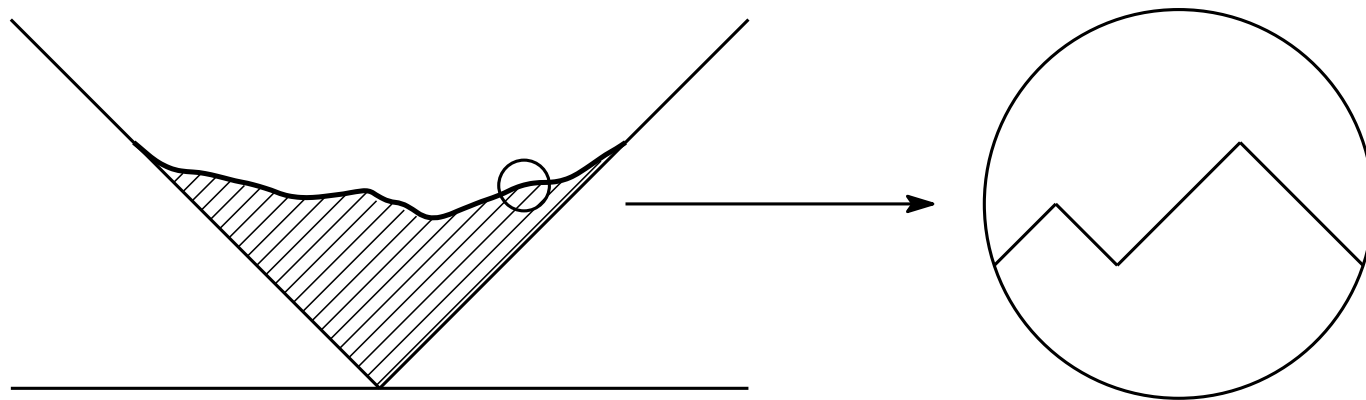


On a dynamic model for limit profiles and
their Gaussian fluctuations in
group-theoretical ensembles of Young diagrams

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Ensemble of Young diagrams \mathbb{Y}_n (n : number of boxes)

$$|\mathbb{Y}_n| = \frac{e^{\pi\sqrt{2n/3}}}{4\sqrt{3n}} \left(1 + O\left(\frac{1}{\sqrt{n}}\right)\right)$$

Pick up a diagram of \mathbb{Y}_n at random according to probability \mathbb{P}_n .

Derivation of a macroscopic shape (if any) and its time evolution under appropriate **scaling limit** as $n \rightarrow \infty$

$$\lambda \in \mathbb{Y}_n \longleftrightarrow \text{profile } \lambda(x) \longrightarrow \lambda^{\sqrt{n}}(x) = \frac{1}{\sqrt{n}} \lambda(\sqrt{n}x)$$

Probability \mathbb{P}_n determines nature of the model.

► \mathbb{Y}_n labels the equivalence classes of irreducible representations of \mathfrak{S}_n (symmetric group of degree n).

⇒ representation-theoretical ensemble of Young diagrams

► Random structure originating from irreducible decomposition or branching rule

Plancherel measure $\mathbb{M}_{\text{Pl}}^{(n)}(\lambda) = \frac{(\dim \lambda)^2}{n!}, \quad \lambda \in \mathbb{Y}_n$

• irreducible decomposition of the bi-regular representation of \mathfrak{S}_n :

$$L_2(\mathfrak{S}_n) \cong \bigoplus_{\lambda \in \mathbb{Y}_n} V^\lambda \otimes V^{\bar{\lambda}}$$

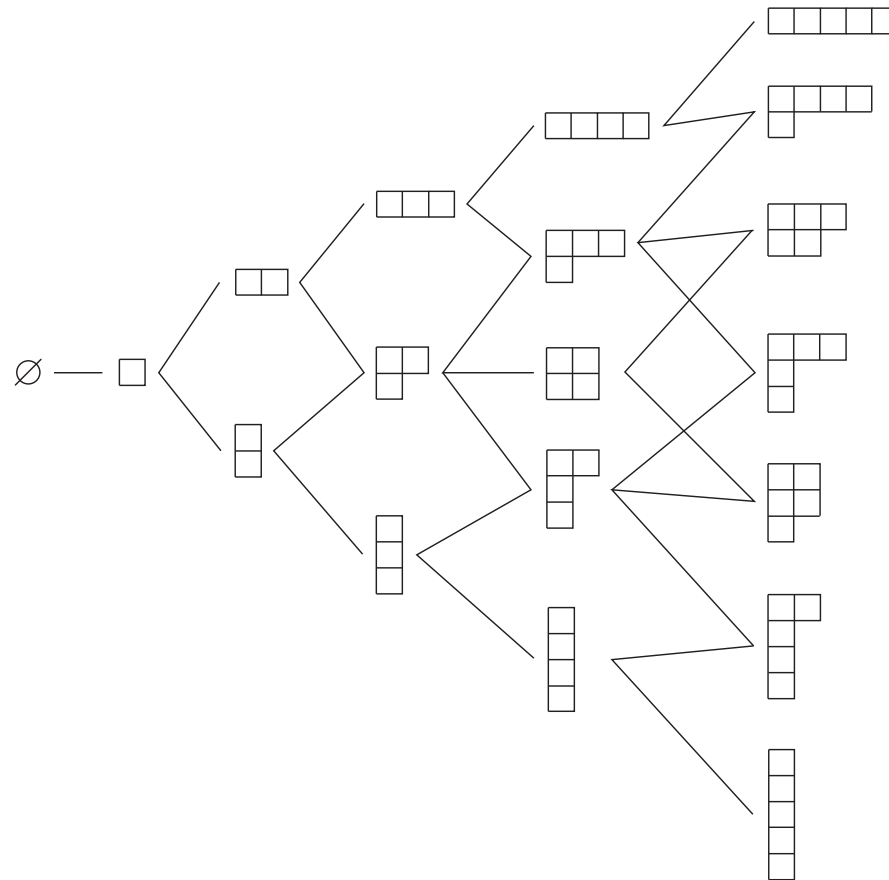
• Robinson–Schensted correspondence

$$\mathfrak{S}_n \cong \{(P, Q) \mid P, Q : \lambda\text{-type standard tableaux}, \lambda \in \mathbb{Y}_n\}$$

length of the longest increasing subsequence in x = length of the first row of λ

Young graph

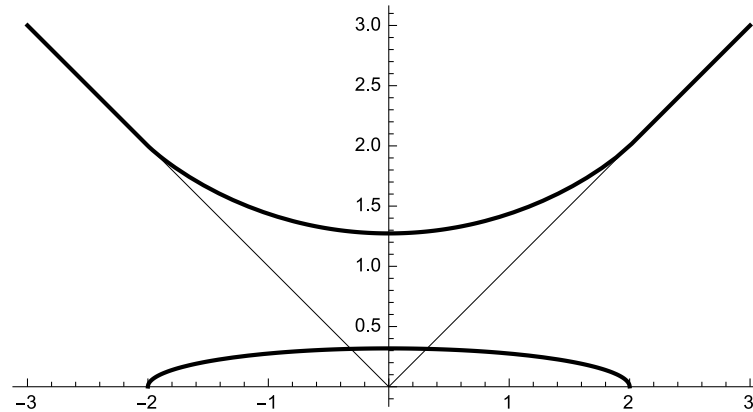
(number of paths from \emptyset = dimension of the irreducible representation)



Static model for the Plancherel ensemble $(\mathbb{Y}_n, \mathbb{M}_{\text{Pl}}^{(n)})$

Vershik – Kerov (1977), Logan – Shepp (1977)

$$\Omega(x) = \begin{cases} \frac{2}{\pi} \left(x \arcsin \frac{x}{2} + \sqrt{4 - x^2} \right), & |x| \leq 2 \\ |x|, & |x| > 2 \end{cases}$$



$$\lim_{n \rightarrow \infty} \mathbb{M}_{\text{Pl}}^{(n)} \left(\left\{ \lambda \in \mathbb{Y}_n \mid \sup_{x \in \mathbb{R}} |\lambda^{\sqrt{n}}(x) - \Omega(x)| \geq \epsilon \right\} \right) = 0 \quad (\forall \epsilon > 0)$$

Other representation-theoretical ensembles of Young diagrams

— Kerov, Biane,

Approximate factorization property for ensemble of Young diagrams

— a weak ergodicity

- positive-definite function on \mathfrak{S}_n corresponding to probability $\mathbb{M}^{(n)}$ on \mathbb{Y}_n

$$f^{(n)}(x) = \sum_{\lambda \in \mathbb{Y}_n} \mathbb{M}^{(n)}(\lambda) \tilde{\chi}^\lambda(x), \quad x \in \mathfrak{S}_n$$

- For $x, y \in \mathfrak{S}_n$ such that $\text{supp } x \cap \text{supp } y = \emptyset$

$$f^{(n)}(xy) - f^{(n)}(x)f^{(n)}(y) = o\left(n^{-\frac{|x|+|y|}{2}}\right)$$

where $|x|$ is the minimal number of transpositions needed to present x

Case of the Plancherel measure : $f^{(n)} = \delta_e$

Gaussian fluctuation

Fundamental fact :

► **Kerov's central limit theorem** for the Plancherel measure ([Kerov 1993](#))

$$\left\{ n^{\frac{k}{2}} \tilde{\chi}_{(k, 1^{n-k})}^{\lambda} \right\}_{k=2,3,\dots} \quad \text{on} \quad (\mathbb{Y}_n, \mathbb{M}_{\mathbb{P}^1}^{(n)})$$
$$\xrightarrow[n \rightarrow \infty]{} \{X_k\}_{k=2,3,\dots} : \text{independent, } X_k \sim N(0, k)$$

Extended to many directions

• profile ω vs its transition measure \mathfrak{m}_ω

e.g. \mathfrak{m}_Ω is the standard Wigner (semicircle) distribution :

$$\mathfrak{m}_\Omega(dx) = \frac{1}{2\pi} \sqrt{4 - x^2} \mathbf{1}_{[-2,2]}(x) dx$$

Markov transform :

$$\begin{aligned} \frac{1}{z} \exp \left\{ \int_{\mathbb{R}} \frac{1}{x-z} \left(\frac{\omega(x) - |x|}{2} \right)' dx \right\} &= \int_{\mathbb{R}} \frac{1}{z-x} \mathfrak{m}_{\omega}(dx) \\ &= \sum_{n=0}^{\infty} \frac{M_n(\mathfrak{m}_{\omega})}{z^{n+1}}, \quad z \in \mathbb{C}^+ \end{aligned}$$

For Young diagram $\lambda = (x_1 < y_1 < x_2 < \cdots < y_{r-1} < x_{r-1} < x_r)$
(x_i : valley, y_i : peak),

$$\frac{(z - y_1) \cdots (z - y_{r-1})}{(z - x_1) \cdots (z - x_r)} = \int_{\mathbb{R}} \frac{1}{z-x} \mathfrak{m}_{\lambda}(dx) = \frac{\mu_1}{z - x_1} + \cdots + \frac{\mu_r}{z - x_r}$$

We know $\mathfrak{m}_{\lambda\sqrt{n}} - \mathfrak{m}_{\Omega} \xrightarrow[n \rightarrow \infty]{} 0,$

Then $\sqrt{n}(\mathfrak{m}_{\lambda\sqrt{n}} - \mathfrak{m}_{\Omega}) \xrightarrow[n \rightarrow \infty]{} ?$ Ivanov – Olshanski (2002)

Interpretation of Kerov's CLT

$$M_n(\mu) = \int_{\mathbb{R}} x^n \mu(dx) = \langle x^n, \mu \rangle$$

$$\Sigma_k(\lambda) = n(n-1)\cdots(n-k+1) \tilde{\chi}_{(k, 1^{n-k})}^\lambda \quad (\lambda \in \mathbb{Y}_n)$$

- Algebra of functions on \mathbb{Y} : $\langle M_n(\mathbf{m}_\lambda) \rangle_n = \langle R_n(\mathbf{m}_\lambda) \rangle_n = \langle \Sigma_k(\lambda) \rangle_k$

free cumulant-moment formula, Kerov polynomial

$$\langle x^{2p}, \sqrt{n}(\mathbf{m}_{\lambda\sqrt{n}} - \mathbf{m}_\Omega) \rangle = \sum_{j=1}^{p-1} \binom{2p}{p-j-1} n^{-\frac{2j+1}{2}} \Sigma_{2j+1}(\lambda) + (\text{remainder}),$$

$$\langle x^{2p-1}, \sqrt{n}(\mathbf{m}_{\lambda\sqrt{n}} - \mathbf{m}_\Omega) \rangle = \sum_{j=1}^{p-1} \binom{2p-1}{p-j-1} n^{-j} \Sigma_{2j}(\lambda) + (\text{remainder})$$

on \mathbb{Y}_n (remainder : w.r.t. $\mathbb{M}_{\mathbb{P}1}^{(n)}$)

→ Inversion!

$$n^{-\frac{k}{2}} \Sigma_k(\lambda) = \langle (\text{Chebyshev polynomial}), \sqrt{n}(\mathbf{m}_{\lambda\sqrt{n}} - \mathbf{m}_{\Omega}) \rangle + (\text{remainder})$$

LHS $\xrightarrow[n \rightarrow \infty]{} (\text{independent})$ Gaussian r.v. by Kerov's CLT

Chebyshev polynomials form ONB in $L^2\left((-2, 2), \frac{1}{\pi\sqrt{4-x^2}} dx\right)$

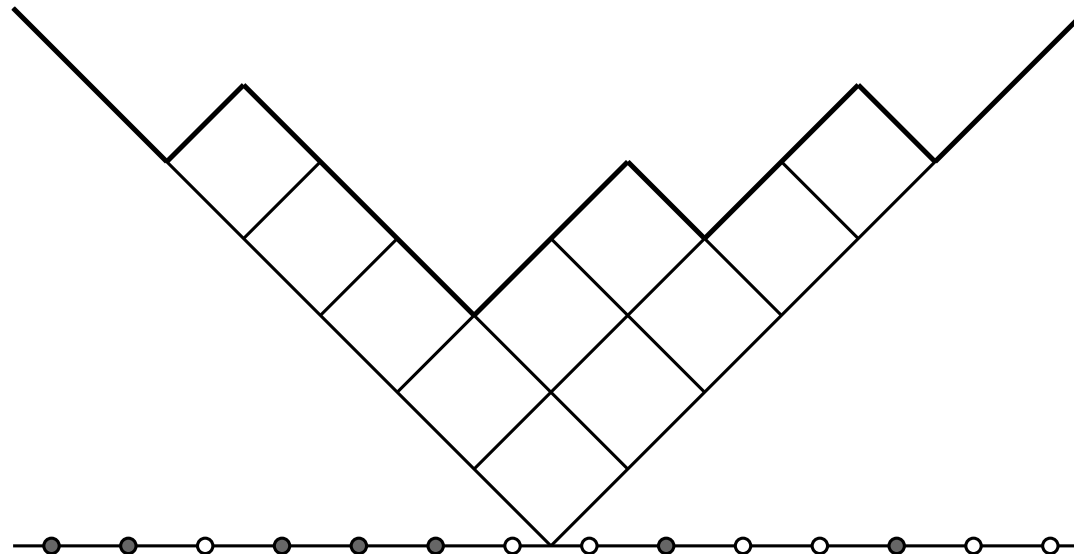
RHS : Coefficient of a random Fourier series

► $\sqrt{n}(\mathbf{m}_{\lambda\sqrt{n}} - \mathbf{m}_{\Omega})$ converges to

$$\begin{aligned} & \sum (\text{Gaussian r.v.}) \times (\text{Chebyshev polynomials}) \frac{1}{\pi\sqrt{4-x^2}} \\ &= \sum_{k=3}^{\infty} \sqrt{k-1} \xi_{k-1} T_k\left(\frac{x}{2}\right) \frac{1}{\pi\sqrt{4-x^2}} \end{aligned}$$

$\{\xi_{k-1}\}$: independent standard Gaussian, $T_k(\cos \theta) = \cos k\theta$

Dynamic model — scaling limit under not only growth of Young diagrams but also transition between diagrams (Markov chain)



1 step transition :

Move one box (peak \rightarrow valley)

\iff twice adjacent flips $\bullet \iff \circ$ in Maya diagram

Canonical setting

$$P_{\lambda, \nu}^{\downarrow} = \begin{cases} \frac{\dim \nu}{\dim \lambda}, & \nu \nearrow \lambda, \\ 0, & \text{otherwise} \end{cases}, \quad P_{\nu, \mu}^{\uparrow} = \begin{cases} \frac{\dim \mu}{(|\nu|+1) \dim \nu}, & \nu \nearrow \mu, \\ 0, & \text{otherwise} \end{cases}$$

Irreducible decomposition of restriction/induction of representations

$$\text{Res}_{\mathfrak{S}_{n-1}}^{\mathfrak{S}_n} \pi^\lambda \cong \bigoplus_{\nu \in \mathbb{Y}_{n-1}: \nu \nearrow \lambda} \pi^\nu, \quad \text{Ind}_{\mathfrak{S}_{n-1}}^{\mathfrak{S}_n} \pi^\nu \cong \bigoplus_{\mu \in \mathbb{Y}_n: \nu \nearrow \mu} \pi^\mu$$

The Plancherel measure $\mathbb{M}_{\mathbb{P}^1}^{(n)}$ on \mathbb{Y}_n is kept invariant by transition probability $P^{(n)} = P^\downarrow P^\uparrow$.

Continuous time Markov chain $(X_s^{(n)})_{s \geq 0}$ on \mathbb{Y}_n

► macroscopic time t , microscopic time $s = tn$, spatial rescale by \sqrt{n}

initial distribution $M_0^{(n)}$



distribution at time tn $M_t^{(n)}$



stationary distribution $M_{PI}^{(n)}$



ω_0

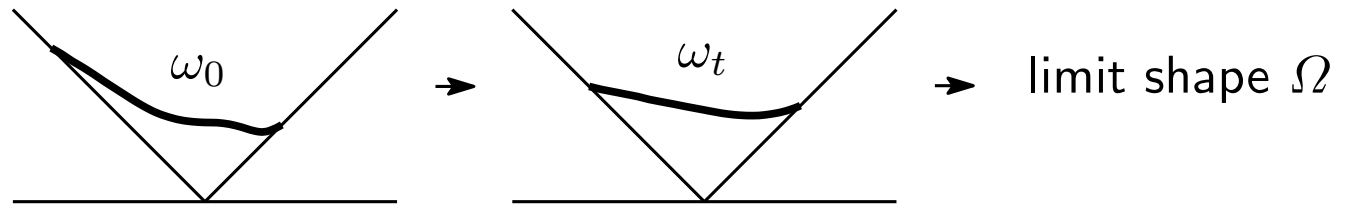
$1/\sqrt{n}, n \rightarrow \infty$



ω_t



Ω



Funaki – Sasada : [CMP 2010]

Grand canonical

- probability on \mathbb{Y} $\mu^\epsilon(\lambda) = Z^{-1} \epsilon^{|\lambda|}$, $\lambda \in \mathbb{Y}$ ($0 < \epsilon < 1$)
such that $\mathbb{E}_{\mu^\epsilon} [|\lambda|] = N^2$ ($\implies \lim_{N \rightarrow \infty} \epsilon = 1$)
- number of boxes ± 1 at 1 step transition (at random for peaks or valleys)
— μ^ϵ kept invariant

Continuous time Markov chain \longrightarrow distribution ν_s^N at time s

► Rescale for time $s = tN^2$, for space $\lambda^N(x) = \frac{1}{N} \lambda(Nx)$

Assumption : weak LLN under initial ensemble ν^N : $\lim_{N \rightarrow \infty} \lambda^N = \psi_0$

Result : $\forall t > 0$, weak LLN under $\nu_{tN^2}^N$: $\lim_{N \rightarrow \infty} \lambda^N = \psi_t$

PDE satisfied by $\psi_t(x)$ is obtained.

Theorem [Publ. RIMS 2015] (canonical setting)

Assumption : Initial ensemble $\{(\mathbb{Y}_n, \mathbb{M}_0^{(n)})\}_n$ satisfies approximate factorization property.

Result : $\forall t > 0$, $\mathbb{M}_t^{(n)}$ denoting distribution of $X_{tn}^{(n)}$, the ensemble $\{(\mathbb{Y}_n, \mathbb{M}_t^{(n)})\}_n$ at time t also satisfies approximate factorization property, and weak LLN : $\forall \epsilon > 0$,

$$\lim_{n \rightarrow \infty} \mathbb{M}_t^{(n)} \left(\left\{ \lambda \in \mathbb{Y}_n \mid \sup_{x \in \mathbb{R}} |\lambda^{\sqrt{n}}(x) - \omega_t(x)| \geq \epsilon \right\} \right) = 0$$

holds. Macroscopic shape ω_t at t is characterized by

$$\mathfrak{m}_{\omega_t} = (\mathfrak{m}_{\omega_0})_{e^{-t}} \boxplus (\mathfrak{m}_{\Omega})_{1-e^{-t}}$$

(free convolution of free compressions of Kerov transition measures).

Time evolution of the distributions could be described through their Stieltjes transforms

► Stieltjes transform of semicircle distribution μ_t of mean 0 and variance t

$$\begin{aligned} g(t, z) &= \int_{\mathbb{R}} \frac{1}{z - x} \mu_t(dx) = \int_{-2\sqrt{t}}^{2\sqrt{t}} \frac{1}{z - x} \frac{\sqrt{4t - x^2}}{2\pi t} dx \\ &= \frac{z - \sqrt{z^2 - 4t}}{2t} \end{aligned}$$

satisfies PDE :

$$\frac{\partial g}{\partial t} = -g \frac{\partial g}{\partial z}$$

- ▶ PDE describing time evolution of transition measure \mathfrak{m}_{ω_t}

$$G(t, z) = \int_{\mathbb{R}} \frac{1}{z - x} \mathfrak{m}_{\omega_t}(dx),$$

$$\frac{\partial G}{\partial t} = \frac{1}{G} \frac{\partial G}{\partial z} + G - G \frac{\partial G}{\partial z}, \quad t > 0, z \in \mathbb{C}^+$$

- ▷ PDE for $\omega(t, x) = \omega_t(x) \dots$

$$c \frac{\partial \omega}{\partial t} = \frac{\partial^2 \omega}{\partial x^2} - \frac{4}{\pi^2} \left(\omega - x \frac{\partial \omega}{\partial x} \right)^{-1} \quad ??$$

(Constraint of constant area ?)

- ▶ ODE for Ω :

$$\frac{\partial^2 \Omega}{\partial x^2} - \frac{4}{\pi^2} \left(\Omega - x \frac{\partial \Omega}{\partial x} \right)^{-1} = 0$$

Case of Funaki – Sasada

In French style

time evolution $\partial_t \psi = \partial_u \left(\frac{\partial_u \psi}{1 - \partial_u \psi} \right) + \frac{\pi}{\sqrt{6}} \frac{\partial_u \psi}{1 - \partial_u \psi}$

$t \rightarrow \infty$ $e^{-(\pi/\sqrt{6})u} + e^{-(\pi/\sqrt{6})\psi(u)} = 1$

In Russian style

time evolution $\frac{\partial \phi}{\partial t} = \frac{1}{2} \frac{\partial^2 \phi}{\partial x^2} - \frac{\pi}{4\sqrt{3}} \left(1 - \left(\frac{\partial \phi}{\partial x} \right)^2 \right)$

$t \rightarrow \infty$ $\phi(x) = \frac{2\sqrt{3}}{\pi} \log \left(e^{(\pi/2\sqrt{3})x} + e^{-(\pi/2\sqrt{3})x} \right)$

Although PDE of time evolution for ω_t is unknown, we have a solution in some sense (through Markov transform)

► Relation between profile ω_t and its Kerov transition measure \mathfrak{m}_{ω_t}
(recall the Markov transform)

$$\int_{\mathbb{R}} \frac{1}{z-x} \mathfrak{m}_{\omega_t}(dx) = \frac{1}{z} \exp \left\{ \int_{\mathbb{R}} \frac{1}{x-z} \left(\frac{\omega_t(x) - |x|}{2} \right)' dx \right\}, \quad z \in \mathbb{C}^+$$

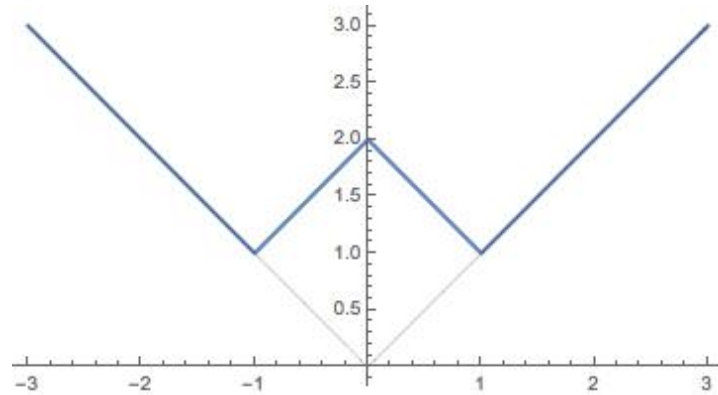
Procedure of computing ω_t from \mathfrak{m}_{ω_t} :

free convolution, free compression \longrightarrow free cumulants

\longrightarrow Voiculescu R -transform \longrightarrow Stieltjes transform

\longrightarrow taking log \longrightarrow inversion of Stieltjes transform \longrightarrow integration!

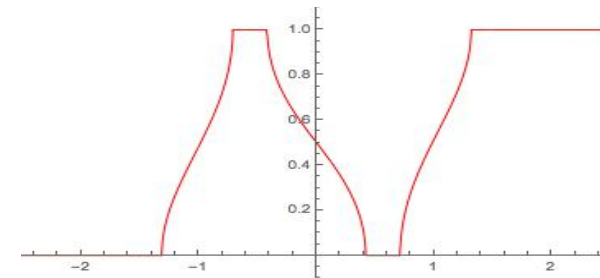
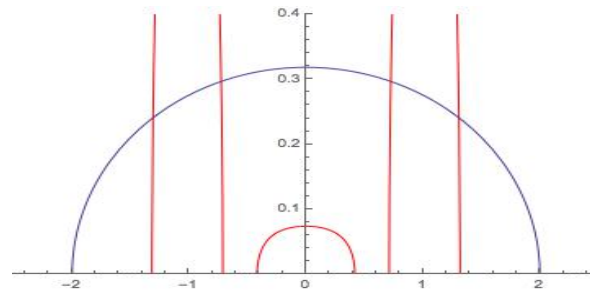
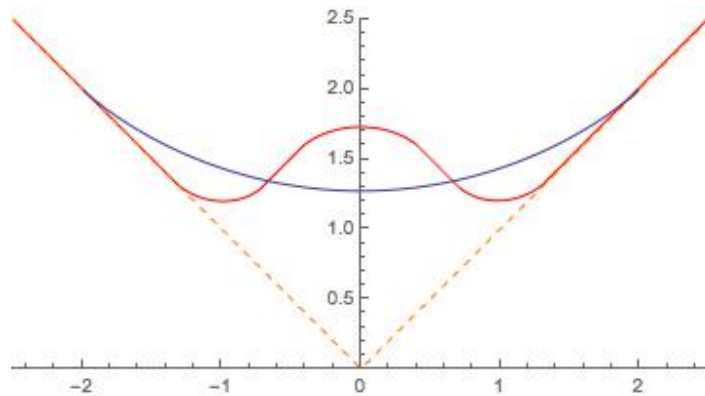
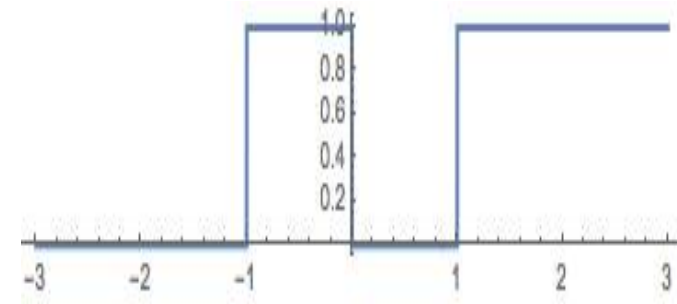
profile



transition measure

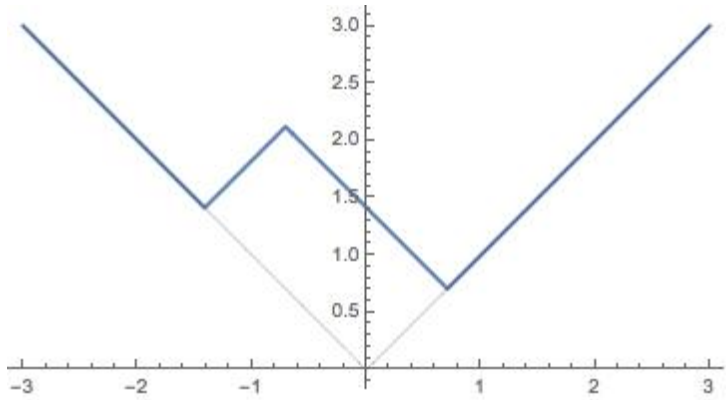
$$\frac{1}{2} \delta_{-1} + \frac{1}{2} \delta_1$$

Rayleigh function



by T.Hasebe

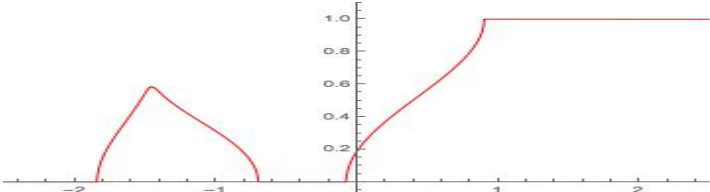
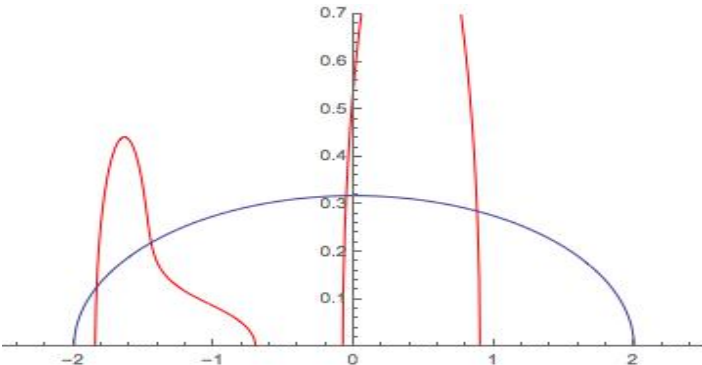
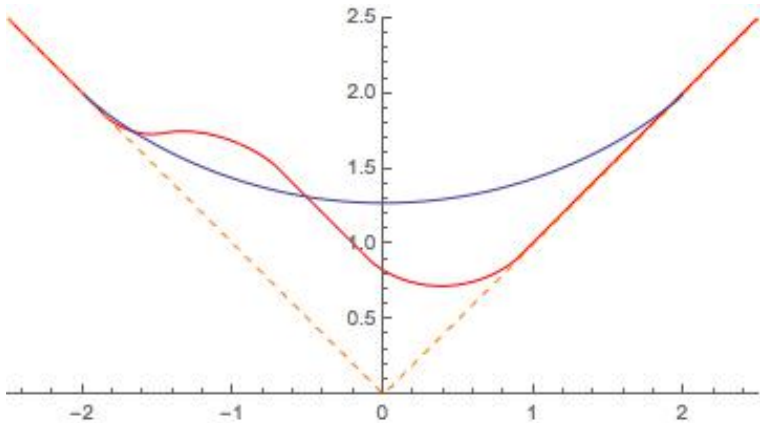
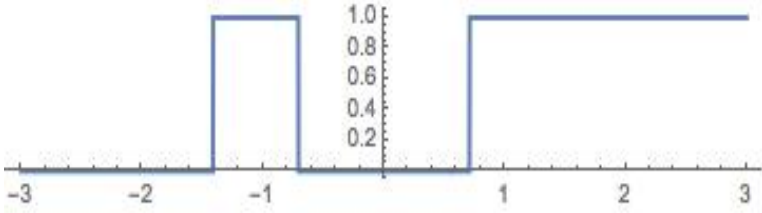
profile



transition measure

$$\frac{1}{3}\delta_{-\sqrt{2}} + \frac{2}{3}\delta_{1/\sqrt{2}}$$

Rayleigh function



Dynamic model : initial $\rightarrow\rightarrow\rightarrow$ Plancherel

Fluctuation for other (non-Plancherel) ensembles

Śniady (2005) “character factorization property”

► $(\mathbb{Y}_n, \mathbb{M}^{(n)})$, C : cumulant functional w.r.t. $E_{\mathbb{M}^{(n)}}$

Assume $C[\Sigma_{j_1}, \dots, \Sigma_{j_k}] = O\left(n^{\frac{j_1 + \dots + j_k - k + 2}{2}}\right)$. Then

$\left\{ \sqrt{n} \left(n^{-\frac{j+1}{2}} \Sigma_j - E_{\mathbb{M}^{(n)}} \left[n^{-\frac{j+1}{2}} \Sigma_j \right] \right) \right\}_{j \geq 2} \xrightarrow[n \rightarrow \infty]{} \{X_j\} : \text{Gaussian, mean 0}$

In our model :

$\sqrt{n}(\mathbf{m}_{\lambda\sqrt{n}} - \mathbf{m}_{\omega_t})$ on $(\mathbb{Y}_n, \mathbb{M}_t^{(n)}) \xrightarrow[n \rightarrow \infty]{} \text{fluctuation at macroscopic time } t$

► $\left\{ \langle x^j, \sqrt{n}(\mathbf{m}_{\lambda\sqrt{n}} - \mathbf{m}_{\omega_t}) \rangle \right\}_j \xrightarrow[n \rightarrow \infty]{} \text{Gaussian system with mean 0}$

► Covariance has complicated t -dependence, vanishes as $t \rightarrow \infty$.

Grand canonical setting $\mathbb{Y} = \bigsqcup_{n=0}^{\infty} \mathbb{Y}_n$

Poissonization of the Plancherel measure

$$\mathbb{M}_{\text{PP}}^{(\xi)} = \sum_{n=0}^{\infty} \frac{e^{-\xi} \xi^n}{n!} \mathbb{M}_{\text{Pl}}^{(n)}, \quad \xi > 0$$

is kept invariant under transition probability $P^{(\xi)}$ on \mathbb{Y} :

$$P^{(\xi)} = \alpha_{\xi}(n) P^{\uparrow(n)} + (1 - \alpha_{\xi}(n)) P^{\downarrow(n)},$$

$$\alpha_{\xi}(n) = \int_0^1 \xi e^{-\xi x} (1-x)^n dx$$

Continuous time Markov chain $(X_s^{(\xi)})_{s \geq 0}$

► Rescale for time $t\xi$, for space $\frac{1}{\sqrt{\xi}} \lambda(\sqrt{\xi}x)$ ($\lambda \in \mathbb{Y}$)

Behavior as $\xi \rightarrow \infty$

Method of proofs

- profile of Young diagram $\lambda \longleftrightarrow$ transition measure \mathfrak{m}_λ
- method of symmetric functions (generators, generating functions)
- free cumulant $R_k(\mathfrak{m}_\lambda)$ vs irreducible character value χ^λ at cycle
 \longleftarrow Kerov polynomial
- diagonalizing transition probability by using irreducible characters
- estimate of ensemble expectation by using approximate factorization property

END