

2014年 11月26日 ~ 29日

基本工集中講義.

□ 一次元可積分確率過程 □

KPZ = 関する概観

ランダム行列

TASEP

Schur process

KPZ equation

Duality and  $\beta$ -TASEP

Macdonald process.

Barrier

What is the KPZ equation?

The KPZ equation is not really well-defined

Explicit formulas

⑩ Ballistic deposition of order 1  $\Rightarrow$   $\beta = 2, \dots, \tau_2, \dots$

KPZ equation (Kardar-Parisi-Zhang '86)

$$\partial_t^2 h(x, t) = \frac{1}{2} \lambda (\partial_x h(x, t))^2 + \nu \partial_x^2 h(x, t) + \sqrt{D} \eta(x, t)$$

$$\downarrow \rho \quad \langle \eta(x, t) \eta(x', t) \rangle =$$

$$\partial_t^2 h(x, t) = \frac{1}{2} (\partial_x h(x, t))^2 + \frac{1}{2} \partial_x^2 h(x, t) + \eta$$

$$\partial_t^2 h(x, t) = \frac{1}{2} \partial_x^2 h(x, t) + \eta(x, t)$$

Edwards-Wilkinson

Add nonlinearity  $(\partial_x h(x, t))^2 \Rightarrow$  KPZ  $\uparrow$  noise

$$\partial_t^2 h \approx v \sqrt{1 + (\partial_x h)^2} \approx v + \frac{v}{2} (\partial_x h)^2 + \dots$$

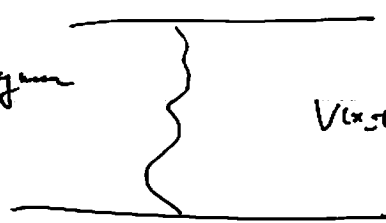
RG argument ( $\beta = \frac{1}{3}$ )

Cole-Hopf transformation

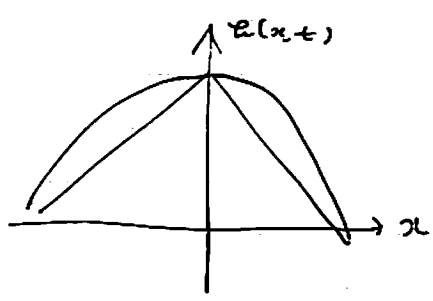
$$z(x,t) = \exp(u(x,t))$$

$$\Rightarrow \partial_t z(x,t) = \frac{1}{2} \partial_x^2 z(x,t) + \eta(x,t) z(x,t)$$

ξ  
directed polymer



$V(x,t) \Rightarrow \eta(x,t)$   
randomize



narrow edge

$$\delta_0 = \lim_{\delta \rightarrow 0} c_{\delta} e^{-|z|/\delta}$$

$$\eta(x,t) = \frac{dB(x,t)}{dt}$$

$$dz(x,t) = \frac{1}{2} \frac{\partial^2 z(x,t)}{\partial t^2} dt + z(x,t) \times dB(x,t)$$

$$dB(x,t) dB(x',t) = \delta(x-x') dt$$

Stratonovich

$$z(x,t) \circ dB(x,t) = z(x,t) \cdot dB(x,t) + \frac{1}{2} dz(x,t) dB(x,t)$$

$$dZ(x,t) = \frac{1}{2} \partial_x^2 Z(x,t) dt + Z(x,t) dB(x,t)$$

well-defined  
 Cole-Hopf sol.  $u = \log Z(x,t)$  (Lighthill - Jacobi)

$$\partial_t u(x,t) = \frac{1}{2} (\partial_x u(x,t))^2 + \frac{1}{2} \partial_x^2 u(x,t) - \omega + \eta(x,t)$$

§ 2. Explicit formula.

$Z(x,0) = \delta(x)$  (narrow wedge)

$$\left\langle e^{-e^{u(x,t)} + \frac{t}{2t} - \delta_t^5} \right\rangle = \det(1 - K_{s,t})$$

where  $\delta_t = (t/2)^{1/3}$  and  $K_{s,t}$

$$K_{s,t}(x,y) = \int_{-\infty}^{\infty} d\lambda \frac{A_c(x+\lambda) A_c(y+\lambda)}{e^{\delta_t(s-\lambda)} - 1}$$

Then

$$u(x,t) = -\frac{x^2}{2t} - \frac{1}{12} \delta_t^3 + \delta_t \frac{x}{t} \quad (\delta_t = (t/2)^{1/3})$$

$$F_{s,t}(s) = P(\beta_t \leq s) =$$

$$1 - \int_{-\infty}^{\infty} \exp\left(-e^{\delta_t(s-u)}\right) \left\{ \det(1 - P_u(B_t - P_{A_c}) P_u) - \det(1 - P_u B_t P_u) \right\} du$$

$$P_{A_c}(x,y) = A_c(x) A_c(y) \quad - P_u = \text{proj. onto } [u, \infty)$$

$$B_t(x,y) = \int_{-\infty}^{\infty} \frac{A_c(x+\lambda) A_c(y+\lambda)}{e^{\delta_t \lambda} - 1}$$

$$\left( \begin{array}{l} \text{GUE} \quad P(H) dH \propto e^{-\text{tr} H^2} d(H) \\ 2^{-1} \Delta(x), 2 \pi e^{-x^2} \end{array} \right)$$

GUE TW-diff.

$$\lim_{N \rightarrow \infty} P \left( \frac{\lambda_{\max} - \sqrt{2N}}{2^{-1/2} N^{-1/6}} \leq r \right) = F_2(r) = \det(I - P_r K_2 P_r)$$

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

Dotsenko - Le Douaral - Calabrese

$$Z(x, t) = E \left[ e^{\sum_0^t \eta(\text{brs}, t-s) ds} Z(\text{brs}, 0) \right]$$

↓

$$\langle (Z_N(x, t))^N \rangle = \langle z | e^{-H_N t} | 0 \rangle$$

$$H_N = - \frac{1}{2} \sum_{j=1}^N \frac{\partial^2}{\partial x_j^2} - \frac{1}{2} \sum_{j \neq k}^N f(x_j - x_k)$$

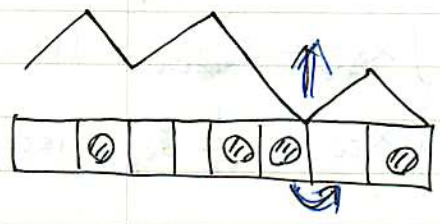
S-Bose gas.  
attractive.

$$\langle e^{-e^{2(t_0 t)} + \frac{t}{2t} - \delta_t s} \rangle = \sum_{N=0}^\infty \frac{(-e^{-\delta_t s})^N}{N!} \langle Z_N^N(t_0 t) \rangle e^{N \frac{\delta_t s}{2}}$$

$$\langle Z^N \rangle \sim e^{N^3} !$$

TASEP or ASEP

$N(x, t)$  = Integrated current at  $(x, x+1)$  up to time  $t$   
 ↑  
 height



ASEP phase  
 ↑  
 ASEP

ASEP is 2D  
 $1=1$  2D in 2D  
 界面と壁

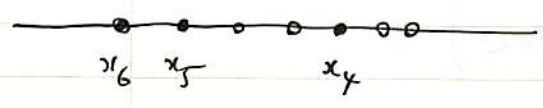
ASEP  $\Rightarrow$  KPZ (weakly asymmetric limit)

$\delta$ -TASEP

Borodin - Corwin

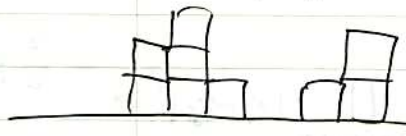
Aparticle  $i$  hops with rate  $1 - \delta^{x_i - x_{i-1}}$

$\delta \rightarrow 1$   $2D$  KPZ



$\delta$ -TAZRP

(1998 TS, Wadati)



$\delta$  - Boron operators

Flat case (replica) (Le Doussal, Calabrese)

Multiple case (replica) (Dotsenko)

Arndt - Heigel - Ritterberg. (1988)

Halpin - Healy (2012)

# 1. Random matrices

## 1.1. Gaussian ensembles

$\mathcal{H}_N^{(\beta)}$  = the set of  $\begin{pmatrix} \text{real sym.} \\ \text{hermittian} \end{pmatrix}$  matrices of size  $N$ .

$\{\xi_{ij}\}_{i,j=1}^{\infty}$  i.i.d.  $N(0,1)$

$\beta=1$

$\beta=2$

def. GOE

$X \in \mathcal{H}_N^{(1)}$

$$\begin{cases} X_{ij} = \xi_{ij} & (1 \leq i < j \leq N) \\ X_{ii} = \frac{1}{\sqrt{2}} \xi_{ij} & (1 \leq i \leq j \leq N) \end{cases}$$

GOE

$X \in \mathcal{H}_N^{(2)}$

$$\begin{cases} X_{ii} = \frac{1}{\sqrt{2}} \xi_{ii} & (1 \leq i \leq N) \\ X_{ij} = \frac{1}{2} (\xi_{ij} + i \xi_{ij}') & (1 \leq i < j \leq N) \end{cases}$$

Rem:  $P_p(x) \propto e^{-\frac{\beta}{2} \text{Tr}(X^2)}$   $\xi \xi \xi \xi \xi \xi - \xi \xi$

$x_i, 1 \leq i \leq N$  (eigenvalues)

### Prop (Joint distribution)

The density is

$$P_{N\beta}(x_1, \dots, x_N) = \frac{1}{\sum_N(\beta)} \prod_{1 \leq i < j \leq N} |x_i - x_j|^\beta e^{-\frac{\beta}{2} \sum_{i=1}^N x_i^2} \quad (\beta=1,2)$$

↑  
normalization

∴ For GOE,

$X = O \Lambda^t O$

$\Lambda = \text{diag}(x_1, \dots, x_N), \quad t O O = I$

$$\begin{cases} {}^t 0 dX 0 = S_0 \cdot X - X \cdot S_0 + dX \\ S_0 = {}^t 0 d 0 \end{cases}$$

$$\begin{aligned} \text{Tr}(dX)^2 &= 2 \text{Tr}(S_0 \cdot \Lambda \cdot S_0 \cdot \Lambda) \\ &\quad - 2 \text{Tr}(S_0 \cdot \Lambda^2 \cdot S_0) + \text{Tr}(d\Lambda)^2 \\ &= 2 \sum_{k < l} S_{0_{kl}} S_{0_{kl}} (\Lambda_k - \Lambda_l)^2 + \sum_k (d\Lambda_k)^2 // \end{aligned}$$

⑩. GUE の場合  $\Rightarrow \beta = 2$  かつ  $\beta = 1$ .

⑪ Largest e.v. dist

$$P_{N\beta} [x_{\max} \leq u] = \frac{1}{Z_{N\beta}} \int_{(-\infty, u]^N} \prod_{i < j} |x_i - x_j|^\beta e^{-\sum_{i=1}^N x_i^2} dx_1 \dots dx_N$$

↑  
largest eigenvalue

For GUE<sub>2</sub> ( $\beta = 2$ ),

$$P_{N2} [x_{\max} \leq u] = \frac{1}{Z_{N2}} \int \Delta(x) \cdot \Delta(x) \left( \prod_{i < j} 1_{(-\infty, u]}(x_i) e^{-x_i^2} \right) dx_1 \dots dx_N$$

$$\Delta(x) = \prod_{i < j} (x_i - x_j)$$



1.2. Fredholm determinant

$\mu$ : a measure on  $\mathbb{R}$  s.t.  $\int_{\mathbb{R}} \mu(dx) < \infty$  (mainly  $\mu(dx) = \rho(x) dx$ )

Kernel  $K$ : a measurable fun.  $K: \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{C}$  s.t.

$$\|K\| = \sup_{(x,y) \in \mathbb{R}^2} |K(x,y)| < \infty$$

Def. The Fredholm determinant associated with the kernel is defined as

$$\det(I-K)_{L^2(\mathbb{R}, \mu)} = \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \int_{\mathbb{R}} \dots \int_{\mathbb{R}} \left( \det K(x_i, x_j)_{1 \leq i, j \leq n} \right) d\mu(x_1) \dots d\mu(x_n)$$

Let us consider a function in the form of a product of determinants.

$$Q_N(x_1, \dots, x_N) = \frac{1}{Z_N} \det(f_i(x_j))_{i,j=1}^N - \det(g_i(x_j))_{i,j=1}^N$$

where  $f_i, g_i (1 \leq i \leq N) \mathbb{R} \rightarrow \mathbb{C}$

s.t.  $\max_i \sup_{x \in \mathbb{R}} (|f_i(x)|, |g_i(x)|) < \infty$

and  $\int f_i(x) g_j(x) d\mu(x) = \delta_{ij}$ .

Prop. For  $A \in \mathcal{B}(\mathbb{R})$

$$\int_{A^N} Q_N(x_1, \dots, x_N) d\mu(x_1) \dots d\mu(x_N) = \det(I-K)_{L^2(\mathbb{R}, \frac{1}{A} d\mu)}$$

where  $K(x,y) = \sum_{i=1}^N f_i(x) g_i(y)$

$$\text{Prop } P_{N,2} [x_{\max} \leq s] = \det(I - K)_{L^2(y, \infty)}$$

$$\text{where } K_{N,2}(x, y) = \sum_{n=0}^{N-1} \frac{H_n(x) H_n(y)}{\sqrt{2} n! 2^n} e^{-\frac{x^2+y^2}{2}}$$

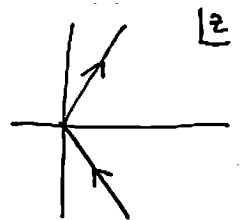
$$\text{and } \int_{-\infty}^{\infty} H_n(x) H_m(x) e^{-x^2} dx = \sqrt{2} n! 2^n \delta_{nm}$$

### 1.3. Tracy - Widom distribution

Def (Airy kernel)

$$K_2(x, y) = \frac{A_2(x) A_2'(y) - A_2'(x) A_2(y)}{x-y}$$

$$A_2(x) = \frac{1}{2\pi i} \int_c e^{\frac{z^3}{3} - xz} dz$$



GUE TW dist

$$F_2(s) = \det(I - K_2)_{L^2(s, \infty)}$$

$$\text{Thm. } \lim_{N \rightarrow \infty} P_{N,2} \left[ (x_{\max} - \sqrt{2N}) \sqrt{2} N^{1/6} \leq s \right] = F_2(s)$$

(Similar for  $\beta=1$ )

# 1.9. GUE Dyson's BM

Time dependent GUE

$N(0,1) \rightarrow$  BM

$$\{B_{ij}, \tilde{B}_{ij}\}_{1 \leq i < j \leq N} : \text{i.i.d. std BM}$$

Def. (Time-dep. GUE  $X = [0, \infty) \rightarrow \mathcal{J}e_N^{(2)}$ )

$$X_{jj} = B_{jj} \quad (1 \leq j \leq N)$$

$$X_{ij} = \frac{1}{\sqrt{2}} (B_{ij} + \sqrt{-1} \tilde{B}_{ij}) \quad (1 \leq i < j \leq N)$$

Thm. (Dyson)  $\{X_i(t)\}$  satisfies

$$dX_i = dW_i + \sum_{j \neq i} \frac{1}{X_i - X_j} dt$$

where  $W_i, 1 \leq i \leq N$  are i.i.d. std. BMs.

$$X(t+dt) = \begin{pmatrix} X_1(t) & & 0 \\ & \ddots & \\ 0 & & X_N(t) \end{pmatrix} \Rightarrow X(t+dt) = \begin{pmatrix} X_1(t) + dB_1(t) & & \\ & \ddots & \\ * & & * \end{pmatrix}$$

1b.  $N=2$   $\Rightarrow$   $\frac{d}{dt} \frac{1}{X_1 - X_2}$  check to 5.

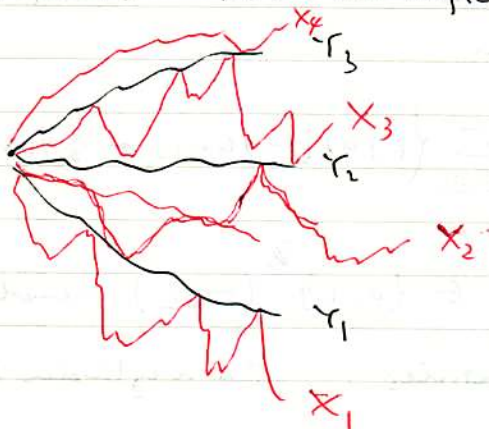
Gelfand - Tsetlin

## 1.5. BM on GT

$Y_j$  ( $1 \leq j \leq N$ ) Dyson's BM with  $N$  particles starting from the origin.

$$i.e. \quad dY_j = dB_j + \sum_{k \neq j} \frac{1}{Y_j - Y_k} dt$$

$X_j$  ( $1 \leq j \leq N+1$ ) BM but reflected by  $Y_{j-1}$  and  $Y_j$



$$dX_j = dB_j + dL_j^- - dL_j^+$$

local time  $X_j = Y_{j-1}$ ,  $X_j = Y_j$

Thm. (Warren)

$X_j$  ( $1 \leq j \leq N+1$ ) is DBM with  $N+1$  particles.

$$\begin{array}{ccc} x_j^{(3)} & & x_j^{(2)} & & x_j^{(1)} \\ & \bullet & & \bullet & \\ x_2^{(2)} & & & & x_1^{(2)} \\ & \bullet & & \bullet & \\ & & & & x_1^{(1)} \end{array}$$

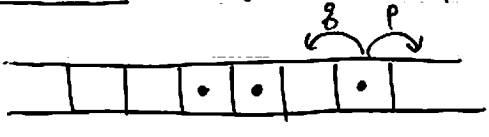
$$y_i^{(k+1)} \geq x_i^{(k)} \geq x_{i+1}^{(k)}$$

## 2. TASEP

### 2.1. ASEP

asymmetric simple exclusion process

Informally



$$p, q \geq 0 \quad pq \neq 0$$

( rescaling of time  
 $\Rightarrow$  summing  $p+q=1$   
 $p=1$  )

More

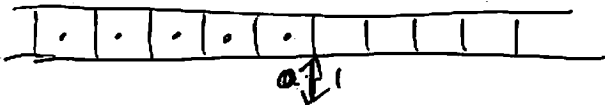
Informally, the process can be constructed using the generator

$$L f(\eta) = \sum_{x \in \mathbb{Z}} \left\{ p \eta(x) (1 - \eta(x+1)) + q \eta(x+1) (1 - \eta(x)) \right\} \left( f(\eta^{(x, x+1)}) - f(\eta) \right)$$

where  $\eta = (\eta_x)_{x \in \mathbb{Z}} \in \{0, 1\}^{\mathbb{Z}} (= \Omega)$  denotes a configuration of particles.  $f$  is a cylinder function on  $\Omega$

and  $\eta^{x, x+1}$  denotes a configuration s.t.  $\eta(x)$  and  $\eta(x+1)$  are swapped in  $\eta$ .

Step initial condition



$$N(t) = \# (\text{particles on } x \geq 1)$$

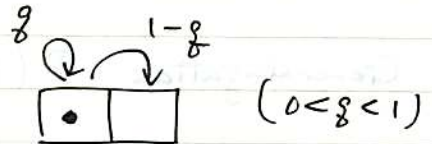
Thm. (Johansson 2000)

$$\mathbb{P}(N(t) > N) = \frac{1}{2^N} \int_{[0,t]^N} \prod_{i < j} (x_i - x_j)^2 \prod_{j=1}^N (e^{-x_j} dx_j)$$

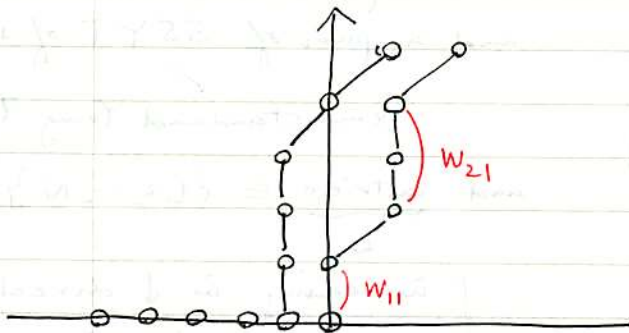
Thm.

$$\lim_{t \rightarrow \infty} \mathbb{P} \left( \frac{N(t) - \frac{t}{2}}{2^{-1/3} t^{1/3}} \geq -s \right) = F_2(s)$$

Main argument of the proof.



Discrete time TASEP



$$w = \{w_{ij}\}_{i,j \geq 1}$$

waiting time

$w_{ij}$  = waiting time for  $i$ -th hop of  $j$ -th particle  
(after it becomes possible)

i.i.d. geometric

$$\mathbb{P}(w_{ij} = n) = (1-g) \cdot g^n$$

$$W = \begin{bmatrix} w_{11} & w_{12} \\ w_{21} & w_{22} \end{bmatrix} \text{ path}$$

$$G(M, N) = \max_{\pi} \left( \sum_{(i,j) \in \pi} w_{ij} \right)$$

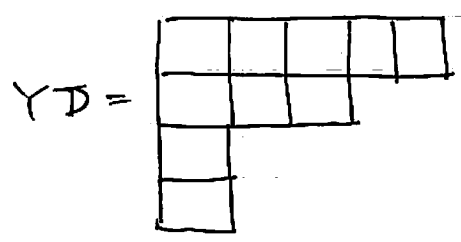
↓  
down-right path from (1,1) to (M,N)

$G(M, N) + M + N - 1 =$  time at which  $N$ th particle has made  $M$ -th hop.

[ Prop  $P(N(t) \geq N) = P(G(N, N) + 2N - 1 \leq t)$  ]

Generalization  $P(w_{ij} = n) = (1 - a_i b_j) (a_i b_j)^n \quad n = 0, 1, 2, \dots$

RSK correspondence = 1:1 map between  $N \times N$ -matrix and a pair of SSYT of some shape



semi standard Young tableaux  
and entries  $\in \{1, 2, \dots, N\}$   
↓  
{ increasing in ↓ direction  
non-decreasing in → direction

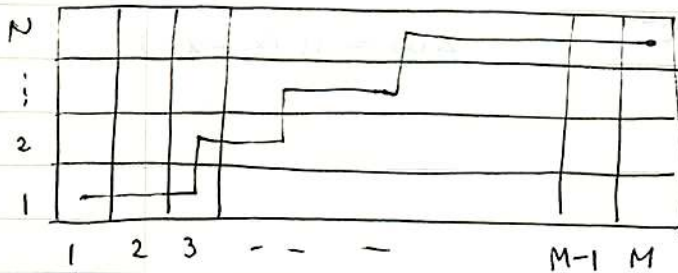
partition  $\lambda = (\lambda_1, \lambda_2, \dots, \lambda_N)$ ,  $\lambda_i \in \mathbb{Z}_+$ ,  $\lambda_1 \geq \lambda_2 \geq \dots$

In addition,  $G(N, N) = \lambda_1 = \frac{1}{2} S_{\lambda}(a) S_{\lambda}(b)$

The mass on  $\lambda$  is

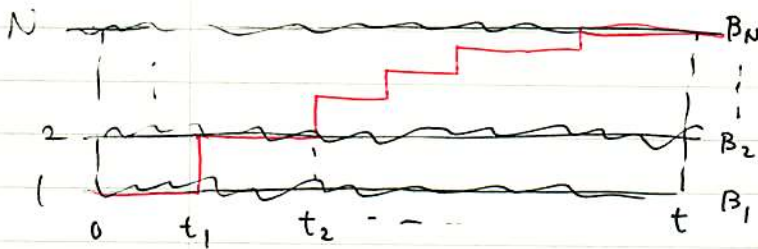
$$\sum_{\substack{T = \text{SSYT} \\ \text{of shape } \lambda}} \underbrace{a_1^{\#(1 \text{ in } T)} \dots a_N^{\#(N \text{ in } T)}}_{S_{\lambda}(a_1, a_2, \dots, a_N)} \times \sum_{\substack{T = \text{SSYT} \\ \text{of shape } \lambda}} b_1^{\#(1 \text{ in } \tilde{T})} \dots b_N^{\#(N \text{ in } \tilde{T})}$$

### 2.3. zero temperature directed polymer



$$G(M, N) : \max$$

$$(M \rightarrow \infty)$$



i.i.d. BM.

$$E = \sum_{i=1}^N (B_i(t_i) - B_i(t_{i-1})) \quad t_0 := 0$$

$$\max_{0=t_0 < t_1 < \dots < t_N \leq t} E \quad \stackrel{d}{=} \quad \chi_{\max}^{\text{QUE}} \quad (\text{Baryshnikov})$$



No. ....

Date .....

復習

5 = 3 " 4 IT 5 | Dyson BM

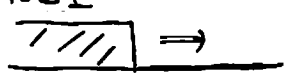
GUE

$$\Delta(x) = \prod_i e^{-x_i^2}$$

$$\Delta(x) = \prod (x_j - x_i)$$

↓  
det f - det g

TASEP



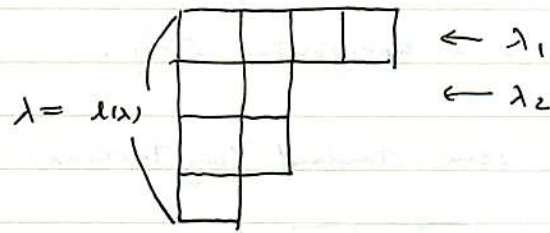
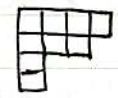
$$\frac{1}{2} S_\lambda(a) S_\lambda(b) \leftarrow \text{Selmer measure}$$

### 3. Schur process

#### 3.1. Schur function

partition  $\lambda = (\lambda_1, \lambda_2, \dots, \lambda_N, 0, \dots)$   $\lambda_j \in \mathbb{Z}_+$   
 $(\lambda_1 \geq \lambda_2 \geq \dots)$   $1 \leq j \leq N$

ex.  $\lambda = (4, 2, 2, 1) \leftrightarrow \lambda' = (4, 3, 1, 1)$



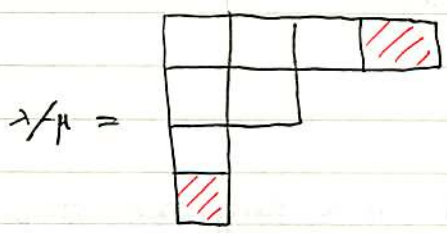
$$|\lambda| = \lambda_1 + \lambda_2 + \dots + \lambda_N$$

skew Young diagram

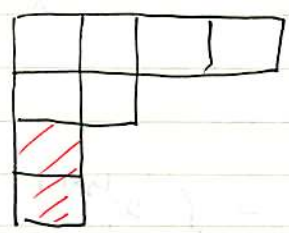
$\lambda, \mu$  -- Young diagram s.t.  $\lambda \supset \mu \iff \lambda_i \geq \mu_i \ (\forall i)$   
 $\lambda/\mu = \lambda - \mu \quad |\lambda/\mu| = |\lambda| - |\mu|$

$\lambda/\mu$  is a horizontal strip

(def) In each column,  $\lambda/\mu$  has at most one box.



horizontal



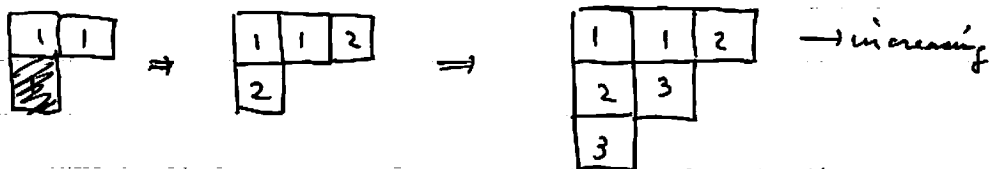
not horizontal

column-strict (skew) Young diagram tableaux is a sequence of partitions

$$\mu = \lambda^{(0)} \subset \lambda^{(1)} \subset \dots \subset \lambda^{(M)} = \lambda$$

s.t.  $\lambda^{(i)} / \lambda^{(i-1)}$  ( $1 \leq i \leq M$ ) is a horizontal strip.

If  $\mu = \emptyset$ , this is equivalent to semi-standard Young tableaux.



↓ strictly increasing

Def. (the (skew) Schur function for a skew YD  $\lambda/\mu$ )

$$S_{\lambda/\mu}(x_1, \dots, x_N) = \sum_{T = \text{column strict tableaux of shape } \lambda/\mu} x^T$$

$$x^T = \prod_{i=1}^N x_i^{|\lambda^{(i)} / \lambda^{(i-1)}|}$$

When  $N=1$ ,

$$S_{\lambda/\mu}(x_1) = \begin{cases} x_1^{|\lambda/\mu|} & \text{If } \lambda/\mu \text{ is a horizontal strip} \\ & (\text{i.e., } \mu_i \leq \lambda_i \leq \mu_{i-1}) \\ 0 & \text{otherwise} \end{cases}$$

$$S_{\lambda'/\mu'}(x) = \begin{cases} x_1^{|\lambda'/\mu'|} & \text{if } \lambda'/\mu' \text{ is a horizontal strip.} \\ & \text{i.e. } (\lambda_j = \mu_j \text{ or } \lambda_j = \mu_j + 1) \\ & \text{and} \\ & \lambda_j \geq \mu_j \\ 0 & \text{otherwise} \end{cases}$$

Involution:  $\omega$  s.t.  $\omega S_{\lambda/\mu}(x) = S_{\lambda'/\mu'}(x)$

Some relations for Schur functions

Write  $x = (x_1, \dots, x_n)$ ,  $y = (y_1, \dots, y_m)$

Define  $\Pi(x, y) = \prod_{i=1}^n \prod_{j=1}^m (1 - x_i y_j)^{-1}$

(cf. Macdonald)

$$\sum_{\mu} S_{\mu/\lambda}(x) S_{\mu/\nu}(y) = \sum_{\tau} S_{\nu/\tau}(x) S_{\lambda/\tau}(y) \Pi(x, y)$$

$$\sum_{\lambda} S_{\lambda}(x) S_{\mu/\lambda}(y) = S_{\mu}(x, y) = S_{\lambda}(x_1, \dots, x_n, y_1, \dots)$$

By applying  $\omega_y$ ,

$$\sum_{\mu} S_{\mu/\lambda}(x) S_{\mu'/\nu'}(y) = \sum_{\tau} S_{\nu/\tau}(x) S_{\lambda'/\tau'}(y) \underbrace{\omega_y \Pi(x, y)}$$

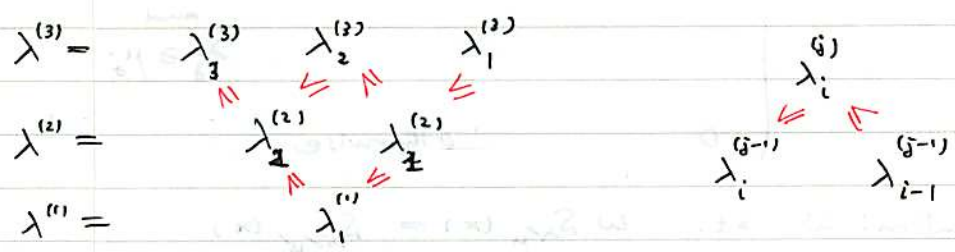
$\omega_x \omega_y$

$$\prod_{i=1}^n \prod_{j=1}^m (1 + x_i y_j)$$

$$\sum_{\lambda} S_{\lambda}(x) S_{\mu'/\lambda'}(y) = S_{\mu'}(x, y)$$

### 3.2. A dynamics on Gelfand-Tsetlin pattern

$$GT_N = \left\{ \lambda_i^{(j)}, 1 \leq i \leq j \leq N \mid \begin{array}{l} \lambda_i^{(j)} \leq N \\ \lambda_i^{(j-1)} \leq \lambda_i^{(j)} \leq \lambda_{i-1}^{(j-1)} \end{array} \right\}$$



Each  $\lambda^{(j)} = \{ \lambda_i^{(j)} \mid 1 \leq i \leq j \}$  is a YD.

① Initial condition  $\lambda_i^{(j)}(0) = 0 \quad (\forall i, j)$

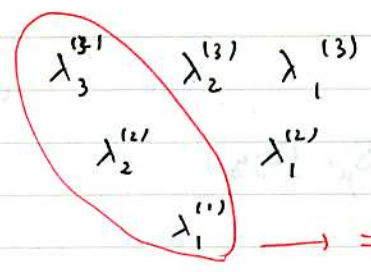
Dynamics order  $\lambda^{(1)} \rightarrow \lambda^{(2)} \rightarrow \dots \rightarrow \lambda^{(N)}$

Basically

$$\lambda_i^{(j)}(t+1) = \lambda_i^{(j)}(t) + \begin{cases} 1 & \text{w.p. } p \\ 0 & \text{w.p. } 1-p \end{cases} \quad p \in (0, 1)$$

If  $\lambda_i^{(j-1)}(t+1) = \lambda_i^{(j)}(t)$ , then  $\lambda_i^{(j)}(t+1) = \lambda_i^{(j)}(t)$ . (blocking)

If  $\lambda_i^{(j-1)}(t+1) = \lambda_i^{(j)}(t) + 1$ , then  $\lambda_i^{(j)}(t+1) = \lambda_i^{(j)}(t) + 1$  (pushing)



$\rightarrow \Rightarrow \text{not } \exists \text{ } \lambda_i^{(j)} \text{ that is "KIE" TASEP}$

### 3.3. Schur process

Prop: One step transition probability for this dynamics is given by

$$G(\underline{\mu}, \underline{\lambda}) = \mathbb{P}(\underline{\lambda}^{(t+1)} = \underline{\lambda} \mid \underline{\lambda}^{(t)} = \underline{\mu})$$

$$= \prod_{j=1}^N \frac{S_{\lambda^{(j)}/\lambda^{(j-1)}}(1) S_{\lambda^{(j)}/\mu^{(j)}}(r)}{\Delta(\lambda^{(j-1)}, \mu^{(j)})}$$

with  $r = \frac{p}{1-p}$  and

$$\Delta(\lambda, \mu) = \sum_{\nu} S_{\nu/\lambda}(1) S_{\nu/\mu}(r)$$

Ex. ( $N=2$ )

$$\mathbb{P}\left(\underline{\lambda}^{(t+1)} = \begin{pmatrix} \lambda_2^{(2)} & \lambda_1^{(2)} \\ \lambda_1^{(1)} \end{pmatrix} \mid \underline{\lambda}^{(t)} = \begin{pmatrix} \mu_2^{(2)} & \mu_1^{(2)} \\ \mu_1^{(1)} \end{pmatrix}\right)$$

$$\stackrel{?}{=} \frac{S_{\lambda^{(1)}/\lambda^{(1)}}(1) S_{\lambda^{(2)}/\mu^{(2)}}(r)}{\Delta(\lambda^{(1)}, \mu^{(2)})} \times \frac{S_{\lambda^{(2)}(1)} S_{\lambda^{(2)}/\mu^{(2)}}}{\Delta(\emptyset, \mu^{(1)})}$$

$$\left( (\mu_2^{(2)}, \mu_1^{(2)}) \Rightarrow (\lambda_2^{(2)}, \lambda_1^{(2)}) \right) \quad \left( \mu_1^{(1)} \Rightarrow \lambda_1^{(1)} \right)$$

$$\mu_1^{(1)} \Rightarrow \lambda_1^{(1)}$$

$$S_{\lambda^{(1)}}(1) = \begin{cases} 1 & \lambda_i^{(1)} = 0 \quad (i \geq 2) \\ 0 & \text{otherwise} \end{cases}$$

$$S_{\lambda^{(2)}/\mu^{(2)}}(r) = \begin{cases} r & \lambda_1^{(2)} = \mu_1^{(1)} + 1, \mu_2^{(2)} \leq \lambda_2^{(2)} \quad (i \geq 2) \\ 1 & \lambda_1^{(2)} = \mu_1^{(1)}, \mu_2^{(2)} \leq \lambda_2^{(2)} \quad (i \geq 2) \\ 0 & \text{otherwise} \end{cases}$$

$$\Delta(\phi, \mu^{(n)}) = \begin{cases} 1+r & \mu_i^{(n)} = 0 \ (i \geq 2) \\ 0 & \text{otherwise} \end{cases}$$

$$\frac{S_{\lambda^{(n+1)}} S_{\lambda^{(n)}/\mu^{(n)}}(r)}{\Delta(\phi, \mu^{(n)})} = \begin{cases} \frac{r}{1+r} = p & \text{if } \lambda_1^{(n+1)} = \mu_1^{(n)}, \mu_i^{(n)} = \lambda_i^{(n)} = 0 \ (i \geq 2) \\ \frac{1}{1+r} = 1-p & \text{if } \lambda_1^{(n+1)} = \mu_1^{(n)}, \mu_i^{(n)} = \lambda_i^{(n)} = 0 \ (i \geq 2) \\ 0 & \text{otherwise} \end{cases}$$

Thm. The prob. that at time  $t$  the config. is  $\underline{\lambda}$  is

$$P_{1,r}(\underline{\lambda}, t) = \frac{S_{\lambda^{(1)}} S_{\lambda^{(2)}/\lambda^{(1)}} \dots S_{\lambda^{(n)}/\lambda^{(n-1)}} S_{\lambda^{(n)}/\underline{\lambda}}(r)}{(1+r)^{tN}}$$

Here we check

$$\sum_{\underline{\mu}} P_{1,r}(\underline{\mu}, t) G(\underline{\mu}, \underline{\lambda}) = P_{1,r}(\underline{\lambda}, t+1)$$

for  $N=2$ .

Indeed,

$$\sum_{\mu^{(1)}, \mu^{(2)}} P_{1,r}(\mu^{(1)}, \mu^{(2)}, t) G(\mu^{(1)}, \mu^{(2)}, \lambda^{(1)}, \lambda^{(2)})$$

$$= \sum_{\mu^{(1)}, \mu^{(2)}} \frac{S_{\mu^{(1)}} S_{\mu^{(2)}/\mu^{(1)}}(r)}{(1+r)^{2t}} \cdot \frac{S_{\lambda^{(1)}/\lambda^{(2)}} S_{\lambda^{(2)}/\mu^{(2)}}(r) S_{\lambda^{(1)}} S_{\lambda^{(2)}/\mu^{(1)}}}{\Delta(\lambda^{(1)}, \mu^{(1)}) \cdot \Delta(\phi, \mu^{(1)})}$$

$$\begin{aligned}
 &= \sum_{\mu^{(2)}} \frac{S_{\mu^{(2)}}(r, \dots, r)^t}{(1+r)^{2t}} \cdot \frac{S_{\lambda^{(2)}/\lambda^{(1)}}(1) S_{\lambda^{(1)}/\mu^{(1)}}(r) S_{\lambda^{(1)}}(1)}{(1+r)^2} \\
 &= \frac{S_{\lambda^{(1)}}(1) S_{\lambda^{(2)}/\lambda^{(1)}}(1) S_{\lambda^{(1)}}(r, \dots, r)^{t+1}}{(1+r)^{2(t+1)}}
 \end{aligned}$$

One can consider a similar process on  $G_{TN}$  s.e. prob. is given by.

$$P_{a,b}(\lambda, t) = \frac{1}{2} S_{\lambda^{(1)}}(a_1) S_{\lambda^{(2)}/\lambda^{(1)}}(a_2) \dots S_{\lambda^{(N)}/\lambda^{(N-1)}}(a_N) \times S_{\lambda}(b_1, \dots, b_t)$$

Schur process. (  $a=1, b=r$  a  $z \neq 1$  M.C. a prob. )

$\Downarrow \Rightarrow z = \lambda^{(N)}$   $T = \text{it } \mathbb{R}^3 \text{ etc}$

$$\frac{1}{2} S_{\lambda}(a) S_{\lambda}(b) \quad (\text{Schur meas.})$$



## 4. KPZ equation

$$\frac{\partial h}{\partial t} = \frac{\lambda}{2} \left( \frac{\partial h}{\partial x} \right)^2 + \nu \frac{\partial^2 h}{\partial x^2} + \sqrt{D} \eta(x, t)$$

$\eta$ : Gaussian white noise with  $\langle \eta(x, t) \eta(x', t') \rangle = \delta(x-x') \delta(t-t')$

### 4.1. Scaling

$$\text{Set } \alpha = \frac{\lambda D^{1/2}}{(2\nu)^{3/2}}$$

$$\text{For } \tilde{x} = \alpha^2 x, \quad \tilde{t} = 2\nu \alpha^4 t, \quad \tilde{h} = \frac{\lambda}{2\nu} h$$

$$\Rightarrow \frac{\partial \tilde{h}}{\partial \tilde{t}} = \frac{1}{2} \left( \frac{\partial \tilde{h}}{\partial \tilde{x}} \right)^2 + \frac{1}{2} \frac{\partial^2 \tilde{h}}{\partial \tilde{x}^2} + \eta(\tilde{x}, \tilde{t})$$

### 4.2. Cole-Hopf solution

We can "define"

$$h(x, t) = \log Z(x, t)$$

where  $Z(x, t)$  is the solution to SHE

$$dZ = \frac{1}{2} \frac{\partial^2}{\partial x^2} Z dt + Z \cdot dB(x, t)$$

One can modify the noise as

$$\langle \eta_k(x, t) \eta_k(x', t') \rangle = C_k(x-x') \delta(t-t')$$

SHE

$$\frac{\partial Z_k}{\partial t} = \frac{1}{2} \frac{\partial^2}{\partial x^2} Z_k + \eta_k Z_k$$

$$C_k(x) \xrightarrow{k \rightarrow \infty} \delta_0(x)$$

$$\text{Then } \lim_{k \rightarrow \infty} Z_k = Z$$

$$\text{For } h_k = \log Z_k, \quad \frac{\partial h_k}{\partial t} = \frac{1}{2} \left( \frac{\partial h_k}{\partial x} \right)^2 + \frac{1}{2} \frac{\partial^2 h_k}{\partial x^2} + \eta_k - \frac{1}{2} C_k(0)$$

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### 4.3. Feynman - Kac Formula

#### 4.3.1. usual formula.

The solution to

$$\frac{\partial u}{\partial t} = \frac{1}{2} \frac{\partial^2 u}{\partial x^2} + V(x) \cdot u, \quad u(x, 0) = f(x).$$

is written as

$$u(x, t) = E_x \left[ f(b_t) \exp \int_0^t V(b_s) ds \right]$$

$b_s = \text{std. BM s.t. } b_0 = x.$

#### 4.3.2. SHE case

Bertini -

$$Z_\kappa(x, t) = E \left[ \exp \int_0^t \gamma_\kappa(b(s), t-s) ds \times Z_\kappa(b(t), 0) \right] \times e^{-\frac{1}{2} \kappa |x|^2 t}$$

$$Z(x, t) := \lim_{\kappa \rightarrow \infty} Z_\kappa(x, t) \quad \text{"the limit"}$$

#### 4.4. Replica method

Random Ising model

$$H = - \sum_{\langle i,j \rangle} J_{ij} s_i s_j \quad \leftarrow \text{random} \quad s_i \in \{-1, 1\}$$

partition function

$$Z = \sum_{s_i \in \{-1, 1\}} e^{-\beta H}$$

Free energy  $F = \log Z.$

$$\downarrow$$

$$\langle \log Z \rangle_{\mathcal{R}} \quad \text{average over } \mathcal{J} = (J_{ij})$$

In some cases, it is easier to compute  $\langle Z^N \rangle$

#### ① Replica trick

$$\langle \log Z \rangle = \lim_{N \rightarrow 0} \frac{\langle Z^N \rangle - 1}{N}$$

#### Moment problem

For a prob. density  $p(x)$ ,  $x \in \mathbb{R}$  the  $n$ -th moment

$$m_n = \int_{\mathbb{R}} t^n p(t) dt \quad n = 0, 1, \dots$$

Is it possible to recover  $p(t)$  from  $\{m_n\}$ ?

Generating function

$$g(z) = \sum_{n=0}^{\infty} \frac{m_n}{z^{n+1}} = \sum_{n=0}^{\infty} \frac{1}{z^{n+1}} \int_{\mathbb{R}} t^n p(t) dt$$

$$\doteq \int_{\mathbb{R}} \frac{1}{z-t} p(t) dt$$

$$p(x) = \frac{1}{2\pi i} \oint_{\varepsilon < 0} \lim_{\varepsilon \rightarrow 0} (g(x+i\varepsilon) - g(x-i\varepsilon))$$

But, in general,  $p(t)$  is not determined by  $\{m_n\}$ .

Consider a r.v.  $t \sim N(0,1)$ , put  $Z = e^t$ .

$Z$  is distributed as log normal

$$P(Z \in dz) = \frac{1}{\sqrt{2\pi} z} e^{-\frac{(\log z)^2}{2}} \mathbb{1}_{z>0} dz$$

It is easy to see that

$$\langle Z^n \rangle = \int_0^{\infty} z^n \frac{1}{\sqrt{2\pi} z} e^{-\frac{(\log z)^2}{2}} dz = e^{\frac{n^2}{2}}$$

For any a.s.t.  $|a| \leq 1$

$$\frac{1}{\sqrt{2\pi} z} e^{-\frac{(\log z)^2}{2}} (1 + a \sin(2\pi \log z)) \mathbb{1}_{z>0}$$

gives the same moment  $e^{\frac{n^2}{2}}$ .

$$m_n = e^{-\frac{n^2}{2}} \quad n \geq 2.$$

$$f(z) = \sum_{n=0}^{\infty} \frac{1}{z^{n+1}} \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{nx - \frac{nx^2}{2}} dx$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{x^2}{2}} \frac{dx}{z - e^x}$$

$$= \frac{1}{\sqrt{2\pi}} \int_0^{\infty} \frac{dz}{z} e^{-\frac{(z)^2}{2}} \frac{1}{z - z}$$

$$\begin{aligned} \therefore \frac{P(z-dz)}{dz} &= \frac{1}{2\pi i} \left( f(z+i\varepsilon) - f(z-i\varepsilon) \right) \\ &= \frac{1}{\sqrt{2\pi} z} e^{-\frac{(z)^2}{2}} dz. \end{aligned}$$

KP2

$$\langle z^n \rangle \sim e^{\frac{n^3}{3}} = \int_{\mathbb{R}} A_i(x) e^{nx} dx$$

4.5.  $\delta$ -Bose gas

For KPZ,

$$\left\langle e^{-e^{2(x,t) + \frac{t}{2\epsilon} - \chi_t}} \right\rangle = \sum_{N=0}^{\infty} \frac{(-e^{-\frac{t}{2\epsilon} + \frac{t}{2\epsilon}})^N}{N!} \langle Z(x,t)^N \rangle$$

Using Feynman-Kac representation,

$$Z_K(x,t) = E_{x_1} \left[ \exp\left(\int_0^t \gamma_K(b(s), t-s) ds\right) \cdot Z_K(b(t), 0) \right] \times e^{-\frac{1}{2} C_K(0)t}$$

$$= \left\langle E_{x_1} \dots E_{x_N} \left[ \prod_{i=1}^N \exp\left(\int_0^t \gamma_K(b_i(s_i), t-s_i) ds_i\right) \cdot Z_K(b_i(t_i), 0) \right] e^{-\frac{1}{2} C_K(0)t} \right\rangle$$

$$\left( \left\langle \prod_{i=1}^N \exp\left(\int_0^t \gamma_K(b_i(s_i), t-s_i) ds_i\right) \right\rangle = e^{\frac{1}{2} \sum_{j \neq k}^N C_K(x_j - x_k)} \right)$$

$$E_{x_1} \dots E_{x_N} \left[ e^{\int ds \frac{1}{2} \sum_{j \neq k}^N C_K(b_j(s) - b_k(s))} \prod_{i=1}^N Z(b_i(t_i), 0) \right]$$

$$= \langle x_1 \dots x_N | e^{-H_{NK}t} | 0, 0, \dots, 0 \rangle$$

Therefore,  $G_N(x_1, \dots, x_N; t)$  satisfies

$$\begin{cases} \frac{\partial G_N(x_1, \dots, x_N; t)}{\partial t} = -H_{NK} G_N(x_1, \dots, x_N; t) \\ G_N(x_1, \dots, x_N; 0) = \prod_{i=1}^N \delta(x_i) \left\langle \prod_{i=1}^N Z(x_i, 0) \right\rangle \end{cases}$$

$$\text{where } H_{NK} = \frac{1}{2} \sum_{j \neq k}^N \frac{\partial^2}{\partial x_j^2} - \frac{1}{2} \sum_{j \neq k}^N C_K(x_j - x_k)$$

As  $\kappa \rightarrow \infty$ ,

$$G_N(x_1, \dots, x_N; t) = \left\langle \prod_{i=1}^N \mathcal{Z}(x_i; t) \right\rangle$$

satisfies

$$\frac{\partial G_N}{\partial t} = -H_N G_N, \quad H_N = \frac{-1}{2} \sum_{j=1}^N \frac{\partial^2}{\partial x_j^2} - \frac{1}{2} \sum_{j \neq k} f(x_j - x_k)$$

$$G_N(\cdot, t=0) = \left\langle \prod_{i=1}^N \mathcal{Z}(x_i; 0) \right\rangle$$

$\rightarrow$  wedge  $\prod_{i=1}^N \delta(x_i)$

Hamiltonian of (attractive) Bose gas.

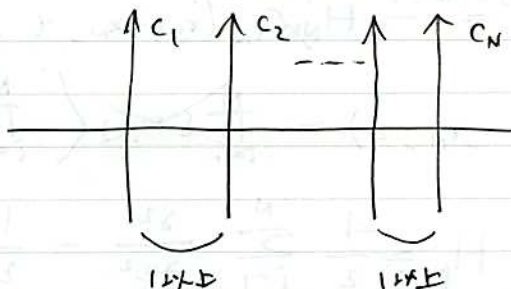
$$\frac{\partial G}{\partial t} = \frac{1}{2} \sum_{j=1}^N \frac{\partial^2}{\partial x_j^2} G$$

$$x_1 < x_2 < \dots < x_N$$

$$(\partial_{x_i} - \partial_{x_{i+1}} - 1) G \Big|_{x_{i+1} = x_i + 0} = 0 \quad (1 \leq i \leq N-1)$$

For wedge  $\mathcal{Z}(x, 0) = \delta(x)$

$$\left\langle \prod_{i=1}^N \mathcal{Z}(x_i; t) \right\rangle = \left( \frac{1}{2\pi i} \right)^N \int_{C_1} \dots \int_{C_N} \prod_{1 \leq A < B \leq N} \frac{z_A - z_B}{z_A - z_B - 1} e^{\frac{1}{2} \sum_{j=1}^N z_j^2 t + \sum_{j=1}^N x_j z_j} dz_1 \dots dz_N$$



$$H_N = -\frac{1}{2} \sum_j \frac{z^2}{2x_j^2} - \frac{1}{2} \sum_{j \neq k} \delta(x_j - x_k)$$

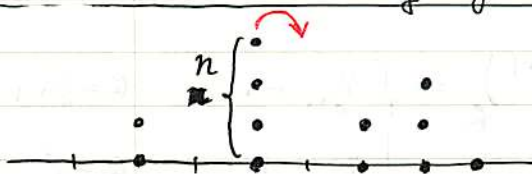
$$\text{wedge} \left\langle \prod_i z(x_i, 0) \right\rangle = \prod_i \delta(x_i)$$

$$\left\langle \prod_i z(x_i, t) \right\rangle = \int \dots \int \prod_{A < B} \frac{z_A - z_B}{z_A - z_B - 1} \prod_i e^{\frac{1}{2} z_j^2 t + x_j z_j}$$

## 5. $q$ TASEP and duality

$q$  TAZRP

### 5.1. Process totally asymmetric zero range process



only jump to the right

$$\text{jump rate } [n] = \frac{1 - q^n}{1 - q} \quad (0 \leq q < 1)$$

$$[n]! = [n] \cdot [n-1] \cdots [1]$$

$$(\alpha; q)_n = (1 - \alpha)(1 - \alpha q) \cdots (1 - \alpha q^{n-1})$$

Rem.  $q \rightarrow 1$  indep.  
 $q = 0$  rate 1 (equivalent to TASEP)

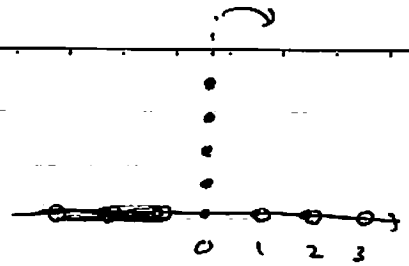
Stationary meas. at 1-site

$$P(\eta_j = n) = \left( \frac{\alpha}{1 - q}; q \right)_\infty \frac{\alpha^n}{[n]!}$$



Initial condition

$$\eta_0 = +\infty, \quad \eta_j = 0 \quad \text{if } \begin{matrix} j < 0 \\ j \geq 1 \end{matrix}$$



Height function

$$h_j = \sum_{k=j}^{\infty} \eta_k \quad (j \geq 1)$$

Generator

$$If(z) = \sum_{j \in \mathbb{Z}} [\eta_j] (f(z^{j, j+1}) - f(z_j))$$

$$(\eta^{j, j+1})_k = \begin{cases} \eta_{j+1} + 1 & k = j+1 \\ \eta_j - 1 & k = j \\ \eta_k & k \neq j, j+1 \end{cases}$$

## 5.2. Result

Thm.

$$\left\langle \frac{1}{(g^{\eta w}; g)_{\infty}} \right\rangle = \det (I + K_g)_{L^2(C_w)},$$

where the kernel

$$K_g(w, w') = \frac{1}{2\pi i} \int_{-i\infty + \delta}^{i\infty + \delta} \Gamma(-s) \Gamma(1+s) (-s)^s g_{w, w'}(g^s) ds$$

$$g_{w, w'}(x) = \frac{1}{xw - w'} \left( \frac{(xw; g)_{\infty}}{(w; g)_{\infty}} \right)^N e^{tw(x-1)}$$

$C_w$  -- small contour around 1



$$z_j = q^{h_j}$$

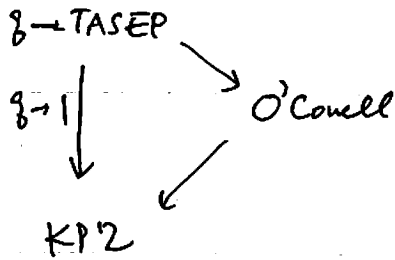
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$$\text{KP2 } a z z \quad e^{-e^{a + \frac{t}{z^k} - x_t s}}$$

$$\langle e^{-e^{a + \frac{t}{z^k} - x_t s}} \rangle = \det(I - K_{t,s})$$

$$K_{t,s}(x,y) = \int_{\mathbb{R}} \frac{A_i(x+\lambda) A_i(y+\lambda)}{e^{\frac{t}{\lambda}(\lambda-s)} + 1} d\lambda$$



$$dz_j = (z_{j+1} - z_j) dt + z_j dB_j$$

Rem. 1 By the  $q$ -binomial thm,

$$\frac{1}{(x; q)_{\infty}} = \sum_{n=0}^{\infty} \frac{x^n}{(q; q)_{\infty}}$$

LHS is written as

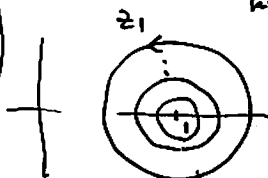
$$\sum_{n=0}^{\infty} \frac{z^n}{(q; q)_{\infty}} \langle q^{n z_N} \rangle \quad z_N = q^{t_N}$$

The theorem is based on the following formula for  $\langle q^{n z_N} \rangle$

Prop.

$$\langle q^{n z_N} \rangle = (-1)^n q^{\frac{n(n-1)}{2}} \int \frac{dz_1}{2\pi i} \dots \int \frac{dz_n}{2\pi i} \prod_{k < l} \frac{z_k - z_l}{z_k - q z_l} \times \prod_{k=1}^n (1 - z_k)^{-N} \frac{e^{-t z_k}}{z_k}$$

( $\gamma_k$  = contains  $\{q \gamma_k\}_{k \geq k}$  and 1) but not 0



prop  $\rightarrow$  Thm

Prop. For a meromorphic function  $f(z)$ ,  $A$  = a fixed set of singularities of  $f$ ,

Assume  $\bigcap_{m=1}^n A \cap A = \emptyset$  ( $\forall m \geq 1$ )

$$\begin{aligned} & \frac{(-1)^n}{2\pi i} q^{\frac{n(n-1)}{2}} \int \dots \int \prod_{1 \leq A < B \leq n} \frac{z_A - z_B}{z_A - qz_B} \frac{f(z_1) \dots f(z_n)}{z_1 \dots z_n} dz_1 \dots dz_n \\ &= (q; q)_n \sum_{\substack{\lambda \vdash n \\ \lambda = 1^{m_1} 2^{m_2} \dots}} \frac{1}{m_1! m_2! \dots (2\pi i)^{l(\lambda)}} \int \dots \int \det \left[ \frac{1}{w_i q^{\lambda_j} - w_j} \right]_{i,j=1}^{l(\lambda)} \\ & \quad \times \prod_{j=1}^{l(\lambda)} f(w_j) f(qw_j) \dots f(q^{j-1} w_j) dw_j \end{aligned}$$

$z_j$  contains  $\{qz_k\}_{k>j}$ ,  $A$  but not 0

$w_j$  contains  $A$

For  $q$  TAZRP we apply this prop with

$$f(z) = \left( \frac{1}{1-z} \right)^N e^{-tz}$$

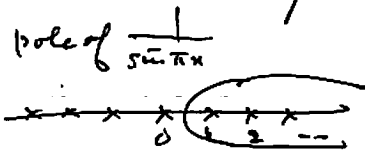
Then we can proceed as

$$\left\langle \frac{1}{(q; q)_{2N}; q} \right\rangle = \sum_{n=0}^{\infty} \sum_{L=0}^{\infty} \frac{1}{m_1! m_2! \dots} \left( \frac{1}{2\pi i} \right)^L \int \dots \int \begin{cases} \sum u_i = L \\ \sum i m_i = n \end{cases}$$

$$= \sum_{L=0}^{\infty} \frac{1}{L!} \sum_{n_1, \dots, n_L \geq 1} \int \dots \int \det \left[ \frac{1}{g^{n_i} w_i - w_j} \right]_{i,j=1}^L \prod (1-g)^{n_j} \zeta^{n_j} \\ \times f(w_j) \dots f(g^{n_j-1} w_j) \frac{dw_j}{2\pi i}$$

$$= \det(I + K)_{L^2(\mathbb{Z}_{\geq 0} \times \mathbb{C}_w)}$$

$$\left( K(n_1, w_1; n_2, w_2) = \frac{1}{g^{n_1} w_1 - w_2} \zeta^{n_1} f(w_1) \dots f(g^{n_1-1} w_1) \right) \\ = \zeta^{n_1} g_{w_1, w_2}(g^{n_1})$$

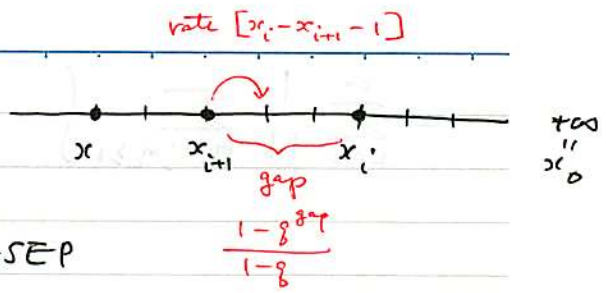


$$K(w_1, w_2) = \sum_{n \geq 1} \zeta^n g_{w_1, w_2}(g^n)$$

$$= \frac{1}{2\pi i} \int_{-i\infty + \delta}^{i\infty + \delta} \Gamma(-s) \Gamma(1+s) (-\zeta)^s g_{w_1, w_2}(g^s) ds$$

### 5.3. Duality

$\delta$ -TASEP  $\xrightarrow{\delta \downarrow 0}$  TASEP



$$\eta_j(t) = x_j(t) - x_{j+1}(t) - 1 \text{ is } \delta\text{-TASEP}$$

step init. cond. for  $\delta$ -TASEP

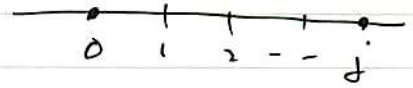
$$x_j(0) = -j, \quad j \geq 1$$

corresponds to  $\eta_0(0) = +\infty$  and  $\eta_j(0) = 0$  ( $j \geq 1$ )

Moreover, 
$$x_j(t) + j = \sum_{k=j}^{\infty} \eta_k(t) = L_j(t)$$

Slightly different version of the  $\delta$ -TASEP

~~There is~~ no hopping once a particle enters  $0$  rate  $[\xi_j]$



Define

$$D(x, \xi) = \prod_{j=0}^n \delta^{(x_j + j) \xi_j} = \begin{cases} 0 & \text{if } \xi_0 > 0 \\ \prod_{j=1}^n \delta^{(x_j + j) \xi_j} & \text{if } \xi_0 = 0 \end{cases}$$

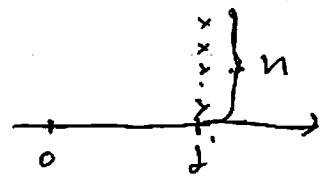
$\uparrow$   $\delta$ -TASEP       $\uparrow$   $\delta$ -TASEP

Prop (duality) ( $\forall n$ )

$$\left[ E^x [D(x(t), \xi)] = E^{\xi} [D(x, \xi(t))] \right]$$

$x(0) = x, \quad \xi(0) = \xi$

Cor.

$$\langle \rho^{n(x_j + j)} \rangle = E^{\xi(0)} [ 1_{\xi_0(t)=0} ]$$


= Prob [ starting from no particle has arrived 0 at time t ]

$$= \sum_{1 \leq y_1 \leq y_2 \leq \dots \leq y_N} P(y; t | y_0; t)$$

$$\begin{cases} y = (y_1, \dots, y_N) \\ y_0 = (j, \dots, j) \end{cases}$$

transition prob. for n particles  
 ρ-TAZRP (3)

$$P(y; t | y_0, 0) = \int \bar{\Psi}_z^r(y) \bar{\Psi}_z^l(y_0) e^{\Lambda_z t} d\mu(z)$$

$$\Lambda_z = \text{eigenvalue} \quad \Lambda_z = - \sum_{i=1}^n z_i$$

$\bar{\Psi}_z^{r,l}(y)$  : right, left eigenfunction of  $I_n$

z = label

ρ-TASEP

$d\mu(z)$  = (spectral) measure.

By Bethe ansatz for  $z = (z_1, \dots, z_n)$

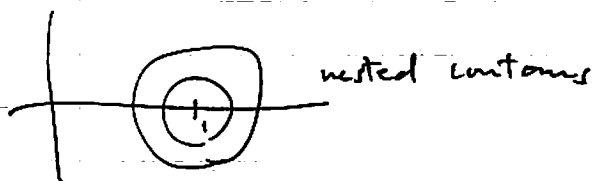
$$\bar{\Psi}_z^l(y) = \sum_{\sigma \in S_n} \prod_{1 \leq B < A \leq n} \frac{z_{\sigma(A)} - \rho z_{\sigma(B)}}{z_{\sigma(A)} - z_{\sigma(B)}} \prod_{j=1}^n (1 - z_{\sigma_j})^{-y_j}$$

$$\bar{\Psi}_z^r(y) = \frac{1}{\rho^{|y|}} \sum_{\sigma \in S_n} \prod_{1 \leq B < A \leq n} \frac{z_{\sigma(A)} - \rho^{-1} z_{\sigma(B)}}{z_{\sigma(A)} - z_{\sigma(B)}} \prod_{j=1}^n (1 - z_{\sigma_j})^{y_j}$$

where  $C_j(y)$  is appeared for diagonal  $y_i = y_j$ .

Thm. (Plancherel formula for  $n$ -particle  $\beta$ TASEP)

$$\int \frac{dz_1}{2\pi i} \dots \int \frac{dz_n}{2\pi i} \prod_{1 \leq A < B \leq n} \frac{z_A - z_B}{z_A - \beta z_B} \prod_{i=1}^n (1 - z_i)^{-y_i - 1} \sum_{\lambda} f^{(\lambda)} \overline{\Psi}_2^r(\lambda) = f(y)$$



By shrinking contours,

Cor (completeness)

$$\sum_{\lambda \vdash n} \int \dots \int d\mu_\lambda(w) \prod_{j=1}^{e(\lambda)} \frac{1}{(w_j; \beta)_{\lambda_j}} \overline{\Psi}_{w(\lambda)}^r \Psi_{w(\lambda)}^e = \delta_{xy}$$

where  $w(\lambda) = (w_1, \beta w_1, \dots, \beta^{\lambda_1-1} w_1, w_2, \dots, \beta^{\lambda_2-1} w_2, \dots, \beta^{\lambda_e-1} w_e)$

$$d\mu_\lambda(w) = \frac{(1-\beta)^n (-1)^n \beta^{-\frac{n^2}{2}}}{n_1! n_2! \dots} \det \left[ \frac{1}{w_i \beta^{\lambda_i} - w_j} \right]_{i,j=1}^{e(\lambda)} \prod_{i=1}^{e(\lambda)} w_i \beta^{\lambda_i} \beta^{\frac{\lambda_i^2}{2}} \frac{dw_i}{2\pi i}$$

## 6. Macdonald process

A generalization of Schur process

(Shew)

### 6.1. Macdonald function / polynomial

$q, t$  : indeterminate

Define  $f(u) = \frac{(tu; q)_\infty}{(qu; q)_\infty}$

$$\varphi_{\lambda/\mu} = \prod_{1 \leq i < j \leq \ell(\lambda)} \frac{f(q^{\lambda_i - \lambda_j} t^{\delta - i}) f(q^{\mu_i - \mu_{j+1}} t^{\delta - i})}{f(q^{\lambda_i - \mu_j} t^{\delta - i}) f(q^{\mu_i - \lambda_{j+1}} t^{\delta - i})}$$

$$\psi_{\lambda/\mu} = \prod_{1 \leq i < j \leq \ell(\lambda)} \frac{f(q^{\mu_i - \mu_j} t^{\delta - i}) f(q^{\lambda_i - \lambda_{j+1}} t^{\delta - i})}{f(q^{\lambda_i - \mu_j} t^{\delta - i}) f(q^{\mu_i - \lambda_{j+1}} t^{\delta - i})}$$

The (shew) Macdonald  $x = (x_1, \dots, x_N)$

$$P_{\lambda/\mu}(x) = \sum_T \varphi_T x^T, \quad \varphi_T = \prod_{i=1}^N \varphi_{\lambda^{(i)} / \mu^{(i)}}$$

$P_{\lambda/\mu}(x; q, t)$  column strict tableaux  
with  $\lambda^{(0)} = \mu, \lambda^{(N)} = \lambda$

$$Q_{\lambda/\mu}(x) = \sum_T \psi_T x^T, \quad \psi_T = \prod_{i=1}^N \psi_{\lambda^{(i)} / \mu^{(i)}} (\dots)$$

Involution

$$\omega P_{\lambda/\mu}(x; q, t) = Q_{\lambda/\mu}(x; \frac{t}{q}, \frac{q}{t})$$

$$\omega Q_{\lambda/\mu}(x; q, t) = P_{\lambda/\mu}(x; t, q)$$

$q = t \Rightarrow$  Schur



When  $N=1$ ,

$$P_{\lambda/\mu}(x_1) = \begin{cases} \psi_{\lambda/\mu} x_1^{|\lambda/\mu|} & \lambda/\mu : \text{horizontal strip} \\ 0 & \text{otherwise} \end{cases}$$

$$Q_{\lambda/\mu}(x_1) = \begin{cases} \psi_{\lambda/\mu} x_1^{|\lambda/\mu|} & \lambda/\mu = \text{c.s.} \\ 0 & \text{otherwise} \end{cases}$$

### Some relations for Macdonald

Define  $\Pi(x, y) := \frac{\prod_{i=1}^N \prod_{j=1}^M (tx_i y_j; \delta)_{\infty}}{\prod_{i=1}^N \prod_{j=1}^M (x_i y_j; \delta)_{\infty}} \quad \delta = (y_1, \dots, y_M)$

$$\omega_y \Pi(x, y) = \prod_{i=1}^N \prod_{j=1}^M (1 + x_i y_j)$$

Then,

$$\sum_{\mu} P_{\mu/\lambda}(x) Q_{\mu/\nu}(y) = \sum_{\tau} P_{\nu/\tau}(x) Q_{\lambda/\tau}(y)$$

$$\sum_{\lambda} P_{\lambda}(x) P_{\mu/\lambda}(y) = P_{\mu}(x, y).$$

## 6.2. Macdonald process

measure on  $GT_N$

$$M_{a,b}(\lambda) = \frac{1}{Z} P_{\lambda^{(1)}}(a_1) P_{\lambda^{(2)}/\lambda^{(1)}}(a_2) \dots P_{\lambda^{(N)}/\lambda^{(N-1)}}(a_N) \begin{cases} Q_{\lambda'}(b) \\ Q_{\lambda^{(N)}}(b) \end{cases}$$

$$\begin{cases} a = (a_1, a_2, \dots, a_N) \\ b = (b_1, b_2, \dots, b_M) \end{cases}$$

$$Z = \Pi(a; b)$$

$$\left( \text{'a.e.' } Z = \omega_b \Pi(a, b) \right)$$

Marginal for  $\lambda = \lambda^{(N)}$  is the Macdonald measure.

$$\frac{1}{Z} P_{\lambda}(a) Q_{\lambda}(b)$$

Dynamics  $G(\mu, \lambda)$  ( $\Pi_{SS} \rightarrow \Pi_{Pa}$ )  
Soln.

t=0  $q$ -Whittaker meas.

$$\begin{cases} a = (a_1, \dots, a_N) & a_i = 1 \\ b = (b_1, \dots, b_M) & b_i = r. \end{cases}$$

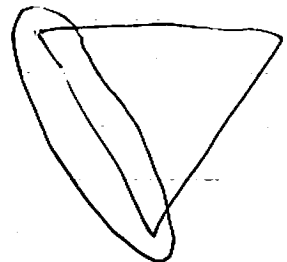
$$Q_{\lambda/\mu'}(x; q, t=0)$$

$$= \sum_T \varphi_T' x^T$$

$$\varphi_{\lambda/\mu}' = \prod_{i \geq 1} (1 - q^{\mu_i - \mu_{i+1}})$$

$$\lambda_i = \mu_i$$

$$\lambda_{i+1} = \mu_{i+1} + 1$$



$q$ -TASEP

B.C. discrete time  $q$ -TASEP

### 6.3. Expectations

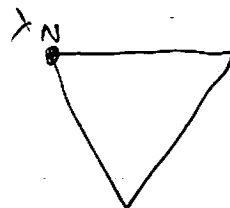
Macdonald measure

$$z^{-1} P_\lambda(x) Q_\lambda(y)$$

$$\left( \begin{aligned} z &= \sum_{\lambda} P_\lambda(x) Q_\lambda(y) \\ &= \pi(x, y) \end{aligned} \right)$$

Want to compute

$$\left\langle \frac{q^{|\lambda|} P_\lambda}{F(\lambda)} \right\rangle = \frac{1}{\pi(x, y)} \sum_{\lambda} F(\lambda) P_\lambda(x) Q_\lambda(y)$$



If  $D P_\lambda = d_\lambda P_\lambda$ ,  
 $\uparrow$   
 $q^\lambda$

$$\langle d_\lambda \rangle = \frac{1}{\pi(x, y)} \sum_{\lambda} \frac{d_\lambda P_\lambda(x) Q_\lambda(y)}{D P_\lambda(x)}$$

$$= \frac{D \pi(x, y)}{\pi(x, y)}$$

$$\langle d_\lambda^n \rangle = \frac{D^n \pi(x, y)}{\pi(x, y)}$$

Macdonald's difference operator

$$(D_N^r f)(x_1, x_2, \dots, x_N)$$

$$:= \sum_{\substack{I \subset \{1, 2, \dots, N\} \\ |I| = r}} t^{\frac{r(r-1)}{2}} \prod_{\substack{I \in I \\ j \notin I}} \frac{t x_i - x_j}{x_i - x_j} f(x_1, \dots, x_N)$$

Prop.

$$(D_N^r P_\lambda)(x_1, \dots, x_N) = e_r(\delta^{\lambda_1} t^{N-1}, \delta^{\lambda_2} t^{N-2}, \dots, \delta^{\lambda_N}) P_\lambda(x_1, \dots, x_N)$$

$$\text{where } e_r(x_1, \dots, x_N) = \sum_{1 \leq i_1 < \dots < i_r \leq N} x_{i_1} \dots x_{i_r}$$

Then,

$$\langle e_r(\delta^{\lambda_1} t^{N-1}, \dots, \delta^{\lambda_N}) \rangle = \frac{D_N^r \Pi(x; y)}{\Pi(x; y)}$$

$$= \frac{1}{(2\pi i)^r r!} \int \dots \int \det \left[ \frac{1}{t z_k - z_l} \right]_{k, l=1}^r \prod_{j=1}^r \left( \prod_{m=1}^r \frac{t z_j - x_m}{z_j - x_m} \right) \frac{f(\delta z_j)}{f(z_j)} dz_j$$

As  $t \rightarrow 0$

$$\lim_{t \rightarrow 0} t^{-\frac{r(r-1)}{2}} e_r(\delta^{\lambda_1} t^{N-1}, \dots, \delta^{\lambda_N}) = \delta^{\lambda_1 + \dots + \lambda_{N-r+1}} (= \delta^{\lambda_N})$$

For continuous time  $\delta$ -FASEP  $f(z) = \left(\frac{1}{1-z}\right)^N e^{-tz}$   $v=1$

$$\Rightarrow \langle \delta^{n\lambda_N} \rangle = \frac{(-1)^n \delta^{\frac{n(n-1)}{2}}}{(2\pi i)^n} \int \dots \int \prod_{1 \leq A < B \leq n} \frac{z_A - z_B}{z_A - \beta z_B} \prod_{j=1}^n \frac{f(z)}{f(\beta z)} \frac{dz_j}{z_j}$$

$$\Rightarrow \left\langle \frac{1}{(\delta^{\lambda_N}; \delta)_\infty} \right\rangle = \det(I+K)$$