u, v, X, Y) &= -\phi_{2\tau_1 - e^{-2L}/2}(u - X) T_{\tau_1, \tau_2}^h(u, v) \phi_{e^{2L}/2 - 2\tau_2}(Y - v), \\
Q_2(u, v, X, Y) &= \phi_{2\tau_1 - e^{-2L}/2}(u - X) T_{\tau_1, \tau_2}^h(u, v) \phi_{e^{2L}/2 - 2\tau_2}(Y + v), \\
Q_3(u, v, X, Y) &= \phi_{2\tau_1 - e^{-2L}/2}(u + X) T_{\tau_1, \tau_2}^h(u, v) \phi_{e^{2L}/2 - 2\tau_2}(Y - v), \\
Q_4(u, v, X, Y) &= -\phi_{2\tau_1 - e^{-2L}/2}(u + X) T_{\tau_1, \tau_2}^h(u, v) \phi_{e^{2L}/2 - 2\tau_2}(Y + v).
\end{aligned} \tag{5.16}$$

By simplifying the exponential prefactor with the denominator on the right-hand side of (5.4), one gets

$$\begin{aligned}
R_{-L,L}(x, y) &= e^{(y^2 - x^2)/2 + L - \sqrt{2}r(e^L y - e^{-L}x) + \sinh(2L)r^2} \\
&\times \int_{\mathbb{R}} du \int_{\mathbb{R}} dv \sum_{j=0}^4 Q_j \left(u, v, e^{-L}x - \frac{(1 + e^{-2L})r}{\sqrt{2}}, e^L y - \frac{(1 + e^{2L})r}{\sqrt{2}} \right). \tag{5.17}
\end{aligned}$$

Note that one has changed the domain of integration for u and v to \mathbb{R} because of the term which corresponds to Q_0 . In the terms which correspond to Q_1 – Q_4 , $T_{\tau_1, \tau_2}^h(u, v)$ is 0 if u or v is positive by (2.10). With these notations, the kernel on the left-hand side of (5.13) using both representations (5.14)–(5.15) of the harmonic oscillator functions φ_n is equal to

$$\begin{aligned}
&(\varphi^* R_{-L,L} e^{2LD} \varphi)_{n,m} \\
&= \int_{\mathbb{R}} dx \int_{\mathbb{R}} dy \varphi_n(x) R_{-L,L}(x, y) e^{2Lm} \varphi_m(y) \\
&= \sqrt{\frac{m! 2^n}{n! 2^m}} \frac{2}{(2\pi i)^2} \int_{\mathbb{R}} dx \int_{\mathbb{R}} dy \int_{i\mathbb{R}} dw \oint_{\Gamma_0} dz \int_{\mathbb{R}} du \int_{\mathbb{R}} dv e^{w^2 - 2wx} w^n e^{-\sqrt{2}r(e^L y - e^{-L}x)} \\
&\times e^{\sinh(2L)r^2 + L} \sum_{j=0}^4 Q_j \left(u, v, e^{-L}x - \frac{(1 + e^{-2L})r}{\sqrt{2}}, e^L y - \frac{(1 + e^{2L})r}{\sqrt{2}} \right) e^{2Lm} \frac{e^{-z^2 + 2zy}}{z^{m+1}}. \tag{5.18}
\end{aligned}$$

Doing the change of variables

$$X = e^{-L}x - \frac{(1 + e^{-2L})r}{\sqrt{2}}, \quad Y = e^L y - \frac{(1 + e^{2L})r}{\sqrt{2}}, \quad W = e^L w, \quad Z = e^{-L}z, \tag{5.19}$$

one obtains

$$\begin{aligned}
(5.18) &= \sqrt{\frac{m! 2^n}{n! 2^m}} \frac{e^{Lm}}{e^{Ln}} \frac{2}{(2\pi i)^2} \int_{\mathbb{R}} dX \int_{\mathbb{R}} dY \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \int_{\mathbb{R}} du \int_{\mathbb{R}} dv e^{W^2 e^{-2L}} \\
&\times e^{-2W(X + \frac{(1 + e^{-2L})r}{\sqrt{2}})} W^n e^{-\sqrt{2}r(Y - X) - \sinh(2L)r^2} \sum_{j=0}^4 Q_j(u, v, X, Y) \frac{e^{-Z^2 e^{2L} + 2Z(Y + \frac{(1 + e^{2L})r}{\sqrt{2}})}}{Z^{m+1}}. \tag{5.20}
\end{aligned}$$

The integral with respect to X and Y in (5.20) can be computed, since they are

Gaussian integrals. One has

$$\begin{aligned} \int_{\mathbb{R}} dX \phi_{2\tau_1 - e^{-2L}/2}(u \pm X) e^{-2WX + \sqrt{2r}X} &= e^{(4\tau_1 - e^{-2L})(\sqrt{2r} - 2W)^2/4 \mp (\sqrt{2r} - 2W)u}, \\ \int_{\mathbb{R}} dY \phi_{e^{2L}/2 - 2\tau_2}(v \pm Y) e^{2ZY - \sqrt{2r}Y} &= e^{(e^{2L} - 4\tau_2)(\sqrt{2r} - 2Z)^2/4 \pm (\sqrt{2r} - 2Z)v}. \end{aligned} \quad (5.21)$$

Then putting the definitions (5.16) into (5.20), using (5.21) and the notation (2.13), one gets

$$\begin{aligned} &(\varphi^* R_{-L,L} e^{2LD} \varphi)_{n,m} \\ &= \sqrt{\frac{m! 2^n e^{Lm}}{n! 2^m e^{Ln}}} \frac{2}{(2\pi i)^2} \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \int_{\mathbb{R}} du \int_{\mathbb{R}} dv \frac{W^n e^{\tau_1(\sqrt{2r} - 2W)^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau_2(\sqrt{2r} - 2Z)^2 - \sqrt{2r}Z}} \\ &\quad \times (f_W(u) \phi_{2(\tau_2 - \tau_1)}(v - u) g_Z(v) - (f_W(u) - f_W(-u)) T_{\tau_1, \tau_2}^h(u, v) (g_Z(v) - g_Z(-v))). \end{aligned} \quad (5.22)$$

By using (8.3) of Lemma 8.1, one can see that the integral of the first term on the right-hand side of (5.22) up to conjugation is

$$\begin{aligned} &\frac{2}{(2\pi i)^2} \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \int_{\mathbb{R}} du \int_{\mathbb{R}} dv \frac{W^n e^{\tau_1(\sqrt{2r} - 2W)^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau_2(\sqrt{2r} - 2Z)^2 - \sqrt{2r}Z}} f_W(u) \phi_{2(\tau_2 - \tau_1)}(v - u) g_Z(v) \\ &= \mathbb{1}(n, m). \end{aligned} \quad (5.23)$$

Comparing (5.22) and (5.23) with (2.20) and (2.11)–(2.12) completes the proof. \square

6 Direct derivation of the correlation kernel

In this section, we prove Theorem 2.5 where the correlation kernel of N non-intersecting Brownian bridges conditioned to stay below a constant level is determined. The direct proof of the correlation kernel follows the line of [56] where the correlation kernel for non-intersecting Brownian bridges were computed without further conditioning.

Let us define the functions

$$\tilde{\Phi}_t^i(x) = \frac{1}{2^n \sqrt{\pi}} \left(\frac{1-t}{t} \right)^{\frac{i+1}{2}} \left(e^{-\frac{x^2}{2t}} H_i \left(\frac{x}{\sqrt{2t(1-t)}} \right) - e^{-\frac{(2r-x)^2}{2t}} H_i \left(\frac{2r-x}{\sqrt{2t(1-t)}} \right) \right), \quad (6.1)$$

$$\tilde{\Psi}_t^j(x) = \frac{1}{j!} \left(\frac{t}{1-t} \right)^{\frac{j}{2}} \left(e^{-\frac{x^2}{2(1-t)}} H_j \left(\frac{x}{\sqrt{2t(1-t)}} \right) - e^{-\frac{(2r-x)^2}{2(1-t)}} H_j \left(\frac{2r-x}{\sqrt{2t(1-t)}} \right) \right) \quad (6.2)$$

for $t \in [0, 1]$, $x < r$ and i integer where H_i is the i th Hermite polynomial. For any $0 \leq t_1 < t_2 \leq 1$ and $x, y < r$, let

$$\tilde{T}_{t_1, t_2}(x, y) = \frac{1}{\sqrt{2\pi(t_2 - t_1)}} \left(e^{-\frac{(x-y)^2}{2(t_2 - t_1)}} - e^{-\frac{(2r-x-y)^2}{2(t_2 - t_1)}} \right) \quad (6.3)$$

be the free evolution kernel of a Brownian motion below level r .

Proposition 6.1. *Let $0 < t_1 < \dots < t_k < 1$ be times and $x_1^{(l)} < \dots < x_N^{(l)}$ be positions, $l = 1, \dots, k$. Then the joint density of N non-intersecting Brownian bridges conditioned to stay below level r for $[0, 1]$ at times t_i and positions $x_j^{(l)}$ is proportional to*

$$\det \left(\tilde{\Phi}_{t_1}^{i-1}(x_j^{(1)}) \right)_{i,j=1}^N \prod_{l=1}^{k-1} \det \left(\tilde{T}_{t_l, t_{l+1}}(x_i^{(l)}, x_j^{(l+1)}) \right)_{i,j=1}^N \det \left(\tilde{\Psi}_{t_k}^{i-1}(x_j^{(k)}) \right)_{i,j=1}^N. \quad (6.4)$$

Proof of Proposition 6.1. We follow the usual strategy to get N non-intersecting Brownian bridges which start and end at 0. We let them start and end at positions $-\varepsilon, -2\varepsilon, \dots, -N\varepsilon$, and then we will let $\varepsilon \rightarrow 0$. By a Karlin–McGregor type formula, their joint density is given by

$$\det \left(\tilde{T}_{0,t_1}(-i\varepsilon, x_j^{(1)}) \right)_{i,j=1}^N \prod_{l=1}^{k-1} \det \left(\tilde{T}_{t_l, t_{l+1}}(x_i^{(l)}, x_j^{(l+1)}) \right)_{i,j=1}^N \det \left(\tilde{T}_{t_k, 1}(x_i^{(k)}, -j\varepsilon) \right)_{i,j=1}^N. \quad (6.5)$$

The product of $k - 1$ determinants in the middle in (6.4) and in (6.5) is the same. The general (i, j) entry of the first determinant in (6.5) is

$$\begin{aligned} \tilde{T}_{0,t_1}(-i\varepsilon, x_j^{(1)}) &= \frac{1}{\sqrt{2\pi t_1}} \left(e^{-\frac{(x_j^{(1)} + i\varepsilon)^2}{2t_1}} - e^{-\frac{(2r - x_j^{(1)} + i\varepsilon)^2}{2t_1}} \right) \\ &= \frac{e^{-\frac{i^2 \varepsilon^2}{2t_1}}}{\sqrt{2\pi t_1}} e^{-\frac{(x_j^{(1)})^2}{2t_1}} \left(1 - \frac{i\varepsilon x_j^{(1)}}{t_1} + \frac{1}{2} \frac{i^2 \varepsilon^2 (x_j^{(1)})^2}{t_1^2} \pm \dots \right) \\ &\quad - \frac{e^{-\frac{i^2 \varepsilon^2}{2t_1}}}{\sqrt{2\pi t_1}} e^{-\frac{(2r - x_j^{(1)})^2}{2t_1}} \left(1 - \frac{i\varepsilon(2r - x_j^{(1)})}{t_1} + \frac{1}{2} \frac{i^2 \varepsilon^2 (2r - x_j^{(1)})^2}{t_1^2} \pm \dots \right) \end{aligned} \quad (6.6)$$

where we used Taylor expansion in the last step.

By elementary row operations with the matrix in the first determinant in (6.5), one obtains

$$\begin{aligned} \det \left(\tilde{T}_{0,t_1}(-i\varepsilon, x_j^{(1)}) \right)_{i,j=1}^N \\ = c(\varepsilon) \left[\det \left(e^{-\frac{(x_j^{(1)})^2}{2t_1}} (x_j^{(1)})^{i-1} - e^{-\frac{(2r - x_j^{(1)})^2}{2t_1}} (2r - x_j^{(1)})^{i-1} \right)_{i,j=1}^N + \mathcal{O}(\varepsilon) \right] \end{aligned} \quad (6.7)$$

where $c(\varepsilon)$ is a constant which does not depend on the $x_j^{(1)}$ variables. (Notice that $c(\varepsilon)$ depends on ε asymptotically as $\varepsilon^{N(N-1)/2}$, but it is unimportant for the proposition.) The determinant on the right-hand side of (6.7) is already independent of ε , hence it is also the factor which appears in the $\varepsilon \rightarrow 0$ limit. By further row manipulations in the determinant on the right-hand side of (6.7), one can turn the monomials $(x_j^{(1)})^{i-1}$ and $(2r - x_j^{(1)})^{i-1}$ into any polynomials of degree $i - 1$, but with the same polynomial for both terms. In particular, by choosing the $(i - 1)$ st Hermite polynomial with rescaled argument $x \mapsto H_{i-1}(x/\sqrt{2t(1-t)})$, one gets that the determinant on the right-hand side of (6.7) is proportional to the first factor in (6.4). The argument for the last determinant is the same, hence the proof is complete. \square

Proposition 6.2. *With the relation (2.9) between the variables t_1, t_2 and τ_1, τ_2 and with (2.24) between x_i and u_i , one has the following equality of the conjugated functions*

$$\Phi_\tau^n(u) = e^{-\frac{r^2}{2} - \frac{(x-r)^2}{2(1-t)}} \tilde{\Phi}_t^n(x), \quad (6.8)$$

$$\Psi_\tau^n(u) = e^{\frac{r^2}{2} + \frac{(x-r)^2}{2(1-t)}} \tilde{\Psi}_t^n(x), \quad (6.9)$$

$$T_{\tau_1, \tau_2}(u_1, u_2) = \sqrt{2(1-t_1)(1-t_2)} e^{\frac{(x_1-r)^2}{2(1-t_1)} - \frac{(x_2-r)^2}{2(1-t_2)}} \tilde{T}_{t_1, t_2}(x_1, x_2). \quad (6.10)$$

With Proposition 6.2, proving Theorem 2.5 is easy.

Proof of Theorem 2.5. The correlation kernel can be directly obtained from the general formula given in [38], since the joint density of N non-intersecting Brownian bridges conditioned to stay below level r is given by (6.4) where the functions which appear in the determinants satisfy

$$\int_{-\infty}^r dx \frac{\tilde{\Phi}_{t_1}^i(x)}{2^{1/4}\sqrt{1-t_1}} \tilde{T}_{t_1, t_2}(x, y) = \frac{\tilde{\Phi}_{t_2}^i(y)}{2^{1/4}\sqrt{1-t_2}}, \quad (6.11)$$

$$\int_{-\infty}^r dy \tilde{T}_{t_1, t_2}(x, y) \frac{\tilde{\Psi}_{t_2}^j(y)}{2^{1/4}\sqrt{1-t_2}} = \frac{\tilde{\Psi}_{t_1}^j(x)}{2^{1/4}\sqrt{1-t_1}}, \quad (6.12)$$

$$\int_{-\infty}^r dx \frac{\tilde{\Phi}_t^i(x)}{2^{1/4}\sqrt{1-t_1}} \frac{\tilde{\Psi}_t^j(x)}{2^{1/4}\sqrt{1-t_2}} = (\mathbb{1} - K_0)(i, j) \quad (6.13)$$

which is a direct consequence of Proposition 2.2 knowing the relations proved in Proposition 6.2. Hence the extended kernel can be written for $x_1, x_2 \leq r$ as

$$K_{\text{ext}}(t_1, x_1; t_2, x_2) = -\mathbb{1}_{\tau_1 < \tau_2} \tilde{T}_{t_1, t_2}(x_1, x_2) + \sum_{n, m=0}^{N-1} \frac{\tilde{\Psi}_{t_1}^n(x_1)}{2^{1/4}\sqrt{1-t_1}} (\mathbb{1} - K_0)^{-1}(n, m) \frac{\tilde{\Phi}_{t_2}^m(x_2)}{2^{1/4}\sqrt{1-t_2}}. \quad (6.14)$$

Due to Proposition 6.2, one can write the correlation kernel in terms of the variables τ_i, u_i according to (2.24) since the extra factor $1/\sqrt{2(1-t_1)(1-t_2)}$ is the volume element. This proves that the correlation kernel in terms of the the natural variables τ_i, u_i is given by (2.25). It is also consistent with the definition (2.23) of the correlation kernel, which finishes the proof. \square

For the proof of Proposition 6.2, the following representations are useful.

Proposition 6.3. *The functions Φ_τ^n and Ψ_τ^m admit the following representations in terms of Hermite polynomials*

$$\Phi_\tau^n(u) = \frac{1}{2^{2n+1}\tau^{\frac{n+1}{2}}\sqrt{\pi}} e^{-\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right)^2} H_n\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right) e^{2\tau r^2 + \sqrt{2}ru} - (u \leftrightarrow -u) \quad (6.15)$$

and

$$\Psi_\tau^m(u) = \frac{(2\sqrt{\tau})^m}{m!} H_m\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right) e^{-2\tau r^2 - \sqrt{2}ru} - (u \leftrightarrow -u). \quad (6.16)$$

The notation $(u \leftrightarrow -u)$ means that we have the same term with u replaced by $-u$.

Proof of Proposition 6.3. Expanding the exponent of (2.11) and doing the change of variables $2\sqrt{\tau}W = w$, one gets

$$\begin{aligned} \Phi_\tau^n(u) &= \frac{1}{\pi i} \int_{i\mathbb{R}} dW W^n e^{4\tau W^2 - (1+4\tau)\sqrt{2}rW + 2\tau r^2} e^{-2uW + \sqrt{2}ru} - (u \leftrightarrow -u) \\ &= \frac{1}{(2\sqrt{\tau})^{n+1}} \frac{1}{\pi i} \int_{i\mathbb{R}} dw w^n e^{w^2 - 2\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right)w + 2\tau r^2 + \sqrt{2}ru} - (u \leftrightarrow -u). \end{aligned} \quad (6.17)$$

By (2.3) and (5.14), one has

$$\frac{1}{\pi i} \int_{i\mathbb{R}} dw w^n e^{w^2 - 2xw} = \frac{1}{2^n \sqrt{\pi}} e^{-x^2} H_n(x). \quad (6.18)$$

Then the integral on the right-hand side of (6.17) can be expressed with Hermite polynomials using (6.18) which immediately yields (6.15).

The representation (6.16) is proved similarly. With the change of variables $2\sqrt{\tau}Z = z$ in (2.12), one obtains

$$\begin{aligned} \Psi_\tau^m(u) &= \frac{1}{2\pi i} \oint_{\Gamma_0} dZ Z^{-(m+1)} e^{-4\tau Z^2 + (1+4\tau)\sqrt{2r}Z - 2\tau r^2} e^{2uZ - \sqrt{2r}u} - (u \leftrightarrow -u) \\ &= (2\sqrt{\tau})^m \frac{1}{2\pi i} \oint_{\Gamma_0} dz z^{-(m+1)} e^{-z^2 + 2\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right)z - 2\tau r^2 - \sqrt{2r}u} - (u \leftrightarrow -u). \end{aligned} \quad (6.19)$$

Using (2.3) and the representation (5.15) yields

$$\frac{1}{2\pi i} \oint_{\Gamma_0} dz \frac{e^{-z^2 + 2zx}}{z^{n+1}} = \frac{1}{n!} H_n(x) \quad (6.20)$$

Then the two integrals on the right-hand side of (6.19) are rewritten with (6.20) which proves (6.16). \square

Proof of Proposition 6.2. We proceed by direct computation. To prove (6.8), one can first rearrange the right-hand side to get

$$\begin{aligned} &e^{-\frac{r^2}{2} - \frac{(x-r)^2}{2(1-t)}} \tilde{\Phi}_t^n(x) \\ &= C_{n,t} e^{-\frac{r^2}{2} - \frac{(x-r)^2}{2(1-t)}} \left(e^{-\frac{x^2}{2t}} H_n\left(\frac{x}{\sqrt{2t(1-t)}}\right) - e^{-\frac{(2r-x)^2}{2t}} H_n\left(\frac{2r-x}{\sqrt{2t(1-t)}}\right) \right) \\ &= C_{n,t} \left(e^{-\frac{x^2}{2t(1-t)}} H_n\left(\frac{x}{\sqrt{2t(1-t)}}\right) e^{\frac{rx}{1-t} + \frac{(t-2)r^2}{2(1-t)}} - e^{-\frac{(2r-x)^2}{2t(1-t)}} H_n\left(\frac{2r-x}{\sqrt{2t(1-t)}}\right) e^{-\frac{rx}{1-t} + \frac{(t+2)r^2}{2(1-t)}} \right) \end{aligned} \quad (6.21)$$

where $C_{n,t} = \frac{1}{2^n \sqrt{\pi}} \left(\frac{1-t}{t}\right)^{\frac{n+1}{2}}$.

Next we rewrite the right-hand side of (6.21) in terms of the variables τ and u . Using (2.24), one has

$$\frac{x}{\sqrt{2t(1-t)}} = \frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}, \quad \frac{2r-x}{\sqrt{2t(1-t)}} = \frac{(1+4\tau)r}{2\sqrt{2\tau}} - \frac{u}{2\sqrt{\tau}} \quad (6.22)$$

and

$$\pm \frac{rx}{1-t} + \frac{(t \mp 2)r^2}{2(1-t)} = 2\tau r^2 \pm \sqrt{2r}u. \quad (6.23)$$

By substituting (2.9), (6.22) and (6.23) on the right-hand side of (6.21), one exactly gets the representation (6.15), which proves (6.8).

Similarly,

$$\begin{aligned}
& e^{\frac{r^2}{2} + \frac{(x-r)^2}{2(1-t)}} \tilde{\Psi}_t^n(x) \\
&= \frac{1}{n!} \left(\frac{t}{1-t} \right)^{\frac{n}{2}} e^{\frac{r^2}{2} + \frac{(x-r)^2}{2(1-t)}} \left(e^{-\frac{x^2}{2(1-t)}} H_n \left(\frac{x}{\sqrt{2t(1-t)}} \right) - e^{-\frac{(2r-x)^2}{2(1-t)}} H_n \left(\frac{2r-x}{\sqrt{2t(1-t)}} \right) \right) \\
&= \frac{1}{n!} \left(\frac{t}{1-t} \right)^{\frac{n}{2}} \left(H_n \left(\frac{x}{\sqrt{2t(1-t)}} \right) e^{-\frac{rx}{1-t} - \frac{(t-2)r^2}{2(1-t)}} - H_n \left(\frac{2r-x}{\sqrt{2t(1-t)}} \right) e^{\frac{rx}{1-t} - \frac{(t+2)r^2}{2(1-t)}} \right).
\end{aligned} \tag{6.24}$$

Then by (2.9), (6.22) and (6.23), one can write the right-hand side of (6.24) in terms of the variables τ and u . Comparing this with (6.16), (6.9) is proved.

Finally, by definition (6.3) and by using (2.24),

$$\begin{aligned}
& \sqrt{2(1-t_1)(1-t_2)} e^{\frac{(x_1-r)^2}{2(1-t_1)} - \frac{(x_2-r)^2}{2(1-t_2)}} \tilde{T}_{t_1, t_2}(x_1, x_2) \\
&= \sqrt{\frac{(1-t_1)(1-t_2)}{\pi(t_2-t_1)}} e^{(1-t_1)u_1^2 - (1-t_2)u_2^2} \left(e^{-\frac{((1-t_1)u_1 - (1-t_2)u_2)^2}{t_2-t_1}} - e^{-\frac{((1-t_1)u_1 + (1-t_2)u_2)^2}{t_2-t_1}} \right) \\
&= \sqrt{\frac{(1-t_1)(1-t_2)}{\pi(t_2-t_1)}} \left(e^{-\frac{(1-t_1)(1-t_2)}{t_2-t_1}(u_1-u_2)^2} - e^{-\frac{(1-t_1)(1-t_2)}{t_2-t_1}(u_1+u_2)^2} \right).
\end{aligned} \tag{6.25}$$

By noticing that

$$4(\tau_2 - \tau_1) = \frac{t_2 - t_1}{(1-t_1)(1-t_2)}, \tag{6.26}$$

the proof is complete. \square

7 Asymptotics

This section is devoted to the proof of Theorem 2.6. To this end, we start with a lemma which contains the asymptotic properties of the harmonic oscillator functions which are necessary for further proofs.

Lemma 7.1. *For the n th harmonic oscillator function φ_n , one has*

$$\lim_{n \rightarrow \infty} 2^{-1/4} n^{1/12} \varphi_n \left(\sqrt{2n} + \frac{sn^{-1/6}}{\sqrt{2}} \right) = \text{Ai}(s) \tag{7.1}$$

uniformly on any compact subset of \mathbb{R} for s . Further, for any $c > 0$, there are s_0 and n_0 such that for any $s \geq s_0$ and $n \geq n_0$,

$$\left| 2^{-1/4} n^{1/12} \varphi_n \left(\sqrt{2n} + \frac{sn^{-1/6}}{\sqrt{2}} \right) \right| \leq e^{-cs}. \tag{7.2}$$

There is a universal constant C such that for any $n \geq 1$,

$$\sup_{x \in \mathbb{R}} \left| 2^{-1/4} n^{1/12} \varphi_n(x) \right| \leq C. \tag{7.3}$$

Proof of Lemma 7.1. The formula (7.1) is well-known, see e.g. [52]. It can be also seen in Lemma 5.8 of [31], while (7.2) can be derived directly from Lemma 5.9 of [31]. Using Definition 5.7 of [31], one has

$$\begin{aligned}
\alpha_n(0, s) &= n^{1/3} e^{3n/2+sn^{1/3}} \frac{1}{2\pi i} \int_{i\mathbb{R}} dw e^{nw^2/2+(2n+sn^{1/3})w} (-w)^n \\
&= 2^{n-1/2} n^{-n/2-1/6} e^{3n/2+sn^{1/3}} \frac{1}{\pi i} \int_{i\mathbb{R}} dW e^{W^2-2(\sqrt{2n}+sn^{1/3}/\sqrt{2})W} W^n \\
&= 2^{n-1/2} n^{-n/2-1/6} e^{3n/2+sn^{1/3}} \sqrt{\frac{n!}{2^n}} \pi^{-1/4} e^{(\sqrt{2n}+sn^{1/3}/\sqrt{2})^2/2} \varphi_n \left(\sqrt{2n} + \frac{sn^{-1/6}}{\sqrt{2}} \right) \\
&= 2^{-1/4} n^{1/12} \varphi_n \left(\sqrt{2n} + \frac{sn^{-1/6}}{\sqrt{2}} \right)
\end{aligned} \tag{7.4}$$

with the change of variables $W = -\sqrt{n/2}w$ in the second equality, with the use of (5.14) in the third and by Stirling's formula in the last one. Hence the results of [31] apply and one gets (7.1) and (7.2) with $c = 1$. By inspecting the proof of Lemma 5.9 in [31], one can realize that the terms which appear in the integral representation of $\beta_t(r, s)$ in (5.40)–(5.42) of [31] are bounded by a large constant times $\exp(-s^{3/2})$ which is less than e^{-cs} for any c if s is large enough. By the last remark after (5.47) in the proof of Proposition 5.9 in [31], one gets that the same bound applies for $\alpha_t(r, s)$ as required.

Finally, (7.3) is an easy consequence of the detailed bound obtained in [44], see also (A.54) of [28] where $p_k(x) = H_k(x)$ (except for a small typo: $2^{2/k}$ should be $2^{k/2}$). By replacing $x/\sqrt{2N}$ by x and by (2.3), one exactly gets (7.3). \square

Then one has the following limits as $N \rightarrow \infty$ and bounds for the functions which appear in the kernel of N non-intersecting Brownian bridges.

Proposition 7.2. *Consider the scaling*

$$u = \frac{UN^{-1/6}}{\sqrt{2}}, \quad n = N - \xi N^{1/3}, \quad m = N - \zeta N^{1/3} \tag{7.5}$$

as well as (2.32) for t_i, r, x_i . Then as $N \rightarrow \infty$, it holds

$$\lim_{N \rightarrow \infty} \left(\frac{2}{N} \right)^{\frac{n}{2}} \frac{N^{-1/6}}{\sqrt{2}} e^{N + \frac{RN^{1/3}}{2}} \Phi_\tau^n(u) = \widehat{\Phi}_T^\xi(U), \tag{7.6}$$

$$\lim_{N \rightarrow \infty} \left(\frac{N}{2} \right)^{\frac{m}{2}} N^{1/3} e^{-N - \frac{RN^{1/3}}{2}} \Psi_\tau^m(u) = \widehat{\Psi}_T^\zeta(U). \tag{7.7}$$

For the rescaled kernel the convergence

$$\lim_{N \rightarrow \infty} \left(\frac{2}{N} \right)^{\frac{n-m}{2}} N^{1/3} K_0(n, m) = \widehat{K}_0(\xi, \zeta) \tag{7.8}$$

holds.

Furthermore, for U, V in a compact interval and for any $c > 0$, there is a $C = C(c)$

such that the bounds

$$\left| \left(\frac{2}{N} \right)^{\frac{n}{2}} \frac{N^{-1/6}}{\sqrt{2}} e^{N + \frac{RN^{1/3}}{2}} \Phi_{\tau}^n(u) \right| \leq C e^{-c\xi}, \quad (7.9)$$

$$\left| \left(\frac{N}{2} \right)^{\frac{m}{2}} N^{1/3} e^{-N - \frac{RN^{1/3}}{2}} \Psi_{\tau}^m(u) \right| \leq C e^{-c\zeta}, \quad (7.10)$$

$$\left| \left(\frac{2}{N} \right)^{\frac{n-m}{2}} N^{1/3} K_0(n, m) \right| \leq C e^{-c(\xi+\zeta)} \quad (7.11)$$

hold for ξ and ζ uniformly in $[0, N^{2/3}]$.

Theorem 2.6 is now an easy consequence of Proposition 7.2.

Proof of Theorem 2.6. It is enough to prove (2.33) in terms of the variables τ_i, u_i , that is,

$$\lim_{N \rightarrow \infty} \frac{N^{-1/6}}{\sqrt{2}} K^{\text{ext}}(\tau_1, u_1; \tau_2, u_2) = \widehat{K}^{\text{ext}}(T_1, U_1; T_2, U_2). \quad (7.12)$$

First of all, (2.32), (7.5) and Brownian scaling give

$$\frac{N^{-1/6}}{\sqrt{2}} T_{\tau_1, \tau_2}(u_1, u_2) = T_{T_1, T_2}(U_1, U_2). \quad (7.13)$$

Further, by the uniform decay properties (7.9)–(7.11) in ξ and ζ , in the sum for n and m in (2.25), dominated convergence can be used. We thus replace the rescaled functions $\Psi_{\tau_1}^n(u_1)$, $\Phi_{\tau_2}^m(u_2)$ and the rescaled resolvent of the kernel K_0 in (2.25) according to (7.6)–(7.8). The conjugations and prefactors exactly cancels. We turn the Riemann sum into an integral and by dominated convergence, we obtain (2.33). \square

Proof of Theorem 2.9. First note that the left-hand side of (2.41) is equal to the left-hand side of (2.21) for

$$h(t) = \sqrt{N} + \frac{H(T)N^{-1/6}}{2}, \quad t = \frac{1 + TN^{-1/3}}{2}. \quad (7.14)$$

Thus we can apply Theorem 2.4. Under the same scaling as in the proof of Theorem 2.6, in particular with $u_i = \frac{U_i N^{-1/6}}{\sqrt{2}}$ we get that

$$\frac{N^{-1/6}}{\sqrt{2}} K_{\tau_1, \tau_2}(u_1, u_2) \rightarrow K_{T_1, T_2}(U_1, U_2). \quad (7.15)$$

Further, inserting $u_i = \frac{U_i N^{-1/6}}{\sqrt{2}}$ into T^h given by (2.10) and by using Brownian scaling, we obtain

$$\lim_{N \rightarrow \infty} \frac{N^{-1/6}}{\sqrt{2}} T_{\tau_1, \tau_2}^h(u_1, u_2) = T_{T_1, T_2}^{H-R}(U_1, U_2). \quad (7.16)$$

The convergence of the Fredholm determinant follows from the bounds of Proposition 7.2, in the same way as for the convergence of the Fredholm determinant of Theorem 2.6. \square

Proof of Proposition 7.2. In the representation of Φ in terms of Hermite polynomials (6.15), we use (2.3). Then we get

$$\Phi_\tau^n(u) = \frac{\sqrt{n!}}{2^{\frac{3n}{2}+1}\tau^{\frac{n+1}{2}}\pi^{\frac{1}{4}}} e^{-\frac{1}{2}\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right)^2} \varphi_n\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right) e^{2\tau r^2 + \sqrt{2}ru} - (u \leftrightarrow -u). \quad (7.17)$$

Using the scaling of the variables (2.32), (7.5), as well as (2.9), one has

$$\begin{aligned} \frac{(1+4\tau)r}{2\sqrt{2\tau}} \pm \frac{u}{2\sqrt{\tau}} &= \sqrt{2N} + \frac{(T^2 + R \pm U)N^{-1/6}}{\sqrt{2}} \mp \frac{TUN^{-1/2}}{\sqrt{2}} + o(N^{-1/2}), \\ 2\tau r^2 \pm \sqrt{2}ru &= \frac{N}{2} + TN^{2/3} + \left(T^2 + \frac{R}{2} \pm U\right) N^{1/3} + RT + T^3 + o(1). \end{aligned} \quad (7.18)$$

Further, by the scaling (2.32), (7.5) and (2.9),

$$(4\tau)^{\frac{n+1}{2}} = e^{TN^{2/3} + T^3/3 - T\xi + o(1)}. \quad (7.19)$$

Finally, Stirling's formula leads to

$$\sqrt{n!} = N^{\frac{n}{2}} (1 - \xi N^{-2/3})^{\frac{N - \xi N^{1/3}}{2}} e^{-\frac{N}{2} - \frac{\xi N^{1/3}}{2} + o(1)} (2\pi N)^{\frac{1}{4}} = (2\pi)^{\frac{1}{4}} N^{\frac{n}{2} + \frac{1}{4}} e^{-\frac{N}{2} + o(1)}. \quad (7.20)$$

Plugging (7.18)–(7.20) into (7.17), one has

$$\begin{aligned} \Phi_\tau^n(u) &= \left(\frac{N}{2}\right)^{\frac{n}{2}} \sqrt{2} N^{1/6} e^{-N - \frac{RN^{1/3}}{2} + o(1)} e^{\frac{2}{3}T^3 + (R + \xi + U)T} \\ &\quad \times 2^{-1/4} N^{1/12} \varphi_{N - \xi N^{1/3}}\left(\sqrt{2N} + \frac{(T^2 + R + U)N^{-1/6}}{\sqrt{2}}\right) - (U \leftrightarrow -U). \end{aligned} \quad (7.21)$$

On the other hand, by (7.1) from Lemma 7.1, for any $\xi > 0$ fixed,

$$\lim_{N \rightarrow \infty} 2^{-1/4} N^{1/12} \varphi_{N - \xi N^{1/3}}\left(\sqrt{2N} + \frac{sN^{-1/6}}{\sqrt{2}}\right) = \text{Ai}(s + \xi). \quad (7.22)$$

Using the notation (2.30), this proves (7.6).

The proof of (7.7) is similar. Using (2.3) in (6.16) gives

$$\Psi_\tau^m(u) = \frac{2^{\frac{3m}{2}}\tau^{\frac{m}{2}}\pi^{\frac{1}{4}}}{\sqrt{m!}} e^{\frac{1}{2}\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right)^2} \varphi_m\left(\frac{(1+4\tau)r}{2\sqrt{2\tau}} + \frac{u}{2\sqrt{\tau}}\right) e^{-2\tau r^2 - \sqrt{2}ru} - (u \leftrightarrow -u). \quad (7.23)$$

Substituting (7.18)–(7.20) with n replaced by m , one has

$$\begin{aligned} \Psi_\tau^m(u) &= \left(\frac{2}{N}\right)^{\frac{m}{2}} N^{-1/3} e^{N + \frac{RN^{1/3}}{2} + o(1)} e^{-\frac{2}{3}T^3 - (R + \zeta + U)T} \\ &\quad \times 2^{-1/4} N^{1/12} \varphi_{N - \zeta N^{1/3}}\left(\sqrt{2N} + \frac{(T^2 + R + U)N^{-1/6}}{\sqrt{2}}\right) - (U \leftrightarrow -U) \end{aligned} \quad (7.24)$$

which proves (7.7) by using (2.30).

To show (7.8), one uses (8.4). By comparing the definitions (2.11)–(2.12) with (7.21) and (7.24), one can write the kernel as

$$\begin{aligned} K_0(n, m) &= \left(\frac{N}{2}\right)^{\frac{n-m}{2}} 2^{-1/2} N^{-1/6} e^{T(\xi-\zeta)+o(1)} \\ &\quad \times \int_{\mathbb{R}} dU e^{2TU} \varphi_{N-\xi N^{1/3}} \left(\sqrt{2N} + \frac{(T^2 + R + U)N^{-1/6}}{\sqrt{2}} \right) \\ &\quad \times \varphi_{N-\zeta N^{1/3}} \left(\sqrt{2N} + \frac{(T^2 + R - U)N^{-1/6}}{\sqrt{2}} \right) \end{aligned} \quad (7.25)$$

where we made the change of variables $u = UN^{-1/6}/\sqrt{2}$. For any $c > 0$, there is a uniform constant $C = C(c)$ such that

$$\left| 2^{-1/4} N^{1/12} \varphi_{N-\xi N^{1/3}} \left(\sqrt{2N} + \frac{sN^{-1/6}}{\sqrt{2}} \right) \right| \leq C e^{-c(\xi+s)} \quad (7.26)$$

for $s > 0$, because of (7.2) with $n = N - \xi N^{1/3}$.

If n does not grow to infinity with $N \rightarrow \infty$, then by definition (2.3) with $x = \sqrt{2N} + sN^{-1/6}/\sqrt{2}$, the harmonic oscillator function on the left-hand side of (7.26) is of order e^{-N} which is even smaller than the right-hand side. Using (7.3), the left-hand side of (7.26) is at most a uniform constant for $s \leq 0$. Hence we can use dominated convergence in (7.25) and conclude that

$$\begin{aligned} \left(\frac{2}{N}\right)^{\frac{n-m}{2}} N^{1/3} K_0(n, m) &\rightarrow e^{T(\xi-\zeta)} \int_{\mathbb{R}} dU e^{2TU} \text{Ai}(T^2 + R + \xi + U) \text{Ai}(T^2 + R + \zeta - U) \\ &= 2^{-1/3} \text{Ai}(2^{-1/3}(2R + \xi + \zeta)). \end{aligned} \quad (7.27)$$

In the last step we use the following identity: for any $s_1, s_2, t \in \mathbb{R}$,

$$\int_{\mathbb{R}} d\lambda e^{t\lambda} \text{Ai}(s_2 + \lambda) \text{Ai}(s_1 - \lambda) = 2^{-1/3} e^{\frac{1}{2}(s_1-s_2)t} \text{Ai} \left(2^{-1/3} \left(s_1 + s_2 - \frac{t^2}{2} \right) \right), \quad (7.28)$$

which follows from (A.5)–(A.6) of [11] using the notation (A.1) in [11]. This proves (7.8).

Using the uniformity of the bound (7.26) in ξ , the exponential bounds in ξ and ζ which can be given for (7.21), (7.24) and for (7.25) yield (7.9)–(7.11). This completes the proof. \square

8 Proof of lemmas

In this section, we give the proofs of all those propositions and lemmas which were found to be technical to give immediately. For the proof of Proposition 2.2 we will use the following lemma.

Lemma 8.1. *With the notation (2.13), for any $u, v \in \mathbb{R}$, one has*

$$\int_{\mathbb{R}} du f_W(u) \phi_{2(\tau_2-\tau_1)}(v-u) = e^{(\tau_2-\tau_1)(\sqrt{2}r-2W)^2} f_W(v), \quad (8.1)$$

$$\int_{\mathbb{R}} dv \phi_{2(\tau_2-\tau_1)}(v-u) g_Z(v) = e^{(\tau_2-\tau_1)(\sqrt{2}r-2Z)^2} g_Z(u). \quad (8.2)$$

Further, for any $\tau > 0$ and integers n and m ,

$$\frac{2}{(2\pi i)^2} \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \int_{\mathbb{R}} du \frac{W^n e^{\tau(\sqrt{2r-2W})^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau(\sqrt{2r-2Z})^2 - \sqrt{2r}Z}} f_W(u) g_Z(u) = \mathbb{1}(n, m), \quad (8.3)$$

$$\frac{2}{(2\pi i)^2} \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \int_{\mathbb{R}} du \frac{W^n e^{\tau(\sqrt{2r-2W})^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau(\sqrt{2r-2Z})^2 - \sqrt{2r}Z}} f_W(u) g_Z(-u) = K_0(n, m). \quad (8.4)$$

Proof of Proposition 2.2. We substitute the definition (2.15) of T_{τ_1, τ_2} and by combining terms after the change of variables $u \rightarrow -u$, one gets

$$\begin{aligned} & \int_{\mathbb{R}_-} du (f_W(u) - f_W(-u)) T_{\tau_1, \tau_2}(u, v) \\ &= \int_{\mathbb{R}} du f_W(u) \phi_{2(\tau_2 - \tau_1)}(v - u) - \int_{\mathbb{R}} du f_W(u) \phi_{2(\tau_2 - \tau_1)}(v + u) \\ &= e^{(\tau_2 - \tau_1)(\sqrt{2r-2W})^2} (f_W(v) - f_W(-v)) \end{aligned} \quad (8.5)$$

where (8.1) was used in the second equality. This proves (2.16). The proof of (2.17) is similar. The identity (2.18) immediately follows from (8.3)–(8.4) after the combination of the terms which appear in (2.11)–(2.12) and by the change of variables $u \rightarrow -u$. \square

Proof of Lemma 8.1. The identities (8.1) and (8.2) are Gaussian integrals which are straightforward to compute.

To show (8.3), one separates the integral with respect to u restricted to \mathbb{R}_- and to \mathbb{R}_+ . We can suppose that Γ_0 is so small that $\text{Re}(Z) \in (-1, 1)$ along $Z \in \Gamma_0$. Then in the integral on \mathbb{R}_- , one can deform the W contour to $-1 + i\mathbb{R}$, and with this, the integral with respect to u can be computed as

$$\int_{\mathbb{R}_-} du f_W(u) g_Z(u) = \int_{\mathbb{R}_-} du e^{2(Z-W)u} = \frac{1}{2(Z-W)} \quad (8.6)$$

since $\text{Re}(Z - W) > 0$ for any $Z \in \Gamma_0$ and $W \in -1 + i\mathbb{R}$. Similarly on \mathbb{R}_+ , one deforms the W contour to $1 + i\mathbb{R}$, and then

$$\int_{\mathbb{R}_+} du f_W(u) g_Z(u) = \int_{\mathbb{R}_+} du e^{2(Z-W)u} = -\frac{1}{2(Z-W)} \quad (8.7)$$

since $\text{Re}(Z - W) < 0$ in this case. By joining the two integration contours for W and by Cauchy's theorem, one gets

$$\begin{aligned} & \frac{2}{(2\pi i)^2} \int_{\mathbb{R}} du \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \frac{W^n e^{\tau(\sqrt{2r-2W})^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau(\sqrt{2r-2Z})^2 - \sqrt{2r}Z}} f_W(u) g_Z(u) \\ &= \frac{1}{2\pi i} \oint_{\Gamma_0} dZ \text{Res} \left(\frac{W^n e^{\tau(\sqrt{2r-2W})^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau(\sqrt{2r-2Z})^2 - \sqrt{2r}Z}}; W = Z \right) \\ &= \frac{1}{2\pi i} \oint_{\Gamma_0} dZ Z^{n-m-1} = \mathbb{1}(n, m) \end{aligned} \quad (8.8)$$

which shows (8.3).

In the same way, (8.4) follows by

$$\begin{aligned}
& \frac{2}{(2\pi i)^2} \int_{\mathbb{R}} du \int_{i\mathbb{R}} dW \oint_{\Gamma_0} dZ \frac{W^n e^{\tau(\sqrt{2r-2W})^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau(\sqrt{2r-2Z})^2 - \sqrt{2r}Z}} f_W(u) g_Z(-u) \\
&= \frac{1}{2\pi i} \oint_{\Gamma_0} dZ \operatorname{Res} \left(\frac{W^n e^{\tau(\sqrt{2r-2W})^2 - \sqrt{2r}W}}{Z^{m+1} e^{\tau(\sqrt{2r-2Z})^2 - \sqrt{2r}Z}}; W = \sqrt{2r} - Z \right) \\
&= K_0(n, m).
\end{aligned} \tag{8.9}$$

This completes the proof of the lemma. \square

Proof of Proposition 2.8. First observe that due to definition (2.31),

$$\frac{\partial}{\partial R} \widehat{K}_0(\xi, \zeta) = \left(\frac{\partial}{\partial \xi} + \frac{\partial}{\partial \zeta} \right) \widehat{K}_0(\xi, \zeta). \tag{8.10}$$

By writing the resolvent of \widehat{K}_0 as a Neumann series and by applying (8.10) to each term of the series, one obtains

$$\frac{\partial}{\partial R} (\mathbb{1} - \widehat{K}_0)^{-1}(\xi, \zeta) = \left(\frac{\partial}{\partial \xi} + \frac{\partial}{\partial \zeta} \right) (\mathbb{1} - \widehat{K}_0)^{-1}(\xi, \zeta) - (\mathbb{1} - \widehat{K}_0)^{-1} \widehat{K}_0(\xi, 0) (\mathbb{1} - \widehat{K}_0)^{-1} \widehat{K}_0(0, \zeta). \tag{8.11}$$

Further, by (2.30),

$$\frac{\partial}{\partial R} \widehat{\Psi}_T^\xi(U) = \frac{\partial}{\partial \xi} \widehat{\Psi}_T^\xi(U) \quad \text{and} \quad \frac{\partial}{\partial R} \widehat{\Phi}_T^\zeta(U) = \frac{\partial}{\partial \zeta} \widehat{\Psi}_T^\zeta(U). \tag{8.12}$$

Now one can take the derivative of the kernel \widehat{K}^{ext} in (2.34) with respect to R . Using (8.11) and (8.12), the proposition follows by direct computation. \square

Proof of Lemma 4.2. For a function h of the form (4.9), one can define approximating functions $h_\varepsilon \in H^1([0, 1])$ for any small $\varepsilon > 0$ such that as ε decreases to 0, the functions increasingly approach h . With other words, $h_\varepsilon(x) \rightarrow h(x)$ increasingly as $\varepsilon \rightarrow 0$ for any $x \in [0, 1]$.

Then the events $E_\varepsilon = \{B_N(t) < h_\varepsilon(t) \text{ for } t \in [0, 1]\}$ increase to $E_0 = \{B_N(t) < h(t) \text{ for } t \in [0, 1]\}$ as $\varepsilon \rightarrow 0$, hence $\mathbf{P}(E_\varepsilon) \rightarrow \mathbf{P}(E_0)$ by the continuity of the measure. Similarly, the events $\widetilde{E}_\varepsilon = \{\widetilde{b}(\tau) \leq \widetilde{h}_\varepsilon(\tau) \text{ for } \tau \in (\tau_1, \tau_2)\}$ which appear in (2.10) used in the definition (2.20) of K_N^h increase to $\widetilde{E}_0 = \{\widetilde{b}(\tau) \leq \widetilde{h}(\tau) \text{ for } \tau \in (\tau_1, \tau_2)\}$ as $\varepsilon \rightarrow 0$, since the functions $\widetilde{h}_\varepsilon$ increase to \widetilde{h} pointwise as $\varepsilon \rightarrow 0$, see (2.9). Hence $\mathbf{P}(\widetilde{E}_\varepsilon) \rightarrow \mathbf{P}(\widetilde{E}_0)$.

To complete the proof, the convergence of the corresponding Fredholm determinants on the right-hand side of (2.19) has to be shown. From $\mathbf{P}(\widetilde{E}_\varepsilon) \rightarrow \mathbf{P}(\widetilde{E}_0)$, one has the pointwise convergence of the operators in the Fredholm determinant. On the other hand, by Lemma 4.1, K_N^h is a trace class operator for any function h , i.e. the Fredholm determinant series converges absolutely, hence the corresponding Fredholm determinants on the right-hand side of (2.19) converge as $\varepsilon \rightarrow 0$ by dominated convergence. \square

Proof of Lemma 5.1. This proof follows the lines of the proof of Lemma 2.3 in [49]. We first rewrite the operator $e^{-2LD} - R_{-L,L}$ as follows. We substitute (5.6) into the definitions (5.3) and (5.4) and we use the identity

$$-\frac{(e^L y - e^{-L} x)^2}{e^{2L} - e^{-2L}} + \frac{(e^L y - e^{-L} x - (e^{2L} - e^{-2L}) \frac{x}{\sqrt{2}})^2}{e^{2L} - e^{-2L}} = \frac{(e^{2L} - e^{-2L}) r^2}{2} - \sqrt{2r} (e^L y - e^{-L} x) \tag{8.13}$$

to simplify the exponential factors. Then one has the decomposition

$$e^{-2LD} - R_{-L,L} = \Gamma_1 \Gamma_2 \Gamma_3 \quad (8.14)$$

where

$$\Gamma_1(x, u_1) = e^{-x^2/2 + \sqrt{2}e^{-L}xr + u_1^2/8\tau_1} T_{e^{-2L}/4, \tau_1} \left(e^{-L}x - \frac{(1 + e^{-2L})r}{\sqrt{2}}, u_1 \right) \mathbb{1}_{u_1 \leq H_1}, \quad (8.15)$$

$$\Gamma_2(u_1, u_2) = e^{-u_1^2/(8\tau_1) + u_2^2/8\tau_k} T_{\tau_1, \tau_k}^{\tau_i, H_i}(u_1, u_2), \quad (8.16)$$

$$\Gamma_3(u_2, y) = \mathbb{1}_{u_2 \leq H_k} T_{\tau_k, e^{2L}/4} \left(u_2, e^L y - \frac{(1 + e^{2L})r}{\sqrt{2}} \right) e^{-u_2^2/8\tau_k + y^2/2 + L - \sqrt{2}e^L y r + (e^{2L} - e^{-2L})r^2/2}. \quad (8.17)$$

The extra conjugation by $e^{u_1^2/8\tau_1}$ and by $e^{u_2^2/8\tau_k}$ was introduced because in this way all the operators $\Gamma_1, \Gamma_2, \Gamma_3$ have finite norm as shown below. Next we decompose the error term as $\Omega_L = \Omega_L^1 + \Omega_L^2$ with

$$\Omega_L^1 = P_{\sqrt{2}r \cosh L} (e^{-2LD} - R_{-L,L}) \bar{P}_{\sqrt{2}r \cosh L}, \quad (8.18)$$

$$\Omega_L^2 = (e^{-2LD} - R_{-L,L}) P_{\sqrt{2}r \cosh L}. \quad (8.19)$$

We bound the trace norm of

$$\tilde{\Omega}_L = e^{LD} K_{\text{Herm}, N} \Omega_L^1 e^{LD} K_{\text{Herm}, N} + e^{LD} K_{\text{Herm}, N} \Omega_L^2 e^{LD} K_{\text{Herm}, N} \quad (8.20)$$

as follows. One has by (8.14) and (8.18) that

$$\begin{aligned} & \|e^{LD} K_{\text{Herm}, N} \Omega_L^1 e^{LD} K_{\text{Herm}, N}\|_1 \\ & \leq \|e^{LD} K_{\text{Herm}, N} P_{\sqrt{2}r \cosh L} \Gamma_1\|_2 \|\Gamma_2\|_{\text{op}} \|\Gamma_3 \bar{P}_{\sqrt{2}r \cosh L} e^{LD} K_{\text{Herm}, N}\|_2. \end{aligned} \quad (8.21)$$

By definition, one can write the square of the first Hilbert–Schmidt norm as

$$\begin{aligned} & \|e^{LD} K_{\text{Herm}, N} P_{\sqrt{2}r \cosh L} \Gamma_1\|_2^2 \\ & = \sum_{n, m=0}^{N-1} \int_{\mathbb{R}} dx \int_{-\infty}^{H_1} dy \int_{\sqrt{2}r \cosh L}^{\infty} dw \int_{\sqrt{2}r \cosh L}^{\infty} dz \\ & \quad \times e^{L(n+m)} \varphi_n(x) \varphi_m(x) \varphi_n(w) \varphi_m(z) \Gamma_1(w, y) \Gamma_1(z, y) \\ & = \sum_{n=0}^{N-1} e^{2nL} \int_{-\infty}^{H_1} dy \left(\int_{\sqrt{2}r \cosh L}^{\infty} dz \varphi_n(z) \Gamma_1(z, y) \right)^2 \\ & \leq N e^{2(N-1)L} \int_{-\infty}^{H_1} dy \left(\int_{\sqrt{2}r \cosh L}^{\infty} dz \varphi_n(z)^2 \right) \left(\int_{\sqrt{2}r \cosh L}^{\infty} dz \Gamma_1(z, y)^2 \right) \\ & \leq N e^{2(N-1)L} \int_{\sqrt{2}r \cosh L}^{\infty} dz \int_{\mathbb{R}} dy \Gamma_1(z, y)^2 \end{aligned} \quad (8.22)$$

where we used first that the harmonic oscillator functions φ_n are orthonormal, then the Cauchy–Schwarz inequality, and finally the orthonormal property of φ_n again. In the

definition of Γ_1 (8.15) and by comparing it with (2.15), one can give the upper bound

$$\begin{aligned}
\int_{\mathbb{R}} dy \Gamma_1(z, y)^2 &\leq e^{-z^2+2\sqrt{2}e^{-L}zr} \int_{\mathbb{R}} dy e^{\frac{y^2}{4\tau_1}} \phi_{2\tau_1-e^{-2L}/2} \left(y - e^{-L}z + \frac{1+e^{-2L}}{\sqrt{2}}r \right)^2 \\
&= e^{-z^2+2\sqrt{2}e^{-L}zr} \int_{\mathbb{R}} dy \frac{e^{\frac{y^2}{4\tau_1}}}{\pi(4\tau_1-e^{-2L})} \exp \left(-\frac{(y - e^{-L}z + \frac{1+e^{-2L}}{\sqrt{2}}r)^2}{2\tau_1 - e^{-2L}/2} \right) \\
&= e^{-(1+o(1))z^2+o(1)z} \int_{\mathbb{R}} dy \frac{1}{4\pi\tau_1} e^{-(1+o(1))\frac{y^2}{4\tau_1}+(1+o(1))\frac{ry}{\sqrt{2}\tau_1}-\frac{r^2}{4\tau_1}+o(1)yz+o(1)} \\
&= \frac{1}{\sqrt{4\pi\tau_1}} e^{-(1+o(1))z^2+\frac{r^2}{4\tau_1}+o(1)z+o(1)}
\end{aligned} \tag{8.23}$$

by computing the Gaussian integral in the last step. The $o(1)$ above means a term which does neither depend on y nor z and which goes to 0 as $L \rightarrow \infty$. Putting (8.22) and (8.23) together, one obtains

$$\begin{aligned}
\|e^{LD} K_{\text{Herm},N} P_{\sqrt{2}r \cosh L} \Gamma_1\|_2^2 &\leq \frac{N e^{2(N-1)L+\frac{r^2}{4\tau_1}}}{\sqrt{4\pi\tau_1}} \int_{\sqrt{2}r \cosh L}^{\infty} dz e^{-(1+o(1))z^2+o(1)z+o(1)} \\
&\leq \frac{N e^{2NL+\frac{r^2}{4\tau_1}}}{\sqrt{4\pi\tau_1}} e^{-2r^2(\cosh L)^2(1+o(1))} \\
&\leq c_1 e^{2NL-c_2e^{2L}}
\end{aligned} \tag{8.24}$$

with positive constants c_1 and c_2 for L large enough. We used the Chernoff bound on the tail of the normal distribution in the second inequality.

Obtaining a bound on $\|\Gamma_3 \overline{P}_{\sqrt{2}r \cosh L} e^{LD} K_{\text{Herm},N}\|_2^2$ is very similar. There is a difference in the step which corresponds to (8.23). It can be done as follows.

$$\begin{aligned}
\int_{\mathbb{R}} dx \Gamma_3(x, z)^2 &\leq e^{z^2+2L-2\sqrt{2}e^Lzr+(e^{2L}-e^{-2L})r^2} \int_{\mathbb{R}} dx e^{-\frac{x^2}{4\tau_k}} \phi_{e^{2L}/2-2\tau_k} \left(x - e^Lz + \frac{(1+e^{2L})r}{\sqrt{2}} \right)^2 \\
&= e^{2L-2r^2-(1+o(1))z^2+o(1)z} \int_{\mathbb{R}} dx \frac{1}{\sqrt{\pi e^{2L}}} e^{-(1+o(1))\frac{x^2}{4\tau_k}-(1+o(1))2\sqrt{2}xr+o(1)xz+o(1)} \\
&= \frac{1}{\pi e^{2L}} e^{2L+(8\tau_k-2)r^2-(1+o(1))z^2+o(1)z+o(1)}
\end{aligned} \tag{8.25}$$

where the $o(1)$ term are again independent of y and z and they go to 0 as $L \rightarrow \infty$. The computation (8.25) results in a bound

$$\|\Gamma_3 \overline{P}_{\sqrt{2}r \cosh L} e^{LD} K_{\text{Herm},N}\|_2 \leq c_1 e^{NL} \tag{8.26}$$

very similarly as in (8.24). The factor $e^{-c_2e^{2L}}$ is not present due to the fact that the projection $P_{\sqrt{2}r \cosh L}$ is replaced by $\overline{P}_{\sqrt{2}r \cosh L}$.

Finally, the operator norm of Γ_2 can be bounded in the following way.

$$\begin{aligned}
\|\Gamma_2\|_{\text{op}}^2 &\leq \sup_{y \in \mathbb{R}} \int_{\mathbb{R}} dx \Gamma_2(x, y)^2 \leq \sup_{y \in \mathbb{R}} \int_{\mathbb{R}} dx e^{-\frac{x^2}{4\tau_1}+\frac{y^2}{4\tau_k}} \phi_{2(\tau_k-\tau_1)}(y-x)^2 \\
&= \sup_{y \in \mathbb{R}} \frac{1}{2} \sqrt{\frac{\tau_1}{\pi(\tau_k^2-\tau_1^2)}} e^{-\frac{(\tau_k-\tau_1)y^2}{4\tau_k(\tau_1+\tau_k)}} = \frac{1}{2} \sqrt{\frac{\tau_1}{\pi(\tau_k^2-\tau_1^2)}}
\end{aligned} \tag{8.27}$$

by straightforward computation involving a Gaussian integral. Putting (8.21), (8.22), (8.27) and (8.26) together proves that the error corresponding to Ω_L^1 goes to 0 as $L \rightarrow \infty$. The proof for Ω_L^2 can be done similarly. \square

References

- [1] M. Adler, S. Chhita, K. Johansson, and P. van Moerbeke. Tacnode GUE-minor processes and double Aztec diamonds. *Probab. Theory Related Fields*, 162:275–325, 2015.
- [2] M. Adler, J. Delépine, and P. van Moerbeke. Dyson’s nonintersecting Brownian motions with a few outliers. *Comm. Pure Appl. Math.*, 62:334–395, 2010.
- [3] M. Adler, P.L. Ferrari, and P. van Moerbeke. Non-intersecting random walks in the neighborhood of a symmetric tacnode. *Ann. Probab.*, 41:2599–2647, 2013.
- [4] M. Adler, K. Johansson, and P. van Moerbeke. Double Aztec diamonds and the tacnode process. *Adv. Math.*, 252:518–571, 2014.
- [5] P. Bleher and A. Kuijlaars. Large n limit of Gaussian random matrices with external source, Part III: Double scaling limit. *Comm. Math. Phys.*, 270:481–517, 2007.
- [6] A. Borodin. Biorthogonal ensembles. *Nucl. Phys. B*, 536:704–732, 1999.
- [7] A. Borodin. Schur dynamics of the Schur processes. *Adv. Math.*, 228:2268–2291, 2011.
- [8] A. Borodin, I. Corwin, and D. Remenik. Multiplicative functionals on ensembles of non-intersecting paths. *Ann. Inst. H. Poincaré Probab. Statist.*, 51:28–58, 2015.
- [9] A. Borodin and P.L. Ferrari. Random tilings and Markov chains for interlacing particles. *preprint: arXiv:1506.03910*, 2015.
- [10] A. Borodin, P.L. Ferrari, M. Prähofer, T. Sasamoto, and J. Warren. Maximum of Dyson Brownian motion and non-colliding systems with a boundary. *Electron. Comm. Probab.*, 14:486–494, 2009.
- [11] A. Borodin, P.L. Ferrari, and T. Sasamoto. Transition between Airy_1 and Airy_2 processes and TASEP fluctuations. *Comm. Pure Appl. Math.*, 61:1603–1629, 2008.
- [12] A. Borodin and G. Olshanski. Stochastic dynamics related to Plancherel measure. In V. Kaimanovich and A. Lodkin, editors, *AMS Transl.: Representation Theory, Dynamical Systems, and Asymptotic Combinatorics*, pages 9–22. 2006.
- [13] A. Borodin and E.M. Rains. Eynard-Mehta theorem, Schur process, and their Pfaffian analogs. *J. Stat. Phys.*, 121:291–317, 2006.
- [14] E. Cator and L. Pimentel. On the local fluctuations of last-passage percolation models. *Stoch. Proc. Appl.*, 125:879–903, 2015.
- [15] F. Colomo and A. Pronko. Third-order phase transition in random tilings. *Phys. Rev. E*, 88:042125, 2013.
- [16] F. Colomo and A. Pronko. Thermodynamics of the six-vertex model in an L -shaped domain. *Comm. Math. Phys.*, 339:699–728, 2015.
- [17] F. Colomo, A. Pronko, and A. Sportiello. Generalized emptiness formation probability in the six-vertex model. *arXiv:1605.01700*, 2016.

- [18] F. Colomo and A. Sportiello. Arctic curves of the six-vertex model on generic domains: the Tangent Method. *arXiv:1605.01388*, 2016.
- [19] I. Corwin and A. Hammond. Brownian Gibbs property for Airy line ensembles. *Inventiones mathematicae*, 195:441–508, 2013.
- [20] I. Corwin, J. Quastel, and D. Remenik. Continuum statistics of the Airy₂ process. *Comm. Math. Phys.*, 317:347–362, 2013.
- [21] S. Delvaux. The tacnode kernel: equality of Riemann-Hilbert and Airy resolvent formulas. *preprint: arXiv:1211.4845*, 2012.
- [22] S. Delvaux. Non-Intersecting Squared Bessel Paths at a Hard-Edge Tacnode. *Comm. Math. Phys.*, 324:715–766, 2013.
- [23] S. Delvaux, A. Kuijlaars, and L. Zhang. Critical behavior of non-intersecting Brownian motions at a tacnode. *Comm. Pure Appl. Math.*, 64:1305–1383, 2011.
- [24] S. Delvaux and B. Vető. The hard edge tacnode process and the hard edge Pearcey process with non-intersecting squared Bessel paths. *Random Matrices Theory Appl.*, 04:1550008, 2015.
- [25] P. Desrosiers and P.J. Forrester. A note on biorthogonal ensembles. *J. Approx. Theory*, 152:167–187, 2008.
- [26] F.J. Dyson. A Brownian-motion model for the eigenvalues of a random matrix. *J. Math. Phys.*, 3:1191–1198, 1962.
- [27] B. Eynard and M.L. Mehta. Matrices coupled in a chain. I. Eigenvalue correlations. *J. Phys. A*, 31:4449–4456, 1998.
- [28] P.L. Ferrari. *Shape fluctuations of crystal facets and surface growth in one dimension*. PhD thesis, Technische Universität München, <http://tumb1.ub.tum.de/publ/diss/ma/2004/ferrari.html>, 2004.
- [29] P.L. Ferrari and H. Spohn. Step fluctuations for a faceted crystal. *J. Stat. Phys.*, 113:1–46, 2003.
- [30] P.L. Ferrari and H. Spohn. Domino tilings and the six-vertex model at its free fermion point. *J. Phys. A: Math. Gen.*, 39:10297–10306, 2006.
- [31] P.L. Ferrari, H. Spohn, and T. Weiss. Brownian motions with one-sided collisions: the stationary case. *Electron. J. Probab.*, 20:1–41, 2015.
- [32] P.L. Ferrari and B. Vető. Non-colliding Brownian bridges and the asymmetric tacnode process. *Electron. J. Probab.*, 44:1–17, 2012.
- [33] J. Hägg. Local Gaussian fluctuations in the Airy and discrete PNG processes. *Ann. Probab.*, 36:1059–1092, 2008.
- [34] K. Johansson. Discrete orthogonal polynomial ensembles and the Plancherel measure. *Annals of Math.*, 153:259–296, 2001.

- [35] K. Johansson. Non-intersecting paths, random tilings and random matrices. *Probab. Theory Related Fields*, 123:225–280, 2002.
- [36] K. Johansson. The arctic circle boundary and the Airy process. *Ann. Probab.*, 33:1–30, 2005.
- [37] K. Johansson. Random matrices and determinantal processes. In A. Bovier, F. Dunlop, A. van Enter, F. den Hollander, and J. Dalibard, editors, *Mathematical Statistical Physics, Session LXXXIII: Lecture Notes of the Les Houches Summer School 2005*, pages 1–56. Elsevier Science, 2006.
- [38] K. Johansson. Non-colliding Brownian Motions and the extended tacnode process. *Commun. Math. Phys.*, 319:231–267, 2013.
- [39] S. Karlin and L. McGregor. Coincidence probabilities. *Pacific J.*, 9:1141–1164, 1959.
- [40] M. Katori and H. Tanemura. Symmetry of matrix-valued stochastic processes and noncolliding diffusion particle systems. *J. Math. Phys.*, 45:3058–3085, 2004.
- [41] M. Katori and H. Tanemura. Infinite systems of noncolliding generalized meanders and Riemann-Liouville differintegrals. *Probab. Theory Relat. Fields*, 138:113–156, 2007.
- [42] M. Katori and H. Tanemura. Noncolliding Brownian Motion and Determinantal Processes. *J. Stat. Phys.*, 129:1233–1277, 2007.
- [43] V. Korepin and P. Zinn-Justin. Thermodynamic limit of the six-vertex model with domain wall boundary conditions. *J. Phys. A*, 33:7053–7066, 2000.
- [44] I. Krasikov. New bounds on the Hermite polynomials. *arXiv:math.CA/0401310*, 2004.
- [45] A. B. J. Kuijlaars, A. Martí nez Finkelshtein, and F. Wielonsky. Non-Intersecting Squared Bessel Paths: Critical Time and Double Scaling Limit. *Comm. Math. Phys.*, 308:227–279, 2011.
- [46] T. Nagao. Dynamical Correlations for Vicious Random Walk with a Wall. *Nucl. Phys. B*, 658:373–396, 2003.
- [47] T. Nagao and P.J. Forrester. Multilevel dynamical correlation functions for Dyson’s Brownian motion model of random matrices. *Phys. Lett. A*, 247:42–46, 1998.
- [48] T. Nagao, M. Katori, and H. Tanemura. Dynamical correlations among vicious random walkers. *Phys. Lett. A*, 307:29–33, 2003.
- [49] G.B. Nguyen and D. Remenik. Non-intersecting Brownian bridges and the Laguerre Orthogonal Ensemble. *arXiv:1505.01708*, 2015.
- [50] M. Prähofer and H. Spohn. Scale invariance of the PNG droplet and the Airy process. *J. Stat. Phys.*, 108:1071–1106, 2002.
- [51] M. Reed and B. Simon. *Methods of Modern Mathematical Physics IV: Analysis of Operators*. Academic Press, New York, 1978.

- [52] G. Szegő. *Orthogonal Polynomials*. American Mathematical Society Providence, Rhode Island, 3th edition, 1967.
- [53] C.A. Tracy and H. Widom. Level-spacing distributions and the Airy kernel. *Comm. Math. Phys.*, 159:151–174, 1994.
- [54] C.A. Tracy and H. Widom. Differential equations for Dyson processes. *Comm. Math. Phys.*, 252:7–41, 2004.
- [55] C.A. Tracy and H. Widom. The Pearcey Process. *Comm. Math. Phys.*, 263:381–400, 2006.
- [56] C.A. Tracy and H. Widom. Nonintersecting Brownian Excursions. *Ann. Appl. Prob.*, 17:953–979, 2007.
- [57] P. Zinn-Justin. Six-vertex model with domain wall boundary conditions and one-matrix models. *Phys. Rev. E*, 62:3411–3418, 2000.