

# ECE302 – Quiz 5

Jonathan Lam

April 15, 2021

A random variable  $R$  is observed, and it is known that under  $H_0$ :

$$f_0(r) = f_{R|H_0}(r | H_0) = \begin{cases} \frac{1}{2}, & -1 \leq r \leq 1 \\ 0, & \text{else} \end{cases}$$

and under  $H_1$ :

$$f_1(r) = f_{R|H_1}(r | H_1) = \frac{1}{2}e^{-|r|}$$

Furthermore, assume the priors  $p_0 = \frac{1}{3}$  and  $p_1 = \frac{2}{3}$ .

1. Plot the class conditional p.d.f.'s on one axis. Be sure to label.

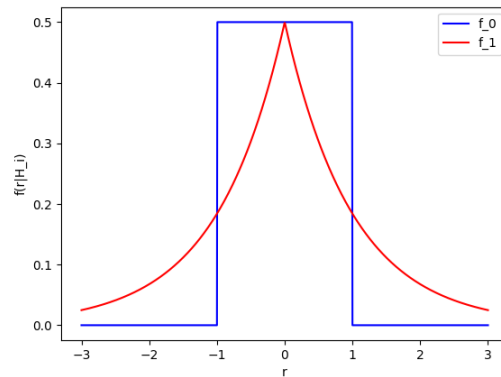


Figure 1: Plot of class conditional densities on the interval  $-3 \leq r \leq 3$

2. Consider a decision rule: If  $|r| > \gamma$  decide  $H_0$  ( $\hat{H} = 0$ ), else decide  $H_1$  ( $\hat{H} = 1$ ). Determine the probability of error for this rule if  $\gamma = \frac{1}{2}$ .

Using the symmetry of the problem, we can focus on the positive half-line. The decision rule then becomes: if  $0 \leq r \leq \frac{1}{2}$ , then choose  $H_1$ ; if  $r > \frac{1}{2}$  choose  $H_0$ . Thus:

$$\begin{aligned} P_{err} &= P(\hat{H} = 1 \cap H_0) + P(\hat{H} = 0 \cap H_1) \\ &= \left[ \int_0^{\frac{1}{2}} f(r | H_0) dr \right] \left( \frac{1}{3} \right) + \left[ \int_{\frac{1}{2}}^{\infty} f(r | H_1) dr \right] \left( \frac{2}{3} \right) \\ &= \left[ \int_0^{\frac{1}{2}} \frac{1}{2} dr \right] \left( \frac{1}{3} \right) + \left[ \int_{\frac{1}{2}}^{\infty} \frac{1}{2} e^{-r} dr \right] \left( \frac{2}{3} \right) \\ &= \frac{1}{12} + \frac{1}{3} e^{-\frac{1}{2}} \approx 0.286 \end{aligned}$$

3. Determine the likelihood ratio test that minimizes the overall probability of error.

Again, we focus only on the positive half-line because of the symmetry. The MAP rule states that the decision boundary should occur when:

$$f(r | H_1)p_1 = f(r | H_0)p_0$$

In other words, choose the decision boundary  $\eta$  such that:

$$\begin{aligned} \left( \frac{1}{2} e^{-\eta} \right) \left( \frac{2}{3} \right) &= \left( \frac{1}{2} \right) \left( \frac{1}{3} \right) \\ e^{-\eta} &= \frac{1}{2} \\ \eta &= \ln 2 \end{aligned}$$

Thus the MAP rule is: for  $|r| > \eta$ , choose  $H_0$ ; otherwise choose  $H_1$ .

4. Now suppose  $p_0$  and  $p_1$  are no longer constrained, but cannot be either 0 or 1. Find which values of  $p_0$  such that the decision rule that minimizes the probability of error always decides the same hypothesis, regardless of the observation.

This situation occurs if the a posteriori probabilities (class conditional multiplied by the priors) don't ever overlap; i.e., the a posteriori probability of  $H_0$  is always greater than  $H_1$ , or vice versa.

It is not possible that the a posteriori probability of  $H_0$  is always higher than  $H_1$ , because it is 0 for parts of the domain where  $H_1$  is not. The opposite is true, if the a posteriori probability of  $H_1$  is always greater than the a posteriori probability of  $H_0$  for  $|r| \leq 1$ , and particularly at  $|r| = 1$ .

$$\left[ \frac{1}{2} e^{-1} \right] (1 - p_0) > \frac{1}{2} p_0 \Rightarrow p_0 < \frac{1}{e + 1}$$

The MAP decision rule always chooses  $H_1$  if  $p_0 < (e + 1)^{-1}$ .