

M2RI Asymptotic Statistics Exercices

Exercice n° 1

Let P a measure on \mathbb{R} . F_1, F_2 two sets of functions from $\mathbb{R} \rightarrow \mathbb{R}$ s.t.
 $\forall f \in F_1 \cup F_2, \int |f| dP < \infty$.

Show that $\forall \epsilon > 0, d_{L^1}(F_1 + F_2, L^1(P), 2\epsilon)$
 $\leq d_{L^1}(F_1, L^1(P), \epsilon) d_{L^1}(F_2, L^1(P), \epsilon)$.

Solution Let $f = f_1 + f_2 \in F_1 + F_2$. There exists $[P_1, \nu_1]$ (resp $[P_2, \nu_2]$)
a bracket containing f_1 (resp f_2) s.t. $\int (\nu_1 - P_1) dP$ (resp. $\int (\nu_2 - P_2) dP$)
 $< \epsilon$.

Then $[P_1 + P_2, \nu_1 + \nu_2]$ is a bracket for $f_1 + f_2$.

Furthermore, $\int (\nu_1 + \nu_2) - (P_1 + P_2) dP \leq \int \nu_1 - P_1 dP + \int \nu_2 - P_2 dP$
 $\leq 2\epsilon$.

Exercice n° 2 $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Unif}([-\epsilon, \epsilon])$.

s.t. $\prod_{i=1}^n (1 + X_i/\sqrt{n})$ converges in distribution and determine the limit.

Hint: we can use that $| \log(1 + X_i/\sqrt{n}) - X_i/\sqrt{n} + \frac{1}{2} X_i^2/n | \leq a_i/n^{3/2}$

where $a_i \geq 0$ and $\max_{1 \leq i \leq n} a_i = O(1)$. (To show if used).

Soln- First, let's apply Taylor-Lagrange to $f(x) = \ln(1+x)$

in 0. $f'(x) = \frac{1}{1+x}$; $f''(x) = \frac{-1}{(1+x)^2}$; $f'''(x) = \frac{2}{(1+x)^3}$

$$\forall i, \exists v \in (0, \frac{X_i}{\sqrt{n}}) \text{ s.t. } \log\left(1 + \frac{X_i}{\sqrt{n}}\right) = \frac{X_i}{\sqrt{n}} - \frac{1}{2} \frac{X_i^2}{n} + \frac{1}{6} \frac{2}{(1+v)^2} \frac{X_i^3}{n^{3/2}}$$

furthermore, since a.s. $|X_i| < 1$,

$$\text{a.s. } \left| \frac{1}{6} \frac{2}{(1+v)^2} X_i^3 \right| \leq \frac{1}{3} \frac{1}{\left(1 - \frac{1}{\sqrt{n}}\right)^2} = O(1).$$

$$\begin{aligned} \text{Then, } \log\left(\prod_{i=1}^n \left(1 + \frac{X_i}{\sqrt{n}}\right)\right) &= \sum_{i=1}^n \log\left(1 + \frac{X_i}{\sqrt{n}}\right) \\ &= \sum_{i=1}^n \frac{X_i}{\sqrt{n}} - \frac{1}{2} \frac{X_i^2}{n} + R_i \\ &\quad \text{with } |R_i| \leq \frac{C}{n^{3/2}} \text{ a.s.} \end{aligned}$$

$$\text{CLT: } \sum_{i=1}^n \frac{X_i}{\sqrt{n}} \xrightarrow{\mathcal{L}} \mathcal{N}(0, \sigma^2) \quad \sigma^2 = \text{Var}(X_i) = \frac{1}{3}$$

$$\begin{aligned} \text{LLN: } - \sum_{i=1}^n \frac{X_i^2}{n} &\xrightarrow{\mathbb{P}} - \mathbb{E}(X_i^2) = -\frac{1}{3} \\ \sum_{i=1}^n R_i &\xrightarrow{\text{a.s.}} 0 \end{aligned}$$

$$\begin{aligned} \text{So, Statist.} \Rightarrow \log\left(\prod_{i=1}^n \left(1 + \frac{X_i}{\sqrt{n}}\right)\right) &\xrightarrow{\text{+CLT}} \mathcal{N}\left(-\frac{1}{6}, \sigma^2\right) \\ &= \mathcal{N}\left(-\frac{1}{6}, \frac{1}{3}\right). \end{aligned}$$

Exercice n° 3:

$(X_1, Y_1), \dots, (X_n, Y_n)$ iid s.t. X_i v.l. on \mathbb{R} with density $f > 0$ continuous and $Y_i = X_i + Z_i$, $Z_i \perp X_i$ and $Z_i \sim \mathcal{N}(0, 1)$.

Assumption: $\mathbb{E}(|X_i|) < \infty$.

$$\hat{\theta} \in \arg \min_{\theta \in [\frac{1}{2}, 2]} \sum_{i=1}^n |Y_i - \theta X_i|$$

1) Show that $\theta \mapsto E(|Y_i - \theta X_i|)$ is continuous on $[\frac{1}{2}, 2]$ and admits a unique minimiser at $\theta_0 = 1$.

Hint: We can use (without proof) that $E(|W+a|) > E(|W|)$ whenever $a \neq 0$ and $W \sim \mathcal{U}(0,1)$.

2) Show that

$$\sup_{\theta \in [\frac{1}{2}, 2]} \left| \frac{1}{n} \sum_{i=1}^n |Y_i - \theta X_i| - E(|Y_i - \theta X_i|) \right| \xrightarrow{P} 0.$$

3) Conclude that $\hat{\theta}_n \xrightarrow{P} 1$.

Solution:

1) $|Y_i - \theta X_i| \leq |Y_i| + |\theta| |X_i| \leq |Y_i| + 2 |X_i|$

and $E(|Y_i| + 2 |X_i|) < \infty$.

So by continuity under the integral, $\theta \mapsto E(|Y_i - \theta X_i|)$ is continuous.

furthermore, if $S \in [-\frac{1}{2}, 1)$,

$$E(|Y_i - (1+S)X_i|) = E(|X_i + 2 - (1+S)X_i|)$$

$$= E(|2 - SX_i|) = E(E(|2 - SX_i| | X_i))$$

$$> E(|2, 1|) \text{ if } S \neq 0$$

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So it is minimal for $S = 0$ and this minimiser is unique.

2) $F = \{f_\theta(x, y) = |y - \theta x| : \theta \in [\frac{1}{2}, 2]\}$.

$$G(x, y) = |y| + 2|x|$$

$$\forall f \in F, \int |f| dP \leq \int |G| dP < \infty.$$

Lipschitz property $|f_\theta(x, y) - f_{\theta'}(x, y)|$
 $= ||y - \theta x| - |y - \theta' x||$
 $\leq |\theta - \theta'| |x|$

Brinkots Construction: Let $\varepsilon > 0$, $\Delta := \frac{\varepsilon}{4E(|X|)}$

$$\theta_k = \frac{1}{2} + k\Delta, \quad k = 0, 1, \dots, K \quad K \leq \frac{2 - \frac{1}{2}}{\Delta}$$

$$[l_k, u_k] := [f_{\theta_k} - \Delta|x|, f_{\theta_k} + \Delta|x|].$$

- $\mathbb{P}(|\theta - \theta_k| \leq \Delta, f_\theta \in [l_k, u_k])$.

- $\int |u_k - l_k| dP = E(2\Delta|X|) = \varepsilon/2 < \varepsilon$

$$\text{So, } d_{\varepsilon}(F, L', \varepsilon) \leq K+1 < +\infty.$$

So F is Glivenko Cantelli.

3) The consistency theorem applies and $\hat{\theta}_n \xrightarrow{P} \theta_0 = 1$.