

3. Schur measures

Previously we considered random infinite tableaux  $\phi \uparrow \lambda_1 \uparrow \lambda_2 \uparrow \dots$  with the property of central measures:

- conditionally to  $\lambda_n \uparrow \lambda_{n+1} \uparrow \dots$ , the law of  $T = \phi \uparrow \lambda_1 \uparrow \dots \uparrow \lambda_n$  is uniform on  $ST(\lambda_n)$ .

$\Leftrightarrow$  the measures  $\pi_n$  of the random partitions  $\lambda_n$  satisfy

$$\frac{\pi_n(\lambda)}{|ST(\lambda)|} = \sum_{\Lambda: \lambda \uparrow \Lambda} \frac{\pi_{n+1}(\Lambda)}{|ST(\Lambda)|}$$

We claim that the extremal central measures are related to Schur functions.

## 1. Further properties of Schur functions.

① Pieri rule:  $p_1 \cdot s_\lambda = \sum_{\Lambda: \lambda \uparrow \Lambda} s_\Lambda$

proof: With a sufficient number  $N$  of variables: ( $N \geq |\lambda| + 1$ ).

$$p_1(x_1, \dots, x_N) a_{\lambda + e_N}(x_1, \dots, x_N) = \sum_{j=1}^N \sum_{\sigma \in S(N)} x_j \varepsilon(\sigma) \frac{N}{\prod_{i=1}^N x^{\lambda_i + N - \sigma(i)}}$$

$$= \sum_{j=1}^N a_{(\lambda + e_j) + e_N}(x_1, \dots, x_N)$$

0 if adding a box on the  $j$ -th row of  $\lambda$  does not produce an integer partition.

$$= \sum_{\Lambda: \lambda \uparrow \Lambda} a_{\Lambda + e_N}(x_1, \dots, x_N) \quad \square$$

Therefore, if  $\Psi: \text{Sym} \rightarrow \mathbb{R}$  is a morphism of algebras with

1.  $\Psi(p_1) = 1$
  2.  $\Psi(s_\lambda) \geq 0 \quad \forall \lambda \in \mathcal{Y}$
- then  $\pi_n(\lambda) = |ST(\lambda)| \Psi(s_\lambda)$  is a central measure.

## ② Hall scalar product on Sym.

The Hall scalar product is the unique  $\langle \cdot, \cdot \rangle$  on Sym which makes  $(s_\lambda)_{\lambda \in \mathcal{Y}}$  an orthonormal basis.

lemma: Consider two graded basis  $(u_\lambda)_{\lambda \in \mathcal{Y}}$  and  $(v_\lambda)_{\lambda \in \mathcal{Y}}$  of Sym. We have  $\langle u_\lambda | v_\rho \rangle = \delta_{\lambda\rho}$  iff  $\sum_{\lambda \in \mathcal{Y}} u_\lambda(x) v_\lambda(y) = \prod_{i,j} \frac{1}{1-x_i y_j}$ .

Indeed, the formal power series rewrites in terms of Schur functions:

$$u_\lambda(x) = \sum_{|\mu| = |\lambda|} U_{\lambda\mu} s_\mu(x) \quad \text{LHS}$$

$$v_\rho(y) = \sum_{|\ell| = |\rho|} V_{\rho\ell} s_\ell(y) \quad ; \quad = \sum_{\lambda, \mu} U_{\lambda\mu} V_{\lambda\rho} s_\mu(x) s_\rho(y)$$

$$\text{and } \langle u_\lambda | v_\rho \rangle = \sum_{\mu} U_{\lambda\mu} V_{\rho\mu} = (U^t V)_{\lambda\rho} = \prod_{i,j} \frac{1}{1-x_i y_j} \quad \text{iff } (U^t V)_{\mu\nu} = \mathbb{1}_{\mu\nu}$$

$\Leftrightarrow U = V^{-1}$  is an orthogonal matrix.

example 1:  $\sum_{\lambda} h_\lambda(x) m_\lambda(y) = \sum_{\mathbb{I}} h_{\mathbb{I}}(x) y^{\mathbb{I}} = \prod_j \left( \sum_k h_k(x) (y_j)^k \right) = \prod_{i,j} \frac{1}{1-x_i y_j}$ .

example 2: set  $z_\lambda = \prod_{s \geq 1} s^{m_s(\lambda)} m_s(\lambda)!$

$$\text{Then, } \sum_{\lambda} \frac{p_\lambda(x) p_\lambda(y)}{z_\lambda} = \sum_{\lambda} \frac{p_\lambda(xy)}{z_\lambda} = \exp\left(\sum_{k=1}^{\infty} \frac{p_k(xy)}{k}\right) = \exp\left(\sum_{i,j} \log\left(\frac{1}{1-x_i y_j}\right)\right) = \prod_{i,j} \frac{1}{1-x_i y_j}$$

③ expansion of Schur functions on semistandard tableaux:

Denote  $x^T$  the product of variables  $x_i$ ,  $i$  appearing in a cell of a semistandard tableau  $T$ .

$$\text{Then: } s_\lambda(x_1, x_2, \dots, x_N) = \sum_{T \in \text{SST}(\lambda, N)} x^T.$$

proof: The Hall scalar product allows one to define skew Schur functions  $s_{\lambda/\mu}$  with  $|\lambda| \geq |\mu|$ :

$$\forall e, \langle s_{\lambda/\mu} | s_e \rangle = \langle s_\lambda | s_\mu s_e \rangle; \deg s_{\lambda/\mu} = |\lambda| - |\mu| \text{ if } s_{\lambda/\mu} \neq 0.$$

We have a generalisation of the Jacobi-Trudy formula:

$$s_{\lambda/\mu} = \det(h_{\lambda_i - i - \mu_j + j})$$

(this implies that if  $\mu \not\subseteq \lambda$ , then  $s_{\lambda/\mu} = 0$ ).

$$\begin{aligned} \text{We have } \sum_{\lambda} s_{\lambda/\mu}(x) s_\lambda(y) &= \sum_{\lambda, e} \langle s_{\lambda/\mu} | s_e \rangle s_e(x) s_\lambda(y) \\ &= \sum_{\lambda, e} \langle s_\lambda | s_\mu s_e \rangle s_e(x) s_\lambda(y) \\ &= \sum_e s_e(x) s_\mu(y) s_e(y) = C(x, y) s_\mu(y). \\ &= \sum_{\nu} h_\nu(x) m_\nu(y) s_\mu(y) \end{aligned}$$

$$\Rightarrow \sum_{\lambda} s_{\lambda/\mu}(x) a_{\lambda + e_N}(y_1, \dots, y_N) = \sum_{\Gamma} h_{\Gamma}(x) y^{\Gamma} a_{\mu + e_N}(y_1, \dots, y_N)$$

$$\begin{aligned} \Rightarrow s_{\lambda/\mu}(x) &= [y^{\lambda + e_N}] \left( \sum_{\substack{\Gamma \\ \sigma \in \text{SST}(N)}} e_{\sigma} y^{\Gamma + \sigma(\mu + e_N)} h_{\Gamma}(x) \right) \\ &= \det(h_{(\lambda_i - i) - (\mu_j - j)}). \end{aligned}$$

$$= (\text{with a bit more work}) \det(e_{(\lambda'_i - i) - (\mu'_j - j)}).$$

$H(t)E(-t) \rightsquigarrow$

In particular, if  $X = \{x_i\}$ , then  $s_{\lambda \cup \mu}(X) = \mathbb{1}_{\lambda \cup \mu}$  is an horizontal strip  $(X_i)^{\lambda \cup \mu}$

This implies the expansion, because  $s_\lambda(X+Y) = \sum_{\mu} s_{\lambda \cup \mu}(X) s_\mu(Y)$ .

proof:  $s_\lambda(X+Y) = [s_\lambda(Z)] \sum_{\lambda} s_\lambda(X+Y) s_\lambda(Z)$

$$= [s_\lambda(Z)] (C(X+Y, Z))$$

$$= [s_\lambda(Z)] (C(X, Z) C(Y, Z))$$

$$= [s_\lambda(Z)] \left( \sum_{\nu, \mu} s_\nu(X) s_\mu(Y) s_\nu(Z) s_\mu(Z) \right)$$

$$= \sum_{\mu} s_{\lambda \cup \mu}(X) s_\mu(Y). \quad \square$$

2. The Thoma simplex indices extremal central measures.

$\Omega = \{(\alpha, \beta) \text{ pair of non increasing nonnegative sequences with } \sum_{i=1}^{\infty} \alpha_i + \beta_i = 1, 0 \leq \alpha_i, \beta_i\}$

We set  $p_\lambda(\alpha, \beta) = 1$   
 $p_{k \geq 2}(\alpha, \beta) = \sum_{i=1}^{\infty} (\alpha_i)^k + (-1)^{k-1} (\beta_i)$

This defines a morphism of algebras  $\text{Sym} = \mathbb{R}[p_1, p_2, \dots] \rightarrow \mathbb{R}$  with  $\Psi_{\alpha, \beta}(p_i) = 1$ .

Theorem:  $\forall (\alpha, \beta) \in \Omega, \forall \lambda \in Y, \Psi_{\alpha, \beta}(s_\lambda) = s_\lambda(\alpha, \beta) \geq 0$ .

At this point, it is very hard to give a complete proof of this fact; some arguments rely on the connection between Schur functions and the representations of  $S(n)$ .

Sketch of proof. 1) the Schur positivity is kept by addition of morphisms on the power sums  $p_{k \geq 2}$ .

This is related to the fact that a product of two Schur functions is a non-negative linear combination of Schur functions.

e) if  $X = \alpha = (\alpha_1, \alpha_2, \dots)$  is a summable sequence with positive coefficients, then  $pk_{\geq 1}(\alpha) = \sum_{i=1}^{\infty} (\alpha_i)^k$  is Schur positive, because we then have  $s_{\lambda}(\alpha) = \sum_{\tau \in \text{SST}(\lambda)} \alpha^{\tau} \geq 0$ .

3) if  $X = \beta$  is a summable sequence with positive coefficients  
 $pk_{\geq 1}(\beta) = (-1)^{k-1} \sum_{i=1}^{\infty} (\beta_i)^k$   
 then  $s_{\lambda}(\beta) \geq 0 \quad \forall \lambda \in \mathcal{Y}$ .

Indeed, by stability by sum, it suffices to treat the case of  $\beta = (\beta_1) = (\beta_1, 0, \dots)$ .

$$\begin{aligned} s_{\lambda}(\beta) &= [s_{\lambda}(z)] \left( \sum_{\lambda} s_{\lambda}(\beta) s_{\lambda}(z) \right) \\ &= [s_{\lambda}(z)] \left( \exp \left( \sum_{k=1}^{\infty} (-1)^{k-1} \beta^k \frac{pk(z)}{k} \right) \right) \\ &= [s_{\lambda}(z)] \left( \prod_{i=1}^{\infty} (1 + \beta z^i) \right) \\ &= [s_{\lambda}(z)] \left( \sum_{k=0}^{\infty} \beta^k e_k(z) \right) \end{aligned}$$

However:  $h_k(z) = s_{(k)}(z)$  (clear by Jacobi-Trudy)

$$e_k(z) = s_{\lambda^k}(z)$$

$$\sum_{\substack{1 \leq i_1 < \dots < i_k \leq k \\ \det(\epsilon_i^{k-1-j})_{1 \leq i, j \leq k} = z_1 z_2 \dots z_k \text{ if } z = (z_1, \dots, z_k)}} (\det(z_i^{k-j}))_{1 \leq i, j \leq k} = e_k(z)$$

$$\text{So } s_{\lambda}(\beta) = \mathbb{1}_{(\lambda=1^k)} \beta^k \geq 0.$$

a)  $(\alpha, \beta) \in \mathcal{Q}$  corresponds to the specialisation  $\alpha + \beta + \gamma E$   
 with  $pk_1(E) = 1$   
 $pk_{\geq 2}(E) = 0$ .

Schur positive:  $s_{\lambda}(E) = \frac{|\text{ST}(\lambda)|}{n!}$  by using the Pieri rule.

□

Kerov-Vershik ring theorem: let  $\Psi: \text{Sym} \rightarrow \mathbb{R}$  be a linear function such that

- $\Psi(p_\lambda s_\lambda) = \sum_{\Delta: \lambda \triangleright \Delta} s_\Delta$  Pieri rule
- $\Psi(s_\lambda) \geq 0 \quad \forall \lambda$ .
- $\Psi(p_1) = 1$ .

We associate to  $\Psi$  the central measure  $\pi_\Psi(\lambda) = |\text{ST}(\lambda)| \Psi(s_\lambda)$ .

Then,  $(\pi_\Psi)_{n \in \mathbb{N}}$  is an extremal central measure iff  $\Psi$  is a morphism of algebras:  
 $\Psi(fg) = \Psi(f)\Psi(g) \quad \forall f, g \in \text{Sym}$ .

proof:  $\Rightarrow$  Suppose that  $\Psi$  corresponds to an extremal central measure. We can view  $\Psi$  as a linear map on the quotient ring  $A = \text{Sym}/(p_1 - 1)$ , positive on the cone  $K$  spanned by the projections of the Schur functions.

Let  $a, b \in A$ . We want to prove that  $\Psi(ab) = \Psi(a)\Psi(b)$ . The ring  $A$  is spanned by  $K$ , so we can assume that  $a, b \in K$ . Also, since

$$(p_1)^n = \sum_{\lambda \in \mathcal{Y}(n)} |\text{ST}(\lambda)| s_\lambda$$

$$1_A = \sum_{\lambda \in \mathcal{Y}(n)} |\text{ST}(\lambda)| \overline{s_\lambda}, \quad \text{every element of } K \text{ is smaller than a multiple of } 1_A.$$

So we can assume that  $1_A - b \in K$ . Let us then write:

$$\Psi(a) = \Psi(ab) + \Psi(a(1_A - b)).$$

1st case:  $\Psi(b) \neq 0, \Psi(1_A - b) \neq 0$ . Then,

$$\Psi(a) = \Psi_b(a) + \Psi_{1_A - b}(a) \quad \text{with} \quad \Psi_c(a) = \frac{\Psi(ac)}{\Psi(c)}.$$

two maps on  $A$  corresponding to central measures.

By extremality,  $\Psi = \Psi_b$ , so  $\Psi(ab) = \Psi(a)\Psi(b)$ .

2nd case:  $\Psi(b) = 0$  or  $\Psi(1_A - b) = 0$ . Let us treat the first case.

We then have  $a \leq \lambda 1_A$ , so  $ab \leq \lambda b$  and  $\Psi(ab) \leq \lambda \Psi(b) = 0$ .

$\Leftarrow$  Suppose that  $\Psi$  is multiplicative. We decompose  $\Psi$  as an integral over the set of extremal points: Choquet theorem

$$\Psi = \int_{\text{Ext}} \phi \, \mathbb{P}[d\phi]. \quad (\text{This decomposition is not unique in general!})$$

For  $\lambda \in \mathcal{Q}$ , set  $X_\lambda = \phi(s_\lambda)$  under  $\mathbb{P}$ . We have  $\mathbb{E}[X_\lambda] = \Psi(s_\lambda)$ .

We also have:

$$\mathbb{E}[(X_\lambda)^2] = \mathbb{E}[\phi(s_\lambda)\phi(s_\lambda)]$$

$$\stackrel{\substack{\text{(extremal points yield} \\ \text{multiplicative functions)}}}{=} \mathbb{E}[\phi(s_\lambda^2)] = \Psi(s_\lambda^2) = \Psi(s_\lambda)^2 = \mathbb{E}[X_\lambda]^2.$$

So,  $\text{var } X_\lambda = 0$ , and  $X_\lambda$  is constant.

$\Rightarrow \mathbb{P}$  is concentrated on one point, and  $\Psi$  is extremal  $\square$ .

missing: the interpretation of the parameters  $\alpha$  and  $\beta$  w.r.t. the sequence of random partitions  $(\mathcal{A}_n)_{n \in \mathbb{N}}$ .

### 3. Schur measures and their determinantal expression.

The Cauchy formula leads one to investigate a new family of measures on partitions.

Suppose given  $X, Y$  morphisms of algebras  $\text{Sym} \rightarrow \mathbb{R}$  with  $s_\lambda(X) \geq 0$   
 $s_\lambda(Y) \geq 0$   
 $\forall \lambda \in \mathcal{Q}$ .

If the series  $\sum_{k=1}^{\infty} \frac{p_k(X)p_k(Y)}{k}$  is convergent, then

$$\sum_{\lambda \in \mathcal{Q}} s_\lambda(X)s_\lambda(Y) = \exp\left(\sum_{k=1}^{\infty} \frac{p_k(X)p_k(Y)}{k}\right) = c(X, Y).$$

Definition: The Schur measure with parameters  $X, Y \in \mathbb{R}_+ \Omega$  is the



probability measure on  $\mathcal{Y} = \bigsqcup_{n \in \mathbb{N}} \mathcal{Y}(n)$  given by

$$\mathbb{P}_{X, Y}(\omega) = \frac{s_X(\lambda) s_Y(\nu)}{C(X, Y)}$$

example: Fix  $\theta > 0$ , and consider the control measure associated to a parameter  $w = (\alpha, \beta) \in \Omega$ .

$$\pi_n(\lambda) = |\text{ST}(\lambda)| s_X(\alpha, \beta).$$

Let  $\mathbb{P}_{\theta, w}$  be the Poissonised version of the measures  $\pi_n$ :

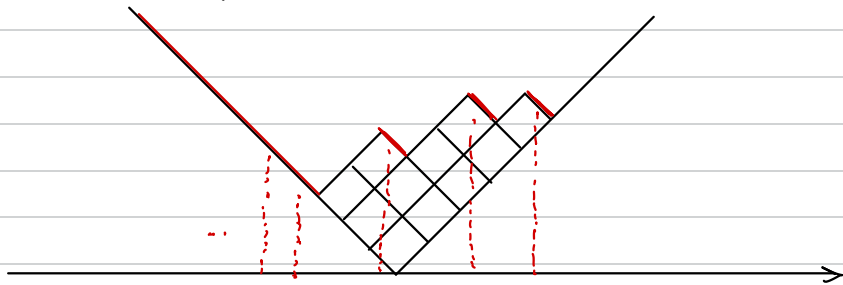
$$\begin{aligned} \mathbb{P}_{\theta, w}[\lambda] &= \text{Poisson}(\theta)[n] \pi_n[\lambda] \\ &= e^{-\theta} \frac{\theta^n}{n!} |\text{ST}(\lambda)| s_X(w). \end{aligned}$$

This is the Schur measure associated to the parameters  $X = \theta E$  and  $Y = w$ .

example: In particular, the Poissonised Plancherel measure

$$\text{PPE}_{\theta}[\lambda] = e^{-\theta} \frac{\theta^n}{n!} \frac{|\text{ST}(\lambda)|^2}{n!} \text{ is } \mathbb{P}_{\theta E, E} = \mathbb{P}_{\sqrt{\theta} E, \sqrt{\theta} E}.$$

We associate to a partition  $\lambda \in \mathcal{Y}$  its descent set  $D(\lambda) = (\lambda_i - i + \frac{1}{2})_{i \geq 1}$



This is a subset of  $\mathbb{Z}' = \mathbb{Z} + \frac{1}{2}$  with the property

$$|\mathbb{Z}'_+ \cap D(\lambda)| = |\mathbb{Z}'_- \cap D(\lambda)| < +\infty.$$

Theorem (Okounkov, 2000). Consider a random partition  $\lambda \sim \mathbb{P}_{X, Y}$ . The random set of descents  $D(\lambda)$  forms a determinantal point process on  $\mathbb{Z}'$ , meaning that:

$$\forall \{x_1, \dots, x_n\} \subset \mathbb{Z}'$$

$$\mathbb{P}[\{x_1, \dots, x_n\} \subset D_\lambda] = \det(K_{X,Y}(x_i, x_j))_{1 \leq i, j \leq n}.$$

The kernel  $K_{X,Y}$  has an explicit generating function:

$$\mathcal{H}(z, w) = \sum_{x, y \in \mathbb{Z}'} K_{X,Y}(x, y) z^x w^{-y} = \frac{\sqrt{zw}}{z-w} \exp\left(\sum_{k=-\infty}^{\infty} \frac{\rho_k(X)}{k} (z^k - w^k) - \sum_{k=-\infty}^{\infty} \frac{\rho_k(Y)}{k} (z^{-k} - w^{-k})\right)$$

example: For the Poissonised Plancherel measure  $\text{PPl}_0$ ,  $X = Y = \sqrt{0} E$

$$\mathcal{H}(z, w) = \frac{\sqrt{zw}}{z-w} \exp(\sqrt{0}((z-w) - (z^{-1} - w^{-1}))).$$