

## Homework 6: Hypothesis Testing - II

CR07: Selected Topics Information Theory (Fall 2019). ENS de Lyon  
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Consider two probability spaces  $(\mathcal{X}^n, \mathcal{F}, P_0)$  and  $(\mathcal{X}^n, \mathcal{F}, P_1)$ , with  $\mathcal{X}$  a discrete set such that  $|\mathcal{X}| = t < \infty$ ,  $n \in \mathbb{N}$  and  $\mathcal{F}$  the largest  $\sigma$ -algebra containing  $\mathcal{X}^n$ . Consider also two probability spaces  $(\mathcal{X}, \mathcal{G}, Q_0)$  and  $(\mathcal{X}, \mathcal{G}, Q_1)$ , where  $\mathcal{G}$  is the largest  $\sigma$ -algebra containing  $\mathcal{X}$ . Within this context, consider the hypotheses

$$H_0 : X \sim P_0 \tag{1a}$$

$$H_1 : X \sim P_1, \tag{1b}$$

where for all  $\mathbf{x} = (x_1, x_2, \dots, x_n) \in \mathcal{X}^n$  and for all  $i \in \{0, 1\}$ ,

$$P_i(\{\mathbf{x}\}) = \prod_{t=1}^n Q_i(\{x_t\}). \tag{2}$$

For all  $(i, j) \in \{0, 1\}^2$ , let  $C_{ij} > 0$  be the cost of accepting hypothesis  $H_i$ , when the ground-truth is hypothesis  $H_j$ . Let also  $\pi_i \in [0, 1]$  be the *a priori* probability with which hypothesis  $H_i$  is accepted. Hence, given a test  $T : \mathcal{X}^n \rightarrow \{0, 1\}$  the Bayesian risk is

$$R(T) = \pi_0 (C_{00}P_0(\mathcal{X}_0) + C_{10}P_0(\mathcal{X}_1)) + (1 - \pi_0) (C_{01}P_1(\mathcal{X}_0) + C_{11}P_1(\mathcal{X}_1)). \tag{3}$$

The test that minimizes the Bayesian risk is such that for all  $\mathbf{x} \in \mathcal{X}^n$ :

$$T_n(\mathbf{x}) = \mathbb{1}_{\{\mathbf{x} \in \mathcal{X}_1\}}, \tag{4}$$

where

$$\mathcal{X}_1 = \left\{ \mathbf{x} \in \mathcal{X}^n : \frac{P_1(\{\mathbf{x}\})}{P_0(\{\mathbf{x}\})} > \frac{\pi_0 (C_{10} - C_{00})}{(1 - \pi_0) (C_{01} - C_{11})} \right\}. \tag{5}$$

- i. Prove that the acceptance region of hypothesis  $H_1$ , i.e., the set  $\mathcal{X}_1$  in (5), can be written as:

$$\mathcal{X}_1 = \left\{ \mathbf{x} \in \mathcal{X}^n : D(P_{\mathbf{x}} \| Q_0) - D(P_{\mathbf{x}} \| Q_1) > \frac{1}{n} \log \left( \frac{\pi_0 (C_{10} - C_{00})}{(1 - \pi_0) (C_{01} - C_{11})} \right) \right\}. \tag{6}$$

2. Prove that the probability of false alarm  $P_0(\mathcal{X}_1)$  satisfies:

$$\lim_{n \rightarrow \infty} -\frac{1}{n} \log P_0(\mathcal{X}_1) = D(P^*||Q_0), \quad (7)$$

where  $P^* \in \Delta(\mathcal{X})$  satisfies

$$\begin{aligned} D(P^*||Q_0) &= \min_{P \in \Delta(\mathcal{X})} D(P||Q_0) \\ \text{s.t. } & D(P||Q_0) - D(P||Q_1) > \frac{1}{n} \log \left( \frac{\pi_0(C_{10}-C_{00})}{(1-\pi_0)(C_{01}-C_{11})} \right). \end{aligned} \quad (8)$$

3. What is the exact probability measure  $P^*$  in (7)?  
 4. Prove that the probability of misdetection  $P_1(\mathcal{X}_0)$  satisfies:

$$\lim_{n \rightarrow \infty} -\frac{1}{n} \log P_1(\mathcal{X}_0) = D(Q^*||Q_1), \quad (9)$$

where  $Q^* \in \Delta(\mathcal{X})$  satisfies

$$\begin{aligned} D(Q^*||Q_1) &= \min_{Q \in \Delta(\mathcal{X})} D(Q||Q_1) \\ \text{s.t. } & D(Q||Q_0) - D(Q||Q_1) < \frac{1}{n} \log \left( \frac{\pi_0(C_{10}-C_{00})}{(1-\pi_0)(C_{01}-C_{11})} \right). \end{aligned} \quad (10)$$

5. What is the exact probability measure  $Q^*$  in (9)?  
 6. Provide an upper-bound on the Bayesian risk in (3) of the optimal test in (4) in the case in which the number of observations is finite ( $n < \infty$ ).  
 7. Provide an upper-bound on the Bayesian risk in (3) of the optimal test in (4) in the case in the asymptotic regime of observations ( $n \rightarrow \infty$ ). HINT: provide the largest achievable exponent of the Bayesian risk.  
 8. What is the impact of the prior  $(\pi_0, 1 - \pi_0)$  in the upper-bounds on the Bayesian risk in both the asymptotic and non-asymptotic regimes?  
 9. Prove that under certain conditions, the biggest achievable exponent of the Bayesian risk is identical to the Chernoff information  $C(Q_0, Q_1)$ , that is,

$$C(Q_0, Q_1) = - \min_{\lambda \in [0,1]} \log \left( \sum_{a \in \mathcal{X}} (Q_0(a))^\lambda (Q_1(a))^{1-\lambda} \right). \quad (11)$$

10. Describe a few applications of the Chernoff Information in computer science, e.g., image segmentation, edge detection, etc.