

# Asymptotics of characters and large Young diagrams

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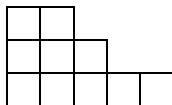
# Context

## Question

Asymptotic behavior of some models of random Young diagrams ?

In other terms:

- For each  $n$ , we give a probability measure on Young diagrams of size  $n$ .



A Young diagram of size 10.

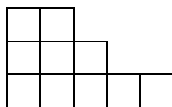
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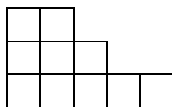
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For some measures, **representation theory of symmetric groups** is a good tool to answer these questions.

# Outline of the talk

- 1 Representations of symmetric groups and functions on Young diagrams.
- 2  $q$ -Plancherel measure
  - Plancherel measure
  - A  $q$ -deformation
  - Asymptotic behavior

# Irreducible character values

## Fact

Irreducible representations  $(V, \rho)$  of the symmetric group are (canonically) indexed by Young diagrams  $\lambda$  of size  $n$ .

If  $\sigma \in S_n$ , we look at its *normalized trace* on  $V_\lambda$ , i.e. :

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

# Polynomial functions on the set of Young diagrams

Let  $\sigma \in \mathfrak{S}_k$ . We define the following function on **all** Young diagrams:

$$\text{Ch}_\sigma(\lambda) = \begin{cases} n^{\downarrow k} \chi^\lambda(\tilde{\sigma}) & \text{if } \lambda \vdash n \geq k \\ 0 & \text{if } \lambda \vdash n < k \end{cases}$$

where  $n^{\downarrow k} = n(n-1)\dots(n-k+1)$

and  $\tilde{\sigma}$  is the image of  $\sigma$  by the canonical inclusion  $S_k \hookrightarrow S_n$  (we just add fixed points to have a permutation in  $S_n$ ).

## Theorem

*The random variables  $\text{Ch}_\sigma$  span linearly a  $\mathbb{C}$ -algebra  $\mathcal{O}$ .*

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

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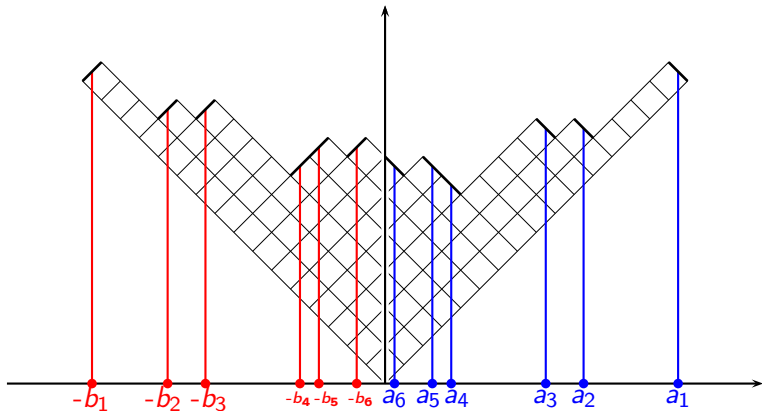
*The random variables  $\text{Ch}_\sigma$  span linearly a  $\mathbb{C}$ -algebra  $\mathcal{O}$ .*

We will describe an algebraic basis of this algebra.



# Frobenius coordinates and their power sums

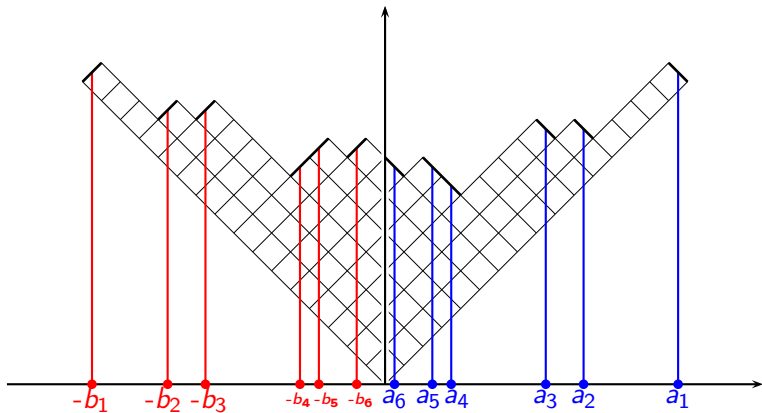
If  $\lambda$  is a Young diagram, define its Frobenius coordinates  $(a_i, b_i), 1 \leq i \leq h$  as follows:



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

# 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:



$$p_m(\mathbb{F}_\lambda) := \sum a_i^m - (-b_i)^m$$

# Character values in terms of Frobenius coordinates

Fix a permutation  $\sigma \in S_k$ .

Denote  $l_1, \dots, l_r$  the length of its cycles.

## Theorem

For all Young diagrams  $\lambda$ ,

$$\text{Ch}_\sigma(\lambda) = \prod_{j=1}^r p_{l_j}(\mathbb{F}_\lambda) + P_\sigma(p_1(\mathbb{F}_\lambda), p_2(\mathbb{F}_\lambda), \dots),$$

where  $P_\sigma$  is a polynomial in variables  $p_1, p_2, \dots$  of degree smaller than  $k$  (by definition,  $\deg(p_m) = m$ ) which does not depend on  $\lambda$ .

Ex:  $\text{Ch}_{(1\ 2\ 3\ 4)} = p_4 - 4p_2 \cdot p_1 + 11/2 p_2$ .

$\text{Ch}_{(1\ 2\ 3)(4\ 5)} = p_3 \cdot p_2 + 6p_4 - 3/2 p_2 \cdot p_1^2 - 67/4 p_2 p_1 + 21p_2$ .

## Character values in terms of Frobenius coordinates

Fix a permutation  $\sigma \in S_k$ .

Denote  $\ell_1, \dots, \ell_r$  the length of its cycles.

### Theorem

For all Young diagrams  $\lambda$ ,

$$Ch_\sigma(\lambda) = \prod_{j=1}^r p_{\ell_j}(\mathbb{F}_\lambda) + P_\sigma(p_1(\mathbb{F}_\lambda), p_2(\mathbb{F}_\lambda), \dots),$$

where  $P_\sigma$  is a polynomial in variables  $p_1, p_2, \dots$  of degree smaller than  $k$  (by definition,  $\deg(p_m) = m$ ) which does not depend on  $\lambda$ .

- Explicit formula in the one-cycle case (Wasserman, 1981);
- Easy to extend to the general case using Faharat-Higman algebra (1957).

## Inverting the previous formula

Consequence: if we define

$$\mathcal{O}_d = \text{Vect} \left( \bigcup_{1 \leq k \leq d} \bigcup_{\sigma \in S_k} \text{Ch}_\sigma \right),$$

then  $\mathcal{O} = \bigcup \mathcal{O}_d$  is a filtered algebra.

$p_m(\mathbb{F}_\lambda)$  can be expressed as a linear combination of  $\text{Ch}_\sigma$ . Ex:

$$p_3 = \text{Ch}_{(1\ 2\ 3)} + 3/2 \text{Ch}_{(1)(2)} + 1/4 \text{Ch}_{(1)};$$

$$p_2^2 = \text{Ch}_{(1\ 2)(3\ 4)} + 4\text{Ch}_{(1\ 2\ 3)} + 2\text{Ch}_{(1)(2)}.$$

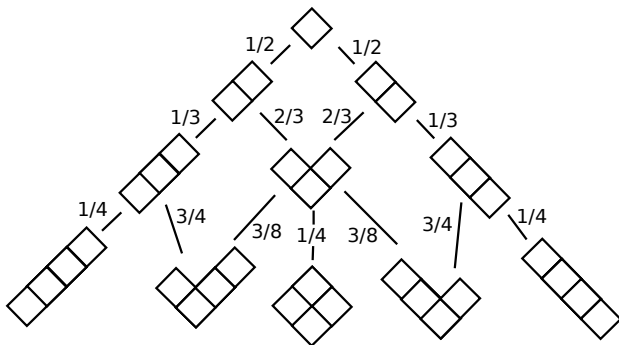
$$\left( \lambda \mapsto \prod p_{m_i}(\mathbb{F}_\lambda) \right) = \text{Ch}_\sigma + \text{smaller degree terms},$$

where  $\sigma$  is a permutation with cycles of length  $m_1, \dots, m_r$ .

# The Plancherel measure

$\mathcal{P}_n$  : a measure on Young diagrams of size  $n$ .

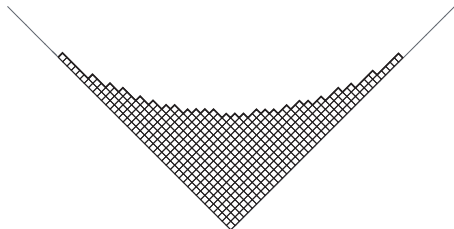
1. can be defined by a Markov process:



# The Plancherel measure

$\mathcal{P}_n$  : a measure on Young diagrams of size  $n$ .

Example of large random Young diagram:



limit shape: Kerov and Vershik / Logan and Shepp (1977);

Fluctuations: Kerov(1993), Ivanov-Olshanski (2003).

# The Plancherel measure

$\mathcal{P}_n$  : a measure on Young diagrams of size  $n$ .

2. can be defined, using representation theory:

- $\mathcal{P}_n(\{\lambda\}) = \frac{(\dim V_\lambda)^2}{n!}$ .

- Recall:

$$\mathbb{C}[\mathfrak{S}_n] \simeq \bigoplus_{\lambda \vdash n} V_\lambda^{\dim V_\lambda}$$

(action on  $\mathbb{C}[\mathfrak{S}_n]$ :  $\tau \cdot (\sum_\sigma c_\sigma \sigma) = \sum_\sigma c_\sigma \tau \circ \sigma$ .)

- In this context :

$$\mathcal{P}_n(\{\lambda\}) = \frac{\dim(\text{isotypic component of type } \lambda)}{\dim \mathbb{C}[\mathfrak{S}_n]}$$



## Normalized character values have simple expectations!

Let  $\sigma \in \mathfrak{S}_k$ . If  $n \geq k$  and  $\lambda \vdash n$ , recall that

$$\text{Ch}_\sigma(\lambda) = n(n-1)\dots(n-k+1) \cdot \chi^\lambda(\tilde{\sigma})$$

$\text{Ch}_\sigma$  can be seen as a *random variable*. Let us compute its expectation:

$$\begin{aligned} \mathbb{E}_{\mathcal{P}_n}(\text{Ch}_\sigma) &= \frac{n^{\downarrow k}}{n!} \sum_{\lambda \vdash n} (\dim V_\lambda) \cdot \text{Tr}_{V_\lambda}(\tilde{\sigma}) \\ &= \frac{n^{\downarrow k}}{n!} \text{Tr}\left(\bigoplus_{\lambda \vdash n} V_\lambda^{\dim V_\lambda}\right)(\tilde{\sigma}) = \frac{n^{\downarrow k}}{n!} \text{Tr}_{\mathbb{C}[\mathfrak{S}_n]}(\tilde{\sigma}) = n^{\downarrow k} \text{tr}_{\mathbb{C}[\mathfrak{S}_n]}(\tilde{\sigma}) \end{aligned}$$

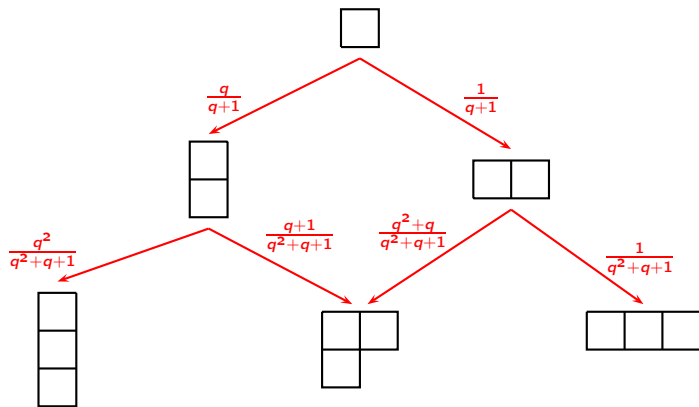
Last expression is easy to evaluate:

$$\mathbb{E}_{\mathcal{P}_n}(\text{Ch}_\sigma) = n^{\downarrow k} \delta_{\sigma, \text{Id}_k}$$

# The $q$ -Plancherel measure

$q\mathcal{P}_n$  : a measure on Young diagrams of size  $n$  (we assume  $q < 1$ ).

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1. example of a large random diagram



# The q-Plancherel measure

$q\text{-}\mathcal{P}_n$  : a measure on Young diagrams of size  $n$  (we assume  $q < 1$ ).

2. can be defined, using representation theory of Hecke algebras:

- Similarly to Plancherel measure, one has

$$\mathbb{E}_{q\text{-}\mathcal{P}_n}(\chi^{q,\bullet}(T_\sigma)) = 0 \quad \text{if } \sigma \neq 0,$$

where  $\chi^{q,\lambda}$  is the character of the irreducible representation of the Hecke algebra.

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- One can translate this with usual characters:

$$\mathbb{E}_{q\text{-}\mathcal{P}_n}(Ch_\sigma) = \frac{(1-q)^k}{\prod_j (1-q^{\ell_j})} n^{\downarrow k}.$$

# Filtration and order of magnitude

## Lemma

$$x \in \mathcal{O}_d \Rightarrow \mathbb{E}(x) = O(n^d)$$

Proof: true on the generating family  $\text{Ch}_\sigma$ .

Application:

$$\mathbb{E}_{q-\mathcal{P}_n}(p_m(\mathbb{F}_\lambda)) \sim \mathbb{E}_{q-\mathcal{P}_n}(\text{Ch}_{(1 \dots m)}) \sim \frac{(1-q)^m}{1-q^m} n^m$$

$$\text{Var}_{q-\mathcal{P}_n}(p_m(\mathbb{F}_\lambda)) = O(n^{2m-1})$$

$$\frac{p_m(\mathbb{F}_\lambda)}{n^m} \text{ converges in probability towards } \frac{(1-q)^m}{1-q^m}.$$

## Convergence of the first rows

But  $\frac{\rho_m(\mathbb{F}_\lambda)}{n^m}$  is the  $(m-1)$ -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  $\frac{(1-q)^m}{1-q^m}$  the  $(m-1)$ -th moment of

$$X_\lambda = \sum_{i \geq 1} q^{i-1} (1-q) \delta_{q^{i-1}(1-q)}$$

We have convergence in probability of the repartition function at each point  $x \neq q^{i-1}(1-q)$

Theorem (F., Méliot 2010)

For every  $i \geq 1$ , in probability

$$\lambda_i/n \longrightarrow_{q-\mathcal{P}} q^{i-1}(1-q)$$

# Remarks

- Quite simple method:
  - 1 Compute expectation of character value;
  - 2 Deduce the convergence of some parameters (easy!);
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  - 2 Deduce the convergence of some parameters (easy!);
  - 3 Translate it on the shape of the Young diagram.
- This result could be deduced directly from step 1 using Martin boundary theory.
- One can also obtain second-order asymptotics.
- Ideas come from a method of Kerov-Ivanov-Olshanski to study fluctuations of Young diagrams under Plancherel measure

# End of the talk

Thanks for listening.

Questions?