

# Disordered systems and Hamilton-Jacobi equations (lecture 3)

## Part 2. Rank-one matrix estimation.

Alternative methods: Lelange, Pislane; Robin Thicis.

### I/ Introduction

**Problem:** Let  $\bar{x} = (\bar{x}_1, \dots, \bar{x}_N)$  be a vector of  $N$  indep. n.v., law  $\mathbb{P}_N = \mathbb{P}^{\otimes N}$ . For some  $t > 0$ , we observe:

$$Y = \sqrt{\frac{2t}{N}} \bar{x} \bar{x}^T + W$$

where  $W = (W_{ij})$  are indep. standard Gaussians.

For instance, we want to understand:

$$\begin{aligned} \text{mmse}_N(t) &= \frac{1}{N^2} \inf_{\hat{\Theta}} \mathbb{E} \left[ \left| \bar{x} \bar{x}^t - \hat{\Theta}(Y) \right|^2 \right] \\ &= \frac{1}{N^2} \mathbb{E} \left[ \left| \bar{x} \bar{x}^t - \mathbb{E} \left[ \bar{x} \bar{x}^t \mid Y \right] \right|^2 \right] \end{aligned}$$

Formally:

$$\mathbb{P}[\bar{x} = x \text{ and } Y = y] = \int d\mathbb{P}_N(x) dy \exp\left(-\frac{1}{2} \left| y - \sqrt{\frac{2t}{N}} x x^t \right|^2\right)$$

$$\mathbb{P}[\bar{x} = x \mid Y] = \frac{\exp\left(-\frac{1}{2} \left| y - \sqrt{\frac{2t}{N}} x x^t \right|^2\right) d\mathbb{P}_N(x)}{\int \exp\left(-\frac{1}{2} \left| y - \sqrt{\frac{2t}{N}} x' x'^t \right|^2\right) d\mathbb{P}_N(x')}$$

**Def:**  $H_N^0(t, x) = \sqrt{\frac{2t}{N}} Y \cdot x x^t - \frac{t}{N} |x|^4$

$$H_N^0(t, x) = \underbrace{\sqrt{\frac{2t}{N}} x \cdot W x}_{\text{like spin glass}} + \underbrace{\frac{2t}{N} (x \cdot \bar{x})^2 - \frac{t}{N} |x|^4}_{\text{help us.}}$$

Denote

$$\langle f(x) \rangle := \frac{\int f(x) \exp(H_N^0(t, x)) d\mathbb{P}_N(x)}{\int \exp(H_N^0(t, x)) d\mathbb{P}_N(x)}$$

*still random*

**Exercise (credit):** show that:

$$\langle f(x) \rangle = \mathbb{E} [f(\bar{x}) \mid Y]$$

**Important ppty (Nishimori):**

$$\mathbb{E} \langle f(x) \rangle = \mathbb{E} [f(\bar{x})]$$

$$\begin{aligned} \mathbb{E} \langle f(x) g(x') \rangle &= \mathbb{E} [\langle f(x) \rangle \langle g(x') \rangle] \\ &= \mathbb{E} [\langle f(x) \rangle \mathbb{E} [g(\bar{x}) \mid Y]] \\ &= \mathbb{E} \langle f(x) g(\bar{x}) \rangle \end{aligned}$$

*indep. copy of  $\bar{x}$  under  $\langle \cdot \rangle$*

$$\mathbb{E} \langle f(x, x') \rangle = \mathbb{E} \langle f(x, \bar{x}) \rangle$$

We need to enrich the model somehow!

We also observe, for some  $h \geq 0$ :

$$\sqrt{2h} \bar{x} + z$$

*standard Gaussian vectn.*

$$H_N(t, h, x) = H_N^0(t, x) + \sqrt{2h} x \cdot z + 2h x \cdot \bar{x} - h |x|^2$$

$$F_N(t, h) = \frac{1}{N} \log \int \exp(H_N(t, h, x)) d\mathbb{P}_N(x)$$

$$\bar{F}_N(t, h) = \mathbb{E} [F_N(t, h)]$$

**Thm:**  $\bar{F}_N \rightarrow f$  solution of:

$$\begin{cases} \partial_t f - (\partial_h f)^2 = 0 & \text{on } \mathbb{R}_+ \times \mathbb{R}_+ \\ f(0, \cdot) = \bar{F}_1(0, \cdot) \end{cases}$$

### II/ Main steps.

**Prop:**  $\partial_t \bar{F}_N - (\partial_h \bar{F}_N)^2 = \frac{1}{N^2} \mathbb{E} \langle (x \cdot \bar{x} - \mathbb{E} \langle x \cdot \bar{x} \rangle)^2 \rangle$

and:

$$\begin{aligned} &\frac{1}{N^2} \mathbb{E} \langle (x \cdot \bar{x} - \mathbb{E} \langle x \cdot \bar{x} \rangle)^2 \rangle \\ &\leq \frac{1}{N} \partial_h^2 \bar{F}_N + \mathbb{E} \left[ (\partial_h F_N - \partial_h \bar{F}_N)^2 \right] \end{aligned}$$

**Lemma:**  $\partial_h \bar{F}_N = \mathbb{E} \left\langle \frac{x \cdot \bar{x}}{N} \right\rangle$

$$\partial_t \bar{F}_N = \mathbb{E} \left\langle \left( \frac{x \cdot \bar{x}}{N} \right)^2 \right\rangle$$

Itô calculus without Itô.

$$\mathbb{E} \left[ \exp(\sqrt{2t} X - t) \right] = 1$$

$$S_0 \mathbb{E} \left[ \left( \frac{1}{\sqrt{2t}} X - 1 \right) \exp(\dots) \right] = 0 \quad \left. \begin{array}{l} \nearrow \\ \searrow \end{array} \right\} \partial_t$$

$$\int x \underbrace{\exp(\sqrt{2t} x - t)}_{\text{...}} \exp\left(-\frac{x^2}{2}\right) dx$$

$$= \int \sqrt{2t} \exp(\dots) \exp\left(-\frac{x^2}{2}\right) dx$$

Ex (credit) 1) prove Lemma 1 using GIP & Nishimii

(2) Relate mmse with  $\partial_t \bar{F}_N$ .

Second part of prop.

A a r.v. Notice that:

$$\mathbb{E} \langle (A - \mathbb{E} \langle A \rangle)^2 \rangle$$

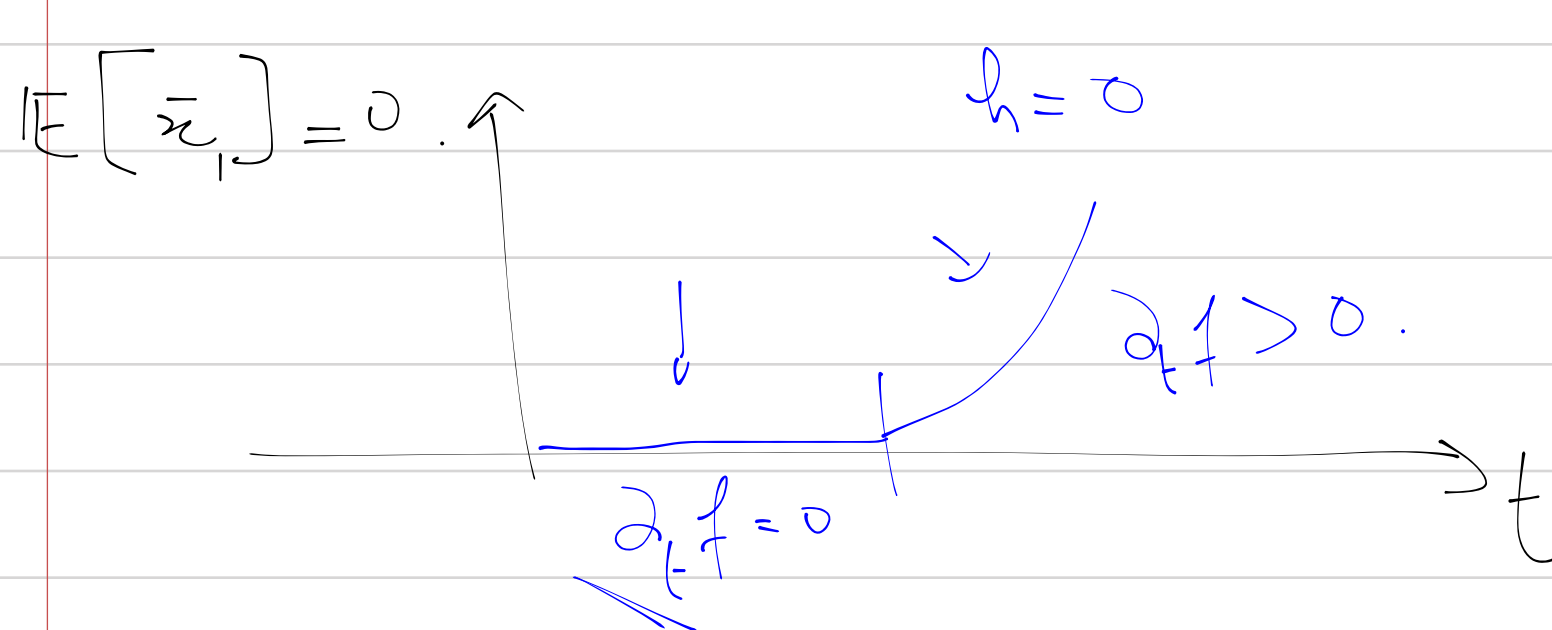
$$= \mathbb{E} \langle (A - \langle A \rangle)^2 \rangle \quad \left. \begin{array}{l} \nearrow \\ \searrow \end{array} \right\} \frac{1}{N} \partial_h^2 \bar{F}_N$$

$$+ \mathbb{E} \left[ \underbrace{(\langle A \rangle - \mathbb{E} \langle A \rangle)^2}_{\partial_h \bar{F}_N - \partial_h \bar{F}_N} \right]$$

New ingredient:

Concentration estimate (Efron-Stein)

$$\mathbb{E} \left[ \left| \bar{F}_N - \bar{F}_N \right|^2 \right] \text{ small.}$$



$$\left\| \partial_t f - \left( \partial_h f \right)^2 \right.$$

$$\left. \right| \leq \frac{C}{N} \left( \partial_h^2 f \right)^{1/2}$$