

# **Stein's method, logarithmic Sobolev and transport inequalities**

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This talk is mainly based on the material developed in the following two papers:

- I. Nourdin, G. Peccati and Y. Swan (2014): Entropy and the fourth moment phenomenon, *Journal of Functional Analysis* **266**, no. 5, 3170–3207.
- M. Ledoux, I. Nourdin and G. Peccati (2014): Stein's method, logarithmic Sobolev and transport inequalities. Submitted.

# **PART I: Stein's method**

Let  $F$  be a given real random variable with, say, mean zero and variance one.

Let  $N \sim N(0, 1)$ .

In many situations of interest, we may expect the law of  $F$  to be close of that of  $N$ , and one is interested in quantifying it. How to formalize this ?

**Stein's lemma.** Assume  $h : \mathbb{R} \rightarrow [0, 1]$  is continuous. Consider  $\varphi_h$  defined, for  $x \in \mathbb{R}$ , as

$$\begin{aligned}\varphi_h(x) &= e^{\frac{x^2}{2}} \int_{-\infty}^x (h(a) - E[h(N)]) e^{-\frac{a^2}{2}} da \\ &= -e^{\frac{x^2}{2}} \int_x^{\infty} (h(a) - E[h(N)]) e^{-\frac{a^2}{2}} da.\end{aligned}$$

Then  $\varphi_h$  is  $C^1$ , satisfies

$$\varphi_h'(x) - x\varphi_h(x) = h(x) - E[h(N)]$$

and is such that  $\|\varphi_h'\|_{\infty} \leq 2$ .

**A bound on the total variation distance.** One has

$$\begin{aligned}
 d_{TV}(F, N) &:= \sup_{A \in \mathcal{B}(\mathbb{R})} \left| P[F \in A] - P[N \in A] \right| \\
 &\leq \sup_{h: \mathbb{R} \rightarrow [0,1]} \left| E[h(A)] - E[h(N)] \right| \\
 &= \sup_{h: \mathbb{R} \rightarrow [0,1] \in C^0} \left| E[h(A)] - E[h(N)] \right| \quad (\text{Lusin}) \\
 &\leq \sup_{\varphi \in C^1: \|\varphi'\|_\infty \leq 2} \left| E[\varphi'(F)] - E[F\varphi(F)] \right| \quad (\text{Stein}).
 \end{aligned}$$

Now the question is: how to relate  $E[\varphi'(F)]$  and  $E[F\varphi(F)]$ ?

**Definition.** We say that  $\tau_F : \mathbb{R} \rightarrow \mathbb{R}$  is a *Stein factor* for  $F$  if

$$E[F\varphi(F)] = E[\tau_F(F)\varphi'(F)]$$

for all test function  $\varphi$ . The *Stein discrepancy* is

$$S(F|N) = \sqrt{E[(1 - \tau_F(F))^2]} = \sqrt{\text{Var}(\tau_F(F))}.$$

**Theorem.** One has  $d_{TV}(F, N) \leq 2S(F|N)$ .

**Examples:** 1. If  $F \sim N(0, 1)$  then  $\tau_F(x) = 1$  is a Stein factor for  $F$ . We then have  $S(F|N) = 0$ .

2. If  $F_n = \frac{1}{\sqrt{n}}(X_1 + \dots + X_n)$  with  $X_i$  iid, centered and unit variance, then  $\tau_{F_n} = \frac{1}{n} \sum_{i=1}^n E[\tau_X(X_i)|F_n]$  is a Stein factor for  $F_n$ , so that

$$S(F_n|N)^2 \leq \frac{1}{n} \text{Var}(\tau_X(X)).$$

3. If  $F = I_p(f)$  is a multiple integral of order  $p$ , then  $\tau_F = E[\langle DF, -DL^{-1}F \rangle | F]$  is a Stein factor for  $F$ , implying in turn that

$$S(F|N)^2 \leq \frac{p-1}{3p} (E[F^4] - 3).$$



**Multivariate extension.** Let  $F$  be a centered random vector of  $\mathbb{R}^d$ . Let  $N \sim N_d(0, \text{Id})$ .

**Definition.** 1) A measurable matrix-valued map on  $\mathbb{R}^d$

$$x \mapsto \tau_F(x) = \left\{ \tau_F^{ij}(x) : i, j = 1, \dots, d \right\}$$

is said to be a *Stein matrix* for  $F$  if  $\tau_F^{ij}(F) \in L^1(\Omega)$  for every  $i, j$  and, for every smooth  $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$ ,

$$E[F \cdot \nabla \varphi(F)] = E[\langle \tau_F(F), \text{Hess}(\varphi)(F) \rangle_{\text{HS}}],$$

with  $\langle A, B \rangle_{\text{HS}} = \sum_{i,j=1}^d a_{ij} b_{ij}$ .

2) The *Stein discrepancy* is  $S(F | N)$ , with

$$S^2(F | N) = E\|\tau_F(F) - \text{Id}\|_{\text{HS}}^2.$$

# **PART II: Logarithmic Sobolev inequality**

Let  $F$  be any random vector of  $\mathbb{R}^d$  whose law, noted  $\nu$ , is absolutely continuous (wrt Lebesgue).

Let  $N \sim N_d(0, \text{Id})$  and denote its law by  $\gamma$ , that is,  $d\gamma(x) = (2\pi)^{-d/2} e^{-|x|^2/2} dx$  on  $\mathbb{R}^d$ .

Let  $h = \frac{d\nu}{d\gamma}$ .

**Definitions.** 1) The *relative entropy* of  $F$  with respect to  $N$  is

$$\mathbf{H}(F | N) = \int_{\mathbb{R}^d} h \log h \, d\gamma (= \text{Ent}_\gamma(h)).$$

2) The *Fisher information* of  $F$  with respect to  $N$  is

$$\mathbf{I}(F | N) = \int_{\mathbb{R}^d} \frac{|\nabla h|^2}{h} \, d\gamma (= \mathbf{I}_\gamma(h)).$$

The classical **logarithmic Sobolev inequality** with respect to the standard Gaussian measure  $\gamma$  indicates that

$$\mathbf{H}(F | N) \leq \frac{1}{2} \mathbf{I}(F | N).$$

**The HSI inequality** (Ledoux, Nourdin, Peccati). One has

$$H(F | N) \leq \frac{1}{2} S^2(F | N) \log \left( 1 + \frac{I(F | N)}{S^2(F | N)} \right).$$

Since  $\log(1+x) \leq x$ , our inequality is a new improved form of the logarithmic Sobolev inequality.

# **PART III: Transport inequalities**

**Wasserstein quadratic distance.** Given two probability measures  $\nu$  and  $\mu$  on the Borel sets of  $\mathbb{R}^d$  whose marginals are square integrable, we define the *quadratic Wasserstein distance* between  $\nu$  and  $\mu$  as the quantity

$$W_2(\nu, \mu) = \inf_{\pi} \left( \int_{\mathbb{R}^d \times \mathbb{R}^d} |x - y|^2 d\pi(x, y) \right)^{1/2}$$

where the infimum runs over all probability measures  $\pi$  on  $\mathbb{R}^d \times \mathbb{R}^d$  with marginals  $\nu$  and  $\mu$ .

Let  $F$  be any random vector of  $\mathbb{R}^d$  whose law, noted  $\nu$ , is absolutely continuous (wrt Lebesgue).

Let  $N \sim N_d(0, \text{Id})$  and denote its law by  $\gamma$ , that is,  $d\gamma(x) = (2\pi)^{-d/2} e^{-|x|^2/2} dx$  on  $\mathbb{R}^d$ .

**Talagrand inequality:** One has the quadratic transportation cost inequality:

$$W_2(F, N)^2 \leq 2 H(F | N).$$

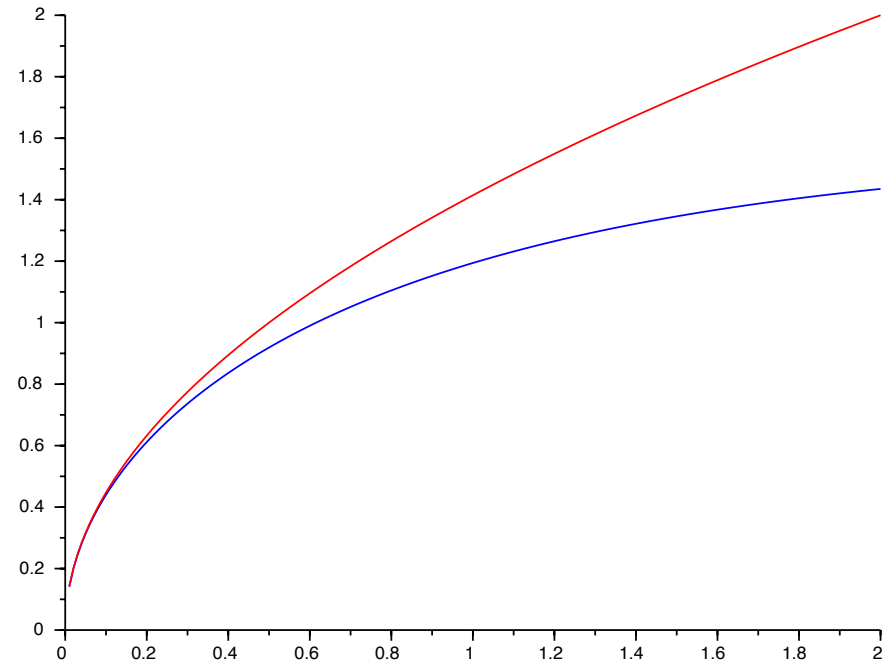
**WSH inequality** (Ledoux, Nourdin, Peccati):

$$W_2(F, N) \leq S(F | N) \arccos\left(e^{-\frac{H(F|N)}{S^2(F|N)}}\right).$$



**WSH implies Talagrand.** This is because

$$\arccos(e^{-r}) \leq \sqrt{2r}, \quad r \geq 0.$$



**Can we also recover the celebrated HWI inequality?**

**Otto and Villani:**

$$H(F|N) \leq W_2(F, N)\sqrt{I(F|N)} - \frac{1}{2}W_2(F, N)^2.$$

(It implies the log-Sobolev inequality since  $xy - \frac{1}{2}y^2 \leq \frac{1}{2}x^2$ .)

## Can we consider further distributions?

On the basis of the Gaussian example, we are able to address the issue of HSI inequalities for distributions on  $\mathbb{R}^d$ ,  $d \geq 1$ , that are not necessarily Gaussian.

Our basic ingredient is a semigroup approach *à la* Bakry-Emery. As such, one can deal with the family of invariant measures of second order differential operators.

These include gamma and beta distributions, as well as families of log-concave measures as illustrations.

But we will not give the details here!

**REST OF THE TALK:  
some elements of proof**

Denote by  $\tau_F$  a Stein matrix associated with  $F$ , having distribution  $d\nu = h d\gamma$ .

For every  $t > 0$ , set  $d\nu_t = P_t h d\gamma$ , and write  $v_t = \log P_t h$ , with  $(P_t)_{t \geq 0}$  the Ornstein-Uhlenbeck semigroup associated with  $d\gamma(x) = (2\pi)^{-d/2} e^{-|x|^2/2} dx$  on  $\mathbb{R}^d$ .

We will make intensive use of the following key inequalities.

**\* Integrated de Bruijn's formula**

$$H(F | N) = \text{Ent}_\gamma(h) = \int_0^\infty I_\gamma(P_t h) dt.$$

**\* Exponential decay of Fisher information**

$$I_\gamma(P_t h) \leq e^{-2t} I_\gamma(h) = e^{-2t} I(F | N).$$

\* **Linking I and S (crucial!)**

$$I_\gamma(P_t h) = \frac{e^{-2t}}{\sqrt{1 - e^{-2t}}} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \left[ (\tau_F(x) - \text{Id})y \cdot \nabla v_t(e^{-t}x + \sqrt{1 - e^{-2t}}y) \right] d\nu(x) d\gamma(y).$$

As a consequence,

$$I_\gamma(P_t h) = \int_{\mathbb{R}^d} |\nabla v_t|^2 d\nu \leq \frac{e^{-4t}}{1 - e^{-2t}} S^2(F | N).$$

\* **Exponential decay of Stein discrepancy**

$$S(\nu_t | \gamma) \leq e^{-2t} S(F | N) (\leq S(F | N)).$$

**Proof of the HSI inequality.** The idea is to bound the Fisher information  $I_\gamma(P_t h)$  differently for  $t$  around 0 and away from 0.

We can write, for every  $u > 0$ ,

$$\begin{aligned} H(F | N) &= \int_0^u I_\gamma(P_t h) dt + \int_u^\infty I_\gamma(P_t h) dt \\ &\leq I(F | N) \int_0^u e^{-2t} dt + S^2(F | N) \int_u^\infty \frac{e^{-4t}}{1 - e^{-2t}} dt \\ &\leq \frac{1}{2} I(F | N) (1 - e^{-2u}) + \frac{1}{2} S^2(F | N) (-e^{-2u} - \log(1 - e^{-2u})). \end{aligned}$$

Optimizing in  $u$  (set  $1 - e^{-2u} = r \in (0, 1)$ ) leads to the desired inequality

$$H(F | N) \leq \frac{1}{2} S^2(F | N) \log \left( 1 + \frac{I(F | N)}{S^2(F | N)} \right).$$

**Proof of the WSH inequality.** For any  $t \geq 0$ , recall  $d\nu_t = P_t h d\gamma$  (in particular,  $\nu_0 = \nu$  and  $\nu_t \rightarrow \gamma$  as  $t \rightarrow \infty$ ). The HSI inequality applied to  $\nu_t$  yields that

$$H(\nu_t | \gamma) \leq \frac{1}{2} S^2(\nu_t | \gamma) \log \left( 1 + \frac{I(\nu_t | \gamma)}{S^2(\nu_t | \gamma)} \right).$$

Now,  $S^2(\nu_t | \gamma) \leq S^2(\nu | \gamma)$  (due to the exponential decay of the Stein discrepancy) and  $r \mapsto r \log \left( 1 + \frac{s}{r} \right)$  is increasing for any fixed  $s$ . It follows that

$$H(\nu_t | \gamma) \leq \frac{1}{2} S^2(\nu | \gamma) \log \left( 1 + \frac{I(\nu_t | \gamma)}{S^2(\nu | \gamma)} \right).$$

By exponentiating both sides, this inequality is equivalent to:



$$\sqrt{I(\nu_t | \gamma)} \leq \frac{I(\nu_t | \gamma)}{S(\nu | \gamma) \sqrt{e^{\frac{2H(\nu_t | \gamma)}{S^2(\nu | \gamma)}} - 1}}.$$

But Otto and Villani showed in their pathbreaking paper that

$$\frac{d^+}{dt} W_2(\nu, \nu_t) \leq \sqrt{I(\nu_t | \gamma)}.$$

Moreover, one has the de Bruijn identity:

$$I(\nu_t | \gamma) = -\frac{d}{dt} H(\nu_t | \gamma).$$

Plugging everything together leads to

$$\begin{aligned} \frac{d^+}{dt} W_2(\nu, \nu_t) &\leq \sqrt{I(\nu_t | \gamma)} \leq -\frac{\frac{d}{dt} H(\nu_t | \gamma)}{S(\nu | \gamma) \sqrt{e^{\frac{2H(\nu_t | \gamma)}{S^2(\nu | \gamma)}} - 1}} \\ &= -\frac{d}{dt} \left( S(\nu | \gamma) \arccos \left( e^{-\frac{H(\nu_t | \gamma)}{S^2(\nu | \gamma)}} \right) \right). \end{aligned}$$

In other words,

$$\frac{d^+}{dt} \left( W_2(\nu, \nu_t) + S(\nu | \gamma) \arccos \left( e^{-\frac{H(\nu_t | \gamma)}{S^2(\nu | \gamma)}} \right) \right) \leq 0,$$

and we get the WSH inequality, namely

$$W_2(F, N) \leq S(F | N) \arccos \left( e^{-\frac{H(F|N)}{S^2(F|N)}} \right),$$

after integrating between  $t = 0$  and  $t = \infty$ .