

UVA Summer School PS2

July 9th 2024

1. **Exercise 6.1** Starting the PNG from the narrow wedge initial condition \mathfrak{d}_0 , i.e.

$$h(0, x) = \begin{cases} 0, & x = 0 \\ -\infty, & x \neq 0 \end{cases}, \quad (1)$$

we will show that the one-point marginal $h(t, 0)$ is equivalent to a Poissonized version of the longest increasing subsequence problem¹.

The longest increasing subsequence problem is as follows: consider the problem of finding Lipschitz-1 paths going from $(0, 0)$ to $(t, 0)$ which pick up a maximal number of space-time Poisson points along the way. All these paths lie inside the square R with vertices $(0, 0)$, $(t/2, t/2)$, $(t/2, -t/2)$ and $(t, 0)$, and

- Show that the maximal number of points which they can pick up is exactly $h(t, 0)$.

Now rotate the picture by -45° , let N denote the number of Poisson points inside the square corresponding to R (so that N is a $\text{Poisson}[t]$ random variable), and order these N points according to their x coordinate. The y coordinates of these points define a random permutation σ of $\{1, \dots, N\}$, which is clearly chosen uniformly from S_N .

- Show that $h(t, 0)$ is nothing but the length of the longest increasing subsequence in σ .

2. **Exercise 6.8** Recall some notions from the lecture. $E_{a,u}$ is expectation of the height function, which jumps down for η particles and up for ζ particles, i.e. $h_x - h_{x-1} = \zeta_x - \eta_x$ with $h_a = u$ with respect to the product measure from Exercise 4.14 (problem 2 from Monday) on the interval $\{a, a+1, \dots, b\}$. Note that we are dealing with the conditional measure that $h(b) = v$, multiplied by the probability that $h(b) = v$, so we can write it as $E_{a,u}[(L^* F) \mathbf{1}_{h_b=v}]$, where $L^* = L_{\text{rw}}^* + L_{\text{cr}}^*$. $F(g)$ is the indicator function that $g - h$ is ever less than or equal to 0.

- Show that $E_{a,u}[(L_{\text{cr}}^* F) \mathbf{1}_{h_b=v}] = 0$.

3. **Exercise 4.6** Let Δ be the discrete Laplace operator on \mathbb{Z} , i.e. $\Delta f(x) = f(x+1) - 2f(x) + f(x-1)$, $x \in \mathbb{Z}$.

- Show that

$$e^{x\Delta}(u_1, u_2) = e^{-2x} I_{|u_2 - u_1|}(2x), \quad (2)$$

where $I_n(2x) := \frac{1}{2\pi i} \oint_{\gamma_0} dz e^{x(z+z^{-1})} / z^{n+1}$ is the modified Bessel function of the first kind. Here γ_0 is any simple positively oriented contour around $0 \in \mathbb{C}$.

¹This was the first exact fluctuation found in the KPZ class, by Baik, Deift in Johansson in 1999. They took a limit of a determinantal formula for the distribution which derived by Gessel about 10 years earlier (it took the decade to realize that it was a formula for this model) and obtained in the limit the same GUE Tracy-Widom distribution which had also been discovered, again about 10 years earlier, for the asymptotic distribution of the top eigenvalue of a matrix from the Gaussian Unitary Ensemble.

4. **Exercise 6.17** [Narrow wedge initial data] We are going to work out the distribution for the PNG starting from narrow wedge initial data, which we write:

$$F_\tau(s) = \mathbb{P}(h(t, x) \leq r) = \det(I - K(t, x, r))_{l(\mathbb{Z}_{>0})} \quad (3)$$

The kernel K is defined in class. Since the random walk g can only hit the hypograph of \mathfrak{d}_0 at the origin,

- Show that

$$e^{-t\Delta} P_{x-t, x+t}^{\text{hit}(\mathfrak{d}_0)} e^{-t\Delta} = \bar{\chi}_0, \quad (4)$$

where $\bar{\chi}_0(u) = 1_{u \leq 0}$.

- Following the previous exercise, show that

$$e^{2t\nabla + x\Delta}(u_1, u_2) = e^{-2x} \frac{1}{2\pi i} \oint_{\gamma_0} \frac{dz}{z^{u_2 - u_1 + 1}} e^{t(z - z^{-1}) + x(z + z^{-1})}, \quad (5)$$

where γ_0 is any simple, positively oriented contour around the origin.

When $t > |x|$ this kernel can be expressed in terms of the Bessel function of the first kind, $J_n(x) = \frac{1}{2\pi i} \oint_{\gamma_0} dz e^{x(z - z^{-1})/2} / z^{n+1}$, we have:

$$e^{2t\nabla + x\Delta}(u_1, u_2) = e^{-2x} \left(\frac{t+x}{t-x} \right)^{(u_2 - u_1)/2} J_{u_2 - u_1}(2\sqrt{t^2 - x^2}). \quad (6)$$

- Use this to show that

$$K(u, v) = \left(\frac{t+x}{t-x} \right)^{(u-v)/2} B_s(u, v), \quad B_s(u, v) = \sum_{\ell \leq 0} J_{u-\ell}(2s) J_{v-\ell}(2s)$$

with $s = \sqrt{t^2 - x^2}$.

B_s is called the *discrete Bessel kernel*.

- Show that it is an *integrable* kernel:

$$B_s(u, v) = \frac{s}{u - v} (J_{u-1}(2s) J_v(2s) - J_u(2s) J_{v-1}(2s)).$$

In order to show this, you need the following recurrence relation on J_n :

$$\frac{n}{t} J_n(2t) = J_{n+1}(2t) + J_{n-1}(2t).$$

- Show that a conjugation of kernel reduces the distribution to be:

$$F_r(s) = \det(I - \tau_r B_s \tau_{-r})_{\ell^2(\mathbb{Z}_{>0})}, \quad (7)$$

where $\tau_r f(u) = f(u + r)$.

Note that from [Borodin-Okounkov 2000],

$$\det(I - \tau_r B_s \tau_{-r})_{\ell^2(\mathbb{Z}_{>0})} = e^{-s^2} \det(I_{i-j}(2s))_{i,j=0,\dots,r-1}, \quad (8)$$

where the I_n are modified Bessel functions of the first kind. This latter is Gessel's formula.