

# Polynomial functions on Young diagrams and limit shape of Young diagrams

Valentin Féray  
joint work with Pierre-Loïc Méliot

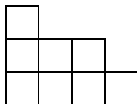
CNRS, Laboratoire Bordelais  
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Seminar über stochastische Prozesse  
Universität Zürich, ETH



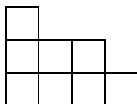
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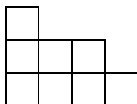
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- Young diagrams  $\lambda \vdash n$ :



- We consider a random Young diagram of size  $n$  (the measure is not uniform, it will be described later).



- Question: asymptotic behaviour of the lengths of its first rows, its first columns, ...
- Tool: representation theory of symmetric groups.

# Outline of the talk

- 1 Model of random Young diagrams
- 2 Polynomial functions on Young diagrams
- 3 How to prove asymptotic results?
- 4 A few remarks

# What is a representation?

$S_n$ : group of permutations of elements  $1, 2, \dots, n$ .

A representation of  $S_n$  is a couple  $(V, \rho)$  where:

- $V$  is a finite dimensional  $\mathbb{C}$  vector space;
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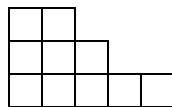
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## Theorem

*Irreducible representations of  $S_n$  are in bijection with partitions  $\lambda \vdash n$  or, equivalently, Young diagrams  $\lambda$  with  $n$  boxes.*

Ex:  $n = 10$ ,  
 $\lambda = (5, 3, 2)$ .





# Characters

Let  $(V, \rho)$  be a representation of  $S_n$ . We define:

$$\chi^\rho(\sigma) = \frac{\text{Tr}(\rho(\sigma))}{\dim(V)} \text{ for } \sigma \in S_n.$$

It is the (normalized) **character** of  $\rho$ .

No loss of information!

$$(\rho, V) \simeq_{S_n} (\rho', V') \Leftrightarrow \chi^\rho = \chi^{\rho'}.$$

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Indeed,

$$\begin{aligned} \chi^\rho(\tau^{-1}\sigma\tau) &= \frac{\text{Tr}(\rho(\tau^{-1}\sigma\tau))}{\dim(\lambda)} \\ &= \frac{\text{Tr}(\rho(\tau)^{-1} \cdot \rho(\sigma) \cdot \rho(\tau))}{\dim(\lambda)} = \frac{\text{Tr}(\rho(\sigma))}{\dim(\lambda)} \end{aligned}$$

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## Theorem

*The characters  $\chi^\lambda$  of irreducible representations  $\rho^\lambda$  of the symmetric group  $S_n$  form an (orthogonal) basis of the space of central functions on  $S_n$ .*

# Some central functions on $\bigcup S_n$

Let  $\omega = (\alpha_1, \alpha_2, \dots; \beta_1, \beta_2, \dots, \gamma)$  with

- $\alpha$  and  $\beta$  infinite non-increasing sequences;
- $\alpha_i, \beta_i, \gamma \geq 0$ ;
- $\sum_i \alpha_i + \sum_i \beta_i + \gamma = 1$ .

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Define

$$p_1(\omega) = 1$$

$$p_k(\omega) = \sum a_i^k + (-1)^{k-1} \sum b_i^k$$

$$p_{(\mu_1, \mu_2, \dots)}(\omega) = \prod_j p_{\mu_j}(\omega).$$

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We will consider the following central function on  $\bigcup S_n$ .

$$F_\omega(\sigma) = p_{t(\sigma)}(\omega),$$

where  $t(\sigma)$  is the cycle-type of  $\sigma$ .

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$$\omega = ((1 - q), (1 - q)q, (1 - q)q^2, \dots; 0, 0, \dots; 0)$$

$$\begin{aligned} F_{\omega}((1 \ 2 \ 3)(4 \ 5)) &= \left( \sum_{i \geq 0} ((1 - q)q^i)^3 \right) \left( \sum_{i \geq 0} ((1 - q)q^i)^2 \right) \\ &= \frac{(1 - q)^5}{(1 - q^3)(1 - q^2)} \\ &= F_{\omega}((1 \ 2 \ 3)(4 \ 5)(6)) \end{aligned}$$

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$$\omega = (0, 0, \dots; 0, 0, \dots; 1)$$

$$F_{\omega}(\sigma) = \begin{cases} 1 & \text{if } \sigma = \text{Id}; \\ 0 & \text{else.} \end{cases}$$

## Probability measure on $\mathcal{Y}_n$

Fix a parameter  $\omega$  an integer  $n$ .

$F_\omega/S_n$  is a central function on  $S_n$ . Therefore, as function on  $S_n$ ,

$$F_\omega = \sum_{\lambda \vdash n} c_\lambda \chi^\lambda.$$

Note that

$$\sum_{\lambda \vdash n} c_\lambda = \sum_{\lambda \vdash n} c_\lambda \chi^\lambda(\text{Id}) = F_\omega(\text{Id}) = 1.$$

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### Proposition

$$\forall \lambda, \quad c_\lambda \geq 0.$$

Hence, the  $c_\lambda$  define a probability measure  $\mathbb{P}_n^\omega$  on  $\mathcal{Y}_n$

$$\mathbb{P}_n^\omega(X = \lambda) = c_\lambda.$$

## Pursuing the examples

- $\omega = (0, 0, \dots; 0, 0, \dots; 1)$ .

A classical result in representation theory states that:

$$\mathbb{C}[S_n] \simeq_{S_n} \bigoplus_{\lambda \vdash n} V_\lambda^{\dim(V_\lambda)}$$

We look at the trace of the action of  $\sigma$ :

$$F_\omega = \sum_{\lambda \vdash n} \frac{\dim(V_\lambda)^2}{n!} \chi_\lambda.$$

Hence  $\mathbb{P}_n^\omega(X = \lambda) = \frac{\dim(V_\lambda)^2}{n!}$ . This measure is known as Plancherel measure.

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- $\omega = ((1 - q), (1 - q)q, (1 - q)q^2, \dots; 0, 0, \dots; 0)$ .

$\mathbb{P}_n^\omega$  is in this case a  $q$ -deformation of Plancherel measure already considered by Kerov in another context.



# Main theorem of the talk

## Theorem

Fix a parameter  $\omega$ . For each  $n$ , we pick a random Young diagram  $\lambda^{(n)}$  with the distribution  $\mathbb{P}_{n,\omega}$ . Then, one has the convergences in probability:

$$\forall i, \quad \frac{\lambda_i^{(n)}}{n} \rightarrow \alpha_i$$
$$\forall i, \quad \frac{(\lambda^{(n)})'_i}{n} \rightarrow \beta_i$$

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- proved by Kerov and Vershik in 1981.
- method of proof used here (which gives also information on the fluctuations): F., Méliot (2010) in the case of  $q$ -Plancherel measure.
- Generalization with the same argument to all  $\mathbb{P}_n^\omega$ : Méliot 2011.

# Examples

- $q = 1/2$ ,  $\omega = ((1 - q), (1 - q)q, (1 - q)q^2, \dots; 0, 0, \dots; 0)$ .

$$\frac{\lambda_i}{n} \rightarrow (1/2)^i, \quad \frac{\lambda'_i}{n} \rightarrow 0$$

Here is a random Young diagram of size 200 (computed and drawn by PL Méliot).



$$\lambda = (101, 51, 28, 8, 7, 3, 1, 1).$$

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$$\lambda = (101, 51, 28, 8, 7, 3, 1, 1).$$

- $\omega = (0, 0, \dots; 0, 0, \dots; 1)$  (Plancherel case)

$$\frac{\lambda_i}{n} \rightarrow 0, \quad \frac{\lambda'_i}{n} \rightarrow 0$$

$\Rightarrow$  no big rows and columns (but no precise information!).

# Normalized character values have simple expectations!

Fix  $\sigma \in \mathfrak{S}_n$ . Let us consider the random variable:

$$X_\sigma(\lambda) = \chi^\lambda(\sigma) = \frac{\text{Tr}(\rho_\lambda(\sigma))}{\dim V_\lambda}.$$

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Let us compute its expectation:

$$\mathbb{E}_{\mathbb{P}_n^\omega}(X_\sigma) = \sum_{\lambda \vdash n} c_\lambda \chi^\lambda(\sigma) = F_\omega(\sigma),$$

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Strategy: express other functions of Young diagrams in terms of  $X_\sigma$ .

## Definition of normalized characters

Let us define

$$\text{Ch}_\mu : \begin{array}{l} \mathcal{Y} \rightarrow \mathbb{C}; \\ \lambda \mapsto |\lambda|(|\lambda| - 1) \dots (|\lambda| - k + 1) \chi^\lambda(\sigma), \end{array}$$

where  $k = |\mu|$ .

and  $\sigma$  is a permutation in  $S_{|\lambda|}$  of cycle type  $\mu 1^{|\lambda|-k}$ .

In particular,

$$\text{Ch}_\mu(\lambda) = 0 \quad \text{as soon as } |\lambda| < |\mu|$$

Examples:

$$\text{Ch}_{1^k}(\lambda) = |\lambda|(|\lambda| - 1) \dots (|\lambda| - k + 1)$$

$$\text{Ch}_{(2)}(\lambda) = |\lambda|(|\lambda| - 1) \chi^\lambda((1 \ 2)) = \sum_i (\lambda_i)^2 - (\lambda'_i)^2$$

$$\text{Ch}_{\mu \cup 1}(\lambda) = (|\lambda| - |\mu|) \text{Ch}_\mu(\lambda)$$



# Product of normalized characters

$$\text{Ch}_\mu(\lambda) = |\lambda|(|\lambda| - 1) \dots (|\lambda| - k + 1) \chi^\lambda(\sigma).$$

## Proposition

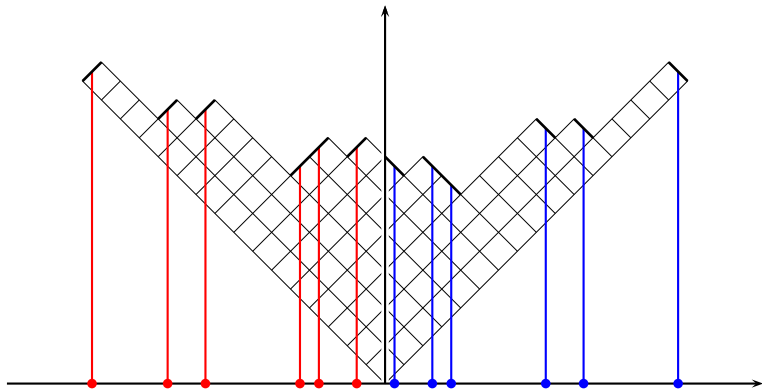
The functions  $\text{Ch}_\mu$ , when  $\mu$  runs over **all** partitions, are linearly independent (over  $\mathbb{C}$ ).

Moreover, they span a subalgebra  $\Lambda^*$  of functions on Young diagrams.

Example:  $\text{Ch}_{(2)} \cdot \text{Ch}_{(2)} = 4 \cdot \text{Ch}_{(3)} + \text{Ch}_{(2,2)} + 2 \text{Ch}_{(1,1)}$ .

# Frobenius coordinates and their power sums

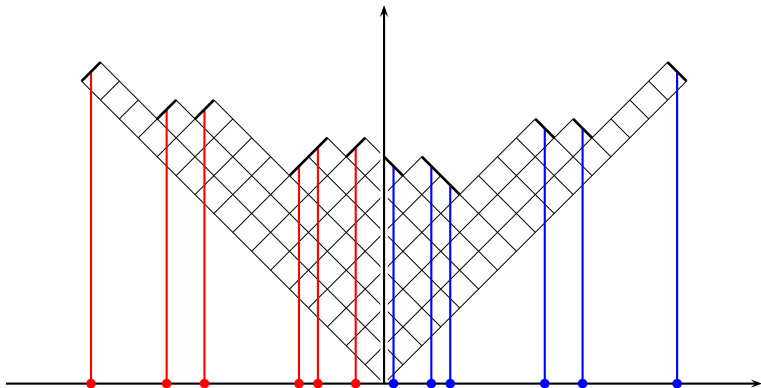
If  $\lambda$  is a Young diagram, define its Frobenius coordinates  $(a_i, b_i), 1 \leq i \leq h$  as follows:



$$b_i = \lambda'_i - i + 1/2 > 0, \quad a_i = \lambda_i - i + 1/2 > 0$$

# Frobenius coordinates and their power sums

If  $\lambda$  is a Young diagram, define its Frobenius coordinates  $(a_i, b_i), 1 \leq i \leq h$  as follows:



$$M_k(\lambda) := \sum a_i^k + (-1)^{k-1} \sum_i b_i^k$$

# An algebraic basis of $\Lambda^*$

Theorem (Kerov, Olshanski, 1994)

$(\lambda \mapsto M_k(\lambda))_{k \geq 1}$  is an algebraic basis of  $\Lambda^*$ .

Example:

$$\text{Ch}_4 = M_4 - 4M_2 \cdot M_1 + \frac{11}{2}M_2.$$

Not very explicit formula for the change of basis

$$\text{Ch}_k = [t^{k+1}] \left\{ -\frac{1}{k} \prod_{j=1}^k (1 - (j - 1/2)t) \cdot \exp \left( \sum_{j=1}^{\infty} \frac{M_j t^j}{j} (1 - (1 - kt)^{-j}) \right) \right\}$$

# Asymptotic change of basis

We consider a gradation on  $\Lambda^*$ :

$$\deg(M_k) = k$$

The highest degree term of the change of basis is easy:

$$\text{Ch}_\mu = \prod_i M_{\mu_i} + \text{smaller degree terms.}$$

# Filtration and order of magnitude

## Lemma

$$x \in \Lambda^* \Rightarrow \mathbb{E}_{\mathbb{P}_n^\omega}(x) = O(n^{\deg(x)})$$

Proof:

$$\begin{aligned} \mathbb{E}_{\mathbb{P}_n^\omega}(\text{Ch}_\mu) &= n(n-1)\dots(n-|\mu|+1)E(\lambda \mapsto \chi^\lambda(\mu)) \\ &= n(n-1)\dots(n-|\mu|+1)\rho_\mu(\omega) \\ &= O(n^{|\mu|}) \end{aligned}$$

so the lemma is true for  $x = \text{Ch}_\mu$ .

As it is a basis, this is enough. □

## Asymptotic behaviour of the $M_k$

Using the previous lemma,

$$\mathbb{E}_{\mathbb{P}_n^\omega}(M_k) = \mathbb{E}_{\mathbb{P}_n^\omega}(\text{Ch}_{(k)}) + O(n^{k-1})$$

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We can also estimate its variance:

$$\begin{aligned} \mathbb{E}_{\mathbb{P}_n^\omega}(M_k^2) - \mathbb{E}_{\mathbb{P}_n^\omega}(M_k)^2 &= \mathbb{E}_{\mathbb{P}_n^\omega}(\text{Ch}_{(k,k)}) - \mathbb{E}_{\mathbb{P}_n^\omega}(\text{Ch}_{(k)})^2 \\ &\quad + O(n^{2k-1}) \\ &= n^{2k} p_{(k,k)}(\omega) - (n^k p_k(\omega))^2 + O(n^{2k-1}) \\ &= O(n^{2k-1}) \end{aligned}$$

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$\frac{M_k(\lambda)}{n^k}$  converges in probability towards  $p_k(\omega)$ .

## End of the proof of the theorem

$\frac{M_k(\lambda)}{n^m}$  is the  $(k - 1)$ -th moment of the probability measure

$$X_\lambda = \sum_{i=1}^d (a_i^*(\lambda)/n) \delta_{(a_i^*(\lambda)/n)} + (b_i^*(\lambda)/n) \delta_{(-b_i^*(\lambda)/n)}.$$

and  $p_k(\omega)$  the  $k - 1$ -th moment of

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$$X_\omega = \gamma \delta_0 + \sum_{i \geq 1} \alpha_i \delta_{\alpha_i} + \beta_i \delta_{-\beta_i}.$$

We have convergence in probability of the repartition function at each point  $x \neq 0, \alpha_i, -\beta_i$ .

$\Rightarrow$  easy to check that it implies the theorem.

# Fluctuations of the $M_k$

- it is easy to compute the dominant term of  $\text{Var}(M_k)$  and  $\text{Cov}(M_k, M_l)$ .
- One can show that the fluctuations of the  $M_k$ 's form a Gaussian vector (using joint cumulants).  
It requires combinatorial computations in  $\Lambda^*$  and some technical tools (introduced by P. Śniady)

# Fluctuations of row and column lengths

Consider the case  $\beta_1 = \beta_2 = \dots = \gamma = 0$ . Let

$$Y_i = \sqrt{n} \left( \frac{\lambda_i}{n} - \alpha_i \right).$$

Then

$$M_k(\lambda) \sim \sum_i \lambda_i^k = n^k \left( \sum \alpha_i^k + \frac{k}{\sqrt{n}} \sum \alpha_i^{k-1} Y_i + \dots \right)$$

i.e.

$$\forall k, \quad k \sum \alpha_i^{k-1} Y_i = \text{fluctuations}(M_k).$$

We can recover the  $Y_i$  only if the  $\alpha_i$  are distinct!

## Fluctuations of row and column lengths (2)

- If the  $\alpha_i$  are distinct, it's working. For instance, if  $\omega = ((1 - q), (1 - q)q, (1 - q)q^2, \dots; 0, 0, \dots; 0)$ ,

Theorem (F., Méliot 2010)

Denote  $Y_{n,q,i}$  the rescaled deviation

$$\sqrt{n} \left( \frac{\lambda_i}{n} - (1 - q) q^{i-1} \right).$$

Then we have convergence of the finite-dimensional laws of the random process  $(Y_{n,q,i})_{i \geq 1}$  towards those of a gaussian process  $(Y_{q,i})_{i \geq 1}$  with:

$$\begin{aligned} \mathbb{E}[Y_{q,i}] &= 0 \quad ; \quad \mathbb{E}[Y_{q,i}^2] = (1 - q) q^{i-1} - (1 - q)^2 q^{2(i-1)} \quad ; \\ \text{cov}(Y_{q,i}, Y_{q,j}) &= -(1 - q)^2 q^{i+j-2}. \end{aligned}$$

- Otherwise, we need another method (cf. Pierre-Loïc's talk).

## Remark on the Plancherel case

- The theorem does not give much information.
- There is a deterministic limit shape (Logan, Shepp 77 and Kerov, Vershik 77). The fluctuations around this shape is known (Ivanov, Kerov, Olshanski 2002).
- They used the same kind of method, replacing power sums of Frobenius coordinates by **free cumulant of the transition measure** (in fact, we have adapted their ideas!).

End of the talk. Thanks!