

Infinite-dimensional stochastic differential equations related to random matrices

- Ginibre RPF, Sine RPF, Bessel RPF, Airy RPF
- General theory for ISDEs:
 - quasi-Gibbs property & log derivative
- Ginibre Interacting Brownian motions
- Palm measures of Ginibre RPF
- Homogenization of diffusion in 2D Coulomb environment

Let $S = \mathbb{R}^d, \mathbb{C}, [0, \infty)$.

S : Configuration space over S

$$S = \left\{ s = \sum_i \delta_{s_i}; s_i \in S, s(|s| < r) < \infty (\forall r \in \mathbb{N}) \right\}$$

μ : RPF over S . i.e. prob meas. on S .

Prob: (1) To construct a *natural* stochastic dynamics

$$\mathbf{X}_t = (X_t^i)_{i \in \mathbb{N}} \quad (\text{labeled dynamics})$$

related to μ , i.e.

$$X_t = \sum_{i \in \mathbb{N}} \delta_{X_t^i} \quad (\text{unlabeled dynamics})$$

is reversible w.r.t. μ .

(2) To find the ∞ -dim. SDE that \mathbf{X}_t satisfies.

- ρ^n is called the n -correlation function of μ w.r.t. Radon m. m if

$$\int_{A_1^{k_1} \times \dots \times A_m^{k_m}} \rho^n(\mathbf{x}_n) \prod_{i=1}^n m(dx_i) = \int_S \prod_{i=1}^m \frac{s(A_i)!}{(s(A_i) - k_i)!} d\mu$$

for any disjoint $A_i \in \mathcal{B}(S)$, $k_i \in \mathbb{N}$ s.t. $k_1 + \dots + k_m = n$.

- μ is called the determinantal RPF generated by (K, m) if its n -correlation fun. ρ^n is given by

$$\rho^n(\mathbf{x}_n) = \det[K(x_i, x_j)]_{1 \leq i, j \leq n}$$

- **Ginibre RPF** $S = \mathbb{C}$. μ_{gin} is generated by (K_{gin}, g)

$$K_{\text{gin}}(x, y) = e^{x\bar{y}} \quad g(dx) = \pi^{-1} e^{-|x|^2} dx$$

Property of Ginibre RPF

(g1) μ_{gin} is translation and rotation invariant

(g2) μ_{gin} is the weak limit of μ_{gin}^N :

the labeled expression $\tilde{\mu}_{\text{gin}}^N$ of μ_{gin}^N is

$$\tilde{\mu}_{\text{gin}}^N = \frac{1}{Z} \prod_{i < j}^N |x_i - x_j|^2 \prod_{k=1}^N g(dx_k) \quad (1)$$

μ_{gin}^N is the determinantal RPF gen. by (K_{gin}^N, g) , where

$$K_{\text{gin}}^N(x, y) = \sum_{i=0}^{N-1} \frac{(x\bar{y})^i}{i!}$$

Non rigorous expression of μ_{gin} as a measure μ_{gin}^- on $\mathbb{C}^{\mathbb{N}}$:

From (g2)

$$\mu_{\text{gin}}^- = \frac{1}{Z} \prod_{i < j}^{\infty} |x_i - x_j|^2 \prod_{k=1}^{\infty} \frac{e^{-|x_k|^2}}{\pi} dx_k \quad (2)$$

From the translation invariance we have another informal expression:

$$\mu_{\text{gin}}^- = \frac{1}{Z} \prod_{i < j}^{\infty} |x_i - x_j|^2 \prod_{k=1}^{\infty} dx_k \quad (3)$$

Which representations are correct?

Non rigorous expression of μ_{gin} as a measure μ_{gin}^- on $\mathbb{C}^{\mathbb{N}}$:

From (g2)

$$\mu_{\text{gin}}^- = \frac{1}{Z} \lim_{r \rightarrow \infty} \prod_{i < j, |x_i|, |x_j| < r}^{\infty} |x_i - x_j|^2 \prod_{k=1}^{\infty} \frac{e^{-|x_k|^2}}{\pi} dx_k \quad (4)$$

From the translation invariance we have another informal expression:

$$\mu_{\text{gin}}^- = \frac{1}{Z} \lim_{r \rightarrow \infty} \prod_{i < j, |x_i - x_j| < r}^{\infty} |x_i - x_j|^2 \prod_{k=1}^{\infty} dx_k \quad (5)$$

Which representations are correct?

Both

Log derivative

- Let μ^1 be the 1-Campbell measure on $\mathbb{R}^d \times S$:

$$\mu^1(A \times B) = \int_A \rho^1(x) \mu_x(B) dx$$

Here $\mu_x(\cdot) = \mu(\cdot - \delta_x | s(x) \geq 1)$ is the Palm m. at x .

- $d_\mu \in L^1(\mathbb{R}^d \times S, \mu^1)$ is called the log derivative of μ if

$$\int_{\mathbb{R}^d \times S} \nabla_x f d\mu^1 = - \int_{\mathbb{R}^d \times S} f d_\mu d\mu^1 \quad \forall f \in C_0^\infty(\mathbb{R}^d) \otimes \mathcal{D}$$

Here ∇_x is the nabla on \mathbb{R}^d , \mathcal{D} is the space of local smooth functions on S with compact support.

- Very informally

$$d_\mu = \nabla_x \log \mu^1$$

- Ginibre RPF: $d_{\mu_{\text{gin}}}$ has plural representations

$$d_{\mu_{\text{gin}}}(x, y) = -2x + 2 \lim_{r \rightarrow \infty} \sum_{|y_i| < r} \frac{x - y_i}{|x - y_i|^2} \quad \text{in } L_{\text{loc}}^2(\mu^1)$$

$$d_{\mu_{\text{gin}}}(x, y) = 2 \lim_{r \rightarrow \infty} \sum_{|x - y_i| < r} \frac{x - y_i}{|x - y_i|^2} \quad \text{in } L_{\text{loc}}^2(\mu^1)$$

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- These correspond to the following:

$$\mu_{\text{gin}}^- = \frac{1}{Z} \lim_{r \rightarrow \infty} \prod_{i < j, |x_i|, |x_j| < r}^{\infty} |x_i - x_j|^2 \prod_{k=1}^{\infty} \frac{e^{-|x_k|^2}}{\pi} dx_k \quad (4)$$

$$\mu_{\text{gin}}^- = \frac{1}{Z} \lim_{r \rightarrow \infty} \prod_{i < j, |x_i - x_j| < r}^{\infty} |x_i - x_j|^2 \prod_{k=1}^{\infty} dx_k \quad (5)$$

- (A1) ρ^k are locally bounded for all $k \in \mathbb{N}$
- (A2) The **log derivative** $d_\mu \in L_{loc}^1(\mu^1)$ exists
- (A3) μ is a quasi-Gibbs measure
- (A4) $\{X_t^i\}$ do not collide each other (**non-collision**)
- (A5) each tagged particle X_t^i never explode (**non-explosion**)

Let $u: S^{\mathbb{N}} \rightarrow S$ such that $u((s_i)) = \sum_i \delta_{s_i}$.

Thm 1. Assume (A1)–(A5). Then $\exists S_0 \subset S$ such that

$$\mu(S_0) = 1, \quad (6)$$

and that, for $\forall s \in u^{-1}(S_0)$, $\exists u^{-1}(S_0)$ -valued pr. $(X_t^i)_{i \in \mathbb{N}}$ and $\exists S^{\mathbb{N}}$ -valued Brownian $m. (B_t^i)_{i \in \mathbb{N}}$ satisfying

$$dX_t^i = dB_t^i + \frac{1}{2} d_\mu(X_t^i, \sum_{j \neq i} \delta_{X_t^j}) dt, \quad (X_0^i)_{i \in \mathbb{N}} = s \quad (7)$$

How to relate the SDE with the equilibrium state μ .

$$dX_t^i = dB_t^i + \frac{1}{2}d\mu(X_t^i, \sum_{j \neq i} \delta_{X_t^j})dt, \quad (X_0^i)_{i \in \mathbb{N}} = \mathbf{s}$$

Thm 2 (O. (JMSJ 09)). (1) *The family of processes $\{(X_t^i)_{i \in \mathbb{N}}\}$ is a diffusion with state space $u^{-1}(S_0)$.*
(2) *The associated unlabeled process $\{X_t\}$ is a diffusion with state space S_0 . Here $X_t = \sum_{i \in \mathbb{N}} \delta_{X_t^i}$.*
(3) *$\{X_t\}$ is reversible w.r.t. μ .*

Remark 1. (1) (A1)–(A5) can be checked for Ginibre RPF ($\beta = 2$), Sine RPFs, Airy RPFs and Bessel RPFs ($\beta = 1, 2, 4$).

(2) We can calculate the log derivatives of these measures.

(3) We have general theorems for quasi-Gibbs property and the log derivatives (O. PTRF, AOP). The statements are too messy to be omitted here.

Examples: By Theorem 1 and 2 we have the following:

Ginibre RPF: When $\mu = \mu_{\text{gin}}$,

$$dX_t^i = dB_t^i + \lim_{r \rightarrow \infty} \sum_{\substack{|X_t^i - X_t^j| < r, \\ j \neq i}} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} dt. \quad (8)$$

and also

$$dX_t^i = dB_t^i - X_t^i dt + \lim_{r \rightarrow \infty} \sum_{\substack{|X_t^j| < r \\ j \neq i}} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} dt. \quad (9)$$

Sine $_{\beta}$ RPF: $S = R$, $\beta = 1, 2, 4$

$$dX_t^i = dB_t^i + \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|X_t^j| < r, j \neq i} \frac{1}{X_t^i - X_t^j} dt$$

Since $d = 1$, we have

$$dX_t^i = dB_t^i + \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|X_t^i - X_t^j| < r, j \neq i} \frac{1}{X_t^i - X_t^j} dt$$

Spohn (1987) considered the case $\beta = 2$:

$$dX_t^i = dB_t^i + \sum_{j \neq i} \frac{1}{X_t^i - X_t^j} dt$$

He constructed the dynamics as a Markov semigroup by Dirichlet form.

Bessel RPF (with Honda):

$S = [0, \infty)$, $\beta = 1, 2, 4$, $a > 1$

$$dX_t^i = dB_t^i + \frac{a}{2X_t^i}dt + \lim_{r \rightarrow \infty} \frac{\beta}{2} \sum_{\substack{|X_t^j| < r \\ j \neq i}} \frac{1}{X_t^i - X_t^j} dt$$

Airy RPF is more complicated.

(Joint work with Tanemura)

quasi-Gibbs m.: Ψ : Ruelle class interaction potential,
 $Q_r = \{|x| \leq r\}$, $\pi_r(s) = s(\cdot \cap Q_r)$, $\pi_r^c(s) = s(\cdot \cap Q_r^c)$

$$\mu_{r,\xi}^m(\cdot) = \mu(\pi_r \in \cdot | s(Q_r) = m, \pi_r^c(s) = \pi_r^c(\xi))$$

- μ is called (Φ, Ψ) -Gibbs m. if it satisfies **DLR eq**:

$$d\mu_{r,\xi}^m = \frac{1}{Z_{r,\xi}} e^{-\mathcal{H}_r - \mathcal{W}_{r,\xi}} \prod_{k=1}^m e^{-\Phi(s_k)} ds_k$$

$$\mathcal{H}_r(s) = \sum_{\substack{s_i, s_j \in Q_r, \\ i < j}} \Psi(s_i - s_j), \quad \mathcal{W}_{r,\xi} = \sum_{\substack{s_i \in Q_r, \\ \xi_j \in Q_r^c}} \Psi(s_i - \xi_j)$$

$$\mu = \frac{1}{Z} e^{-\sum_{i < j}^{\infty} \Psi(x_i - x_j)} \prod_{k=1}^{\infty} e^{-\Phi(x_k)} dx_k \quad (\text{informally})$$

- $\Phi = \Psi = 0$: Poisson rpf: $\Lambda = \frac{1}{Z} \prod_{i \in \mathbb{N}} dx_i$.
 - Ginibre RPF: $\Phi = 0$ $\Psi(x) = -2 \log |x|$
- $\mathcal{W}_{r,\xi}$ diverge, so DLR does not make sense

Gibbs m. Let $\nu_r^m = \prod_{k=1}^m 1_{Q_r}(s_k) e^{-\Phi(s_k)} ds_k$

$$d\mu_{r,\xi}^m = \frac{1}{z_{r,\xi}^m} e^{-\mathcal{H}_r - \mathcal{W}_{r,\xi}} d\nu_r^m \quad (\text{DLR eq})$$

quasi-Gibbs m. $\exists c_{r,\xi}^m$

$$\frac{1}{c_{r,\xi}^m} e^{-\mathcal{H}_r} d\nu_r^m \leq \mu_{r,\xi}^m \leq c_{r,\xi}^m e^{-\mathcal{H}_r} d\nu_r^m$$

- If μ is Ginibre RPF, $\mathcal{W}_{r,\xi}$ and $z_{r,\xi}^m$ diverge. But $e^{-\mathcal{W}_{r,\xi}}/z_{r,\xi}^m$ conv.

$$\frac{1}{c_{r,\xi}^m} \leq \frac{e^{-\mathcal{W}_{r,\xi}}}{z_{r,\xi}^m} \leq c_{r,\xi}^m$$

- Quasi-Gibbs is very mild restriction. If μ is (Φ, Ψ) -quasi-Gibbs m, then μ is also $(\Phi + f, \Psi)$ -quasi Gibbs m for any loc bdd m'able f .

Unlabeled level construction Let \mathbb{D} be the canonical square field on S : $s = \sum_i \delta_{s_i}$, $\mathbf{s} = (s_i)$.

$$\mathbb{D}[f, g](\mathbf{s}) = \frac{1}{2} \sum_i \nabla_{s_i} \tilde{f}(\mathbf{s}) \cdot \nabla_{s_i} \tilde{g}(\mathbf{s})$$

Let \mathcal{D} be the set of local smooth fun with cpt support.

$$\mathcal{E}^\mu(f, g) = \int_S \mathbb{D}[f, g] d\mu$$

Thm 3. (1) *If μ is quasi-Gibbs, then $(\mathcal{E}^\mu, \mathcal{D})$ is closable on $L^2(S, \mu)$.*
 (2) *If $(\mathcal{E}^\mu, \mathcal{D})$ is closable on $L^2(S, \mu)$ and (A.1) is satisfied, then there exists diffusion X_t associated with the closure of $(\mathcal{E}^\mu, \mathcal{D})$ on $L^2(S, \mu)$.*

If μ is Poisson rpf with Lebesgue intensity, then $X_t = \sum_i \delta_{B_t^i}$.

Thm 4. *Ginibre RPF ($\beta = 2$), Sine RPFs, Airy RPFs and Bessel RPFs ($\beta = 1, 2, 4$) are quasi-Gibbs m. for $\Psi(x) = -\beta \log |x|$.*

- The key point of the proof is to use the **small fluctuation property** (SFP) of linear statistics for these measures.
- SFP was established by Soshnikov (Sine, Airy, Bessel RPFs), Shirai (Ginibre RPF).
- Proof consists of several parts:
 - (1) To find a good finite particle approximation $\{\mu^N\}$
 - (2) To prove uniform *small fluctuation* of $\{\mu^N\}$
 - (3) To prove uni bounds of 1 & 2 cor funs of $\{\mu^N\}$
 - (4) To carry out the limiting procedure of d_{μ^N} & quasi-Gibbs property. (General theorems to appear in O. PTRF, AOP)

Related problems:

- Yoo proved that Determinantal RPF with

$$\text{Spec}(K) \subset [0, 1)$$

are *Gibbs* measures. So it is likely all Determinantal RPF are quasi Gibbs measures, i.e., under the condition

$$\text{Spec}(K) \subset [0, 1]$$

To strength Yoo's result like this is important because RPFs in infinite volume appeared in RMT usually satisfy that

$$\text{Spec}(K) = \{0, 1\}$$

- To calculate the log derivative of Determinantal RPFs.

- β ensemble of Sine, Bessel, Airy for general $\beta > 0$:

(Valkó, B.-Világ, B., Ramírez, J.-Rider, B.-Világ, B.)

Good finite approximations are clear: Log gasses.

The problem is to control correlation functions and to prove small fluctuations.

- The spectrum of Gaussian Analytic functions

(Some progress done by Shirai)

- In particular, GAF with Bergmann Kernel

Geometric property of Ginibre RPF

Rider, Goldman, Kostlan, Shirai

Palm meas. For $\mathbf{x} = \{x_1, \dots, x_m\} \subset S^m$ set

$$\mu_{\mathbf{x}} := \mu(\cdot - \sum_{l=1}^m \delta_{x_l} \mid s(\{x_l\}) \geq 1(\forall l))$$

Thm 5 (with Shirai). Let $m, n \in \{0\} \cup \mathbb{N}$. Then

(1) If $m = n$, then $\mu_{\mathbf{x}} \sim \mu_{\mathbf{y}}$. (\sim means ab. cont.)

(2) If $m \neq n$, then $\mu_{\mathbf{x}}$ and $\mu_{\mathbf{y}}$ are singular each other.

Remark: • In case of Gibbs measures, it holds always

$$\mu_{\mathbf{x}} \prec \mu$$

• In this sense Ginibre RPF is similar to periodic RPF.

Thm 6 (with Shirai). Suppose $m = n$. Then for μ_y -a.s.

$$s = \sum_i \delta_{s_i},$$

$$\frac{d\mu_x}{d\mu_y} = \frac{\Delta^m(\mathbf{x}) \det[K_{\text{gin}}(x_i, x_j)]_{i,j=1}^m}{\Delta^m(\mathbf{y}) \det[K_{\text{gin}}(y_i, y_j)]_{i,j=1}^m} \lim_{r \rightarrow \infty} \prod_{|s_i| < b_r} \frac{|x - s_i|^2}{|y - s_i|^2}$$

cpt uni in $\mathbf{x} \in \mathbb{C}^m$.

- $\{b_r\}_{r \in \mathbb{N}}$: $\lim b_r = \infty$
- $|x - s_i| = \prod_{m=1}^m |x_m - s_i|$ for $\mathbf{x} = (x_1, \dots, x_m)$
- $\Delta^m(\mathbf{x}) = \prod_{i < j}^m |x_i - x_j|^2$ if $m \geq 2$, $\Delta^m(\mathbf{x}) = 1$ if $m = 1$.

In particular, if $m = 1$, then

$$\frac{d\mu_x}{d\mu_y} = \frac{e^{-|x|^2}}{e^{-|y|^2}} \lim_{r \rightarrow \infty} \prod_{|s_i| < b_r} \frac{|x - s_i|^2}{|y - s_i|^2}$$

Index of the number of missing particles:

$$D_q = \{z \in \mathbb{C}; |z| < \sqrt{q}\} \quad q \in \mathbb{N}$$

$$F_r(s) = \frac{1}{r} \sum_{q=1}^r (s(D_q) - q). \quad (10)$$

Thm 7. Let S be the configuration space over \mathbb{C} .

Let $m \in \mathbb{N}$. Then for $\mathbf{x} = (x_1, \dots, x_m)$

$$\lim_{r \rightarrow \infty} F_r(s) = -m \quad \text{weakly in } L^2(S, \mu_{\mathbf{x}}) \quad (11)$$

Remark: m is the number of the removed particles.

$$\infty - m \neq \infty$$

Application of stochastic geo. to stochastic dyn.:
homogenization of diffusion in 2D Coulomb environment

Let $s = \sum_i \delta_{s_i} \in S$. Let $X_t^s \in \mathbb{R}^2$ be the solution of

$$dX_t^s = dB_t + \lim_{q \rightarrow \infty} \sum_{i \in \mathbb{N}, |X_t^s - s_i| < q} \frac{X_t^s - s_i}{|X_t^s - s_i|^2} dt$$

Then

$$\lim_{\varepsilon \rightarrow \infty} \varepsilon X_{t/\varepsilon^2}^s = \sqrt{\alpha} B_t \quad \text{in } \mu_{\text{gin},0}\text{-measure} \quad ([O.98])$$

The matrix α is called the effective conductivity.

Thm 8. $\alpha = 0$

The key of the proof is the function F_r in (10):

$$F_r(s) = \frac{1}{r} \sum_{q=1}^r (s(D_q) - q)$$

Representation of α Let $\tilde{\mathcal{E}} : L^2(\mu_{\text{gin},0}) \otimes L^2(\mu_{\text{gin},0}) \rightarrow \mathbb{R}$:

$$\tilde{\mathcal{E}}(\mathbf{f}, \mathbf{g}) = \int_S \frac{1}{2} \sum_{i=1}^2 f_i g_i d\mu_{\text{gin},0} \quad \text{for } \mathbf{f} = (f_1, f_2)$$

$$\tilde{\mathcal{D}} = \overline{\{(D_1 f, D_2 f); f \in \mathcal{D}_0\}}$$

Here D_i is the generator of the translation, e_i unit vector.

There's a unique $\mathbf{u}_i \in \tilde{\mathcal{D}}$ s.t.

$$\tilde{\mathcal{E}}(\mathbf{u}_i, \mathbf{g}) = \tilde{\mathcal{E}}(\mathbf{e}_i, \mathbf{g}) \text{ for all } \mathbf{g} \in \tilde{\mathcal{D}} \quad (12)$$

Thm 9 (O.98). *The effective-diffusion matrix α is given by*

$$\alpha_{ij} = \tilde{\mathcal{E}}(\mathbf{e}_i - \mathbf{u}_i, \mathbf{e}_j - \mathbf{u}_j) \quad (13)$$

Moreover, $\alpha = 0$ if and only if $\mathbf{e}_i \in \tilde{\mathcal{D}}$ for all i .

- We need to prove $\mathbf{e}_1, \mathbf{e}_2 \in \tilde{\mathcal{D}}$
- One can check $(F_r, 0) \in \tilde{\mathcal{D}}$. Hence

$$\mathbf{e}_1 = \lim_{r \rightarrow \infty} (-F_r, 0) \quad \text{weakly in } \tilde{\mathcal{D}}$$

This completes the proof of Thm 8.

Conj: If we replace $\mu_{\text{gin},0}$ by μ_{gin} , then $\alpha > 0$. Indeed, in case of periodic μ , this is the case.