

# Chapter 1

## Empirical Process Theory - Exercises

Shu Shen

**Exercise 18.1** Let  $g(x, \theta) = \mathbb{I}(x \leq \theta)$  for  $\theta \in [0, 1]$  and assume  $X \sim F = \mathcal{U}[0, 1]$ . Let  $N_1(\varepsilon, F)$  be  $L_1$  packing numbers.

(a) Show that  $N_1(\varepsilon, F)$  equal the packing numbers constructed with respect to the Euclidean metric  $d(\theta_1, \theta_2) = |\theta_1 - \theta_2|$ .

(b) Verify that  $N_1(\varepsilon, F) \leq \lceil \frac{1}{\varepsilon} \rceil$ .

**Solution. 18.1**

(a) When  $\theta_1 \geq \theta_2$ ,

$$\begin{aligned} \|g(x, \theta_1) - g(x, \theta_2)\|_{F,1} &= E_F [\|g(x, \theta_1) - g(x, \theta_2)\|] \\ &= \int \|g(x, \theta_1) - g(x, \theta_2)\| dF \\ &= \int (\mathbb{I}(x \leq \theta_1) - \mathbb{I}(x \leq \theta_2)) dF \\ &= \int \mathbb{I}(x \leq \theta_1) dF - \int \mathbb{I}(x \leq \theta_2) dF \\ &= F(\theta_1) - F(\theta_2) \\ &= \theta_1 - \theta_2 = d(\theta_1, \theta_2) \end{aligned}$$

When  $\theta_1 < \theta_2$ , a similar calculation is given  $d(\theta_1, \theta_2)$ .

(b) Set  $N = \frac{1}{\varepsilon}$  as an integer, and let  $\theta_j = \frac{j}{N}$  be the  $\frac{j}{N}$ -th quantile.

$\theta_j$  can be packed into  $[0, 1]$  with each pair satisfying  $d(\theta_{j+1}, \theta_j) = |\theta_{j+1} - \theta_j| = \varepsilon$

Thus  $N_1(\varepsilon, F) \leq \lceil \frac{1}{\varepsilon} \rceil$ .

**Exercise 18.3** Define  $\nu_n(\theta) = \frac{1}{\sqrt{n}} \sum_{i=1}^n X_i \mathbb{I}(X_i \leq \theta)$  for  $\theta \in [0, 1]$  where  $E[X] = 0$  and  $E[X^2] = 1$ .

(a) Show that  $\nu_n(\theta)$  is stochastically equicontinuous.

(b) Find the stochastic process  $\nu(\theta)$  which has asymptotic finite dimensional distributions of  $\nu_n(\theta)$ .

(c) Show that  $\nu_n \xrightarrow{d} \nu$ .

**Solution. 18.3**

(a)  $\nu_n(\theta) = \frac{1}{\sqrt{n}} \sum_{i=1}^n (X_i \mathbb{I}(X_i \leq \theta) - E[X_i \mathbb{I}(X_i \leq \theta)])$ . Suppose  $X_i \sim F$ , where  $F$  can be either continuous or discrete. We can compute the bracket integral to check the complexity of the function  $x \mathbb{I}(x \leq \theta)$  for  $\theta \in [0, 1]$ . Set  $N = \varepsilon^{-1}$  as an integer, and let  $\theta_j = F^{-1}\left(\frac{j}{N}\right)$  be the  $\frac{j}{N}$ -th quantile. We construct a bracket with the lower bound function  $l_j(x) = x \mathbb{I}(x \leq \theta_j)$  and the upper bound function  $u_j(x) = x \mathbb{I}(x \leq \theta_{j+1})$ . Obviously, for any  $\theta \in [0, 1]$ , there exists a pair  $[l_j, u_j]$  such that  $l_j(x) \leq x \mathbb{I}(x \leq \theta) \leq u_j(x)$ . The size of such a bracket is

$$\begin{aligned} \|u_j(x) - l_j(x)\|_{F,2}^2 &= E \left[ [x \mathbb{I}(\theta_j \leq x \leq \theta_{j+1})]^2 \right] \\ &\leq E [\mathbb{I}(\theta_j \leq x \leq \theta_{j+1})] \\ &= F(\theta_{j+1}) - F(\theta_j) \leq \varepsilon \end{aligned}$$

where the inequality holds as  $x \in [\theta_j, \theta_{j+1}] \subseteq [0, 1]$ . Thus  $N_{[\cdot]}(\varepsilon, L_2(F)) \leq N = \varepsilon^{-1} = O(\varepsilon^{-1})$ , we have  $J_{[\cdot]}(1, L_2(F)) < \infty$ . We invoke the textbook's Theorem 18.4 to establish stochastic equicontinuity.

(b) Let  $u_i(\theta) = X_i \mathbb{I}(X_i \leq \theta) - E[X_i \mathbb{I}(X_i \leq \theta)]$ . Denote  $m(\theta) := E[X^2 \mathbb{I}(X \leq \theta)]$  and  $\mu(\theta) = E[X \mathbb{I}(X \leq \theta)]$ .

$$\sigma^2(\theta) := \text{var}[u_i(\theta)] = E[X^2 \mathbb{I}(X \leq \theta)] - (E[X \mathbb{I}(X \leq \theta)])^2 = m^2(\theta) - \mu^2(\theta)$$

and

$$\begin{aligned}
\sigma(\theta_1, \theta_2) &:= \text{Cov}(u_i(\theta_1), u_i(\theta_2)) \\
&= \text{Cov}(X\mathbb{I}(X \leq \theta_1), X\mathbb{I}(X \leq \theta_2)) \\
&= E[X^2\mathbb{I}(X \leq \theta_1)\mathbb{I}(X \leq \theta_2)] - E[X\mathbb{I}(X \leq \theta_1)]E[X\mathbb{I}(X \leq \theta_2)] \\
&= E[X^2\mathbb{I}(X \leq \theta_1 \wedge \theta_2)] - E[X\mathbb{I}(X \leq \theta_1)]E[X\mathbb{I}(X \leq \theta_2)] \\
&= m^2(\theta_1 \wedge \theta_2) - \mu(\theta_1)\mu(\theta_2).
\end{aligned}$$

Since  $u_i(\theta)$  is i.i.d, we have the CLT

$$\begin{pmatrix} v_n(\theta_1) \\ v_n(\theta_2) \end{pmatrix} = \sqrt{n} \sum_{i=1}^n \begin{pmatrix} X\mathbb{I}(X \leq \theta_1) \\ X\mathbb{I}(X \leq \theta_2) \end{pmatrix} \xrightarrow{d} N\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma^2(\theta_1) & \sigma(\theta_1, \theta_2) \\ \sigma(\theta_1, \theta_2) & \sigma^2(\theta_2) \end{pmatrix}\right)$$

The joint distribution is satisfied for any finite  $\theta_1, \theta_2, \dots, \theta_m$ , so

$$(v_n(\theta_1), v_n(\theta_2), \dots, v_n(\theta_m)) \xrightarrow{d} (v(\theta_1), v(\theta_2), \dots, v(\theta_m))$$

for every finite set  $\theta_1, \theta_2, \dots, \theta_m \in \theta \in [0, 1]$

(c) Given (a) and (b), we invoke the textbook's Theorem 18.3 to establish  $v_n \xrightarrow{d} v$ .