

Asymptotic Analysis of Random Matrices and Orthogonal Polynomials

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Multiple orthogonal polynomials

- Given weight functions w_1, \dots, w_r on the real line and $\vec{n} = (n_1, \dots, n_r) \in \mathbb{N}^r$. Notation $|\vec{n}| = n_1 + \dots + n_r$
- The **type II multiple orthogonal polynomial (MOP)** is a monic polynomial $P_{\vec{n}}$ of degree $|\vec{n}|$ such that

$$\left\{ \begin{array}{ll} \int P_{\vec{n}}(x) x^k w_1(x) dx = 0, & k = 0, 1, \dots, n_1 - 1, \\ \int P_{\vec{n}}(x) x^k w_2(x) dx = 0, & k = 0, 1, \dots, n_2 - 1, \\ \vdots & \vdots, \\ \int P_{\vec{n}}(x) x^k w_r(x) dx = 0, & k = 0, 1, \dots, n_r - 1, \end{array} \right.$$

- These are $|\vec{n}|$ conditions for the $|\vec{n}|$ free coefficients of $P_{\vec{n}}$. In typical cases there is existence and uniqueness, but not always.

Type I multiple orthogonality

- **Type I multiple orthogonal polynomials** are r polynomials $A_{\vec{n}}^{(1)}, A_{\vec{n}}^{(2)}, \dots, A_{\vec{n}}^{(r)}$, of degrees

$$\deg A_{\vec{n}}^{(j)} \leq n_j - 1, \quad j = 1, \dots, r$$

- They are such that the linear form

$$Q_{\vec{n}}(x) = A_{\vec{n}}^{(1)}(x)w_1(x) + \dots + A_{\vec{n}}^{(r)}(x)w_r(x)$$

satisfies

$$\begin{cases} \int x^k Q_{\vec{n}}(x) dx = 0, & k = 0, 1, \dots, |\vec{n}| - 2, \\ \int x^k Q_{\vec{n}}(x) dx = 1, & k = |\vec{n}| - 1. \end{cases}$$

Block Hankel matrix

- **Moments** $\mu_j^{(i)} = \int x^j w_i(x) dx$
- $n \times m$ **Hankel matrix** for i th weight

$$H_{n,m}^{(i)} = \left(\mu_{j+k-2}^{(i)} \right)_{j=1,\dots,n,k=1,\dots,m}$$

- **Block Hankel matrix**

$$H_{\vec{n}} = \begin{bmatrix} H_{n,n_1}^{(1)} & \cdots & H_{n,n_r}^{(r)} \end{bmatrix}, \quad n = |\vec{n}|$$

- **Conditions for type I MOPs** give linear system with matrix $H_{\vec{n}}$.
- **Conditions for type II MOP** give linear system with matrix $H_{\vec{n}}^T$.
- **Both type of MOPs** exist if and only if

$$\det H_{\vec{n}} \neq 0.$$

Riemann-Hilbert problem (case $r = 2$)

In the RH problem we look for a 3×3 matrix valued function $Y(z)$ satisfying

RH-Y1 $Y : \mathbb{C} \setminus \mathbb{R} \rightarrow \mathbb{C}^{3 \times 3}$ is analytic.

RH-Y2 Y has boundary values for $x \in \mathbb{R}$, denoted by $Y_{\pm}(x)$, and

$$Y_+(x) = Y_-(x) \begin{pmatrix} 1 & w_1(x) & w_2(x) \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \text{for } x \in \mathbb{R}.$$

RH-Y3 As $z \rightarrow \infty$,

$$Y(z) = \left(I + \mathcal{O} \left(\frac{1}{z} \right) \right) \begin{pmatrix} z^{n_1+n_2} & 0 & 0 \\ 0 & z^{-n_1} & 0 \\ 0 & 0 & z^{-n_2} \end{pmatrix}$$

Solution in terms of type II MOPs

Theorem (Van Assche, Geronimo, K (2001))

RH problem has a unique solution if and only if the type II MOP $P_{\vec{n}}$ uniquely exists.

In that case the first row of Y is given by

$$\begin{pmatrix} P_{\vec{n}}(z) & \frac{1}{2\pi i} \int_{-\infty}^{\infty} \frac{P_{\vec{n}}(s)w_1(s)}{s-z} ds & \frac{1}{2\pi i} \int_{-\infty}^{\infty} \frac{P_{\vec{n}}(s)w_2(s)}{s-z} ds \\ * & * & * \\ * & * & * \end{pmatrix}$$

- Other rows are filled using $P_{\vec{n}-\vec{e}_1}$ and $P_{\vec{n}-\vec{e}_2}$ (if they exist).

Inverse of Y

- The type I MOPs are in the inverse of Y .

$$Y^{-1}(z) = \begin{pmatrix} -\int_{-\infty}^{\infty} \frac{Q_{\vec{n}}(s)}{s-z} ds & * & * \\ 2\pi i A_{\vec{n}}^{(1)}(z) & * & * \\ 2\pi i A_{\vec{n}}^{(2)}(z) & * & * \end{pmatrix}$$

where $Q_{\vec{n}} = A_{\vec{n}}^{(1)} w_1 + A_{\vec{n}}^{(2)} w_2$

- Other columns contain type I MOPs with multi-indices $\vec{n} + \vec{e}_1$ and $\vec{n} + \vec{e}_2$.

Biorthogonal ensembles

- **Probability density function on \mathbb{R}^n of the form**

$$\frac{1}{Z_n} \det [f_i(x_j)]_{i,j=1}^n \cdot \det [g_i(x_j)]_{i,j=1}^n,$$

- **Normalization constant**

$$Z_n = \int_{\mathbb{R}^n} \det [f_i(x_j)]_{i,j=1}^n \cdot \det [g_i(x_j)]_{i,j=1}^n dx_1 \cdots dx_n \neq 0$$

- **By [Andréief \(1883\)](#) identity**

$$Z_n = n! \det M_n, \quad M_n = \left[\int_{-\infty}^{\infty} f_i(x) g_j(x) dx \right]_{i,j=1}^n$$

- **Corollary:** $\det M_n \neq 0$

Correlation kernel

- Biorthogonal ensemble is a **determinantal point process** with correlation kernel

$$K_n(x, y) = \sum_{i=1}^n \sum_{j=1}^n (M_n^{-1})_{ji} f_i(x) g_j(y).$$

- Representation as determinant

$$K_n(x, y) = -\frac{1}{\det M_n} \det \begin{bmatrix} & M_n & \begin{matrix} f_1(x) \\ \vdots \\ f_n(x) \end{matrix} \\ g_1(y) \cdots g_n(y) & & 0 \end{bmatrix}$$

- Perform elementary row and column transformations to transform M_n to the identity matrix I_n

Correlation kernel (cont.)

- After transformation $M_n \mapsto I_n$

$$K_n(x, y) = -\det \begin{bmatrix} & & \phi_1(x) \\ & I_n & \vdots \\ \psi_1(y) & \cdots & \psi_n(y) & 0 \end{bmatrix}$$

with functions ϕ_j and ψ_j satisfying

$$\int_{-\infty}^{\infty} \phi_i(x) \psi_j(x) dx = \delta_{i,j} \quad (\text{biorthogonality})$$

- Also single sum $K_n(x, y) = \sum_{j=1}^n \phi_j(x) \psi_j(y)$

Correlation kernel (cont.)

- **Characterization:**

K_n is the kernel of the **projection operator** onto the linear span of f_1, \dots, f_n , whose kernel is the orthogonal complement of the linear span of g_1, \dots, g_n .

- **Operator**

$$K_n : h \mapsto K_n h, \quad K_n h(x) = \int K_n(x, y) h(y) dy$$

- **Characterization**

$$K_n h = h \quad \text{if } h = \alpha_1 f_1 + \alpha_2 f_2 + \dots + \alpha_n f_n,$$

$$K_n h = 0 \quad \text{if } \int h(x) g_j(x) dx = 0 \quad \text{for } j = 1, \dots, n.$$

Definition

A **multiple orthogonal polynomial (MOP) ensemble** is a biorthogonal ensemble with functions

$$\begin{aligned}f_i(x) &= x^{i-1}, & \text{for } i = 1, \dots, n, \\g_i(y) &= y^{i-1} w_1(y), & \text{for } i = 1, \dots, n_1, \\g_{n_1+i}(y) &= y^{i-1} w_2(y), & \text{for } i = 1, \dots, n_2, \\& \vdots \\g_{n_1+\dots+n_{r-1}+i}(y) &= y^{i-1} w_r(y), & \text{for } i = 1, \dots, n_r.\end{aligned}$$

Here w_1, \dots, w_r are given functions, and n_1, \dots, n_r are non-negative integers such that

$$n = n_1 + \dots + n_r.$$

Block Hankel matrix

- In a MOP ensemble the matrix M_n is the block Hankel matrix

$$M_n = H_{\vec{n}} = \begin{bmatrix} H_{n,n_1}^{(1)} & \cdots & H_{n,n_r}^{(r)} \end{bmatrix}, \quad n = |\vec{n}|$$

- $\det H_{\vec{n}} \neq 0$ and so the MOPs exist.
- The RH problem has a unique solution.

Christoffel Darboux formula

Theorem (Bleher-K (2004) for $r = 2$, Daems-K (2004))

The correlation kernel K_n for the MOP ensemble is given by

$$K_n(x, y) = \frac{1}{2\pi i(x - y)} \times$$
$$(0 \quad w_1(y) \quad \cdots \quad w_r(y)) Y_+^{-1}(y) Y_+(x) \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

Proof for case $r = 2$

- **Assume $r = 2$. Let $L_n(x, y)$ be the right-hand side**

$$L_n(x, y) = \frac{1}{2\pi i(x - y)} \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} Y_+^{-1}(y) Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

We show

- (a) $L_n h = h$ if h is a polynomial of degree $\leq n - 1$,
- (b) $L_n h = 0$ if $\int h(y) y^{j-1} w_i(y) dy = 0$ for $j = 1, \dots, n_i$, and $i = 1, 2$.

Proof of (a)

- Let h be a polynomial of degree $\leq n - 1$.

$$\begin{aligned} L_n(x, y)h(y) &= \frac{h(y)}{2\pi i(x - y)} \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} Y_+^{-1}(y) Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \\ &= \frac{h(y) - h(x)}{2\pi i(x - y)} \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} Y_+^{-1}(y) Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \\ &\quad + \frac{h(x)}{2\pi i(x - y)} \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} Y_+^{-1}(y) Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \end{aligned}$$

- $\int L_n(x, y)h(y)dy$ splits into two integrals.

Proof of (a), first integral

- **First integral has**

$$\underbrace{\frac{h(y) - h(x)}{2\pi i(x - y)}}_{\substack{\text{polynomial in } y \\ \text{of degree } \leq n - 2}} \underbrace{(0 \ w_1(y) \ w_2(y)) Y_+^{-1}(y)}_{\substack{\text{vector with linear forms} \\ \text{of type I MOPs}}} Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

- **Integral with respect to y is 0 for every x because of type I multiple orthogonality.**

Proof of (a), second integral

- **Second integral is**

$$\frac{h(x)}{2\pi i} \int_{-\infty}^{\infty} \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} Y_+^{-1}(y) Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \frac{dy}{x-y}$$

- **From jump condition in RH problem**

$$\begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} Y_+^{-1}(y) = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix} (Y_-^{-1}(y) - Y_+^{-1}(y))$$

- **It remains to prove**

$$\frac{1}{2\pi i} \int_{-\infty}^{\infty} \left[\frac{Y_-^{-1}(y) - Y_+^{-1}(y)}{x-y} Y_+(x) \right]_{1,1} dy = 1.$$

Proof of (a), second integral (cont.)

$$\frac{1}{2\pi i} \int_{-\infty}^{\infty} \left[\frac{Y_-^{-1}(y) - Y_+^{-1}(y)}{x - y} Y_+(x) \right]_{1,1} dy = 1.$$

- **Replace $x \in \mathbb{R}$ by z with $\operatorname{Im} z > 0$.**
- $y \mapsto \left[\frac{Y_-^{-1}(y)}{z - y} Y(z) \right]_{1,1}$ **is analytic in lower half plane and is $O(y^{-n-1})$ as $y \rightarrow \infty$. By Cauchy's theorem**

$$\frac{1}{2\pi i} \int_{-\infty}^{\infty} \left[\frac{Y_-^{-1}(y)}{z - y} Y(z) \right]_{1,1} dy = 0$$

- $y \mapsto \left[\frac{Y_+^{-1}(y)}{z - y} Y(z) \right]_{1,1}$ **has pole in upper half plane and same behavior at infinity. By residue calculation**

$$\frac{1}{2\pi i} \int_{-\infty}^{\infty} \left[\frac{Y_+^{-1}(y)}{z - y} Y(z) \right]_{1,1} dy = -1$$

- **Subtract the two results and then let $z \rightarrow x \in \mathbb{R}$.**

Proof of (b)

- **Assume** $\int h(y)y^{j-1}w_i(y)dy = 0$ for $j = 1, \dots, n_j$, $i = 1, 2$. **We have to prove** $L_n h(x) = 0$
- **We have that** $L_n h(x) =$

$$\frac{1}{2\pi i} \int_{-\infty}^{\infty} h(y) \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} \frac{Y_+^{-1}(y) - Y_+^{-1}(x)}{x - y} Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} dy$$
$$+ \frac{1}{2\pi i} \int_{-\infty}^{\infty} h(y) \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} \frac{Y_+^{-1}(x)}{x - y} Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} dy.$$

- **Second integral is obviously zero.**

- **In first integral we can take out** $Y_+(x) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$.

Proof of (b), (cont.)

- We are left to evaluate

$$\frac{1}{2\pi i} \int_{-\infty}^{\infty} h(y) \begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} \frac{Y_+^{-1}(y) - Y_+^{-1}(x)}{x - y} dy$$

- Second row of Y^{-1} has polynomials of degree $\leq n_1$
- Third row of Y^{-1} has polynomials of degree $\leq n_2$
- Hence for every x , the entries of

$$\begin{pmatrix} 0 & w_1(y) & w_2(y) \end{pmatrix} \frac{Y_+^{-1}(y) - Y_+^{-1}(x)}{x - y}$$

take the form

$$w_1(y)(\text{poly of deg } \leq n_1 - 1) + w_2(y)(\text{poly of deg } \leq n_2 - 1)$$

- This is in the linear span of g_1, \dots, g_n and the integral is zero.

Examples of MOP ensembles

- **Non-intersecting Brownian motions**
- **Non-intersecting squared Bessel paths**
- **Random matrix model with external source**
- **Two matrix model**

Non-intersecting Brownian motions

- **Brownian motion** transition probability density

$$p_t(x, y) = \frac{1}{\sqrt{2\pi t}} e^{-\frac{(y-x)^2}{2t}}$$

- Biorthogonal ensemble of **non-intersecting Brownian motions**

$$\frac{1}{Z_n} \det [p_t(a_i, x_j)]_{i,j=1}^n \cdot \det [p_{T-t}(x_i, b_j)]_{i,j=1}^n$$

with Z_n depending on a_1, \dots, a_n and b_1, \dots, b_n .

- In confluent limit where all $a_j \rightarrow a$ both Z_n and the first determinant tend to 0.
- Take limit using L'Hôpital's rule. First determinant becomes

$$\det \left[\frac{\partial^{i-1}}{\partial a^{i-1}} p_t(a, x_j) \right]_{i,j=1}^n .$$

Non-intersecting Brownian motions

- From $p_t(a, x) = \frac{1}{\sqrt{2\pi t}} e^{-\frac{(x-a)^2}{2t}}$ we get

$$\frac{\partial^{i-1}}{\partial a^{i-1}} p_t(a, x) = \left(\begin{array}{l} \text{polynomial in } x \\ \text{of degree } i-1 \end{array} \right) e^{-\frac{1}{2t}(x^2-2ax)}$$

- Apply appropriate row operations to the determinant and take out common factors from each column

$$\det \left[\frac{\partial^{i-1}}{\partial a^{i-1}} p_t(a, x_j) \right]_{i,j=1}^n \propto \det [x_j^{i-1}]_{i,j=1}^n \cdot \prod_{j=1}^n e^{-\frac{1}{2t}(x_j^2-2ax_j)}$$

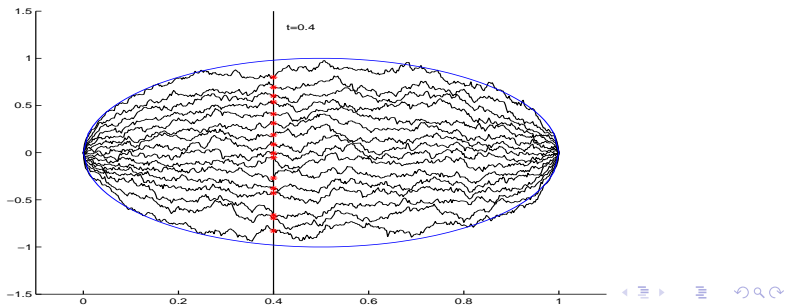
- Similarly when all $b_j \rightarrow b$ we get a second factor

$$\propto \det [x_j^{i-1}]_{i,j=1}^n \cdot \prod_{j=1}^n e^{-\frac{1}{2(\tau-t)}(x_j^2-2bx_j)}$$

Non-intersecting Brownian motions

- In **fully confluent limit** all $a_j \rightarrow a$, all $b_j \rightarrow b$, we find an OP ensemble with quadratic potential (= GUE)

$$\frac{1}{Z_n} \left(\det [x_j^{i-1}]_{i,j=1}^n \prod_{j=1}^n e^{-\frac{1}{2t}(x_j^2 - 2ax_j)} \right) \left(\det [x_j^{i-1}]_{i,j=1}^n \prod_{j=1}^n e^{-\frac{1}{2(T-t)}(x_j^2 - 2bx_j)} \right)$$
$$= \frac{1}{Z_n} \Delta(x)^2 \prod_{j=1}^n e^{-V(x_j)}, \quad V(x) = \frac{T}{t(T-t)} \left(\frac{x^2}{2} - \left((1 - \frac{t}{T})a + \frac{t}{T}b \right)x \right)$$



Non-intersecting Brownian motions

- If n_1 of the b_j 's tend to b_1 and the remaining n_2 tend to b_2 , then we have to treat the first n_1 rows separately from the last n_2 rows in taking the confluent limit.
- The second determinant now becomes

$$\det [g_i(x_j)]_{i,j=1}^n$$

with functions

$$\begin{aligned} g_i(x) &= x^{i-1} e^{-\frac{1}{2(T-t)}(x^2 - 2b_1x)}, & i &= 1, \dots, n_1, \\ g_{n_1+i}(x) &= x^{i-1} e^{-\frac{1}{2(T-t)}(x^2 - 2b_2x)}, & i &= 1, \dots, n_2. \end{aligned}$$

- Together with $\Delta(x) \prod_{j=1}^n e^{-\frac{1}{2t}(x_j^2 - 2ax_j)}$ we now find a **MOP ensemble** with two weights and $\vec{n} = (n_1, n_2)$.

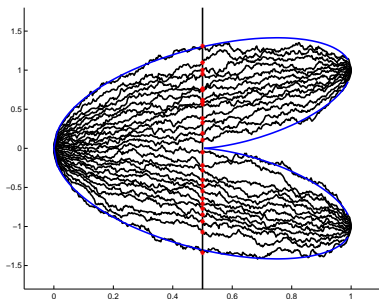
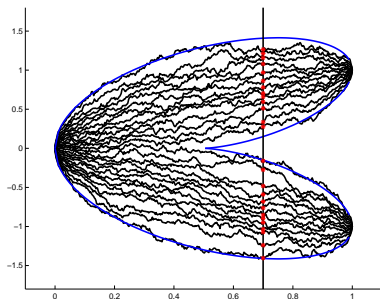
Non-intersecting Brownian motions

- Two Gaussian weights

$$w_i(x) = e^{-V_i(x)}, \quad V_i(x) = \frac{T}{t(T-t)} \left(\frac{x^2}{2} - c_i x \right),$$

where $c_i = (1 - \frac{t}{T})a + \frac{t}{T}b_i$ for $i = 1, 2$.

- Associated MOPs are **multiple Hermite polynomials**



Non-intersecting squared Bessel paths

- Squared Bessel processes is diffusion process on $[0, \infty)$ with transition probability density

$$p_t(x, y) = \frac{1}{2t} \left(\frac{y}{x}\right)^{\alpha/2} e^{-\frac{1}{2t}(x+y)} I_\alpha \left(\frac{\sqrt{xy}}{t}\right), \quad x, y > 0,$$

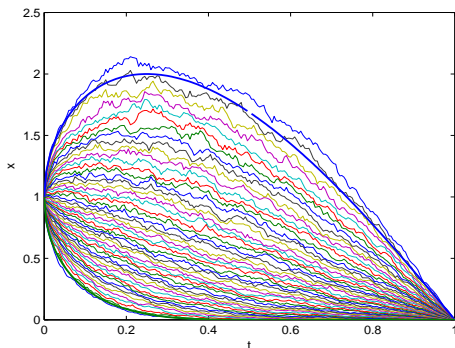
- I_α is the **modified Bessel function** of order $\alpha > -1$.
- In confluent limit all $a_j \rightarrow a$, all $b_j \rightarrow 0$, this leads to a **MOP ensemble** with two weights

$$w_1(x) = x^{\alpha/2} e^{-\frac{Tx}{2t(T-t)}} I_\alpha \left(\frac{\sqrt{ax}}{t}\right)$$

$$w_2(x) = x^{(\alpha+1)/2} e^{-\frac{Tx}{2t(T-t)}} I_{\alpha+1} \left(\frac{\sqrt{ax}}{t}\right)$$

and $n_1 = \lceil n/2 \rceil$, $n_2 = \lfloor n/2 \rfloor$

Non-intersecting squared Bessel paths



- For $a \rightarrow 0$ this further reduces to a **Laguerre unitary ensemble (LUE)**

Random matrix model with external source

- Hermitian matrix model with external source

$$\frac{1}{Z_n} e^{-\text{Tr}(V(M)-AM)} dM$$

- **External source** A is a given Hermitian $n \times n$ matrix
- Joint p.d.f. for eigenvalues

$$P(x_1, \dots, x_n) \propto \Delta(x)^2 \prod_{j=1}^n e^{-V(x_j)} \int_{U(n)} e^{\text{Tr} A U X U^{-1}} dU$$

where $A = \text{diag}(a_1, \dots, a_n)$, $X = \text{diag}(x_1, \dots, x_n)$.

- The integral over the unitary group can be done by the **Harish-Chandra / Itzykson-Zuber** formula.

Random matrix model with external source

- If all a_i and all x_j are **distinct** then

$$\int_{U(n)} e^{\text{Tr} A U X U^{-1}} dU \propto \frac{\det [e^{a_i x_j}]_{i,j=1}^n}{\Delta(a)\Delta(x)}$$

- P.d.f. for eigenvalues

$$\propto \Delta(x) \prod_{j=1}^n e^{-V(x_j)} \frac{\det [e^{a_i x_j}]_{i,j=1}^n}{\Delta(a)}$$

- If some a_i 's coincide, we take the confluent limit.
- If n_1 of the a_j 's tend to c_1 and $n_2 = n - n_1$ to c_2 , then we find **MOP ensemble** with two weights

$$w_1(x) = e^{-(V(x)-c_1x)}, \quad w_2(x) = e^{-(V(x)-c_2x)}.$$

Random matrix model with external source

- **MOP ensemble** with weights

$$w_1(x) = e^{-(V(x)-c_1x)}, \quad w_2(x) = e^{-(V(x)-c_2x)}$$

and $\vec{n} = (n_1, n_2)$.

- In Gaussian case $V(x) = \frac{1}{2}x^2$, the eigenvalues in the external source model have the same joint distribution has the positions of **non-intersecting Brownian motions** with one starting position and two ending positions.
- If V is non-Gaussian then we have something else.

Two matrix model

- The Hermitian **two matrix model**

$$\frac{1}{Z_n} e^{-\text{Tr}(V(M_1)+W(M_2)-\tau M_1 M_2)} dM_1 dM_2$$

is a probability measure on pairs (M_1, M_2) of $n \times n$ Hermitian matrices.

- V and W are polynomial potentials
- $\tau \neq 0$ is a coupling constant

Determinantal point process

- **Explicit formula for joint p.d.f. of the eigenvalues of M_1 and M_2**

$$\frac{1}{(n!)^2} \det \begin{pmatrix} K_{11}(x_i, x_j) & K_{12}(x_i, y_j) \\ K_{21}(y_i, x_j) & K_{22}(y_i, y_j) \end{pmatrix}$$

with 4 kernels that are expressed in terms of **biorthogonal polynomials**

- **Two sequences $(p_j)_j$ and $(q_k)_k$ of monic polynomials that satisfy if $j \neq k$,**

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p_j(x) q_k(y) e^{-(V(x)+W(y)-\tau xy)} dx dy = h_k^2 \delta_{j,k}.$$

Mehta-Shukla (1994), Eynard-Mehta (1998)

Ercolani-McLaughlin (2001)

Bertola-Eynard-Harnad (2002-04)

- The kernels are expressed in terms of these biorthogonal polynomials and **transformed functions**

$$Q_j(x) = \int_{-\infty}^{\infty} q_j(y) e^{-(V(x)+W(y)-\tau xy)} dy,$$
$$P_k(y) = \int_{-\infty}^{\infty} p_k(x) e^{-(V(x)+W(y)-\tau xy)} dx,$$

as follows:

$$K_{11}(x_1, x_2) = \sum_{k=0}^{n-1} \frac{1}{h_k^2} p_k(x_1) Q_k(x_2), \quad K_{12}(x, y) = \sum_{k=0}^{n-1} \frac{1}{h_k^2} p_k(x) q_k(y),$$
$$K_{21}(y, x) = \sum_{k=0}^{n-1} \frac{1}{h_k^2} P_k(y) Q_k(x) - e^{-(V(x)+W(y)-\tau xy)}, \quad K_{22}(y_1, y_2) = \sum_{k=0}^{n-1} \frac{1}{h_k^2} P_k(y_1) q_k(y_2)$$

Biorthogonality

- **Biorthogonality condition for p_n**

$$\int_{-\infty}^{\infty} p_n(x) Q_k(x) dx = 0 \quad \text{for } k = 0, 1, \dots, n-1$$

where $Q_k(x) = e^{-V(x)} \int_{-\infty}^{\infty} q_k(y) e^{-(W(y)-\tau xy)} dy$.

- **Equivalently, we may replace $q_k(y)$ by y^{k-1}**

$$w_k(x) = e^{-V(x)} \int_{-\infty}^{\infty} y^{k-1} e^{-(W(y)-\tau xy)} dy,$$

and $\int_{-\infty}^{\infty} p_n(x) w_k(x) dx = 0$ for $k = 1, \dots, n$.

- **We integrate by parts if $k \geq \deg W$.**

Biorthogonality

- Calculation for $W(y) = \frac{1}{4}y^4$, $k \geq 4$.

$$\begin{aligned}w_k(x) &= e^{-V(x)} \int_{-\infty}^{\infty} y^{k-1} e^{-\left(\frac{1}{4}y^4 - \tau xy\right)} dy \\&= -e^{-V(x)} \int_{-\infty}^{\infty} y^{k-4} e^{\tau xy} d\left(e^{-\frac{1}{4}y^4}\right) \\&= e^{-V(x)} \int_{-\infty}^{\infty} \left((k-4)y^{k-5} + \tau xy^{k-4}\right) e^{-\left(\frac{1}{4}y^4 - \tau xy\right)} dy \\&= (k-4)w_{k-4}(x) + \tau x w_{k-3}(x).\end{aligned}$$

- This leads to **multiple orthogonality**

Multiple orthogonality

Proposition (K-McLaughlin (2005))

Suppose $\deg W = r + 1$. Then the biorthogonal polynomial p_n is a **multiple orthogonal polynomial** with r weights w_1, \dots, w_r , and near-diagonal multi-index (n_1, \dots, n_r) .

If n is a multiple of r then $n_j = \frac{n}{r}$ for every j .

- The eigenvalues of M_1 , when averaged over M_2 , are a **MOP ensemble** with r weights.
- There is a RH problem of size $(r + 1) \times (r + 1)$.
- Asymptotic analysis of this RH problem was done for $W(y) = \frac{1}{4}y^4$ by **Duits-K (2009)** and for $W(y) = \frac{1}{4}y^4 + \frac{\alpha}{2}y^2$ by **Duits-K-Mo (2012)**