EE-806

HOMEWORK SOLUTIONS #1

1. Computer Exercise 1:

For this problem, data was generated as follows:

$$H_0: \qquad y \sim U[-1.5, 1.5] H_1: \qquad y \sim \frac{1}{2}\mathcal{N}(-2, 1) + \frac{1}{2}\mathcal{N}(2, 1)$$

with prior probability, $\pi_0 = 0.3$. This might have been guessed via pdf/pmf approximation using histograms:



There are various ways in which one could design a decision rule. Now that you have learned more about decision theory, do you have some better ways of attacking this problem?

pdf/pmf approximations:

2. Computer Exercise 2:

My program generated the following output. (Deterministic on left, random on right.)



Randomized $\delta(0)$ rule:

>> binary case i)

Always zero: PrE=0.692, PrF=0.000 Always one: PrE=0.308, PrF=1.000 Keep bits: PrE=0.130, PrF=0.198 Flip bits: PrE=0.870, PrF=0.802

case ii)

Always zero: PrE=0.707, PrF=0.000 Always one: PrE=0.293, PrF=1.000 Keep bits: PrE=0.556, PrF=0.198 Flip bits: PrE=0.444, PrF=0.802

case iii)

Always zero: PrE=0.195, PrF=0.000 Always one: PrE=0.805, PrF=1.000 Keep bits: PrE=0.588, PrF=0.605 Flip bits: PrE=0.412, PrF=0.395

case iv)

Always zero: PrE=0.502, PrF=0.000 Always one: PrE=0.498, PrF=1.000 Keep bits: PrE=0.496, PrF=0.494 Flip bits: PrE=0.504, PrF=0.506

Randomized $\delta(1)$ rule:



3. Poor II-1:

The likelihood ratio for this binary channel problem is

$$L(y) = \begin{cases} \frac{\lambda_1}{1-\lambda_0} & y=0\\ \frac{1-\lambda_1}{\lambda_0} & y=1 \end{cases}$$

Under uniform costs, the threshold is $\tau = \frac{\pi_0}{1-\pi_0}$. There are several different ways to approach this problem. I considered the possible rules and derived the Bayes risk for each rule. Thus,

$$\begin{array}{lll} \text{if } y = 0 \quad \text{then } \delta_B(y) = 1 & \text{if } \frac{\lambda_1}{1 - \lambda_0} > \frac{\pi_0}{1 - \pi_0} \\ \Leftrightarrow \text{if } \frac{\lambda_1}{1 - \lambda_0 + \lambda_1} > \pi_0 \\ \text{if } y = 1 & \text{then } \delta_B(y) = 1 & \text{if } \frac{1 - \lambda_1}{\lambda_0} > \frac{\pi_0}{1 - \pi_0} \\ \Leftrightarrow \text{if } \frac{1 - \lambda_1}{1 + \lambda_0 - \lambda_1} > \pi_0 \end{array}$$

Using the notation in the textbook, we can define two quantities:

$$\bar{\pi} = \max\left\{\frac{\lambda_1}{1-\lambda_0+\lambda_1}, \frac{1-\lambda_1}{1+\lambda_0-\lambda_1}\right\}$$
$$\underline{\pi} = \min\left\{\frac{\lambda_1}{1-\lambda_0+\lambda_1}, \frac{1-\lambda_1}{1+\lambda_0-\lambda_1}\right\}$$

From these we can determine all of the possible Bayes rules depending on the values of $\pi_0, \lambda_1, \lambda_0$:

$$\begin{array}{ccc} \pi_0 < \underline{\pi} & \delta_B(y) = 1 \\ \pi_0 > \bar{\pi} & \delta_B(y) = 0 \\ \frac{1-\lambda_1}{1+\lambda_0-\lambda_1} \ge \pi_0 \ge \frac{\lambda_1}{1-\lambda_0+\lambda_1} & \delta_B(y) = y \\ \frac{\lambda_1}{1-\lambda_0+\lambda_1} \ge \pi_0 \ge \frac{1-\lambda_1}{1+\lambda_0-\lambda_1} & \delta_B(y) = 1-y \end{array}$$

Next we determine the associated Bayes risk for each of these rules. Since we are considering the case of uniform costs, the general formula for the risk reduces to $r(\delta) = \pi_0 P_0(\Gamma_1) + (1 - \pi_0)P_1(\Gamma_0)$. For each of the cases above, we'll calculate $P_i(\Gamma_j)$, $i \neq j$ and then the total risk:

You can easily verify that the four risk expressions above are concisely written as:

$$r(\delta) = V(\pi_0) = \min\{(1-\pi_0)\lambda_1, \pi_0(1-\lambda_0)\} + \min\{(1-\pi_0)(1-\lambda_1), \pi_0\lambda_0\}$$

4. Poor II-2(a):

The likelihood ratio is given by

$$L(y) = \frac{p_1(y)}{p_0(y)} = \begin{cases} \frac{3}{2(y+1)} & y \in [0,1]\\ \text{undefined} & y \notin [0,1] \end{cases}.$$

With uniform costs and equal priors, the threshold $\tau = 1$. Noting that $\frac{3}{2(y+1)} \ge 1$ is equivalent to $y \le \frac{1}{2}$, and recalling that the likelihood ratio is only defined over $y \in [0, 1]$ we conclude that

$$\Gamma_1 = [0, 1/2], \quad \Gamma_0 = (1/2, 1]$$

Thus the Bayes rule is given by

$$\delta_B(y) = \begin{cases} 1 & 0 \le y \le 1/2 \\ 0 & 1/2 < y \le 1 \end{cases}$$

The corresponding minimum Bayes risk is

$$r(\delta_B) = \frac{1}{2} \int_{\Gamma_0} p_1(y) dy + \frac{1}{2} \int_{\Gamma_1} p_0(y) dy = \frac{1}{2} \int_0^{1/2} \frac{2}{3} (y+1) dy + \frac{1}{2} \int_{1/2}^1 dy = \frac{11}{24}.$$

5. Poor II-6(a):

Here we have $p_0(y) = p_N(y+s)$ and $p_1(y) = p_N(y-s)$, which gives

$$L(y) = \frac{p_1(y)}{p_0(y)} = \frac{1 + (y+s)^2}{1 + (y-s)^2}.$$

With equal priors and uniform costs, we find that the Bayes rule is

$$\begin{array}{rcl} L(y) & \geq & 1 \\ \Leftrightarrow 1 + (y+s)^2 & \geq & 1 + (y-2)^2 \\ \Leftrightarrow ys & \geq & 0 \\ \Leftrightarrow y & \geq & 0 & (\text{since } s > 0) \end{array}$$

or, in other words,

$$\delta_B(y) = \begin{cases} 1 & \text{if } y \ge 0\\ 0 & \text{if } y < 0 \end{cases}$$

The minimum Bayes risk is then

$$r(\delta_B) = \frac{1}{2} \int_0^\infty \frac{1}{\pi \left[1 + (y+s)^2\right]} dy + \frac{1}{2} \int_{-\infty}^0 \frac{1}{\pi \left[1 + (y-s)^2\right]} dy = \frac{1}{2} - \frac{\tan^{-1}(s)}{\pi}.$$