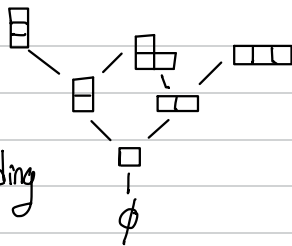


2. Central measures, the Thoma simplex
and symmetric functions

We want to study the Plancherel measure $P_n(\lambda) = \frac{|ST(\lambda)|^2}{n!}$

1. Central measures on the Young graph

Consider the Young graph \mathcal{Y} :



$\lambda \uparrow \Delta$ if Δ can be obtained from λ by adding one cell at the top-right border.

A standard tableau with shape λ corresponds to a sequence $\phi \uparrow \lambda_1 \uparrow \dots \uparrow \lambda_n = \lambda$.
recurrence relation:

$$|ST(\Delta)| = \sum_{\lambda: \lambda \uparrow \Delta} |ST(\lambda)|$$

We can therefore define a transition kernel $p_n^{n+1}: \mathcal{Y}(n+1) \times \mathcal{Y}(n) \rightarrow \mathbb{R}$
 $\Delta, \lambda \mapsto \mathbb{1}_{\lambda \uparrow \Delta} \frac{|ST(\lambda)|}{|ST(\Delta)|}$

Given $\lambda \in \mathcal{Y}(n)$, the transition kernels $p_{n-1}^n, p_{n-2}^{n-1}, \dots, p_0^1$ allow one to choose a uniform standard tableau in $ST(\lambda)$.

Definition: A central measure is a family of probability measures $(\pi_n)_{n \in \mathbb{N}}$ on the levels $\mathcal{Y}(n)$ of the Young graph such that:

$$\forall n, \pi_{n+1} p_n^{n+1} = \pi_n.$$

Proposition: the Plancherel measures form a central measure.

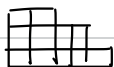
Proof: We have to establish:

$$P_n(\lambda) = \sum_{\Delta: \lambda \uparrow \Delta} P_{n+1}(\Delta) \frac{|ST(\lambda)|}{|ST(\Delta)|}$$

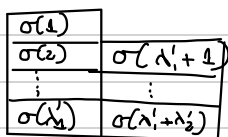
$$\Leftrightarrow \frac{|ST(\lambda)|}{n!} = \sum_{\Delta: \lambda \uparrow \Delta} \frac{|ST(\Delta)|}{(n+1)!} \Leftrightarrow (n+1) \frac{|ST(\lambda)|}{n!} = \sum_{\Delta: \lambda \uparrow \Delta} |ST(\Delta)|.$$

Given an integer partition λ , we denote λ' the conjugate partition given by the sizes of the columns of λ .

ex: $(5, 4, 2)'$ = $(3, 3, 2, 2, 1)$.



We claim that both sides of the equation are equal to the number of permutations $\sigma \in S_{n+1}$ s.t.



... is a semistandard tableau with shape λ



* LHS

indeed, if σ is such a permutation, then:

- $\sigma(n+1)$ can be any integer in $\llbracket 1, n+1 \rrbracket$. ($n+1$ possibilities)
- $P(\sigma_1 \dots \sigma_n)$ is the semi-standard tableau indicated above; knowing $\sigma(n+1)$, there are $|ST(\lambda)|$ possibilities

- $Q(\sigma_1 \dots \sigma_n) =$

λ'_1	$\lambda'_1 + \lambda'_2$
\vdots	\vdots
2	$\lambda'_1 + 1$
1	

 ... (1 possibility).

* RHS

- $Q(\sigma)$ is one of the standard tableaux obtained by adding $n+1$ at the top right boundary of $Q \rightarrow$ one possibility for each Δ with $\lambda \uparrow \Delta$.
- $P(\sigma)$ is arbitrary in $ST(\Delta)$, knowing Δ .

□.

Given a control measure $(\pi_n)_{n \in \mathbb{N}}$, one can use the Kolmogorov extension theorem in order to define a random infinite tableau $T = \phi \uparrow \lambda_1 \uparrow \dots \uparrow \lambda_n \uparrow \dots$ such that:

- 1) $\forall n, \text{law}(\lambda_n) = \pi_n$.
- 2) conditionally to $\lambda_n \uparrow \lambda_{n+1} \uparrow \dots$, λ_{n-1} has distribution $p_{n-1}^n(\lambda_{n-1}, \dots)$

Then: $\forall n$, the subtableau of T formed by the integers $\in \llbracket 1, n \rrbracket$ has uniform law in $ST(\lambda_n)$.

problem: find all the control measures on infinite standard tableaux

\Leftrightarrow all the families (π_n) of probability measures which satisfy the harmonicity property:

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

The set of such functions $\mathcal{Y} \rightarrow \mathbb{R}_+$ is a compact convex set
 \rightarrow it suffices to describe the extremal control measures. (product topology on $\mathbb{R}_+^{\mathcal{Y}}$).

example: $\pi_n = \delta_{(n)}$

$\pi_n = \delta_{(1^n)}$

$\pi_n = \text{Pl}_n$

(we shall see that all these examples are extremal).

Theorem (Thoma, 1964; Kerov-Vershik 1981)

There exists - an algebra of functions $\text{Sym} = \bigoplus_{\lambda \in \mathcal{Y}} \mathbb{R} s_\lambda$

- a compact set Ω

- a way to associate to any $w \in \Omega$ a morphism of algebras $\text{Sym} \rightarrow \mathbb{R}$

$$f \mapsto \psi_w(f)$$

such that any extremal control measure writes uniquely as

$$\pi_n(\lambda) = |\text{ST}(\lambda)| \psi_w(s_\lambda) \text{ for some } w \in \Omega.$$

2. Symmetric functions.

We fix an infinite alphabet of variables $X = \{x_1, x_2, \dots, x_n, \dots\}$.

A **monomial** is a finite product of variables.

ex: $x_1^3 x_2 x_4^2$.

notation: if $I \in \mathbb{N}^{(\mathbb{N}^*)}$ is a sequence of integers eventually equal to 0, we set $x^I = x_1^{i_1} x_2^{i_2} \dots x_s^{i_s}$.

The degree of a monomial is its number of terms. $\deg(x^{\mathbb{I}}) = |\mathbb{I}| = \sum_{s \geq 1} i_s$.

A polynomial is a formal linear combination $\sum c_{\mathbb{I}} x^{\mathbb{I}}$ of monomials which is bounded in degree: $c_{\mathbb{I}} = 0$ for all monomials with $\deg(x^{\mathbb{I}}) > \text{some } d \in \mathbb{N}$.

ex: $2x_1^2 x_3 - x_2^5, x_1 + x_2 + \dots + x_n + \dots$ (not necessarily finite sums!).

We denote $\mathbb{R}[X]$ the real vector space of polynomials. It is graded by $\deg(\sum c_{\mathbb{I}} x^{\mathbb{I}}) = \max(|\mathbb{I}|, c_{\mathbb{I}} \neq 0)$ ($-\infty$ by convention for the 0 polynomial).

The rule $x^{\mathbb{I}} \cdot x^{\mathbb{J}} = x^{\mathbb{I} + \mathbb{J}}$ can be extended by linearity to make $\mathbb{R}[X]$ into a graded real algebra.

$$\deg PQ = \deg P + \deg Q.$$

$$\mathbb{R}[X] = \bigoplus_{n=0}^{\infty} \mathbb{R}_n[X] \rightarrow \text{homogeneous polynomial with degree } n.$$

remark: For any $N \geq 0$, we have a specialisation morphism

$$\phi_N: \mathbb{R}[X] \rightarrow \mathbb{R}[x_1, x_2, \dots, x_N].$$

$$P(x_1, x_2, \dots) \mapsto P(x_1, \dots, x_N, x_{N+1} = 0, x_{N+2} = 0, \dots)$$

$\mathbb{R}[X] = \varprojlim_{N \rightarrow +\infty} \mathbb{R}[x_1, x_2, \dots, x_N]$ with respect to the morphisms

$$\mathbb{R}[x_1, \dots, x_{N+1}] \xrightarrow{\phi_N^{N+1}} \mathbb{R}[x_1, \dots, x_N]. \quad (\text{in the category of graded real algebras})$$

Definition A polynomial $P \in \mathbb{R}[X]$ is a symmetric function if, for any permutation

$$\sigma \in \mathcal{S}(\infty) = \bigcup_{n=1}^{\infty} \mathcal{S}(n),$$

$$P(x_{\sigma(1)}, x_{\sigma(2)}, \dots) = P(x_1, x_2, \dots)$$

$$\Leftrightarrow \forall N, \phi_N(P) \in \mathbb{R}[x_1, \dots, x_N]^{\mathcal{S}(N)}.$$

$\text{Sym} = \{ \text{symmetric functions} \}$ is a graded subalgebra of $\mathbb{R}[X]$.

example 1 : $\forall k \geq 1, p_k(X) = \sum_{i=1}^{\infty} (x_i)^k$ k -th power sum.

For $\lambda \in \mathcal{Y} = \bigcup_{n \in \mathbb{N}} \mathcal{Y}(n)$, we set $p_\lambda(X) = p_{\lambda_1}(X) p_{\lambda_2}(X) \dots p_{\lambda_\ell}(X)$.

p_λ is a symmetric function with degree $|\lambda|$.

example 2 : $\forall k \geq 1, e_k(X) = \sum_{i_1 < i_2 < \dots < i_k} x_{i_1} x_{i_2} \dots x_{i_k}$ k -th elementary symmetric function

$e_\lambda = e_{\lambda_1} e_{\lambda_2} \dots e_{\lambda_\ell}$. $\deg e_\lambda = |\lambda|$

example 3 : $\forall k \geq 1, h_k(X) = \sum_{i_1 \leq i_2 \leq \dots \leq i_k} x_{i_1} x_{i_2} \dots x_{i_k}$ k -th homogeneous symmetric function.

$h_\lambda = h_{\lambda_1} h_{\lambda_2} \dots h_{\lambda_\ell}$. $\deg h_\lambda = |\lambda|$.

We have for instance $h_2(X) = e_2(X) + p_2(X)$
 $= e_{1^2}(X) - e_2(X)$.

Theorem : We have $\text{Sym} = \mathbb{R}[p_1, p_2, \dots]$
 $= \mathbb{R}[e_1, e_2, \dots]$
 $= \mathbb{R}[h_1, h_2, \dots]$.

Proof : ① First, let us show that the three families generate the same algebra.

$$P(t, X) = \sum_{k=1}^{\infty} \frac{p_k(X)}{k} t^k = \sum_{i=1}^{\infty} \sum_{k=1}^{\infty} \frac{x_i^k}{k} t^k = \sum_{i=1}^{\infty} -\log(1 - x_i t)$$

$$= \log \left(\prod_{i=1}^{\infty} \frac{1}{1 - x_i t} \right).$$

$$\text{However, } \prod_{i=1}^{\infty} \frac{1}{1 - x_i t} = \prod_{i=1}^{\infty} \sum_{k_i \geq 0} (x_i t)^{k_i} = \sum_{k=0}^{\infty} h_k(X) t^k \quad (\text{with } h_0 = 1)$$

So, $\log H = P \Rightarrow$ the two families generate the same algebra.

$$\text{We also have } \frac{1}{H(t, X)} = \prod_{i=1}^{\infty} (1 + x_i t) = \sum_{k=0}^{\infty} e_k(X) t^k \quad (\text{with } e_0 = 1)$$

So, $\frac{1}{H(t, X)} = E(t) \Rightarrow$ the two families generate the same algebra.

②. If $x^\Gamma = x_1^{m_1} \dots x_n^{m_n}$, denote $\lambda = t(x^\Gamma)$ the type of x^Γ , which is the integer partition obtained by reordering the integers $m_1 \dots m_n$.

$P \in \text{Sym} \iff$ the coefficient of x^Γ only depends of $t(x^\Gamma)$.

Therefore, a linear basis of Sym consists in the **monomial functions**

$$m_\lambda(x) = \sum_{t(x^\Gamma) = \lambda} x^\Gamma.$$

examples: $m_2(x) = \sum x_i^2$

$$m_{2,1}(x) = \sum_{i \neq j} x_i^2 x_j$$

$$m_{1,1}(x) = \sum_{i < j} x_i x_j.$$

Denote $\lambda < \mu$ if μ can be obtained from λ by adding certain parts, and reordering
 ex: $(5, 3, 3, 1) < (6, 5, 1)$.

Then: $\forall \lambda, p_\lambda(x) = \left(\prod_{s \geq 1} m_s(\lambda)! \right) m_\lambda(x) + \sum_{\mu > \lambda} * m_\mu(x)$

\rightarrow triangular relation, and $(p_\lambda)_{\lambda \in \Gamma_n}$ is a linear basis of Sym .

example: $p_{5,3,3,1} = \sum_{i,j,k,l} x_i^5 x_j^3 x_k^3 x_l$

$$= \sum_{i \neq j \neq k = l} \text{---} + \text{terms with equalities}$$

$$= 2 m_{5,3,3,1} + 2 m_{8,3,1} + 2 m_{6,3,3} + m_{6,5,1}$$

$$+ 2 m_{5,4,1} + m_{11,1} + 2 m_{9,3} + m_{7,5} + m_{12}$$

$$+ 2 m_{6,6} + 2 m_{8,4} \quad \square.$$

A last interesting basis of Sym consists in the **Schur functions**.

3. Schur functions

An **antisymmetric** polynomial in N variables is a polynomial $P \in \mathbb{R}[x_1, \dots, x_N]$ such that

$\forall \sigma \in S(N), P(x_{\sigma(1)}, \dots, x_{\sigma(N)}) = \epsilon(\sigma) P(x_1, \dots, x_N)$.

example: $\Delta(x_1, \dots, x_N) = \det(x_i^{N-j})_{1 \leq i, j \leq N} = \text{Vandermonde determinant}$
 $= \prod_{1 \leq i < j \leq N} (x_i - x_j)$.

lemma 1: if $P \in \text{Antisym}_N$ and $x^{\mathbb{F}}$ appears with a non zero coefficient in P , then $t(x^{\mathbb{F}})$ is an integer partition with its parts all distinct.

\rightarrow a linear basis of Antisym_N consists in the functions $a_{\lambda + e_N} = \sum_{\sigma \in S(N)} \epsilon(\sigma) x^{\sigma(\lambda + e_N)}$
 with $e_N = (N-1, N-2, \dots, 0)$ and $\lambda \in \mathbb{Y}_{\text{length} \leq N}$.

lemma 2: if $P \in \text{Antisym}_N$, then $P = \Delta \Phi$ with $\Phi \in \text{Sym}_N = \mathbb{R}[x_1, \dots, x_N]$.

the map $\psi_N: \text{Sym}_N \rightarrow \text{Antisym}_N$ is a $\frac{N(N-1)}{2}$ graded isomorphism.
 $f \mapsto f \Delta$

We set $s_{\lambda}(x_1, \dots, x_N) = \psi_N^{-1}(a_{\lambda + e_N}) = \frac{a_{\lambda + e_N}(x_1, \dots, x_N)}{\Delta(x_1, \dots, x_N)}$
 $= \frac{\det(x_i^{\lambda_j + N - j})}{\det(x_i^{N-j})}$.

lemma 3: We have $s_{\lambda}(x_1, \dots, x_N, x_{N+1} = 0) = s_{\lambda}(x_1, \dots, x_N)$ if $e(\lambda) \leq N$.

$$\frac{\det \left(\begin{array}{ccc|c} x_i^{\lambda_j + N - j} & & & x \\ \hline & & & \vdots \\ & & & * \\ \hline 0 & \dots & 0 & 1 \end{array} \right)}{\det \left(\begin{array}{ccc|c} x_i^{N-j} & & & * \\ \hline & & & \vdots \\ & & & * \\ \hline 0 & \dots & 0 & 1 \end{array} \right)} = \frac{x_1 \dots x_N}{x_1 \dots x_N} \frac{a_{\lambda + e_N}(x_1, \dots, x_N)}{a_{e_N}(x_1, \dots, x_N)}$$

\Rightarrow There exists symmetric functions $s_{\lambda} \in \mathbb{Y}$ with $\deg s_{\lambda} = |\lambda|$
 $\text{Sym} = \bigoplus_{\lambda \in \mathbb{Y}} \mathbb{R} s_{\lambda}$.

new statement of the Thoma / Kerov-Vershik theorem

$$\Omega = \{(\alpha, \beta) : \alpha, \beta \text{ nonincreasing non-negative sequence with } \sum_{i=1}^{\infty} \alpha_i + \beta_i \leq 1\}$$

Thoma simplex.

$$\text{Set } p_1(\alpha, \beta) = 1 \text{ and } p_{k \geq 2}(\alpha, \beta) = \sum_{i=1}^{\infty} \alpha_i^k + (-1)^k \beta_i^k.$$

As $\text{Sym} = \mathbb{R}[p_1, p_2, \dots]$, this defines a morphism of algebras $\text{Sym} \rightarrow \mathbb{R}$.

- $\forall \lambda \in \mathcal{Y}, s_\lambda(\alpha, \beta) \geq 0$.
- $\pi_n(\lambda) = |\text{ST}(\lambda)| s_\lambda(\alpha, \beta)$ is an extremal control measure
- the corresponding infinite tableau $T = \phi \uparrow \lambda_1 \uparrow \lambda_2 \dots \uparrow \lambda_n \uparrow \dots$ satisfies:

$$\forall k \geq 1, \frac{\lambda_{n,k}}{n} \xrightarrow{\text{p.s.}} \alpha_k \quad ; \quad \frac{\lambda'_{n,k}}{n} \xrightarrow{\text{p.s.}} \beta_k.$$
- All extremal control measures are of this kind.

4. Cauchy determinants and several properties of Schur functions.

Consider for $N \geq 1$ the Cauchy determinant:

$$C_N = \det \left(\frac{1}{1 - x_i y_j} \right)_{1 \leq i, j \leq N}$$

$$\begin{aligned} \text{We have } C_N &= \begin{vmatrix} \frac{1}{1-x_1 y_1} & \dots & \frac{1}{1-x_1 y_N} \\ \vdots & & \vdots \\ \frac{1}{1-x_N y_1} & \dots & \frac{1}{1-x_N y_N} \end{vmatrix} = \begin{vmatrix} \frac{(y_1 - y_N) x_1 \dots (y_{N-1} - y_N) x_1}{1 - x_1 y_1} & \dots & \frac{(y_{N-1} - y_N) x_1}{1 - x_1 y_{N-1}} & 1 \\ \vdots & & \vdots & \vdots \\ \frac{(y_1 - y_N) x_N \dots (y_{N-1} - y_N) x_N}{1 - x_N y_1} & \dots & \frac{(y_{N-1} - y_N) x_N}{1 - x_N y_{N-1}} & 1 \end{vmatrix} \\ &= \frac{1}{\prod_{i=1}^N \prod_{j=1}^N \frac{1}{1-x_i y_j}} \prod_{i=1}^{N-1} (y_i - y_{i+1}) \prod_{j=1}^{N-1} \frac{1}{1-x_N y_j} \begin{vmatrix} (x_1 - x_N) & \dots & (x_1 - x_N) & 0 \\ \frac{(x_1 - x_N)}{1 - x_1 y_1} & \dots & \frac{(x_1 - x_N)}{1 - x_1 y_{N-1}} & \vdots \\ \vdots & & \vdots & \vdots \\ \frac{(x_{N-1} - x_N)}{1 - x_{N-1} y_1} & \dots & \frac{(x_{N-1} - x_N)}{1 - x_{N-1} y_{N-1}} & 0 \end{vmatrix} \\ &\quad \quad \quad * \quad \quad \quad * \quad \quad \quad 1 \end{aligned}$$

$$= \frac{\prod_{i=1}^{N-1} (y_N - y_i)(x_N - x_i)}{\prod_{i=1}^N \prod_{j=1}^{N-1} (1 - x_i y_j)} C_{N-1}.$$

$$\Rightarrow C_N = \frac{\Delta(x_1, \dots, x_N) \Delta(y_1, \dots, y_N)}{\prod_{i,j} (1 - x_i y_j)}.$$

Corollaries: ① the **Cauchy formula**: given two commuting algebras of variables, we have

$$\sum_{\lambda \in \mathcal{Y}} s_\lambda(X) s_\lambda(Y) = \prod_{i,j \geq 1} \frac{1}{1 - x_i y_j}.$$

Indeed, it suffices to show the formula with finite algebras (x_1, \dots, x_N) (y_1, \dots, y_N) .

$$\begin{aligned} \Leftrightarrow C_N &= \sum_{\lambda \in \mathcal{Y}} a_{\lambda + e_N}(X) a_{\lambda + e_N}(Y) \\ &= \sum_{\lambda \in \mathcal{Y}} \left(\sum_{\sigma \in \mathcal{S}(N)} \varepsilon(\sigma) x^{\sigma(\lambda + e_N)} \right) \left(\sum_{\tau \in \mathcal{S}(N)} \varepsilon(\tau) y^{\tau(\lambda + e_N)} \right) \\ &= \sum_{\Gamma \in \mathbb{N}^N} \sum_{\rho \in \mathcal{S}(N)} \varepsilon(\rho) x^{-\Gamma} y^{\rho(\Gamma)} = \sum_{\rho \in \mathcal{S}(N)} \varepsilon(\rho) \prod_{i=1}^N \frac{1}{1 - x_i y_{\rho(i)}} \\ &= C_N. \end{aligned}$$

② the **Jacobi-Trudy formula**:

$$s_\lambda(x) = \det (h_{\lambda_i + j - i})_{1 \leq i, j \leq \ell(\lambda)} \quad \text{with } h_0 = 1 \\ h_{k < 0} = 0.$$

$$\begin{aligned} \text{Indeed, } s_\lambda(x_1, \dots, x_N) &= [y^{\lambda + e_N}] \left(\sum_{\lambda \in \mathcal{Y}} s_\lambda(x_1, \dots, x_N) a_{\lambda + e_N}(y_1, \dots, y_N) \right) \\ &= [y^{\lambda + e_N}] \left(\prod_{i,j} y_j - y_i \prod_{i,j} \frac{1}{1 - x_i y_j} \right) \end{aligned}$$

$$\begin{aligned}
&= [y^{\lambda + e_N}] \sum_{\sigma \in S(N)} \varepsilon(\sigma) y^{\sigma(e_N)} \sum_{I \in \mathbb{N}^N} y^I h_I(x) \\
&= \sum_{\sigma \in S(N)} \varepsilon(\sigma) \prod_{i=1}^N h_{\lambda_i - i + \sigma(i)} \\
&= \det (h_{\lambda_i + j - i}) .
\end{aligned}$$

