## Non-Paramatric Statistics Exercise 3

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## Exercise 1.5

Let  $(X, \mathcal{A})$  be a measurable space and  $\mu$  be a probability measure on X. Moreover, let  $\mathbf{P}$  be a probability measure on X that has a  $\mu$ -density  $h: X \to [0, \infty)$ , and let  $L_{\text{dens}}: X \times \mathbb{R} \to [0, \infty)$  be the corresponding loss function. Compute the excess  $L_{\text{dens}}$ -risks for both  $\mathbf{P}$  and  $\mu$ . Which one is better suited for capturing the intuitive goal of density estimation?

Solution:

$$h$$
 is a  $\mu$ -density of  $\mathbf{P} \Rightarrow \forall A \in \mathcal{A} : \mathbf{P}(A) = \int_A h \, \mathrm{d}\mu$ .

Recall from lecture:  $\forall x \in X, t \in \mathbb{R} : L_{dens}(x,t) = |h(x) - t|$ .

 $\forall x \in X : L_{\text{dens}}(x, h(x)) = 0 \implies \text{The Bayes Risk of } L_{\text{dens}} \text{ related to any measure of X is 0.}$  (\*) Suppose now we have an estimated density h' of h.

This means that: 
$$h' \geq 0 \land \int_X h' d\mu = 1 \land \forall A \in \mathcal{A} : \mathbf{P}'(A) := \int_A h' d\mu$$
.

As the excess  $L_{\text{dens}}$ -risks of  $\mu$ , we obtain:

$$\mathcal{R}_{L_{\mathrm{dens}},\mu}(h') - \mathcal{R}^*_{L_{\mathrm{dens}},\mu} \stackrel{(*)}{=} \mathcal{R}_{L_{\mathrm{dens}},\mu}(h') \stackrel{\mathrm{Def.}}{=} \int_X L_{\mathrm{dens}}(x,h'(x)) \,\mathrm{d}\mu(x) = \int_X |h-h'| \,\mathrm{d}\mu$$

As for the excess  $L_{\text{dens}}$ -risks of  $\mathbf{P}$ , we obtain:

$$\mathcal{R}_{L_{\mathrm{dens}},\mathbf{P}}(h') - \mathcal{R}_{L_{\mathrm{dens}},\mathbf{P}}^* \stackrel{(*)}{=} \mathcal{R}_{L_{\mathrm{dens}},\mathbf{P}}(h') \stackrel{\mathrm{Def.}}{=} \int_X L_{\mathrm{dens}}(x,h'(x)) \, \mathrm{d}\mathbf{P}(x) = \int_X |h - h'| \, \mathrm{d}\mathbf{P} = \int_X |h - h'| \, h \, \mathrm{d}\mu.$$

The excess risk relating to  $\mathbb{P}$ is more suitable for capturing the intuitive goal of density estimation because we can see that compared with that of excess risk under  $\mathbb{P}$ , the integrand of excess risk under  $\mu$  has an additional function h, and thus, the value of the integrand will be larger for  $x \in X : h(x) > 1$ , and smaller for  $x \in X : h(x) < 1$ . This means that the risk penalty for the part, where h is "relatively large", is larger, and the risk penalty for the part, where h is "relatively small", is then smaller.

## Exercise 2.1

Let  $X \neq \emptyset$  and  $\mathfrak{A} = (A_j)_{j \in J}$  be an at most countable partition of X. Describe the space  $\mathcal{L}_0(X)$  of measurable functions  $X \to \mathbb{R}$  for the corresponding  $\sigma$ -algebra  $\mathcal{A} := \sigma(\mathfrak{A})$ .

Solution:

We construct  $\Sigma := \{ S \subseteq X | \exists I \subseteq J : S = \bigcup_{i \in I} A_i \}.$ 

 $\mathfrak{A} = (A_j)_{j \in J}$  is a countable partition of  $X \Leftrightarrow X = \dot{\bigcup}_{j \in J} A_j$ .

 $\Rightarrow$  A simple verification of the axioms of  $\sigma$ -algebra shows that  $\Sigma$  is a  $\sigma$ -algebra containing  $\mathfrak{A}$ .

Meanwhile it is also easy to see that 
$$\Sigma \subseteq \sigma(\mathfrak{A}) \Rightarrow \Sigma = \sigma(\mathfrak{A})$$
. (i)

Moreover, recall the following theorem from measure theory:

Let  $(\Omega_1, \mathcal{F}_1)$ ,  $(\Omega_2, \mathcal{F}_2)$  be messable spaces and  $f: \Omega_1 \to \Omega_2$ .

Also let  $(M_n)_{n\in\mathbb{N}}$  a sequence from  $\mathcal{F}_1$  and at the same time a partition of  $\Omega_1$ .

For  $\forall n \in \mathbb{N}$  denote  $\mathcal{F}_1 \mid M_n := \{A \cap M_n | A \in \mathcal{F}_1\}.$ 

Then it holds that: f is  $\mathcal{F}_1$ - $\mathcal{F}_2$ -measurable  $\Leftrightarrow \forall n \in \mathbb{N} : f|_{M_n}$  is  $\mathcal{F}_1|M_n$ - $\mathcal{F}_2$ -measurable.

 $(A_j)_{j\in J}$  is paarwise disjoint  $\Rightarrow \forall j\in J: \mathcal{A}|A_j=\{\emptyset,A_j\}.$ 

$$\Rightarrow \forall j \in J \,\forall f \in \mathcal{L}_0(X) : f|_{A_j}^{-1}(\mathcal{B}) \in \mathcal{A}|A_j = \{\emptyset, A_j\}. \tag{ii}$$

We also know that  $\forall c \in \mathbb{R} : \{c\} \in \mathcal{B}$ .

With (ii) we can then deduce that 
$$\forall j \in J \ \forall f \in \mathcal{L}_0(X) : f|_{A_i}^{-1}(\{c\}) \in \{\emptyset, A_j\}.$$
 (iii)

Combining all considerations , we conclude that:

$$\mathcal{L}_{0}(X) \stackrel{\text{Def}}{=} \{ f : X \to \mathbb{R} \mid f^{-1}(\mathcal{B}) \subseteq \mathcal{A} \}$$

$$\stackrel{(i)}{=} \{ f : X \to \mathbb{R} \mid \forall B \in \mathcal{B} \exists I_{B} \subseteq J : f^{-1}(B) = \bigcup_{i \in I_{B}} A_{i} \}$$

$$\stackrel{(ii)}{=} \{ f : X \to \mathbb{R} \mid \forall j \in J : f|_{A_{j}}^{-1}(\mathcal{B}) = \emptyset \ \lor \ f|_{A_{j}}^{-1}(\mathcal{B}) = A_{j} \}$$

$$\stackrel{(iii)}{=} \{ f : X \to \mathbb{R} \mid f = \sum_{i \in I} c_{i} \mathbb{1}_{A_{i}} \}.$$