

Spring 2016  
**6.441 - Information Theory**  
**Homework 6**  
Due: Thur, Mar 17, 2016 (in class)  
Prof. Y. Polyanskiy

## 1 Reading (optional)

1. Read [1, Chapters 11,12]

## 2 Exercises

**NOTE:** Each exercise is 10 points. Only 3 exercises per assignment will be graded. If you submit more than 3 solved exercises please indicate which ones you want to be graded.

- 1 Consider a binary hypothesis testing problem:

$$H_0 : X_1, \dots, X_n \text{ i.i.d. with } X_j \sim P_0 \quad (1)$$

$$H_1 : X_1, \dots, X_n \text{ i.i.d. with } X_j \sim P_1 \quad (2)$$

For  $n = 1$  the hypothesis testing region

$$\mathcal{R}(P_0, P_1) \triangleq \bigcup_{P_{Z|X_1}} (\mathbb{P}_0[Z = 0], \mathbb{P}_1[Z = 0])$$

is shown on Fig. 1. Thus, if one wants reliability  $\alpha \triangleq \pi_{0|0} = 0.99$  then the smallest probability of error  $\beta \triangleq \pi_{0|1}$  is  $\approx 0.993$ . By taking more observations (i.e.,  $n \gg 1$ ) the quality of the test can be dramatically improved. Estimate minimal  $n$  required to achieve

$$\pi_{0|0} \geq 0.99 \quad (3)$$

$$\pi_{0|1} \leq 10^{-40}. \quad (4)$$

Describe behavior of the region  $\mathcal{R}(P_0^n, P_1^n)$  as  $n \rightarrow \infty$ .

- 2 Let  $P$  be the uniform distribution on the interval  $[0, 1]$ . Let  $Q$  be the equal mixture of the uniform distribution on  $[0, 1/2]$  and the point mass at 1.

1. Compute the the region  $\mathcal{R}(P, Q)$ .
2. Explicitly describe the tests that achieve the optimal boundary  $\beta_\alpha(P, Q)$ .

- 3 Recall the total variation distance

$$\text{TV}(P, Q) \triangleq \sup_E (P[E] - Q[E]).$$

1. Prove that

$$\text{TV}(P, Q) = \sup_{0 \leq \alpha \leq 1} \{\alpha - \beta_\alpha(P, Q)\}.$$

Explain how to read the value  $\text{TV}(P, Q)$  from the region  $\mathcal{R}(P, Q)$ .

2. (Bayesian criteria) Fix a prior  $\pi = (\pi_0, \pi_1)$  such that  $\pi_0 + \pi_1 = 1$  and  $0 < \pi_0 < 1$ . Denote the optimal Bayesian (average) error probability by

$$P_e \triangleq \inf_{P_{Z|X^n}} \pi_0 \pi_{1|0} + \pi_1 \pi_{0|1}.$$

Prove that if  $\pi = (\frac{1}{2}, \frac{1}{2})$ , then

$$P_e = \frac{1}{2}(1 - \text{TV}(P, Q)).$$

Find the optimal test.

3. Find the optimal test for general prior  $\pi$  (not necessarily equiprobable).  
 4. Why is it always sufficient to focus on deterministic test in order to minimize the Bayesian error probability?  
 4 Function  $\alpha \mapsto \beta_\alpha(P, Q)$  is monotone and thus by Lebesgue's theorem possesses a derivative

$$\beta'_\alpha \triangleq \frac{d}{d\alpha} \beta_\alpha(P, Q).$$

almost everywhere on  $[0, 1]$ . Prove

$$D(P||Q) = - \int_0^1 \log \beta'_\alpha d\alpha.$$

- 5 Let  $P, Q$  be distributions such that for all  $\alpha \in [0, 1]$  we have

$$\beta_\alpha(P, Q) \triangleq \min_{P_{Z|X}: P[Z=0] \geq \alpha} Q[Z=0] = \alpha^2.$$

Find  $\text{TV}(P, Q)$ ,  $D(P||Q)$  and  $D(Q||P)$ .

- 6 We have shown in class that for testing iid products and any fixed  $\epsilon \in (0, 1)$ :

$$\log \beta_{1-\epsilon}(P^n, Q^n) = -nD(P||Q) + o(n), \quad n \rightarrow \infty,$$

which is equivalent to Stein's lemma. Show furthermore that assuming  $V(P||Q) < \infty$  we have

$$\log \beta_{1-\epsilon}(P^n, Q^n) = -nD(P||Q) + \sqrt{nV(P||Q)}Q^{-1}(\epsilon) + o(\sqrt{n}),$$

where  $Q^{-1}(\cdot)$  is the functional inverse of  $Q(x) = \int_x^\infty \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt$  and

$$V(P||Q) \triangleq \text{Var}_P \left[ \log \frac{dP}{dQ} \right].$$

## References

- [1] T. Cover and J. Thomas, *Elements of Information Theory*, Second Edition, Wiley, 2006

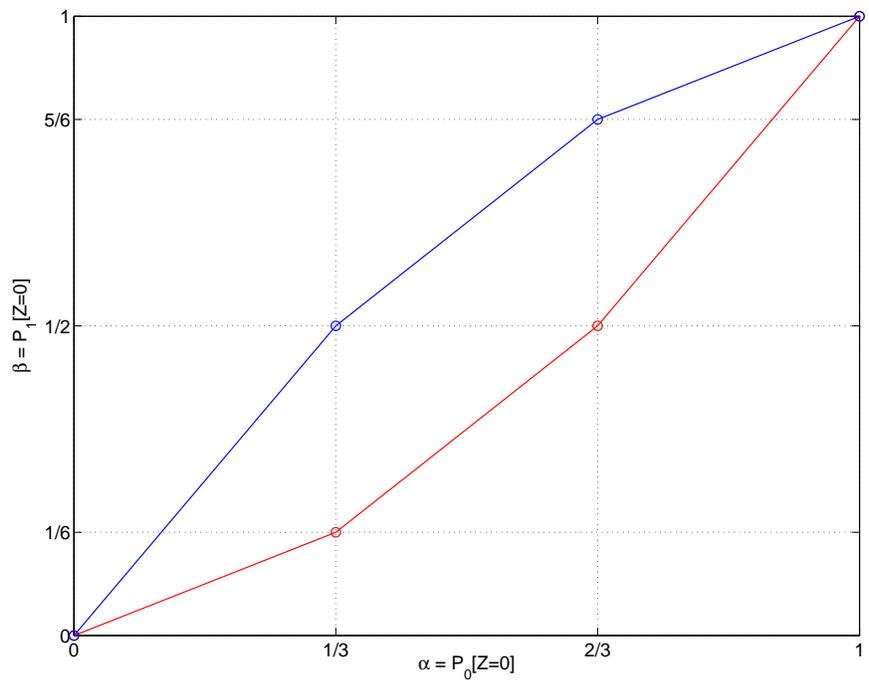


Figure 1: Upper (blue) and lower (red) boundaries of  $\mathcal{R}(P_0, P_1)$  for  $n = 1$

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