

Large random Young diagrams and representation theory

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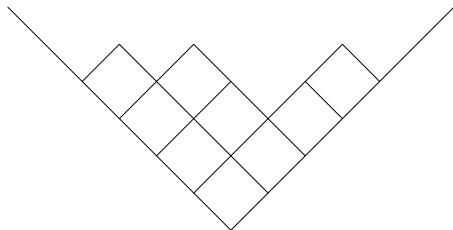
CNRS, Laboratoire Bordelais
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Workshop on Free Probability
and Random Combinatorial Structures
University of Bielefeld (Germany)
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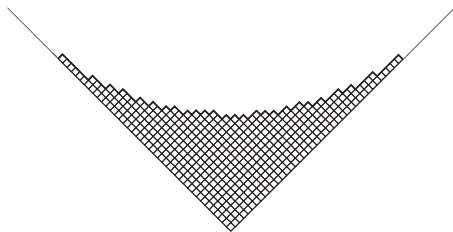
Teaser

Here is, in Russian representation, the Young diagram corresponding to $\lambda = 4, 2, 2, 1$:



Teaser

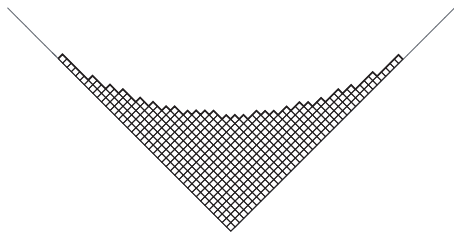
Here is, in Russian representation, a large random Young diagram (taken randomly with Plancherel's distribution):



How does it look like when we choose randomly a large (renormalized) Young diagram?

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How does it look like when we choose randomly a large (renormalized) Young diagram?

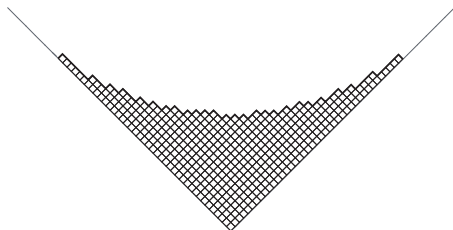
For some measures, representation theory of symmetric groups and free cumulants allow us to find easily answers to this question!

Outline of the talk

- 1 Limit law theorem for Plancherel's measure revisited
- 2 Generalizations to balanced and non-balanced random Young diagrams

Normalized border of a Young diagram

A Young diagram drawn with Russian convention



The Young diagram is determined by the continuous, piecewise affine function ω_λ in black.

Renormalization (area=1):

$$\omega_{\overline{\lambda}}(x) = (1/\sqrt{|\lambda|}) \cdot \omega_\lambda(\sqrt{|\lambda|x}).$$

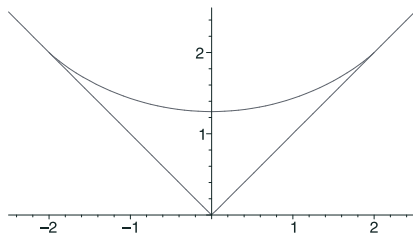
Existence of a limiting curve

Theorem (Logan and Shepp 77, Kerov and Vershik 77)

Let us take randomly (with Plancherel measure) a sequence of Young diagram λ_n of size n . Then, after renormalization, in probability, for the uniform convergence topology on continuous functions, one has:

$$\omega_{\lambda_n} \rightarrow \delta_{\Omega},$$

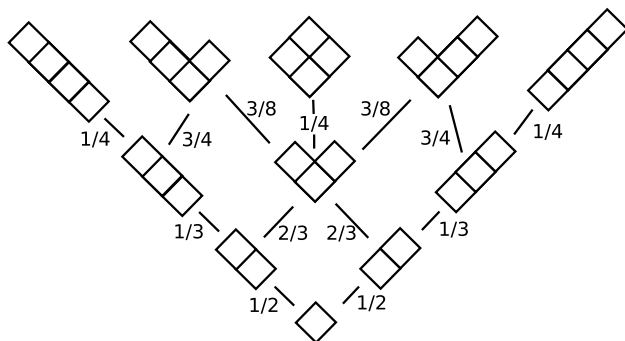
where Ω is an explicit function drawn here:



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 .

2. can be defined, using representation theory:

- Partitions of n index irreducible representations of \mathfrak{S}_n
- Therefore :

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

- In this context :

$$\mathcal{P}_n(\{\lambda\}) = \frac{(\dim V_\lambda)^2}{n!} = \frac{\dim(\text{isotypic component of type } \lambda)}{\dim \mathbb{C}[\mathfrak{S}_n]}$$

Normalized character values have simple expectations!

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

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

Let us compute its expectation:

$$\begin{aligned} \mathbb{E}_{\mathcal{P}_n}(X_\sigma) &= \frac{1}{n!} \sum_{\lambda \vdash n} (\dim V_\lambda) \cdot \text{Tr}(\rho_\lambda(\sigma)) \\ &= \frac{1}{n!} \text{Tr}\left(\bigoplus_{\lambda \vdash n} V_\lambda^{\dim V_\lambda}\right)(\sigma) = \frac{1}{n!} \text{Tr}_{\mathbb{C}[\mathfrak{S}_n]}(\sigma) = \text{tr}_{\mathbb{C}[\mathfrak{S}_n]}(\sigma) \end{aligned}$$

Last expression is easy to evaluate:

$$\mathbb{E}_{\mathcal{P}_n}(X_\sigma) = \delta_{\sigma, \text{Id}_n}$$

And now?

- Character values do not give directly informations on the shape of the diagram. 😞
- Is there some other random variables, linked to the shape of the diagram, which can be expressed in terms of normalized character values?
- Yes, thanks Kerov's and Olshanski's algebra of *polynomial functions on the set of Young diagrams*. 😊

Polynomial functions on the set of Young diagrams

Let $\mu \vdash k$ and $\sigma \in \mathfrak{S}_k$ of type μ . We define:

$$\Sigma_{\mu}(\lambda) = \begin{cases} n(n-1)\dots(n-k+1)\chi^{\lambda}(\sigma) & \text{if } \lambda \vdash n \geq k \\ 0 & \text{if } \lambda \vdash n < k \end{cases}$$

Consequence:

$$\mathbb{E}_{\mathcal{P}_n}(\Sigma_{\mu}) = \begin{cases} n(n-1)\dots(n-k+1) & \text{if } \mu = \mathbf{1}^k \text{ with } k \geq n \\ 0 & \text{else} \end{cases}$$

Theorem

The random variables Σ_{μ} span linearly an algebra denoted $\mathcal{P}ol$.

We will describe an other basis of this algebra.

Moments of transition measure

Let μ_λ be the measure defined by:

$$\int_{\mathbb{R}} \frac{d\mu_\lambda(x)}{z-x} = \frac{1}{z} \exp \left(\int_{\mathbb{R}} \frac{(\omega'(x) - \operatorname{sgn}(x)) dx}{2(z-x)} \right)$$

Theorem (Kerov, Olshanski, 1994)

If $M_k(\mu_\lambda) = \int_{\mathbb{R}} x^k d\mu_\lambda(x)$, one has:

$$\mathcal{Pol} = \mathbb{C}[\lambda \mapsto M_k(\mu_\lambda)_{k \geq 2}]$$

\Rightarrow one has an expansion

$$\prod_j M_{k_j} = \sum_{\mu} c_{\mu} \Sigma_{\mu}.$$

Idea behind the following slides

Problem: we don't have good descriptions of the expansion of $\prod_j M_{k_j}$ in terms of Σ_μ .

But we interested in asymptotics of quantities :

$$M_k(\bar{\lambda}) = \frac{1}{n^{k/2}} M_k(\lambda).$$

We do not need to know the whole expansion.

gradation

We can define a gradation on $\mathcal{P}ol$ by:

$$\deg(M_k) = k$$

Theorem (Biane, 1998)

Σ_μ has degree $|\mu| + \ell(\mu)$ and

$$\Sigma_\mu = \prod_i R_{\mu_i+1} + \text{smaller degree terms,}$$

where $R_k(\lambda)$ is the k -th free cumulant of the measure $d\mu_\lambda$ defined by:

$$M_k = \sum_{\pi \in \text{NCP}(k)} \prod_{b \in \pi} R_{|b|} \quad \text{note that } \deg(R_k) = k.$$

Inverting Biane's theorem

Formula

$$\Sigma_\mu = \prod_i R_{\mu_i+1} + \text{smaller degree terms}$$

can be read as:

One has a triangular change of basis between (Σ_μ) and $(\prod_i R_{\mu_i+1})$.

Therefore,

$$\prod_i R_{\mu_i+1} = \Sigma_\mu + \text{smaller degree terms}$$

Remark: the degree of $X = \sum_\mu c_\mu \Sigma_\mu \in \mathcal{P}ol$ is $\max_{c_\mu \neq 0} \deg(\Sigma_\mu)$.

Therefore,

$$\begin{aligned} \mathbb{E} \left(\prod R_{\mu_j+1}(\bar{\lambda}) \right) &= \sqrt{n}^{-|\mu|-\ell(\mu)} \mathbb{E} \left(\prod R_{\mu_j+1}(\lambda) \right) \\ &= \sqrt{n}^{-|\mu|-\ell(\mu)} \mathbb{E}(\Sigma_\mu) + o(1) \end{aligned}$$

Limits of free cumulants

In particular,

$$\lim_{n \rightarrow \infty} \mathbb{E}_{\mathcal{P}_n}(R_k(\bar{\lambda})) = \begin{cases} 1 & \text{if } k = 2 \\ 0 & \text{else.} \end{cases}$$

and

$$\lim_{n \rightarrow \infty} \text{Var}_{\mathcal{P}_n}(R_k(\bar{\lambda})) = 0.$$

Therefore, in probability,

$$R_k(\bar{\lambda}) \rightarrow \delta_{k,2}$$

$\rightarrow \mu_\lambda$ converges to the semi-circle law.

Limiting curve

Lemma (technical, due to Kerov)

convergence of cumulants \Rightarrow uniform convergence of ω .

Moreover, one can compute ω from the cumulant sequence.

Can be generalized!

- Take a (reducible) family of representations of \mathfrak{S}_n , whose characters are easy to compute. For instance,

$$V = (\mathbb{C}^r)^{\otimes n}, \text{ with } r \sim c \cdot n^\alpha$$

The normalized character is $\chi(\sigma) = r^{\#\text{cycles of } \sigma - n}$

- Consider the associated measures on Young diagram:

$$\begin{aligned} SW_n(\{\lambda\}) &= \frac{\dim(\text{isotypic component of type } \lambda)}{\dim((\mathbb{C}^r)^{\otimes n})} \\ &= \frac{\left| \left\{ \begin{array}{l} \text{standard tableaux} \\ \text{de forme } \lambda \end{array} \right\} \right| \cdot \left| \left\{ \begin{array}{l} \text{semi-standard tableaux} \\ \text{de forme } \lambda \text{ (entries } \leq N) \end{array} \right\} \right|}{r^n} \end{aligned}$$

Can be generalized!

- 1 If $\alpha \geq 1/2$, one can use the same method as Plancherel measure because:

$$\mathbb{E}(\Sigma_\mu) = O(n^{(|\mu|+\ell(\mu))/2}).$$

- 2 if $\alpha > 1/2$, same limit curve than the Plancherel case.
if $\alpha = 1/2$, limit curve is the curve with free cumulants

$$0, 1, c, c^2, \dots$$

(result obtained by Biane, 2001: he also computed an explicit formula for these curves)

- 3 this works in general if $\chi(\rho_1\rho_2) \sim \chi(\rho_1)\chi(\rho_2)$ as soon as ρ_1 and ρ_2 has disjoint support (also a necessary condition).

Second order asymptotics

Recall: for the first-order asymptotics, one has used

$$\prod_i R_{\mu_i+1} = \Sigma_\mu + \text{smaller degree terms}$$

If we know explicitly the next term in the expansion, one can compute the fluctuations of R_k !

- in the case of Plancherel's measure, fluctuations are gaussian: one can deduce the fluctuations of ω_λ around the limit function Ω (Kerov).
- in more generality, P. Śniady has given sufficient conditions for the fluctuations of the R_k 's to be gaussian.

Limits ?

- This methods works only for representations such that:

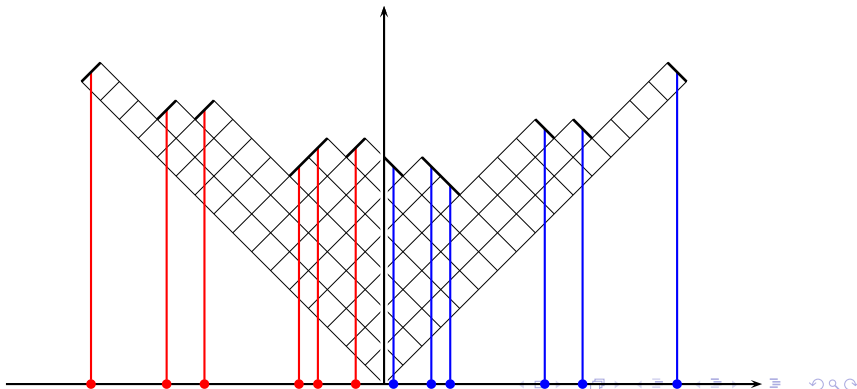
$$\mathbb{E}(\Sigma_\mu) = O(n^{(|\mu|+\ell(\mu))/2}).$$

- If this is not satisfied (for instance, case $\alpha < 1/2$ a few slides ago), the diagrams have quite big (i.e. $\gg \sqrt{n}$) row(s) and/or column(s) (they are not *balanced*).
 $\Rightarrow \bar{\lambda}$ is perhaps not the good renormalization.
- But, still, one would like to describe asymptotically the shape of the diagram.

Power sums of Frobenius coordinates

In the non-balanced case, free cumulants should be replaced by power sums of (modified) Frobenius coordinates:

$$p_k(\lambda) = \sum_{i=1}^d (a_i)^k - (b_i)^k.$$



Power sums of Frobenius coordinates

In the non-balanced case, free cumulants should be replaced by power sums of (modified) Frobenius coordinates:

$$p_k(\lambda) = \sum_{i=1}^d (a_i)^k - (b_i)^k.$$

This intuition comes from the following results:

Properties of the p_i 's

$$\mathcal{Pol} = \mathbb{C}[p_1, p_2, \dots] \text{ (Kerov, Olshanski, 1994)}$$

If λ is not balanced,

$$\Sigma_\mu(\lambda) = \prod_i p_{\mu_i}(\lambda)(1 + o(1)).$$

Motivation for a new gradation

Let us look more precisely to the measure SW_n in the case $\alpha < 1/2$.

Expectation of characters:

$$\begin{aligned}\mathbb{E}(\Sigma_\mu) &= n(n-1)\dots(n-|\mu|+1)(cn^\alpha)^{\ell(\mu)-|\mu|} \\ &\sim c^{\ell(\mu)-|\mu|} n^{\alpha\ell(\mu)-\alpha|\mu|+|\mu|}\end{aligned}$$

As we need a result of kind

$$\mathbb{E}(\Sigma_\mu) = O(n^{\deg(\Sigma)}),$$

we will define a gradation such that:

$$\deg(\Sigma_\mu) = \alpha\ell(\mu) - \alpha|\mu| + |\mu|$$

New gradation

Definition of the gradation

Let us define:

$$\deg_{\alpha}(p_{\mu}) = \alpha l(\mu) - \alpha |\mu| + |\mu|.$$

One has:

- $$\Sigma_{\mu} = p_{\mu} + \text{smaller degree terms.}$$
- If $X \in \mathcal{P}ol$, then $\mathbb{E}_{SW_n}(X) = O(n^{\deg_{\alpha}(X)})$

Convergence of power sums

Same ideas as before:

$$p_\mu = \Sigma_\mu + \text{smaller degree terms.}$$

Therefore:

$$\mathbb{E}_{SW_n}(p_\mu) = (c^{\ell(\mu)-|\mu|} + o(1)) \cdot n^{\alpha\ell(\mu)-\alpha|\mu|+|\mu|}$$

i.e.

$$\lim_{n \rightarrow \infty} \mathbb{E}_{SW_n} \left(\frac{p_k}{n^{\alpha-\alpha k+k}} \right) = \frac{1}{c^{k-1}}$$

$$\lim_{n \rightarrow \infty} \text{Var}_{SW_n} \left(\frac{p_k}{n^{\alpha-\alpha k+k}} \right) = 0$$

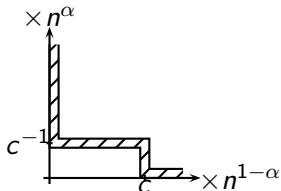
\Rightarrow convergence in probability of *normalized* power sums towards those of measure $c\delta_{c-1}$.

Result on the diagram

After a technical step, one can obtain:

Theorem (F., Méliot, 2010)

With the probability measure described before, $\forall \varepsilon, \eta > 0$, $\exists n_0$ s.t. $\forall n \geq n_0$, the border of the diagram λ_n is, after rescaling and with probability greater than $1 - \varepsilon$, contained in the hatched area of width η :



q -Plancherel measure: definition and motivation

We can also consider a q -deformation Plancherel measure defined by E. Strahov (2008).

Motivations:

- links with Hecke algebras and representations of $GL(n, \mathbb{F}_q)$.
- image by Robinson-Schensted of the distribution $q^{maj(\sigma)} / [n]!$ on permutations.

Definition:

$$\mathbb{E}_{q\text{-}\mathcal{P}_n}(\chi_q(\mathcal{T}_\mu)) = 0,$$

where the χ_q^λ are the irreducible characters of the generic Hecke algebra. Luckily, this can be translated in terms of usual characters:

$$\mathbb{E}_{q\text{-}\mathcal{P}_n}(\Sigma_\mu) = \frac{(1-q)^{|\mu|}}{\prod_i 1 - q^{\mu_i}} n^{\downarrow|\mu|}$$

!!

Asymptotics of q-Plancherel measure

Applying usual method with gradation $\deg_0(p_\mu) = |\mu|$, one obtains

Theorem (F., Méliot, 2010)

In probability, under q-Plancherel measure,

$$\forall k \geq 1, \frac{p_k(\lambda)}{|\lambda|^k} \xrightarrow{M_{n,q}} \frac{(1-q)^k}{1-q^k}.$$

Moreover,

$$\forall i \geq 1, \frac{\lambda_i}{n} \xrightarrow{M_{n,q}} (1-q)q^{i-1};$$

$$\forall i \geq 1, \frac{\lambda'_i}{n} \xrightarrow{M_{n,q}} 0,$$

We also obtained the second-order asymptotics.

Conclusion

- Showing that some parameters of the diagrams converge is very simple!
- - 1 implies immediately convergence of character.
 - 2 can be used to find a continuous limiting object with some extra works.
 - 3 not precise enough to study the first row, except if it has size $\Theta(n)$.
- Perspective: would be interesting to generalize it to other groups and objects...

Many thanks!

Thank you for listening!

Any questions?