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# Regularization properties for the 2D Boltzmann equation

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# 1. The 2D Boltzmann equation

## The mechanism

$$\begin{aligned}V_{k+1} &= V_k + \frac{1}{2}(R_\theta - I)(V_k - v_k) \\ &= \frac{V_k + v_k}{2} + R_\theta \times \frac{V_k - v_k}{2}\end{aligned}$$

with

$$\begin{aligned}V_k &\sim f_k(dv) & v_k &\sim f_k(dv) \\ \theta &\in \left[-\frac{\pi}{2}, \frac{\pi}{2}\right] & \text{with "intensity"} & |V_k - v_k|^\gamma \times \frac{1}{|\theta|^{1+\nu}}\end{aligned}$$

## The analytical equation

$$\partial_t f_t(v) = \int_{R^2} dv_* \int_{\pi/2}^{\pi/2} \left( f_t(v') f_t(v'_*) - f_t(v) f_t(v_*) \right) \frac{|v - v_*|^\gamma}{|\theta|^{1+\nu}} d\theta$$

with

$$v' = \frac{v + v_*}{2} + R_\theta \times \frac{v - v_*}{2}, \quad v'_* = \frac{v + v_*}{2} - R_\theta \times \frac{v - v_*}{2}.$$

## The probabilistic interpretation (Tanaka,1978)

$$V_t = V_0 + \int_0^t \int \frac{1}{2}(R_\theta - I)(V_{s-} - v) \mathbf{1}_{\{u \leq |V_{s-} - v|^\gamma\}} N(ds, du, d\theta, dv)$$

with  $N$  Poisson point measure of compensator

$$\widehat{N}(ds, du, d\theta, dv) = ds du f_s(dv) \frac{d\theta}{|\theta|^{1+\nu}}$$

where  $f_t(dv)$  is the solution of the analytical equation. Then

$$P(V_t \in dv) = f_t(dv).$$

**The apporximaiting eqaution** : we take

$$\widehat{N}_\varepsilon(ds, du, d\theta, dv) = ds du f_t(dv) \frac{d\theta}{\varepsilon \vee |\theta|^{1+\nu}}$$

**Construction** : we take

$$\lambda_\varepsilon = \int_{-\pi/2}^{\pi/2} \frac{d\theta}{\varepsilon \vee |\theta|^{1+\nu}} < \infty,$$

$$\mu_\varepsilon(d\theta) = \frac{1}{\lambda_\varepsilon} \times \frac{d\theta}{\varepsilon \vee |\theta|^{1+\nu}} \quad \text{probability law}$$

and

$$T_{k+1} - T_k \sim \mathbf{1}_{[0,\infty)}(t) \frac{1}{\lambda_\varepsilon} \exp(-\lambda_\varepsilon t)$$

$$\theta_k \sim \mu_\varepsilon(d\theta) \quad v_k \sim f_{T_k}(dv) \quad U_k \sim \mathbf{1}_{[0,1]}(u) du$$

and the approximating equation is

$$V_{T_{k+1}}^\varepsilon = V_{T_k}^\varepsilon + \frac{1}{2} (R_{\theta_k} - I) (V_{T_k}^\varepsilon - v) \mathbf{1}_{\{U_k \leq |V_{T_k}^\varepsilon - v_k|^\gamma\}}$$

**Theorem**

$$E |V_t - V_t^\varepsilon| \leq C \varepsilon^{1+\gamma}$$

## Alternative representation

$$\bar{V}_{T_{k+1}}^\varepsilon = \bar{V}_{T_k}^\varepsilon + \frac{1}{2}(R_{\bar{\theta}_k} - I)(\bar{V}_{T_k}^\varepsilon - v)$$

with

$$\bar{\theta}_k \sim |V_{T_k}^\varepsilon - v_k|^\gamma \times \mu_\varepsilon(d\theta).$$

**Conclusion** :  $\bar{V}_t^\varepsilon$  is a functional of :

$$T_k \sim \exp(\lambda_\varepsilon) \quad \text{colision times}$$

$$v_k \sim f_{T_k}(dv) \quad \text{colision directions}$$

$$\bar{\theta}_k \sim |V_{T_k}^\varepsilon - v_k|^\gamma \times \mu_\varepsilon(d\theta) \quad \text{colision angels}$$

Global Law of the colision angels

$$(\bar{\theta}_1, \dots, \bar{\theta}_n) \sim \prod_{k=1}^n |V_{T_k}^\varepsilon - v_k|^\gamma \times \mu_\varepsilon(d\theta_k)$$

## Main result.

Suppose that

$$f_0(dv) \neq \delta_a \quad \forall a \in R^2$$

1. Suppose also that

$$0 < \nu < \frac{1}{3}, \quad 1 > \gamma > \frac{2\nu(1 + \nu)}{1 - 3\nu}$$

Then for every  $t > 0$

$$f_t(dv) = \phi_t(v)dv$$

with  $\phi_t \in L^2(R^2, dv)$ .

2. Suppose that

$$0 < \nu < \frac{1}{4}, \quad 1 > \gamma > \frac{3\nu(2 + \nu)}{1 - 4\nu}$$

Then  $\phi_t$  continuous.

## 2 Integration by parts formulas

**Definition.** Let  $F$  be a random variable. We say that  $IP_\alpha(F)$  holds if there exists  $H_\alpha(F) \in \cap_{p \geq 1} L^p$  such that

$$E(\partial_\alpha \phi(F)) = E(\phi(F)H_\alpha(F)) \quad \forall \phi \in C_c^\infty(\mathbb{R}^d).$$

**Malliavin Calculus (In a Gaussian framework) :**

For a simple functional  $F_n$  one defines the Malliavin derivative  $DF_n$  and then take the extension : if

$$F_n \rightarrow F, \quad \text{in } L^2$$

Then one defines  $DF = \lim_n DF_n$ . Using such a differential calculus one builds up  $IP_\alpha(F)$ .

**Elementary Integration by parts** : Suppose that

$$F_n = f_n(\Theta_1, \dots, \Theta_n)$$

and

$$(\Theta_1, \dots, \Theta_n) \sim p(\theta_1, \dots, \theta_n) d\theta_1, \dots, d\theta_n.$$

Then, using **elementary integration by parts** one establishes

$$E(\partial_\alpha \phi(F_n)) = E(\phi(F_n) H_\alpha(F_n)) \quad \forall \phi \in C_c^\infty(R^d).$$



**Regularity of the law** : Suppose that  $IP_\alpha(F)$ ,  $|\alpha| \leq d$  hold. Then  $P_F(dx) = p_F(x)dx$ .

**Proof.**

$$|E(\partial_\alpha \phi(F))| = |E(\phi(F)H_\alpha(F))| \leq \|\phi\|_\infty E(|H_\alpha(F)|)$$

and this implies

$$P_F(dy) = p_F(y)dy \quad \text{and} \quad p_F \in C_b.$$

**Alternative approach** : One employes **elementary** integration by parts to bilt  $IP_\alpha(F_n)$  and then

$$\begin{aligned} |E(\partial_\alpha \phi(F))| &= \lim_n |E(\partial_\alpha \phi(F_n))| = \lim_n |E(\phi(F_n)H_\alpha(F_n))| \\ &\leq \|\phi\|_\infty \sup_n E(|H_\alpha(F_n)|). \end{aligned}$$

**Conclusion** : We need

- a)  $F_n \rightarrow F$  in law,
- b)  $\sup_n E(|H_\alpha(F_n)|) < \infty$ .

**In our framework**

$$\sup_n E(|H_\alpha(F_n)|) = \infty.$$

**Why?** We have

$$p(\bar{\theta}_1, \dots, \bar{\theta}_n) = \prod_{k=1}^n |V_{T_k}^\varepsilon - v_k|^\gamma \times \frac{1}{\varepsilon \vee |\theta_k|^{1+\nu}}$$
$$\ln p(\bar{\theta}_1, \dots, \bar{\theta}_n) = \sum_{k=1}^n (\gamma \ln |V_{T_k}^\varepsilon - v_k| - \ln \varepsilon \vee |\theta_k|^{1+\nu})$$

so

$$\partial_{\theta_i} \ln p = -\partial_{\theta_i} \ln \varepsilon \vee |\theta_i|^{1+\nu} + \sum_{k=i+1}^n \gamma \partial_{\theta_i} \ln |V_{T_k}^\varepsilon - v_k|$$

**Approach using the Fourier transform.** We have to prove that

$$\int_{\mathbb{R}} |\xi|^p |\hat{p}(\xi)| d\xi < \infty \quad \text{with} \quad \hat{p}(\xi) = E(\exp(i\xi F)).$$

We denote

$$\hat{p}_n(\xi) = E(\exp(i\xi F_n)), \quad \varepsilon_n = E|F - F_n|.$$

And we write

$$\begin{aligned} |\hat{p}(\xi)| &\leq |\hat{p}(\xi) - \hat{p}_n(\xi)| + |\hat{p}_n(\xi)| \leq |\xi| \varepsilon_n + \frac{1}{|\xi|^k} E(\partial_x^k \exp(i\xi F_n)) \\ &\leq |\xi| \varepsilon_n + \frac{1}{|\xi|^k} E(|H_k(F_n)|). \end{aligned}$$

**Toy example :**

$$\varepsilon_n = \frac{1}{n} \quad \text{and} \quad E(|H_k(F_n)|) = n^2.$$

Then

$$|\hat{p}(\xi)| \leq \frac{|\xi|}{n} + \frac{n^2}{|\xi|^k} \quad \forall \xi \in \mathbb{N}.$$

Equilibrium between  $n$  and  $\xi$  : We take

$$n = |\xi|^\rho$$

Then

$$|\widehat{p}(\xi)| \leq |\xi|^{1-\rho} + \frac{1}{|\xi|^{k-2\rho}} \quad \forall \xi, n \in N.$$

We want

$$1 - \rho = -(k - 2\rho) \quad \Leftrightarrow \rho = \frac{k + 1}{3}$$

Conclusion : if we choose

$$n(\xi) = |\xi|^{\frac{k+1}{3}}$$

then we obtain

$$|\widehat{p}(\xi)| \leq |\xi|^{\frac{-k+2}{3}}$$

so that

$$\int_R |\xi|^p |\widehat{p}(\xi)| d\xi < \infty \quad \forall p \in N.$$

**Conclusion** : We write

$$|\hat{p}_F(\xi)| \leq |\xi| E |F - F_n| + \frac{1}{|\xi|^k} E(|H_k(F_n)|)$$

and we have to estimate in an accurate way

**Approximation rate** :  $E |F - F_n| \rightarrow 0$ ,

**Blow up rate**  $|H_k(F_n)| \rightarrow \infty$ .

We we are lucky we get an **equilibrium**.

## Integration by parts : Toy example

We consider an one dimensional random variable

$$\Theta \sim p(\theta)d\theta$$

and a "functional"

$$F = f(\Theta).$$

Then

$$E(\phi'(F)) = E(\phi(F))$$

**Proof. Step1.**

$$\phi'(F) = \phi'(f(\Theta)) = \partial_{\Theta} (\phi(f(\Theta))) \times g(\Theta) \quad \text{with} \quad g(\Theta) = \frac{1}{f'(\Theta)}.$$

**Warning : non degeneracy :  $f'(\theta) \neq 0$ .**

Step 2.

$$\begin{aligned} E(\phi'(F)) &= \int \partial_{\theta} (\phi(f(\theta))) \times g(\theta) \times p(\theta) d\theta \\ &= - \int \phi(f(\theta)) \times \left( g'(\theta) \times p(\theta) + g(\theta) \times p'(\theta) \right) d\theta \\ &= - \int \phi(f(\theta)) \times \left( g'(\theta) + g(\theta) \times \frac{p'(\theta)}{p(\theta)} \right) \times p(\theta) d\theta \\ &= -E(\phi(F) \left( g'(\Theta) + g(\Theta) \times \frac{p'(\Theta)}{p(\Theta)} \right)) \end{aligned}$$

so that

$$H(F) = - \left( g'(\Theta) + g(\Theta) \times \partial_{\Theta} \ln p(\Theta) \right).$$