

Infinite-dimensional stochastic differential equations related to random matrices

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Motivation

- Soft edge scaling limit and Airy RPFs
- Coulomb stochastic dyn in infinite dimensions

General Theory

- Quasi-Gibbs meas. Log derivative
- Tail triviality • Palm singularity
- Existence and uniqueness of strong solutions of ISDEs
- Uniqueness of Dirichlet forms

Applications & examples

- Application to interacting Brownian motions (IBMs)
- Examples: Sine, Bessel, Airy, Ginibre RPFs
- Homogenization of diffusions in Coulomb environment
- Phase transition conjecture of Ginibre IBMs

Motivation 1

First motivation:
Dynamical soft edge scaling limit
of
Gaussian ensembles

Motivation 1

- The dist of eigen values of the G(O/U/S)E Random Matrices are given by ($\beta = 1, 2, 4$)

$$m_{\beta}^N(d\mathbf{x}_N) = \frac{1}{Z} \prod_{i < j}^N |x_i - x_j|^{\beta} e^{-\frac{\beta}{4} \sum_{i=1}^N |x_i|^2} d\mathbf{x}_N, \quad (1)$$

- The distribution of

$$N^{-1} \sum_{i=1}^N \delta_{x_i/\sqrt{N}} \quad \text{under } m_{\beta}^N$$

converges the semi-circle law

$$\varsigma(x)dx = \frac{1}{2\pi} \sqrt{4 - x^2} dx \quad (2)$$

Sine rpf (Dyson's model)–Bulk scaling limit

$$m_{\beta}^N(dx_N) = \frac{1}{Z} \prod_{i<j}^N |x_i - x_j|^{\beta} e^{-\frac{\beta}{4} \sum_{i=1}^N |x_i|^2} dx_N, \quad \varsigma(x)dx = \frac{1}{2\pi} \sqrt{4 - x^2} dx$$

- Take $x_i = s_i/\sqrt{N}$ in (1) and set

$$\mu_{\text{sin},\beta}^N(ds_N) = \frac{1}{Z} \sum_{i<j}^N |s_i - s_j|^{\beta} \prod_{k=1}^N e^{-\beta |s_k|^2/4N} ds_N \quad (3)$$

- The associated N particle system is given by the SDE: 30p

$$dX_t^i = dB_t^i + \frac{\beta}{2} \sum_{j \neq i}^N \frac{1}{X_t^i - X_t^j} dt - \frac{\beta}{4N} X_t^i dt \quad (4)$$

- So the ass ∞ particle system is given by

$$dX_t^i = dB_t^i + \frac{\beta}{2} \sum_{j \neq i}^{\infty} \frac{1}{X_t^i - X_t^j} dt$$

Airy rpf – Soft edge scaling limit

Airy rpf: $\mu_{\text{Ai},\beta}$ ($S = \mathbb{R}$, $\beta = 1, 2, 4$)

Take the scaling $x_i \mapsto 2\sqrt{N} + s_i N^{-1/6}$ in

$$m_{\beta}^N(d\mathbf{x}_N) = \frac{1}{Z} \prod_{i < j}^N |x_i - x_j|^{\beta} e^{-\frac{\beta}{4} \sum_{i=1}^N |x_i|^2} d\mathbf{x}_N$$

and set

$$\mu_{\text{Ai},\beta}^N(ds_N) = \frac{1}{Z} \prod_{i < j}^N |s_i - s_j|^{\beta} e^{-\frac{\beta}{4} \sum_{i=1}^N |2\sqrt{N} + N^{-1/6}s_i|^2} ds_N.$$

Then $\mu_{\text{Ai},\beta}$ is the TDL of $\mu_{\text{Ai},\beta}^N$:

$$\lim_{N \rightarrow \infty} \mu_{\text{Ai},\beta}^N = \mu_{\text{Ai},\beta}$$

Airy rpf – Soft edge scaling limit

- $\beta = 2 \Rightarrow \mu_{\text{Ai},\beta}$ is the det rpf gen by (K_{Ai}, dx) :

$$K_{\text{Ai}}(x, y) = \frac{\text{Ai}(x)\text{Ai}'(y) - \text{Ai}'(x)\text{Ai}(y)}{x - y}$$

Here $\text{Ai}(\cdot)$ the Airy function such that

$$\text{Ai}(z) = \frac{1}{2\pi} \int_{\gamma} dk e^{i(zk + k^3/3)}, \quad z \in \mathbb{C}. \quad (5)$$

If $\beta = 1, 4$, the correlation func of $\mu_{\text{Ai},\beta}$ are given by similar formula of quaternion determinant.

Airy rpf – Dynamical soft edge scaling limit

- From

$$\mu_{\text{Ai},\beta}^N(ds_N) = \frac{1}{Z} \prod_{i<j} |s_i - s_j|^\beta e^{-\frac{\beta}{4} \sum_{i=1}^N |2\sqrt{N} + N^{-1/6}s_i|^2} ds_N$$

we deduce the SDE of the N particle system:

$$dX_t^i = dB_t^i + \frac{\beta}{2} \sum_{j=1, j \neq i}^N \frac{1}{X_t^i - X_t^j} dt - \frac{\beta}{2} \left\{ N^{1/3} + \frac{1}{2N^{1/3}} X_t^i \right\} dt$$

- **Problem:** What is the limit SDE?

Does $\lim_{N \rightarrow \infty} \left\{ \sum_{j=1, j \neq i}^N \frac{1}{X_t^i - X_t^j} - N^{1/3} \right\}$ converge ?

How to solve the limit SDE?

Airy rpf – Dynamical soft edge scaling limit

Thm 1 (with Tanemura). Let $\beta = 1, 2, 4$. Then:

- the limit ISDE is

$$dX_t^i = dB_t^i + \frac{\beta}{2} \lim_{r \rightarrow \infty} \left\{ \left(\sum_{j \neq i, |X_t^j| < r} \frac{1}{X_t^i - X_t^j} \right) - \int_{|x| < r} \frac{\varrho(x)}{-x} dx \right\} dt$$

$$\varrho(x) = \frac{\sqrt{-x}}{\pi} \mathbf{1}_{(-\infty, 0]}(x)$$

- The above SDE has a unique, strong solution.
- So far the sto dyn related to Airy RPF was constructed only for $\beta = 2$ by Spohn, Johansson, and others by the method of space-time cor funs. This sto dyn is same as the above.
- The labeled dyn $\mathbf{X}_t = (X_t^i)$ is a diffusion with state space $\mathbb{R}^{\mathbb{N}}$.
- The unlabeled dyn $X_t = \sum_i \delta_{X_t^i}$ is reversible w.r.t. $\mu_{\text{airy}, \beta}$.

Second motivation:

How to define

Coulomb random point fields

in infinite volume

How to construct

Coulomb interacting Brownian motions

(Stochastic Coulomb dynamics)

in infinite-dimensions

How to define Coulomb RPFs in infinite volume 1

Let Ψ_c be c -dim Coulomb potential:

$$\Psi_2(x) = -\log |x|, \quad \Psi_c(x) = (2 - c)^{-1} |x|^{2-c} \quad (c \neq 2)$$

Very loosely, translation invariant Coulomb random point fields μ_c with inverse temperature $\beta > 0$:

$$\mu_{c,\beta} \sim \frac{1}{Z} e^{-\beta \sum_{i<j}^{\infty} \Psi_c(x_i - x_j)} \prod_{k=1}^{\infty} dx_k \quad (6)$$

In particular,

$$\mu_{2,\beta} \sim \frac{1}{Z} \prod_{i<j}^{\infty} |x_i - x_j|^{\beta} \prod_{k=1}^{\infty} dx_k \quad (7)$$

(1) How to define Coulomb RPFs (No DLR eq.!)

How to solve c dim Coulomb infinite-dim SDEs:

- $(\mathbb{R}^d)^{\mathbb{N}}$ -valued SDE: $\mathbf{X}_t = (X_t^i)_{i \in \mathbb{N}}$
- Ψ_c is a c -dim Coulomb pot.
- Coulomb ISDE:

$$dX_t^i = dB_t^i - \frac{\beta}{2} \sum_{j=1, j \neq i}^{\infty} \nabla \Psi_c(X_t^i - X_t^j)$$

$$dX_t^i = dB_t^i + \frac{\beta}{2} \sum_{j=1, j \neq i}^{\infty} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^c} dt \quad (c \text{ dim Coulomb}).$$

(2) How to solve Coulomb ISDEs (**No Ito's scheme !**)

I will show, if $(c, d, \beta) = (2, 2, 2), (2, 1, 1), (2, 1, 2), (2, 1, 4)$, then OK.

Quasi-Gibbs measures and Log derivative

How to define Coulomb RPFs in infinite vol 2: Ψ -Quasi-Gibbs meas.

- $S = \mathbb{R}^d$, $S_r = \{|x| \leq r\}$, $S = \{s = \sum_i \delta_{s_i}, s(S_r) < \infty (\forall r)\}$
- $\pi_r, \pi_r^c : S \rightarrow S$, $\pi_r(s) = s(\cdot \cap S_r)$, $\pi_r^c(s) = s(\cdot \cap S_r^c)$
- Let μ be a RPF over S .

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

- Let $\Psi : S \rightarrow \mathbb{R} \cup \{\infty\}$ (interaction).

$$\mathcal{H}_r = \sum_{s_i, s_j \in S_r, i < j} \Psi(s_i - s_j)$$

Def: μ is Ψ -quasi-Gibbs measure if $\exists c_{r,\xi}^m$ s.t.

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

Here $\nu_r^m = \prod_{k=1}^m 1_{S_r}(s_k) ds_k$

- Gibbs measures \Rightarrow Quasi-Gibbs measure .

Coulomb RPFs

$$c_{r,\xi}^m - 1 e^{-\mathcal{H}_r} d\nu_r^m \leq \mu_{r,\xi}^m \leq c_{r,\xi}^m e^{-\mathcal{H}_r} d\nu_r^m \quad (\text{quasi-Gibbs property})$$

Let Ψ_c is the c dim Coulomb potential as before.

- We say μ is Coulomb RPF if μ is Ψ_c -quasi-Gibbs meas.
- The case ($d \leq c < d + 2$) is interesting.
- μ is called strict Coulomb RPF if $c = d$.

Thm 2 (O. AOP 13, O.-Honda, O.-Tanemura).

(1) Ginibre RPF is a $2\Psi_2$ -quasi Gibbs measure.

(2) Sine $_\beta$ RPF are $\beta\Psi_2$ -quasi Gibbs m for $\beta = 1, 2, 4$.

(3) Bessel $_2^a$ RPF is a $2\Psi_2$ -quasi Gibbs m.

(4) Airy $_\beta$ RPF are $\beta\Psi_2$ -quasi Gibbs m for $\beta = 1, 2, 4$.

- **Conjecture:** The following is a quasi Gibbs measure

(1) β -Sine, Bessel, Airy RPFs for all β .

(2) All determinantal RPFs. Zero points of GAFs.

Application of quasi-Gibbs property to dynamics

Let \mathcal{D}_0 be a local, smooth funs on S .

Let $\tilde{f}(s_1, \dots) = f(s)$, where \tilde{f} is symmetric, $s = \sum \delta_{s_i}$.

$$\mathbb{D}[f, g] = \frac{1}{2} \sum_i \nabla_i \tilde{f} \cdot \nabla_i \tilde{g}$$

$$\mathcal{E}^\mu(f, g) = \int_S \mathbb{D}[f, g] \mu(ds) : \text{bilinear form on } \mathcal{D}_0^\mu$$

$$\mathcal{D}_0^\mu = \{f \in \mathcal{D}_0; \mathcal{E}^\mu(f, f) < \infty, f \in L^2(\mu)\}$$

Thm 3. Let μ be Ψ -quasi-Gibbs with upper semi-conti Ψ . Then

(1) $(\mathcal{E}^\mu, \mathcal{D}_0^\mu)$ is closable on $L^2(\mu)$.

If, in addition, 1-cor fun ρ^1 is loc L^1 , and k -density funs σ_r^k on S_r are L^2 for all $r, k \in \mathbb{N}$, then

(2) \exists diffusion $X_t = \sum_i \delta_{X_t^i}$ associate with $(\mathcal{E}^\mu, \mathcal{D}_0^\mu)$.

Log derivative of μ : precise correspondence between RPFs & potentials

- Let μ_x be the (reduced) Palm m. of μ conditioned at x
$$\mu_x(\cdot) = \mu(\cdot - \delta_x | s(x) \geq 1)$$
- Let μ^1 be the 1-Campbell measure on $\mathbb{R}^d \times S$:

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

- $d\mu \in L^1(\mathbb{R}^d \times S, \mu^1)$ 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 bounded, local smooth functions on S .

- Very informally

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

- The log derivative gives the precise correspondence between RPFs and potentials
- We next give examples of log derivatives

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

Thm 4 (O. PTRF 12).

(1) Let μ_{gin} be the Ginibre RPF. Then

$$d^{\mu_{\text{gin}}}(x, s) = \lim_{r \rightarrow \infty} 2 \sum_{|x-s_i| < r} \frac{x - s_i}{|x - s_i|^2}$$

$$d^{\mu_{\text{gin}}}(x, s) = -2x + \lim_{r \rightarrow \infty} 2 \sum_{|s_i| < r} \frac{x - s_i}{|x - s_i|^2}$$

(2) Let $\mu_{\text{sin}, \beta}$ be the Sine $_\beta$ RPF. Suppose $\beta = 1, 2, 4$. Then

$$d^{\mu_{\text{sin}, \beta}}(x, s) = \lim_{r \rightarrow \infty} \beta \sum_{|x-s_i| < r} \frac{1}{x - s_i}$$

Thm 5 (O.-Honda). Let $\mu_{\text{bes},2}^a$ be the Bessel $_2^a$ RPF. Then

$$d^{\mu_{\text{bes},2}^a}(x, s) = \frac{a}{x} + \lim_{r \rightarrow \infty} 2 \sum_{|x-s_i| < r} \frac{1}{x-s_i}$$

Thm 6 (O.-Tanemura). [Airy RPFs: $\mu_{\text{Ai},\beta}$]

Let $\beta = 1, 2, 4$. Then the log derivative $d^{\mu_{\text{Ai},\beta}}$ is

$$d^{\mu_{\text{Ai},\beta}}(x, s) = \beta \lim_{r \rightarrow \infty} \left\{ \left(\sum_{|x-s_i| < r} \frac{1}{x-s_i} \right) - \int_{|x| < r} \frac{\varrho(x)}{-x} dx \right\}$$

Here

$$\varrho(x) = \frac{\sqrt{-x}}{\pi} \mathbf{1}_{(-\infty, 0]}(x)$$

Other important geometric property of RPFs:
Tail triviality & Tail decomposition

Tail triviality of determinantal RPFs & Tail decomp of quasi-Gibbs m

Let $\mathcal{T} = \mathcal{T}(S)$ be the tail σ field of S :

$$\mathcal{T}(S) = \bigcap_{r=1}^{\infty} \sigma[\pi_r^c].$$

Thm 7. *Let μ be a det RPF. Then $\mathcal{T}(S)$ is μ -trivial.*

- Thm 7 is a generalization of that for the discrete determinantal RPFs due to Russel Lyons, Shirai-Takahashi.

Thm 8. *Let μ be a quasi-Gibbs measure. Let $\mu(\cdot|\mathcal{T})$ be the regular conditional probability. Then*

$$\mu(\cdot) = \int_S \mu(\cdot|\mathcal{T})(\xi) \mu(d\xi)$$

and, for μ -a.s. ξ ,

$$\mu(A|\mathcal{T})(\xi) = 1_A(\xi) \quad \text{for any } A \in \mathcal{T}.$$

- Thm 8 is a generalization of that for the discrete Gibbs m due to Georgii.

Palm singularity & ab continuity
of Ginibre RPF

Palm singularity: Def of Ginibre RPF

Let $K(x, y)$ be a kernel, m be a meas.

- ν is called a determinantal rpf generated by (K, m) if its n correlation fun ρ^n w.r.t. m is given by

$$\rho^n(\mathbf{x}_n) = \det[K(x_i, x_j)]_{1 \leq i, j \leq n} \quad (8)$$

- **Ginibre rpf** μ is the det rpf generated by (K_{gin}, g) :

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

- Ginibre RPF is translation invariant.

How to detect the number of missing particles

Problem:

- Let ν be a translation invariant rpf on \mathbb{C} .
- Let $s = \sum_i \delta_{s_i}$ be a sample point under ν .
- Remove **a finite number** of particles from the sample points $\{s_i\}$.
- Can one detect the number of the removed particles?

Answers

Problem:

- Let ν be a translation invariant rpf on \mathbb{C} .
- Let $s = \sum_i \delta_{s_i}$ be a sample point under ν .
- Remove a finite number of particles from the sample points $\{s_i\}$.
- Can one detect the number of the missing particles?

If ν is a periodic rpf, then "Yes".

If ν is a Poisson rpf, then "No".

The Ginibre rpf μ has a property between periodic and Poisson.

- Yes! for this problem. So Ginibre is similar to periodic RPF rather than Poisson RPFs.
- The quasi-Gibbs property implies Ginibre is similar to Poisson RPFs rather than periodic RPFs.

Main Theorems

Palm meas. For a set of m -points $\mathbf{x} = \{x_1, \dots, x_m\}$ let

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

Thm 9 (O.-Shirai). Let $m, n \in \{0\} \cup \mathbb{N}$. Then

(1) If $m = n$, then $\mu_{\mathbf{x}}$ and $\mu_{\mathbf{y}}$ are *mutually ab. cont.*

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

• (2) shows a special property of Ginibre rpf. Indeed,

Λ Poisson rpf $\Rightarrow \Lambda_{\mathbf{x}} = \Lambda$

ν Gibbs meas with Ruelle's class potentials $\Rightarrow \nu_{\mathbf{x}} \prec \nu$

• ν periodic rpf \Rightarrow (2) holds

Main Theorems

Thm 10 (O.-Shirai). Let $m = n$. Then for μ_y -a.s. s

$$\frac{d\mu_x}{d\mu_y} = \frac{1}{Z_{xy}} \lim_{r \rightarrow \infty} \prod_{|s_i| < b_r} \frac{|x - s_i|^2}{|y - s_i|^2} \quad (s = \sum_i \delta_{s_i}) \quad (9)$$

compact uniformly in $x \in \mathbb{C}^m$, $y \in \mathbb{C}^m \setminus \{s_1, \dots, s_m\}$

$$Z_{xy} = \frac{\Delta(y) \det[K_{\text{gin}}(x_i, x_j)]_{i,j=1}^m}{\Delta(x) \det[K_{\text{gin}}(y_i, y_j)]_{i,j=1}^m}$$

$$\Delta(x) = \prod_{i < j}^m |x_i - x_j|^2, \quad |x - s_i| = \prod_{m=1}^m |x_m - s_i|$$

$$\{b_r\}_{r \in \mathbb{N}} : \quad b_r \uparrow \infty$$

Main Theorems

Let $D_{\sqrt{q}} = \{z \in \mathbb{C}; |z| < \sqrt{q}\}$,

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

By definition $s(D_{\sqrt{q}})$ is the number of particles $s = \sum_i \delta_{s_i}$ in the disk $D_{\sqrt{q}}$.

Thm 11 (O.-Shirai). Let $\mathbf{x} = (x_1, \dots, x_m)$.

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

- Thm 11 means we can determine the number of missing particles:

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

Dynamical Theory: infinite dimensional SDEs

General theorems on infinite-dim SDEs

(A1) μ is a Ψ -quasi-Gibbs m with upper-semicont Ψ . \Rightarrow (closability)

(A2) $\sum_{k=1}^{\infty} k\mu(S_r^k) < \infty$, $\sigma_r^k \in L^2(S_r^k, dx) \Rightarrow$ (existence of diffusions)

Here $S_r = \{|x| < r\}$, $S_r^k = \{s(S_r) = k\}$, σ_r^k is k -density fun on S_r .

(A3) The log derivative $d^\mu \in L^1_{loc}(\mu^1)$ exists \Rightarrow (SDE representation)

(A4) $\{X_t^i\}$ do not collide each other \Rightarrow (non-collision)

(A5) each tagged particle X_t^i never explode \Rightarrow (non-explosion)

Thm 12. (O.12(PTRF)) (A1)–(A5) $\Rightarrow \exists S_0 \subset S$ such that

$$\mu(S_0) = 1, \quad (12)$$

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

$$dX_t^i = dB_t^i + \frac{1}{2} d^\mu(X_t^i, \sum_{j \neq i} \delta_{X_t^j}) dt, \quad (X_0^i)_{i \in \mathbb{N}} = s \quad (13)$$

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

existence of strong solution

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

$$H^1 = \{(x, s) \in S \times S; d^\mu(x, s) \text{ is locally Lips cont.}\}$$

Here “locally” means we regard $d^\mu(x, s)$ as symmetric fun on S_r with fixed particles outside S_r^c for $\forall r$ except a capacity zero set. (non-single points, say).

(A6) $\text{Cap}^\mu(H^c) = 0$. Here $H = \{\delta_x + s; (x, s) \in H^1\}$

existence of strong solution

Thm 13 (O.-Tanemura). Assume (A1)–(A6). Then:

(1) *The ISDE*

$$dX_t^i = dB_t^i + \frac{1}{2}d^\mu(X_t^i, \sum_{j \neq i} \delta_{X_t^j})dt,$$

$$(X_0^i)_{i \in \mathbb{N}} = s$$

has *a strong solution* for $s = (s_i) \in S^{\mathbb{N}}$ s.t. $\sum_i \delta_{s_i} \in H$.

(2) *The ass unlabeled diffusion* $X = \sum_i \delta_{X^i}$ *satisfies*

$$P_{\mu_\xi} \circ X_t^{-1} \prec \mu_\xi \quad (\forall t) \quad \text{for } \mu\text{-a.s. } \xi$$

Here $\mu_\xi = \mu(\cdot | \mathcal{T}(S))(\xi)$ is the tail decomposition in Thm 8.

Uniqueness of strong solutions 1

Thm 14 (O.-Tanemura). Assume (A1)–(A6).
Let $\mathbf{X} = (X^i)$ and $\widehat{\mathbf{X}} = (\widehat{X}^i)$ be strong sol of the ISDE

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

on the same Br m. Let $X_t = \sum_i \delta_{X_t^i}$ and $\widehat{X}_t = \sum_i \delta_{\widehat{X}_t^i}$.
Suppose, for μ -a.s. ξ ,

$$P_{\mu_\xi} \circ X_t^{-1} \prec \mu_\xi \text{ and } P_{\mu_\xi} \circ \widehat{X}_t^{-1} \prec \mu_\xi \quad (\forall t)$$

Here $\mu_\xi = \mu(\cdot | \mathcal{T}(S))(\xi)$ (Thm 8). Then

$$\mathbf{X} = \widehat{\mathbf{X}} \text{ a.s.} \quad \text{for } \mu\text{-a.s. } \mathbf{s} = \sum_{i=1}^{\infty} \delta_{s_i}$$

Thm 15 (O.-Tanemura). Assume (A1)–(A7). Here (A7) μ is tail trivial.

Then the strong solution $\mathbf{X} = (X^i)$ such that

$$P_\mu \circ X_t^{-1} \prec \mu \quad \text{for all } t$$

is unique for μ -a.e. $x = \sum_i \delta_{x_i}$. Here X is the unlabeled dynamics of \mathbf{X} :

$$X_t = \sum_i^\infty \delta_{X_t^i}$$

Cor If μ is a determinantal RPF, then the strong, solution of the ISDE that is reversible w.r.t. μ is unique.

- Tail σ -fields of Airy, Sine, Ginibre RPFs with $\beta = 2$ are trivial.

Uniqueness of Dirichlet forms

Let $\mathcal{D}_{\text{poly}}^\mu$ be the closure of the set of polynomials on S such that $\mathcal{E}_1^\mu(f, f) < \infty$. Then

$$\mathcal{D}_{\text{poly}}^\mu \subset \mathcal{D}^\mu$$

because polynomials are local and smooth.

Thm 16 (O.-Tanemura). *Assume (A1)–(A7). Then the Dirichlet form that are extension of $(\mathcal{E}^\mu, \mathcal{D}_{\text{poly}}^\mu)$ is unique.*

In particular, $\mathcal{D}_{\text{poly}}^\mu = \mathcal{D}^\mu$, and Lang's construction and O.'s construction are same.

Remark 1. If (A5) (non-explosion) does not hold. Then Thm 16 does not hold. This is very natural theorem that says the uniqueness of Dirichlet forms is related to the non-explosion problem of tagged problem.

Examples:

Examples

Examples: Gibbs measures

All example below satisfy (A1)–(A6). Hence by Thm 14 we have a strong solution that preserves the tail σ field.

Gibbs measures :

- All Gibbs measures with Ruelle's class potentials (smooth outside the origin) satisfy the assumptions (A.1)–(A.6). Non-collision (A4) does not hold in general. But it always holds for $d \geq 2$ and, for repulsive interaction ψ in $d = 1$.
- In this case, the SDEs become

$$dX_t^i = dB_t^i - \frac{1}{2} \nabla \Phi(X_t^i) dt - \frac{1}{2} \sum_{j \neq i} \nabla \psi(X_t^i - X_t^j) dt. \quad (14)$$

Examples: Ruelle's class potentials 1

Lennard-Jones 6-12 potential:

$$\Phi_{6,12}(x) = |x|^{-12} - |x|^{-6}$$

The corresponding ISDE is:

$$dX_t^i = dB_t^i + \frac{\beta}{2} \sum_{j=1, j \neq i}^{\infty} \left\{ \frac{12(X_t^i - X_t^j)}{|X_t^i - X_t^j|^{14}} - \frac{6(X_t^i - X_t^j)}{|X_t^i - X_t^j|^8} \right\} dt \quad (i \in \mathbb{N})$$

Examples: Ruelle's class potentials 2

Coulomb like potentials (not Coulomb!)

Let $a > d$ and set $\Phi_a(x) = (1/a)|x|^{-a}$.

$$dX_t^i = dB_t^i + \frac{\beta}{2} \sum_{j=1, j \neq i}^{\infty} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^{a+2}} dt \quad (i \in \mathbb{N}). \quad (15)$$

At first glance the ISDE (15) resembles Ginibre IBMs, because these corresponds to the case $a = 0$ in (15). The sums in the drift terms, however, converge absolutely, unlike Coulomb (log) potentials. We emphasize that the structures of the dynamics given by the solutions of (15) and Ginibre IBMs are completely different from each other.

Examples: Ginibre rpf

Ginibre rpf: $\Psi(x) = -\beta \log |x|$ $d = 2$, $\beta = 2$. If $\mu = \mu_{\text{gin},2}$,

$$dX_t^i = dB_t^i + \lim_{r \rightarrow \infty} \sum_{|X_t^i - X_t^j| < r, j \neq i} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} dt \quad (16)$$

and also

$$dX_t^i = dB_t^i - X_t^i dt + \lim_{r \rightarrow \infty} \sum_{\substack{|X_t^j| < r \\ j \neq i}} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} dt. \quad (17)$$

This comes from the plural expressions of $d\mu_{\text{gin},2}$.

For finite N , these SDEs give **different** solution.

But in the limit $N \rightarrow \infty$ give **the same solution** if the initial distribution is closed to Ginibre rpf.

Examples: Bessel rpf–hard edge scaling limit

Bessel RPF (joint work with Honda):

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

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

$\beta = 1, 4$ are in progress.

Examples: sine rpf (Dyson's model)–bulk scaling limit

Sine_β RPF: $S = R$, $\beta = 1, 2, 4$

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

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 semigr by Dirichlet form.

The def of $\mu = \mu_{\text{sin},\beta}$:

$\beta = 2 \Rightarrow \mu_{\text{sin},\beta}$ is the det rpf generated by (K_{sin}, dx) :

$$K_{\text{sin}}(x, y) = \frac{\sin(\pi(x - y))}{\pi(x - y)}$$

$\beta = 1, 4 \Rightarrow$ the correlation funs are given by quaternion det.

Examples: Airy rpf – Soft edge scaling limit

Thm 17 (O.-Tanemura). *Let $\beta = 1, 2, 4$. Then:*

- *The log derivative $d^{\mu_{\text{Ai},\beta}}$ is*

$$d^{\mu_{\text{Ai},\beta}}(x, s) = \beta \lim_{r \rightarrow \infty} \left\{ \left(\sum_{|x-s_i| < r} \frac{1}{x-s_i} \right) - \int_{|x| < r} \frac{\varrho(x)}{-x} dx \right\}$$

Here

$$\varrho(x) = \frac{\sqrt{-x}}{\pi} \mathbf{1}_{(-\infty, 0]}(x)$$

- *Airy rpf $\mu_{\text{Ai},\beta}$ satisfy (A1)–(A6) and the limit ISDE is*

$$dX_t^i = dB_t^i + \frac{\beta}{2} \lim_{r \rightarrow \infty} \left\{ \left(\sum_{j \neq i, |X_t^j| < r} \frac{1}{X_t^i - X_t^j} \right) - \int_{|x| < r} \frac{\varrho(x)}{-x} dx \right\} dt$$

Examples: Airy rpf – Soft edge scaling limit

- The key idea is to take the **rescaled** semi-circle law ς , as the first approximation of the 1-correlation fun $\rho_{\text{Ai},\beta}^{N,1}$.
- Our method can be applied to other soft edge scaling. Our result is the first time to clarify the SDE describing the limit infinite system for the soft edge.

Examples: Airy rpf – Soft edge scaling limit

Thm 18 (O.-Tanemura). Assume $\beta = 2$.

Let us label $X_t^i > X_t^{i+1}$ ($\forall i$). Then :

(1) The top particle X_t^1 is the Airy process $A(t)$ in the sense of Spohn.

(2) The infinite dim stochastic dynamics constructed by Spohn, Johansson & others by the space-time correlation fun is a solution of the prescribed SDE:

$$dX_t^i = dB_t^i + \frac{\beta}{2} \lim_{r \rightarrow \infty} \left\{ \left(\sum_{j \neq i, |X_t^j| < r} \frac{1}{X_t^i - X_t^j} \right) - \int_{|x| < r} \frac{\varrho(x)}{-x} dx \right\} dt$$

Examples: Airy rpf – Soft edge scaling limit

- The SDE gives a kind of Girsanov formula.
- These examples are the first time that the infinite dynamics are constructed for rpf appeared in random matrix theory with $\beta = 1, 4$ even if the bulk and the hard edge as well as the soft edge scaling

In one dimensional system, the method of space-time correlation functions are available (Nagao, Katori-Tanemura, Spohn, and others), but this method is restricted to $\beta = 2$.

- By construction, if the total system start from the Airy $_{\beta}$ rpf $\mu_{\text{Ai},\beta}$, then the distribution of the top particle X_t^1 equals $F_{\beta,edge}(x)$, the β Tracy-Widom distribution, where $\beta = 1, 2, 4$.

A phase transition conjecture for 2D Coulomb stochastic dynamics

Homogenization
and
Phase transition conj of Ginibre IBMs

A phase transition conjecture for 2D Coulomb stochastic dynamics

Let $S = \mathbb{R}^2$. Let $\beta \in [0, \infty)$ and set

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

When $\beta = 2$, then the SDE has a solution, for general β we **assume** the existence of solution and the rpf $\mu_{\text{gin}, \beta}$.

- We tag $X_t^{i_0}$ and investigate the diffusive scaling:

$$\lim_{\epsilon \rightarrow 0} \epsilon X_{t/\epsilon^2}^{i_0} = \sqrt{2\alpha_{\text{self}}[\mu_{\text{gin}, \beta}]} B_t$$

- Assume $X_0^{i_0} = 0$ and $\sum_{i \neq i_0} \delta_{X_0^i} \sim \mu_{\text{gin}, \beta, 0}$.
- $\alpha_{\text{self}}[\cdot]$ is called the self-diffusion matrix.

A phase transition conjecture for 2D Coulomb stochastic dynamics

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

$$\lim_{\epsilon \rightarrow 0} X_{t/\epsilon^2}^{i_0} = \sqrt{2\alpha_{\text{self}}[\mu_{\text{gin},\beta}]} B_t, \quad X_0^{i_0} = 0, \quad \sum_{i \neq i_0} \delta_{X_0^i} \sim \mu_{\text{gin},\beta,0}.$$

Conj: There exist constants $\beta_1 < \beta_2 < \beta_3$ such that

(C1) $\beta < \beta_1 \Rightarrow \alpha_{\text{self}}[\mu_{\text{gin},\beta}] > 0$ (diffusive)

(C2) $\beta_1 < \beta < \infty \Rightarrow \alpha_{\text{self}}[\mu_{\text{gin},\beta}] = 0$ (subdiffusive),

(C3) $\beta_2 < \beta < \infty \Rightarrow X_t^{i_0}$ has an inv prob measure
 $X_t^{i_0} = O(\log t)$ (log behavior)

(C4) $\beta_3 < \beta < \infty \Rightarrow (X_t^i)_{i \in \mathbb{N}}$ form a lattice like system.

Moreover,

$$\beta_1 \sim 1, \quad \beta_2 \sim 2.$$

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

$$dX_t^s = dB_t + \frac{\beta}{2} \lim_{q \rightarrow \infty} \sum_{|X_t^s - s_i| < q} \frac{X_t^s - s_i}{|X_t^s - s_i|^2} dt$$

Let μ be a rpf, and set for a.s. s w.r.t. μ

$$\lim_{\varepsilon \rightarrow \infty} \varepsilon X_{t/\varepsilon^2}^s = \sqrt{\alpha_{\text{eff}}^\beta[\mu]} B_t \quad (18)$$

Thm 19. μ_{per} be a periodic rpf \Rightarrow

(1) $\alpha_{\text{eff}}^\beta[\mu_{\text{per}}] > 0$.

(2) $\alpha_{\text{eff}}^\beta[\mu_{\text{per},0}] > 0$ for $\beta < 1$

$\alpha_{\text{eff}}^\beta[\mu_{\text{per},0}] = 0$, X_t^s has a inv prob m for $\beta > 2$

Rigorous results: homogenization of diffusion in Ginibre env.

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

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

Thm 20. Assume $s \sim \mu_{\text{gin},2,0}$ and set

$$\lim_{\varepsilon \rightarrow \infty} \varepsilon X_{t/\varepsilon^2}^s = \sqrt{\alpha_{\text{eff}}^2[\mu_{\text{gin},2,0}]} B_t$$

Then

$$\alpha_{\text{eff}}^2[\mu_{\text{gin},2,0}] = 0$$

• We use (10) in Thm 11 to prove Thm 20

Conj: The positivity of $\alpha_{\text{eff}}^2[\mu_{\text{gin},2}]$ is an open problem.

Since $\mu_{\text{gin},2}$ is similar to μ_{per} , we should have

$$\alpha_{\text{eff}}^2[\mu_{\text{gin},2}] > 0$$

Observation: self-diffusion of 2D Coulomb system 1

Obs 0: $\mu_{\text{gin},\beta}$ exists for general $\beta > 0$.

Obs 1: Since (by O.-Shirai [2012])

$$\mu_{\text{gin},2} \perp \mu_{\text{gin},2,0} \quad (19)$$

we have for general $\beta > \beta_1$ ($\beta_1 \leq 2$)

$$\mu_{\text{gin},\beta} \perp \mu_{\text{gin},\beta,0}. \quad (20)$$

Let $\mathbf{X}_t = (X_t^i)_{i \in \mathbb{N}}$ be the solution of

$$dX_t^i = dB_t^i + \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|X_t^i - X_t^j| < r, j \neq i} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} dt \quad (21)$$

Note that

$$X_t := \sum_{i \in \mathbb{N}} \delta_{X_t^i} \sim \mu_{\text{gin},\beta} \quad (22)$$

Observation: self-diffusion of 2D Coulomb system 2

Let X_t^1 be the tag particle, and set $Y_t^i = X_t^{i+1} - X_t^1$.
 $Y_t = \sum_{i \in \mathbb{N}} \delta_{Y_t^i}$ is the env seen from the tagged particle.

Obs 2: By (20) and

$$Y_t := \sum_{i \neq i_0} \delta_{Y_t^i} \sim \mu_{\text{gin}, \beta, 0}, \quad (23)$$

we have $X_t^* \in \mathbb{C}$ such that (by Goldman coupling if $\beta = 2$)

$$X_t^* \sim \text{prob } m, \quad Y_t + \delta_{X_t^*} \sim X_t \sim \mu_{\text{gin}, \beta} \quad (24)$$

Observation: self-diffusion of 2D Coulomb system 3

Obs 3:

$$dX_t^1 = dB_t^1 + \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|X_t^1 - X_t^i| < r, i \geq 2} \frac{X_t^1 - X_t^i}{|X_t^1 - X_t^i|^2} dt \quad (25)$$

$$dX_t^1 = dB_t^1 - \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|Y_t^i| < r, i \in \mathbb{N}} \frac{Y_t^i}{|Y_t^i|^2} dt$$

Observation: self-diffusion of 2D Coulomb system 4

Set $Y_t^* = X_t^* - X_t^1$. Then from

$$dX_t^1 = dB_t^1 - \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|Y_t^i| < r, i \in \mathbb{N}} \frac{Y_t^i}{|Y_t^i|^2} dt \quad (26)$$

we have

$$dX_t^1 = dB_t^1 - \frac{\beta}{2} \frac{X_t^1 - X_t^*}{|X_t^1 - X_t^*|^2} dt - \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|Y_t^i| < r, i \in \mathbb{N} \cup \{*\}} \frac{Y_t^i}{|Y_t^i|^2} dt$$

Hence

$$\begin{aligned} dX_t^1 &= dB_t^1 - \frac{\beta}{2} \frac{X_t^1}{|X_t^1|^2} dt \\ &+ \frac{\beta}{2} \left\{ \frac{X_t^1}{|X_t^1|^2} - \frac{X_t^1 - X_t^*}{|X_t^1 - X_t^*|^2} \right\} dt - \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{\substack{|Y_t^i| < r, \\ i \in \mathbb{N} \cup \{*\}}} \frac{Y_t^i}{|Y_t^i|^2} dt \end{aligned} \quad (27)$$

Observation: self-diffusion of 2D Coulomb system 5

$$\begin{aligned}
 dX_t^1 &= dB_t^1 - \frac{\beta}{2} \frac{X_t^1}{|X_t^1|^2} dt \\
 &+ \frac{\beta}{2} \left\{ \frac{X_t^1}{|X_t^1|^2} - \frac{X_t^1 - X_t^*}{|X_t^1 - X_t^*|^2} \right\} dt - \frac{\beta}{2} \lim_{r \rightarrow \infty} \sum_{|Y_t^i| < r, i \in \text{NU}\{*\}} \frac{Y_t^i}{|Y_t^i|^2} dt
 \end{aligned} \tag{28}$$

Obs 4: (1) By homogenization, $\exists \sqrt{2a[\beta]} \leq E$

$$\epsilon \left\{ B_{u/\epsilon^2}^1 - \frac{\beta}{2} \lim_{r \rightarrow \infty} \int_0^{u/\epsilon^2} \sum_{|Y_t^i| < r, i \in \text{NU}\{*\}} \frac{Y_t^i}{|Y_t^i|^2} dt \right\} = \sqrt{2a[\beta]} \hat{B}_u \tag{29}$$

Since X_t^* has inv prob

$$\epsilon \int_0^{u/\epsilon^2} \frac{\beta}{2} \left\{ \frac{X_t^1}{|X_t^1|^2} - \frac{X_t^1 - X_t^*}{|X_t^1 - X_t^*|^2} \right\} dt \sim o(\epsilon) \tag{30}$$

Observation: self-diffusion of 2D Coulomb system 6

Hence we have (approximately)

$$dX_t^1 = \sqrt{2a[\beta]}dB_t^1 - \frac{\beta}{2} \frac{X_t^1}{|X_t^1|^2}dt \quad (31)$$

By the simple calculation ($\beta > \beta_{00}$, \tilde{B}_t is 1D Br m)

$$d|X_t^1| = \sqrt{2a[\beta]}d\tilde{B}_t + \left(a[\beta] - \frac{\beta}{2}\right) \frac{1}{|X_t^1|}dt \quad (32)$$

So the phase transition follows from the one of Bessel processes.

Simulation of Ginibre IBM (2D Coulomb system) and phase transition

Simulation of Coulomb interacting Brownian motions

Simulation of Ginibre IBM (2D Coulomb system) and phase transition

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

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

Here, since $\rho^1 = 1/\pi$, $\alpha = |\{|x| \leq 1\}| \rho^1 = 1$.

- Taking (OU) & (T) into account we take the model:

$$dX_t^i = dB_t^i + \frac{\beta}{2} \left\{ -X_t^i + \sum_{j=1, j \neq i}^N \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} \right\} dt \quad (\text{OUN})$$

$$dX_t^i = dB_t^i + \frac{\beta}{2} \left\{ \sum_{j=1, j \neq i}^N \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} \right\} dt \quad (\text{TN})$$

Simulation: 3D Coulomb system

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

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

We take $\rho^1 = 1$. So $\alpha = |\{|x| \leq 1\}| \rho^1 = 4\pi/3$.

- Taking (OU) & (T) into account we take the model:

$$dX_t^i = dB_t^i + \frac{\beta}{2} \left\{ -\frac{4\pi}{3} X_t^i + \sum_{j=1}^N \sum_{j \neq i} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} \right\} dt \quad (\text{OUN})$$

$$dX_t^i = dB_t^i + \frac{\beta}{2} \left\{ \sum_{j=1}^N \sum_{j \neq i} \frac{X_t^i - X_t^j}{|X_t^i - X_t^j|^2} \right\} dt \quad (\text{TN})$$