

Disordered systems and Hamilton-Jacobi equations

I/ Motivations

Students $i \in \{1, \dots, N\}$
 two dorms ; assignment $\sigma \in \{+1\}^N$

Quality of interaction between i and j : J_{ij}

Suppose (J_{ij}) are indep. standard Gaussian.

We want to maximize $\sigma \mapsto \sum J_{ij} \mathbb{1}_{\{\sigma_i = \sigma_j\}}$

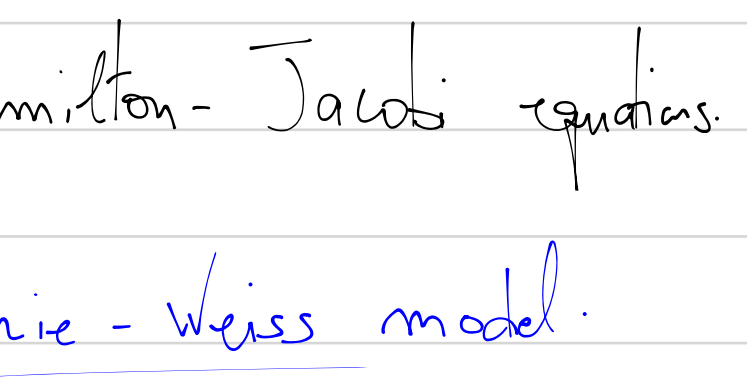
What is $\max_{\sigma \in \{\pm 1\}^N} \sum J_{ij} \sigma_i \sigma_j$ as $N \rightarrow \infty$?

Frustrations

$$\mathbb{E} \frac{1}{N} \log \sum_{\sigma \in \{\pm 1\}^N} \exp \left(\frac{\beta}{\sqrt{N}} \sum_{ij=1}^N J_{ij} \sigma_i \sigma_j \right)$$

\rightarrow complicated formula.
 $N \rightarrow \infty$

Parisi '79 ; Guerra '03 ; Talagrand '06.



Rank-one matrix estimation

We observe a noisy version of a rank-one matrix.

Common thread. Hamilton-Jacobi equations

Part I - Curie-Weiss model.

I/ Definitions.

We want to study the probab. measure that, to each $\sigma \in \{\pm 1\}^N$, associates a weight prop. to

$$\exp \left(\frac{t}{N} \sum_{ij=1}^N \sigma_i \sigma_j + h \sum_{i=1}^N \sigma_i \right)$$

$$\langle f(\sigma) \rangle_{t,h} = \frac{\sum_{\sigma} f(\sigma) \exp(\dots)}{\sum_{\sigma} \exp(\dots)} \quad \begin{matrix} t \geq 0 \\ h \in \mathbb{R} \end{matrix}$$

$$F_N(t, h) = \frac{1}{N} \log \sum_{\sigma} \exp \left(\frac{t}{N} \sum \sigma_i \sigma_j + h \sum \sigma_i \right)$$

Moment generating function

$$\partial_h F_N = \frac{1}{N} \frac{\sum (\sum \sigma_i) \exp(\dots)}{\sum \exp(\dots)} = \frac{1}{N} \langle \sum \sigma_i \rangle$$

$$\partial_t F_N = \frac{1}{N} \langle \frac{1}{N} \sum_{ij} \sigma_i \sigma_j \rangle = \langle \left(\frac{1}{N} \sum \sigma_i \right)^2 \rangle$$

$$\partial_t F_N - (\partial_h F_N)^2 = \langle \left(\frac{1}{N} \sum \sigma_i \right)^2 \rangle - \left(\langle \frac{1}{N} \sum \sigma_i \rangle \right)^2$$

$$\partial_h^2 F_N = \frac{1}{N} \langle \left(\sum \sigma_i \right)^2 \rangle - \frac{1}{N} \left(\langle \sum \sigma_i \rangle \right)^2$$

$$\partial_t F_N - (\partial_h F_N)^2 = \frac{1}{N} \partial_h^2 F_N$$

Moreover: $F_N(0, h) = \frac{1}{N} \log \sum_{\sigma} \exp \left(h \sum_{i=1}^N \sigma_i \right)$

$$= \frac{1}{N} \log \sum_{\sigma} \prod_{i=1}^N \exp(h \sigma_i)$$

$$\hookrightarrow = \frac{1}{N} \log (e^h + e^{-h})^N$$

$$F_N(0, h) = F_1(0, h) = \Psi(h)$$

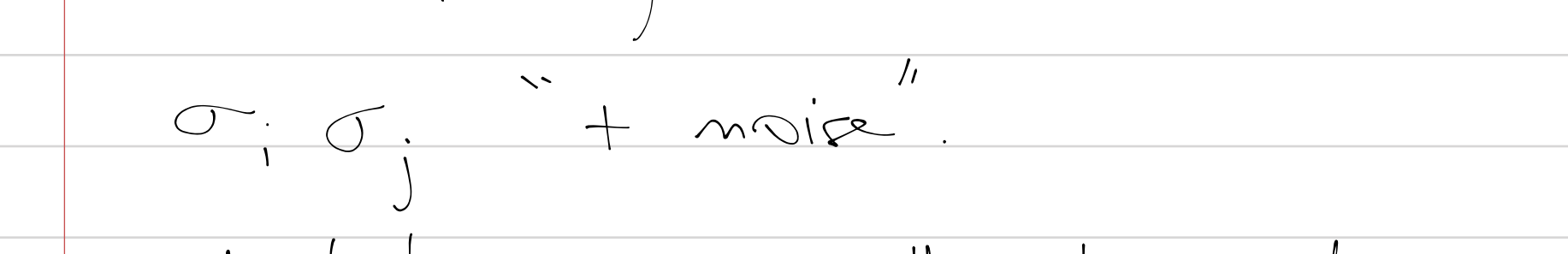
II/ Introduce on Hamilton-Jacobi equations

We need to think about what it means to be a solution of:

$$\partial_t f - \left(\partial_h f \right)^2 = 0 \quad \left(\partial_t f - \|\nabla f\|^2 = 0 \right)$$

Maybe we look for C^1 functions that satisfy the equation.

PB: t small. t large.



$$\int \exp(\dots) dP^{\otimes N}(\sigma)$$

P probab. with bdd support on \mathbb{R}

$$\frac{1}{\sqrt{N}} \sum_{ij=1}^N J_{ij} \sigma_i \sigma_j \sim W(0, N)$$

$$\max_{\sigma \in \{\pm 1\}^N} \left(\mathbb{E} \sigma \right) \sim \sqrt{N} c \sqrt{\log(2^N)}$$

$$\sim cN$$

$$\{1, \dots, N\}$$

$$\sigma_i \in \{\pm 1\}$$

σ_i, σ_j " + noise "

Link between i & j with prob. p_N if $\sigma_i \sigma_j = 1$

q_N if $\sigma_i \sigma_j = -1$

$$N p_N \rightarrow +\infty \quad \left(\begin{matrix} p > q \end{matrix} \right)$$