

Sum of matrix entries of representations of the symmetric group and its asymptotics

Dario De Stavola

7 September 2015

Advisor: Valentin Féray

Affiliation: University of Zürich

Partitions

A partition $\lambda \vdash n$ is a non increasing sequence of positive integers

$$\lambda = (\lambda_1, \dots, \lambda_l)$$

such that $\sum \lambda_i = n$

Example

$$\lambda = (3, 2) \vdash 5$$

$$\lambda = \begin{array}{|c|c|c|} \hline \square & \square & \square \\ \hline \square & \square & \\ \hline \end{array}$$

Representations

A representation of S_n is a morphism

$$\pi: S_n \rightarrow GL(V)$$

where V is finite dimensional \mathbb{C} vector space

Representations

A representation of S_n is a morphism

$$\pi: S_n \rightarrow GL(V)$$

where V is finite dimensional \mathbb{C} vector space

Irreducible representations of $S_n \longleftrightarrow$ partitions $\lambda \vdash n$

Representations

A representation of S_n is a morphism

$$\pi: S_n \rightarrow GL(V)$$

where V is finite dimensional \mathbb{C} vector space

Irreducible representations of $S_n \longleftrightarrow$ partitions $\lambda \vdash n$

$$\pi^\lambda, \quad \dim \lambda := \dim V^\lambda$$

Representations

A representation of S_n is a morphism

$$\pi: S_n \rightarrow GL(V)$$

where V is finite dimensional \mathbb{C} vector space

Irreducible representations of $S_n \longleftrightarrow$ partitions $\lambda \vdash n$

$$\pi^\lambda, \quad \dim \lambda := \dim V^\lambda$$

$$\chi^\lambda(\sigma) = \text{tr}(\pi^\lambda(\sigma)), \quad \hat{\chi}^\lambda(\sigma) = \frac{\text{tr}(\pi^\lambda(\sigma))}{\dim \lambda}$$

Standard Young tableaux

| | | | | |
|----|---|----|----|----|
| 1 | 2 | 8 | 9 | 12 |
| 3 | 5 | 10 | 13 | |
| 4 | 7 | | | |
| 6 | | | | |
| 11 | | | | |

Standard Young tableaux

| | | | | |
|----|---|----|----|----|
| 1 | 2 | 8 | 9 | 12 |
| 3 | 5 | 10 | 13 | |
| 4 | 7 | | | |
| 6 | | | | |
| 11 | | | | |

$\dim \lambda :=$ number of SYT of shape λ

Standard Young tableaux

| | | | | |
|----|---|----|----|----|
| 1 | 2 | 8 | 9 | 12 |
| 3 | 5 | 10 | 13 | |
| 4 | 7 | | | |
| 6 | | | | |
| 11 | | | | |

$\dim \lambda :=$ number of SYT of shape λ

$$\lambda = (3, 2) \Rightarrow \dim \lambda = 5$$

| | | |
|---|---|---|
| 1 | 2 | 3 |
| 4 | 5 | |

| | | |
|---|---|---|
| 1 | 2 | 4 |
| 3 | 5 | |

| | | |
|---|---|---|
| 1 | 3 | 4 |
| 2 | 5 | |

| | | |
|---|---|---|
| 1 | 2 | 5 |
| 3 | 4 | |

| | | |
|---|---|---|
| 1 | 3 | 5 |
| 2 | 4 | |

Plancherel measure

$$\sum_{\lambda \vdash n} (\dim \lambda)^2 = n!$$

Plancherel measure

$$\sum_{\lambda \vdash n} (\dim \lambda)^2 = n!$$

Plancherel measure

To $\lambda \vdash n$ we associate the weight $\frac{\dim \lambda^2}{n!}$

Plancherel measure

$$\sum_{\lambda \vdash n} (\dim \lambda)^2 = n!$$

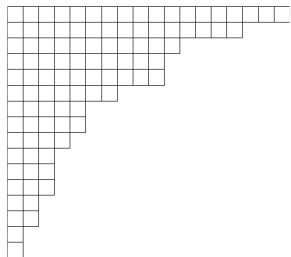
Plancherel measure

To $\lambda \vdash n$ we associate the weight $\frac{\dim \lambda^2}{n!}$

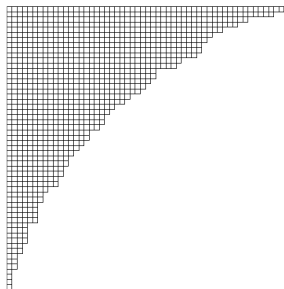
Probability on the set \mathbb{Y}_n of partitions of n

Limit shape

λ distributed with the Plancherel measure and renormalized, then

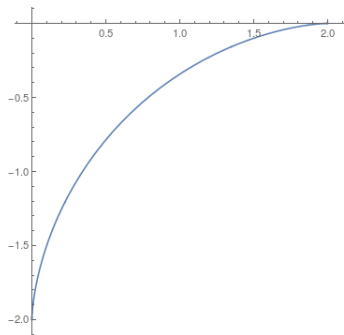


(a)



(b)

Image from D. Romik "The Surprising Mathematics of Longest Increasing Subsequences"



$$\omega_x(\theta) = \left(1 + \frac{2\theta}{\pi}\right) \sin \theta + \frac{2}{\pi} \cos \theta$$

$$\omega_y(\theta) = \left(1 - \frac{2\theta}{\pi}\right) \sin \theta - \frac{2}{\pi} \cos \theta$$

Theorem (Kerov 1999)

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \hat{\chi}_\rho^\lambda \rightarrow \prod_{k \geq 2} k^{m_k(\rho)/2} \mathcal{H}_{m_k(\rho)}(\xi_k)$$

Relations with random matrices

Rows $\lambda_1, \lambda_2, \lambda_3, \dots$ of a random Young diagram

First, second, third, \dots biggest eigenvalues of a Gaussian random Hermitian matrix

Relations with random matrices

Rows $\lambda_1, \lambda_2, \lambda_3, \dots$ of a random Young diagram

First, second, third, \dots biggest eigenvalues of a Gaussian random Hermitian matrix

Same first order asymptotics

Relations with random matrices

Rows $\lambda_1, \lambda_2, \lambda_3, \dots$ of a random
Young diagram

First, second, third, \dots biggest
eigenvalues of a Gaussian random
Hermitian matrix

Same first order asymptotics
Same joint fluctuation (Tracy-Widom law)

Relations with random matrices

Rows $\lambda_1, \lambda_2, \lambda_3, \dots$ of a random Young diagram

First, second, third, \dots biggest eigenvalues of a Gaussian random Hermitian matrix

Same first order asymptotics
Same joint fluctuation (Tracy-Widom law)

Similar tools: *moment method*, link with *free probability theory*

Signed distance

$$d_k(T) = \begin{array}{l} \text{length of northeast path from } k \text{ to } k+1 \\ \text{or } - \text{ length of southwest path from } k \text{ to } k+1 \end{array}$$

Signed distance

$$d_k(T) = \begin{array}{l} \text{length of northeast path from } k \text{ to } k+1 \\ \text{or } - \text{ length of southwest path from } k \text{ to } k+1 \end{array}$$

$$T = \begin{array}{|c|c|c|} \hline 1 & 2 & 3 \\ \hline 4 & 5 & \\ \hline \end{array} \Rightarrow d_3(T) = -3$$

Signed distance

$d_k(T)$ = length of northeast path from k to $k+1$
 or $-$ length of southwest path from k to $k+1$

$$T = \begin{array}{|c|c|c|} \hline 1 & 2 & 3 \\ \hline 4 & 5 & \\ \hline \end{array} \Rightarrow d_3(T) = -3$$

$$(3, 4) \begin{array}{|c|c|c|c|} \hline 1 & 3 & 5 & 7 \\ \hline 2 & 6 & & \\ \hline 4 & & & \\ \hline \end{array} = \begin{array}{|c|c|c|c|} \hline 1 & 4 & 5 & 7 \\ \hline 2 & 6 & & \\ \hline 3 & & & \\ \hline \end{array}$$

Young seminormal representation

$$\pi^\lambda((k, k+1))_{T, \tilde{T}} = \begin{cases} 1/d_k(T) & \text{if } T = \tilde{T} \\ \sqrt{1 - \frac{1}{d_k(T)^2}} & \text{if } (k, k+1)T = \tilde{T} \\ 0 & \text{else} \end{cases}$$

Example

$$\lambda = (3, 2)$$

$$\pi^\lambda((2, 4, 3)) = \pi^\lambda((3, 4)(2, 3)) = \pi^\lambda((3, 4))\pi^\lambda((2, 3))$$

$$= \begin{bmatrix} -1/3 & \sqrt{8/9} & 0 & 0 & 0 \\ \sqrt{8/9} & 1/3 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & -1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & -1/2 & \sqrt{3/4} & 0 & 0 \\ 0 & \sqrt{3/4} & 1/2 & 0 & 0 \\ 0 & 0 & 0 & -1/2 & \sqrt{3/4} \\ 0 & 0 & 0 & \sqrt{3/4} & 1/2 \end{bmatrix}$$

$$= \begin{bmatrix} -1/3 & -\sqrt{2/9} & \sqrt{2/3} & 0 & 0 \\ \sqrt{8/9} & -1/6 & \sqrt{1/12} & 0 & 0 \\ 0 & \sqrt{3/4} & 1/2 & 0 & 0 \\ 0 & 0 & 0 & -1/2 & \sqrt{3/4} \\ 0 & 0 & 0 & -\sqrt{3/4} & -1/2 \end{bmatrix}$$

$$0 \leq u \leq 1$$

Partial trace

$$PT_u^\lambda(\sigma) := \sum_{i \leq u \dim \lambda} \frac{\pi^\lambda(\sigma)_{i,i}}{\dim \lambda}$$

- We would like to refine Kerov's result

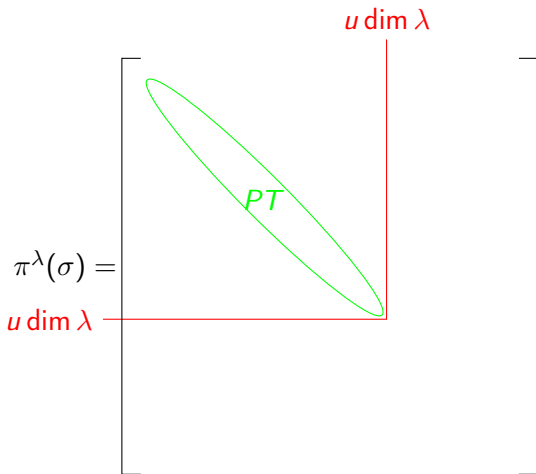
$$0 \leq u \leq 1$$

Partial trace

$$PT_u^\lambda(\sigma) := \sum_{i \leq u \dim \lambda} \frac{\pi^\lambda(\sigma)_{i,i}}{\dim \lambda}$$

- We would like to refine Kerov's result
- The partial trace has been studied in random matrix theory, e.g. for orthogonal random matrices

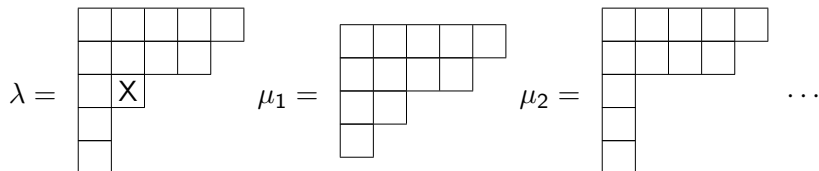
Visually



Decomposition of PT

$$\lambda = \begin{array}{|c|c|c|c|c|} \hline \square & \square & \square & \square & \square \\ \hline \square & \square & \square & \square & \\ \hline \square & \square & & & \\ \hline \square & & & & \\ \hline \square & & & & \\ \hline \end{array}$$

Decomposition of PT



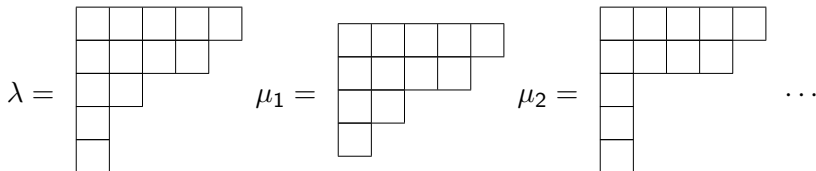
Decomposition of PT

$$\lambda = \begin{array}{|c|c|c|c|c|c|} \hline \square & \square & \square & \square & \square & \square \\ \hline \square & \square & \square & \square & \square & \\ \hline \square & \square & \square & \square & & \\ \hline \square & \square & & & & \\ \hline \square & & & & & \\ \hline \square & & & & & \\ \hline \end{array} \quad \mu_1 = \begin{array}{|c|c|c|c|c|} \hline \square & \square & \square & \square & \square \\ \hline \square & \square & \square & \square & \square \\ \hline \square & \square & \square & \square & \\ \hline \square & \square & \square & & \\ \hline \square & \square & & & \\ \hline \square & & & & \\ \hline \end{array} \quad \mu_2 = \begin{array}{|c|c|c|c|c|} \hline \square & \square & \square & \square & \square \\ \hline \square & \square & \square & \square & \square \\ \hline \square & \square & \square & \square & \\ \hline \square & \square & \square & & \\ \hline \square & \square & & & \\ \hline \square & & & & \\ \hline \square & & & & \\ \hline \end{array} \quad \dots$$

Proposition (DS)

$$PT_u^\lambda(\sigma) = \sum_{i < \bar{k}} \frac{\chi^{\mu_i}(\sigma)}{\dim \lambda} + \text{Rem}$$

Decomposition of PT



Proposition (DS)

$$PT_u^\lambda(\sigma) = \sum_{i < \bar{k}} \frac{\chi^{\mu_i}(\sigma)}{\dim \lambda} + \text{Rem}$$

$$\text{Rem} = \sum_{i \leq \tilde{u} \dim \mu_{\bar{k}}} \frac{\pi^{\mu_{\bar{k}}}(\sigma)_{i,i}}{\dim \lambda} = \frac{\dim \mu_{\bar{k}}}{\dim \lambda} PT_{\tilde{u}}^{\mu_{\bar{k}}}(\sigma)$$

Asymptotics

$$PT_u^\lambda(\sigma) = \sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma) + \text{Rem}$$


Asymptotics

$$PT_u^\lambda(\sigma) = \underbrace{\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma)} + \text{Rem}$$

$$F_{sc}(c) n^{-\frac{|\rho| - m_1(\rho)}{2}} \prod_{k \geq 2} k^{m_k(\rho)/2} \mathcal{H}_{m_k(\rho)}(\xi_k)$$

Asymptotics

$$PT_u^\lambda(\sigma) = \underbrace{\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma)} + \text{Rem}$$


$$A \cdot n^{-\frac{|\rho| - m_1(\rho)}{2}} B$$

Theorem (Kerov 1993)

$$\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \rightarrow A \text{ (deterministic)}$$

Theorem (Kerov 1999)

$$n^{\frac{|\rho| - m_{\mathbf{1}}(\rho)}{2}} \hat{\chi}^{\lambda}(\sigma) \rightarrow B \text{ (random)}$$

Theorem (Kerov 1993)

$$\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \rightarrow A \text{ (deterministic)}$$

Theorem (Kerov 1999)

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \hat{\chi}^\lambda(\sigma) \rightarrow B \text{ (random)}$$

Theorem (DS)

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma) \rightarrow AB$$

Theorem (Kerov 1993)

$$\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \rightarrow A \text{ (deterministic)}$$

Theorem (Kerov 1999)

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \hat{\chi}^\lambda(\sigma) \rightarrow B \text{ (random)}$$

Theorem (DS)

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma) \rightarrow AB$$

The two objects are *asymptotically independent*

First, a definition

Contents

$$c(\square) := \text{col}(\square) - \text{row}(\square)$$

First, a definition

Contents

$$c(\square) := \text{col}(\square) - \text{row}(\square)$$

| | | | | |
|----|----|---|---|---|
| 0 | 1 | 2 | 3 | 4 |
| -1 | 0 | 1 | 2 | |
| -2 | -1 | | | |
| -3 | | | | |
| -4 | | | | |

Jucys-Murphy elements

$$J_k := (1, k) + (2, k) + \dots + (k-1, k) \in Z(\mathbb{C}[S_n])$$

Jucys-Murphy elements

$$J_k := (1, k) + (2, k) + \dots + (k-1, k) \in Z(\mathbb{C}[S_n])$$

$$\pi^\lambda(J_k) = \begin{bmatrix} c_{T_1}(\overline{k}) & & \mathbb{0} \\ & c_{T_2}(\overline{k}) & \\ \mathbb{0} & & \ddots \end{bmatrix}$$

$$\chi^\lambda(J_2 + \dots + J_n)$$

$$\chi^\lambda(J_2 + \dots + J_n) = \sum_{i=2}^n \chi^\lambda(J_i)$$

$$\chi^\lambda(J_2 + \dots + J_n) = \sum_{i=2}^n \chi^\lambda(J_i) = \sum_{i=2}^n \sum_{k=1}^{\dim \lambda} c_{T_k}(\boxed{i})$$

$$\chi^\lambda(J_2 + \dots + J_n) = \sum_{i=2}^n \chi^\lambda(J_i) = \sum_{i=2}^n \sum_{k=1}^{\dim \lambda} c_{T_k}(\boxed{i}) = \dim \lambda \sum_{\square \in \lambda} c(\square)$$

$$\binom{n}{2} \chi^\lambda(\tau) = \chi^\lambda(J_2 + \dots + J_n) = \sum_{i=2}^n \chi^\lambda(J_i) = \sum_{i=2}^n \sum_{k=1}^{\dim \lambda} c_{T_k}(\boxed{i}) = \dim \lambda \sum_{\square \in \lambda} c(\square)$$

$$\binom{n}{2} \chi^\lambda(\tau) = \chi^\lambda(J_2 + \dots + J_n) = \sum_{i=2}^n \chi^\lambda(J_i) = \sum_{i=2}^n \sum_{k=1}^{\dim \lambda} c_{T_k}(\overline{i}) = \dim \lambda \sum_{\square \in \lambda} c(\square)$$

$$\binom{n}{2} \hat{\chi}^\lambda(\text{transposition}) = \sum_{\square \in \lambda} c(\square)$$

Considering $\chi^\lambda(J_2 + \dots + J_n)$ we get

$$\binom{n}{2} \hat{\chi}^\lambda(\text{transposition}) = \sum_{\square \in \lambda} c(\square)$$

Considering $\chi^\lambda(J_2 + \dots + J_n)$ we get

$$\binom{n}{2} \hat{\chi}^\lambda(\text{transposition}) = \sum_{\square \in \lambda} c(\square)$$

Considering $\chi^\lambda \left(\prod_{i=1}^l (J_2^{\nu_i} + \dots + J_n^{\nu_i}) \right)$ we get

Considering $\chi^\lambda(J_2 + \dots + J_n)$ we get

$$\binom{n}{2} \hat{\chi}^\lambda(\text{transposition}) = \sum_{\square \in \lambda} c(\square)$$

Considering $\chi^\lambda \left(\prod_{i=1}^l (J_2^{\nu_i} + \dots + J_n^{\nu_i}) \right)$ we get

$$c_\rho n^{\downarrow(|\rho| - m_1(\rho))} \hat{\chi}_\rho^\lambda = \prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} \right) - \sum_{\tilde{\rho} < \rho} c_{\tilde{\rho}} n^{\downarrow(|\tilde{\rho}| - m_1(\tilde{\rho}))} \hat{\chi}_{\tilde{\rho}}^\lambda$$

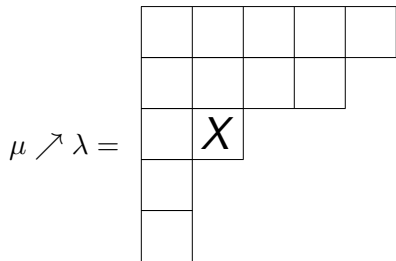
where $\rho_i = \nu_i + 1$

Considering $\chi^\lambda(J_2 + \dots + J_n)$ we get

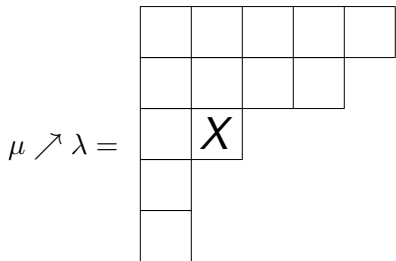
$$\binom{n}{2} \hat{\chi}^\lambda(\text{transposition}) = \sum_{\square \in \lambda} c(\square)$$

Considering $\chi^\lambda \left(\prod_{i=1}^l (J_2^{\nu_i} + \dots + J_n^{\nu_i}) \right)$ we get

$$\hat{\chi}^\lambda(\sigma) n^{\frac{|\rho| - m_1(\rho)}{2}} \sim \prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} \right)$$



$$\hat{\chi}^\mu(\sigma) n^{\frac{|\rho| - m_{\mathbf{1}}(\rho)}{2}} \sim \prod_{i=1}^l \left(\sum_{\square \in \mu} c(\square)^{\nu_i} \right)$$

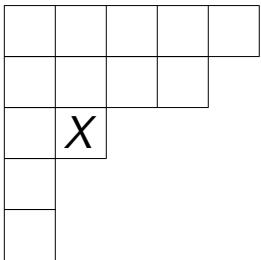


$$\hat{\chi}^\mu(\sigma) n^{\frac{|\rho| - m_{\mathbf{1}}(\rho)}{2}} \sim \prod_{i=1}^l \left(\sum_{\square \in \mu} c(\square)^{\nu_i} \right)$$

||

$$\prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} - c(\boxtimes)^{\nu_i} \right)$$

$\mu \nearrow \lambda =$



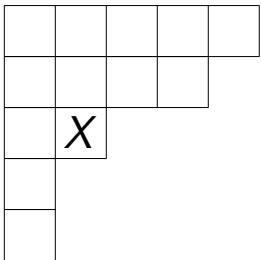
$$\hat{\chi}^\mu(\sigma) n^{\frac{|\rho| - m_{\mathbf{1}}(\rho)}{2}} \sim \prod_{i=1}^l \left(\sum_{\square \in \mu} c(\square)^{\nu_i} \right)$$

$$\parallel$$

$$\prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} - c(\boxtimes)^{\nu_i} \right)$$

$$\wr$$

$$\prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} \right)$$

$$\mu \nearrow \lambda =$$


$$\hat{\chi}^\mu(\sigma) n^{\frac{|\rho| - m_{\mathbf{1}}(\rho)}{2}} \sim$$

$$\prod_{i=1}^l \left(\sum_{\square \in \mu} c(\square)^{\nu_i} \right)$$

$$\parallel$$

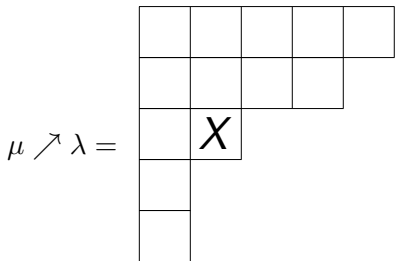
$$\prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} - c(\boxtimes)^{\nu_i} \right)$$

$$\}$$

$$\prod_{i=1}^l \left(\sum_{\square \in \lambda} c(\square)^{\nu_i} \right)$$

$$\}$$

$$\hat{\chi}^\lambda(\sigma) n^{\frac{|\rho| - m_{\mathbf{1}}(\rho)}{2}}$$



$$n^{\frac{|\rho|-m_1(\rho)}{2}} \sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma)$$

}

$$n^{\frac{|\rho|-m_1(\rho)}{2}} \left(\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \right) \hat{\chi}^{\lambda}(\sigma)$$

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \hat{\chi}^{\mu_j}(\sigma)$$

}

$$n^{\frac{|\rho| - m_1(\rho)}{2}} \left(\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} \right) \hat{\chi}^\lambda(\sigma)$$

↓

 $A \cdot B$

Telescopic sum

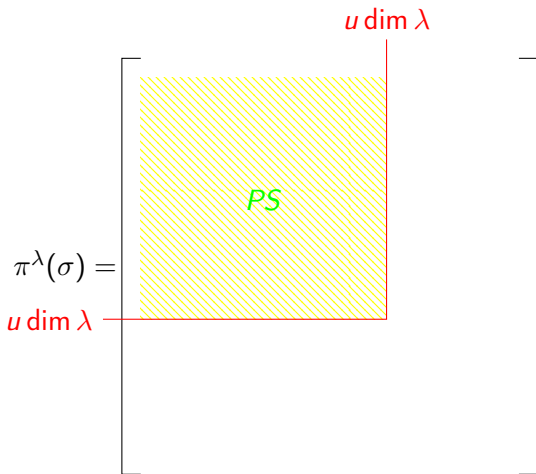
$$PT_u^\lambda(\sigma) = \sum_{j < \bar{k}_1} \frac{\dim \mu_j^{(1)}}{\dim \lambda} \hat{\chi}^{\mu_j^{(1)}}(\sigma) + \sum_{j < \bar{k}_2} \frac{\dim \mu_j^{(2)}}{\dim \lambda} \hat{\chi}^{\mu_j^{(2)}}(\sigma) + \dots$$

Unfortunately, I cannot prove convergence...

Partial sum

$$PS_u^\lambda(\sigma) := \sum_{i,j \leq u \dim \lambda} \frac{\pi^\lambda(\sigma)_{i,j}}{\dim \lambda}$$

Visually



Decomposition of PS

$$\sigma \in S_r$$

$$PS_u^\lambda(\sigma) = \sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} PS_1^{\mu_j}(\sigma) + \text{Rem}$$

Decomposition of PS

$$\sigma \in S_r$$

$$PS_u^\lambda(\sigma) = \underbrace{\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} PS_1^{\mu_j}(\sigma)} + \text{Rem}$$

$$= \sum_{\tau \in S_r} \mathbb{E}_{PL}^r [\hat{\chi}(\tau) PS_1(\sigma)] PT_u^\lambda(\tau)$$

Decomposition of PS

$$\sigma \in S_r$$

$$PS_u^\lambda(\sigma) = \underbrace{\sum_{j < \bar{k}} \frac{\dim \mu_j}{\dim \lambda} PS_1^{\mu_j}(\sigma)} + \text{Rem}$$

$$= \sum_{\tau \in S_r} \mathbb{E}_{PL}^r [\hat{\chi}(\tau) PS_1(\sigma)] PT_u^\lambda(\tau)$$

And we have convergence $PS_u^\lambda(\sigma) \rightarrow u \mathbb{E}_{PL}^r [PS_1(\sigma)]$

THANK YOU