

Probability in High Dimensions

Solutions to Exercises, Week 4

August 10, 2020

2.3 Poincaré Inequalities

Problem 2.8 (Exponential distribution). *Let $\mu(dx) = \mathbf{1}_{x \geq 0} e^{-x} dx$ be the one-sided exponential distribution. In this problem, we will derive two different (and not directly comparable) Poincaré inequalities for the distribution μ .*

a. *Show that*

$$\mathrm{Var}_\mu[f] \leq 2 \mathbf{E}[\xi |f'(\xi)|^2], \quad \xi \sim \mu.$$

b. *Show that*

$$\mathrm{Var}_\mu[f] \leq 4 \mathbf{E}[|f'(\xi)|^2], \quad \xi \sim \mu.$$

c. *Show that the Dirichlet form of the solution of the Cox-Ingersoll-Ross stochastic differential equation*

$$dX_t = 2(1 - X_t)dt + 2\sqrt{X_t}dB_t,$$

which is a Markov process on $\mathbb{R}_+$ with generator

$$\mathcal{L}f(x) = 2(1 - x)f'(x) + 2xf''(x),$$

yields the Poincaré inequality from part (a).

d. *Show that the Dirichlet form of the solution of the stochastic differential equation*

$$dX_t = -\mathrm{sign}(X_t)dt + \sqrt{2}dB_t,$$

which is a Markov process on $\mathbb{R}$ with generator

$$\mathcal{L}f(x) = -\mathrm{sign}(x)f'(x) + f''(x),$$

yields the Poincaré inequality from part (b).

Solution. a. By a simple polar coordinates argument the hint follows, that is $\xi \sim (X^2 + Y^2)/2$ with X, Y i.i.d. $N(0, 1)$.

Let

$$g(x, y) = f\left(\frac{x^2 + y^2}{2}\right).$$

Then

$$\partial_x g(x, y) = f'\left(\frac{x^2 + y^2}{2}\right)x, \quad \partial_y g(x, y) = f'\left(\frac{x^2 + y^2}{2}\right)y.$$

Apply the Gaussian Poincare inequality we get

$$\begin{aligned}
\text{var}_\mu(f) &= \text{var}[g(X, Y)] \\
&\leq \mathbb{E}[\|\nabla g(X, Y)\|^2] \\
&= \mathbb{E}[|f'(\xi)|^2 (x^2 + y^2)] \\
&= 2\mathbb{E}[|f'(\xi)|^2 \xi].
\end{aligned}$$

b. Notice that $g(0) = 0$. Then

$$\begin{aligned}
\text{Var}_\mu[f] &= \text{Var}_\mu[f - f(0)] \\
&\leq \mathbb{E}[(f(\xi) - f(0))^2] \\
&= \mathbb{E}[2f'(\xi)(f(\xi) - f(0))]
\end{aligned}$$

and by Cauchy-Schwarz

$$\leq 2\mathbb{E}[|f'(\xi)|^2]^{1/2} \mathbb{E}[(f(\xi) - f(0))^2]^{1/2}.$$

Therefore

$$\text{Var}_\mu[f] \leq \mathbb{E}[(f - f(0))^2] \leq 4\mathbb{E}[|f'(\xi)|^2],$$

and the result follows.

c. Let $g(x) = xf(x)f'(x)$. Then

$$g'(x) = f(x)f'(x) + xf'(x)^2 + xf(x)f''(x).$$

Thus

$$\begin{aligned}
\int e^{-x} xf(x)f'(x)dx &= \int [-e^{-x}]' xf(x)f'(x)dx \\
&= \int e^{-x} [xf(x)f'(x)]' dx \\
&= \int e^{-x} [ff' + xf'f' + xff''] dx.
\end{aligned}$$

Therefore

$$\begin{aligned}
\langle f, \mathcal{L}f \rangle &= 2 \int e^{-x} f(x) [f'(x) - xf'(x) + xf''(x)] dx \\
&= 2 \int e^{-x} [f(x)f'(x) - xf(x)f'(x) + xf(x)f''(x)] dx \\
&= 2 \int e^{-x} [xf(x)f'(x) - xf'(x)f'(x) - xf(x)f''(x) - xf(x)f'(x) + xf(x)f''(x)] dx \\
&= 2 \int e^{-x} [-xf'(x)f'(x)] dx \\
&= -2\mathbb{E}[\xi|f'(\xi)|^2].
\end{aligned}$$

d. Let $f : \mathbb{R} \rightarrow \mathbb{R}$. Then

$$\langle f, \mathcal{L}f \rangle = \frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} f(x) [-\operatorname{sign}(x) f'(x) + f''(x)] dx.$$

Then

$$\begin{aligned} & \frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} f(x) f''(x) dx \\ &= -\frac{1}{2} \int_{x=-\infty}^{\infty} \left[e^{-|x|} f(x) \right]' f'(x) dx \\ &= -\frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} [-\operatorname{sign}(x) f(x) + f'(x)] f'(x) dx, \end{aligned}$$

and thus

$$\begin{aligned} \langle f, \mathcal{L}f \rangle &= -\frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} f(x) \operatorname{sign}(x) f'(x) dx + \frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} \operatorname{sign}(x) f(x) f'(x) dx \\ &\quad - \frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} [f'(x)]^2 dx \\ &= -\frac{1}{2} \int_{x=-\infty}^{\infty} e^{-|x|} [f'(x)]^2 dx. \end{aligned}$$

Problem 2.9 (Dependent random signs). *Question is too long and is written in the notes.*

Solution. a.

b. For an n -vector $x = (x_1, \dots, x_n)$, write x^{-i} for the $n-1$ -vector $(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n)$. For small enough $h > 0$ and bounded continuous f we have

$$\begin{aligned} P_h f(x) &= \mathbb{E}[f(X_h) \mid X_0 = x] \\ &= \mathbb{E}[f(X_h) \mathbb{1}\{\text{no jump}\} \mid X_0 = x] \\ &\quad + \mathbb{E}[f(X_h) \mathbb{1}\{\text{one jump}\} \mid X_0 = x] + \mathbb{E}[f(X_h) \mathbb{1}\{\text{more than one jump}\} \mid X_0 = x] \\ &= f(x) e^{-nh} + e^{-nh} nh \frac{1}{n} \sum_{i=1}^n \int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_n) \mu(dz_i \mid x^{-i}) + o(h). \end{aligned}$$

Therefore

$$\begin{aligned} \frac{1}{h} [P_h f(x) - f(x)] &= f(x) \frac{e^{-nh} - 1}{h} + \frac{1}{n} e^{-nh} \sum_{i=1}^n \int f(x) \mu(dx_i \mid x) + o(1) \\ &= \sum \delta_i f + o(1). \end{aligned}$$

For the Dirichlet form notice that

$$\begin{aligned} & \langle f, \delta_i g \rangle \\ &= \int \mu(dx_1, \dots, dx_n) f(x_1, \dots, x_n) \int g(x_1, \dots, x_{i-1}, z_i, x_i, \dots, x_n) \mu(dz_i \mid x^{-i}) \\ &= \int \mu(dx^{-i}) \int \mu(dy_i \mid x^{-i}) \int \mu(dz_i \mid x^{-i}) f(x_1, \dots, x_{i-1}, y_i, x_i, \dots, x_n) g(x_1, \dots, x_{i-1}, z_i, x_i, \dots, x_n) \\ &= \langle \delta_i f, \delta_i g \rangle = \langle \delta_i f, g \rangle. \end{aligned}$$

- c. Say first that $i < j$. Given a vector x , let $x^{j,\pm 1}$ be the vector that agrees with x at all entries except perhaps at the j -th entry where $(x^{j,\pm 1})_j := \pm 1$.

$$\begin{aligned}
& \Delta_j \int f d\mu_i \\
&= \max_{x \in \{\pm 1\}^n} \left| \int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_{j-1}, 1, x_{j+1} \dots) \mu(dz_i \mid x^{j,+1}) \right. \\
&\quad \left. - \int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) \mu(dz_i \mid x^{j,-1}) \right| \\
&= \max_{x \in \{\pm 1\}^n} \left| \left[\int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_{j-1}, 1, x_{j+1} \dots) \mu(dz_i \mid x^{j,+1}) \right. \right. \\
&\quad \left. \left. - \int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) \mu(dz_i \mid x^{j,+1}) \right] \right. \\
&\quad \left. - \left[\int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) \mu(dz_i \mid x^{j,-1}) \right] \right. \\
&\quad \left. - \int f(x_1, \dots, x_{i-1}, z_i, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) \mu(dz_i \mid x^{j,+1}) \right] \\
&\leq \int \mu(dz_i \mid x^{j,+1}) |\Delta_j f| \\
&+ \max_{x \in \{\pm 1\}^n} \left| f(x_1, \dots, x_{i-1}, +1, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) [\mu(dz_i = +1 \mid x^{j,+1}) - \mu(dz_i = +1 \mid x^{j,-1})] \right. \\
&\quad \left. - f(x_1, \dots, x_{i-1}, -1, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) \times [\mu(dz_i = +1 \mid x^{j,+1}) - \mu(dz_i = +1 \mid x^{j,-1})] \right| \\
&\leq \Delta_j f \\
&+ \max_{x \in \{\pm 1\}^n} \left| [f(x_1, \dots, x_{i-1}, +1, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots) - f(x_1, \dots, x_{i-1}, -1, x_{i+1}, \dots, x_{j-1}, -1, x_{j+1} \dots)] \right. \\
&\quad \left. \max_{x \in \{\pm 1\}^n} [\mu(dz_i = +1 \mid x^{j,+1}) - \mu(dz_i = +1 \mid x^{j,-1})] \right| \\
&\leq \Delta_j f + \Delta_i f C_{ij}.
\end{aligned}$$

- d. Since $\mathcal{L}f = \lim h^{-1}(P_h f - f)$ we roughly have that $P_h f = f + h\mathcal{L}f + o(h)$. Then applying

$$\begin{aligned}
\Delta_j \left(f + \frac{t}{m} \mathcal{L}f \right) &= \Delta_j f + \frac{t}{m} \Delta_j \frac{P_{t/m} f - f}{t/m} + o\left(\frac{t}{m}\right) \\
&= \Delta_j P_{t/m} f + o\left(\frac{t}{m}\right) \\
&= e^{-t/m} \Delta_j f + \frac{t}{m} \sum_{i=1}^n \Delta_j \left(\int f d\mu_i - f \right) + o\left(\frac{t}{m}\right)
\end{aligned}$$

and using the inequality from the last part

$$\leq \left(1 - \frac{t}{m}\right) \Delta_j f + \sum_{i=1}^n \Delta_i f C_{ij} + o\left(\frac{t}{m}\right).$$

- e. Then using the previous result in vector notation

$$\Delta P_t f = \lim_{m \rightarrow \infty} \Delta(I + t\mathcal{L}t/m)^m f$$

$$\begin{aligned}
&= \lim_{m \rightarrow \infty} \Delta(I + t\mathcal{L}t/m) [(I + t\mathcal{L}t/m)^{m-1} f] \\
&\leq \lim_{m \rightarrow \infty} \Delta [(I + t\mathcal{L}t/m)^{m-1} f] [I - t(I - C)/m] \\
&\leq \lim_{m \rightarrow \infty} \Delta [(I + t\mathcal{L}t/m)(I + t\mathcal{L}t/m)^{m-2} f] [I - t(I - C)/m] \\
&\leq \lim_{m \rightarrow \infty} \Delta [(I + t\mathcal{L}t/m)^{m-2} f] [I - t(I - C)/m]^2 \\
&\leq \cdots \leq \lim_{m \rightarrow \infty} \Delta f [I - t(I - C)/m]^m = \Delta f e^{-t(I-C)}.
\end{aligned}$$

f. It is sufficient to get

$$\mathcal{E}(P_t f, P_t f) \leq \kappa(f) e^{2t(1-\|C\|_{sp})}$$

with $\kappa(f) = \Delta f \cdot (\Delta f)^T$, and then use the implication suggested in the question.

Using (b) we have $\mathcal{E}(P_t f, P_t f) = \sum_i \mu((\delta_i P_t f)^2)$ which is less than $\Delta P_t f (\Delta P_t f)^T$; this is in turn less than $\Delta f e^{-t(I-C)}$ by (e). Hence we have

$$\begin{aligned}
\mathcal{E}(P_t f, P_t f) &\leq \Delta f e^{-t(I-C)} e^{-t(I-C)^T} (\Delta f)^T \leq \Delta f e^{-t(I-C)} e^{-t(I-C)^T} (\Delta f)^T \\
&\leq \Delta f (\Delta f)^T \|e^{-t(I-C)} e^{-t(I-C)^T}\|_{sp}.
\end{aligned}$$

First note that the eigenvalues of e^A are the exponentials of the eigenvalues of A , and that the eigenvalues of e^{-A} are the multiplicative inverses of the eigenvalues of e^A . Hence

$$\lambda_{\max}(e^{-A}) = \lambda_{\min}(e^A) = e^{\lambda_{\min} A} = e^{-\lambda_{\max}(-A)}.$$

In particular, $\lambda_{\max}(e^{-t(I-C)}) = \exp\{-t\lambda_{\max}(C - I)\} = \exp\{-t(\|C\|_{sp} - 1)\} = e^{t(1-\|C\|_{sp})}$. Then using that $\|AB\|_{sp} \leq \|A\|_{sp} \|B\|_{sp}$ (which can be seen directly from the definition of eigenvalues), we have the required result.