

**Partial exam 2 - Wednesday 2 May 2018 - Duration : 60 min***No document, no phone, no computing machine.*Name: SKETCH OF First name: the solution Signature: \_\_\_\_\_

Exercise 1 :	Exercise 2 :	Grade /20 :
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**Exercise 1 (Gaussian Bayes Classifier,  $\approx 8$  pts)**

Suppose you have the following training set with one real-valued input  $X$  and a categorical output  $Y$  that has two values  $A$  and  $B$ .

$X$	0	2	3	4	5	6	7
$Y$	A	A	B	B	B	B	B

1. You must learn the Gaussian Bayes Classifier from this data. Write the parameters of the classifiers in this table :

$\mu_A = 1$	$\sigma_A^2 = 1$	$\Pr(Y = A) = 2/7$
$\mu_B = 5$	$\sigma_B^2 = 2$	$\Pr(Y = B) = 5/7$

Justify your calculation hereafter :

$$\begin{aligned} \mu_A &= \frac{0+2}{2} = 1 & \mu_B &= \frac{3+4+5+6+7}{5} = \frac{25}{5} = 5 \\ \sigma_A^2 &= \frac{(-1)^2 + 1^2}{2} = 1 & \sigma_B^2 &= \frac{(-2)^2 + (-1)^2 + 0^2 + 1^2 + 2^2}{5} = \frac{10}{5} = 2 \\ \Pr(Y=A) &= \frac{2}{7} & \Pr(Y=B) &= \frac{5}{7} \end{aligned}$$

2. Calculate  $\alpha = f_{X|Y}(X=2|Y=A)$  and  $\beta = f_{X|Y}(X=2|Y=B)$ . Do not propose any numerical approximation ; just give a simplified closed form expression.

$$\begin{aligned} \alpha &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1^2}{2}} = \frac{1}{\sqrt{2\pi e}} \\ \beta &= \frac{1}{\sqrt{2\pi \cdot 2}} e^{-\frac{3^2}{2 \cdot 2}} = \frac{e^{-9/4}}{2\sqrt{\pi}} \end{aligned}$$

3. What is the joint probability  $f_{X,Y}(X=2, Y=A)$ ? The answer must be given in terms of  $\alpha$  and  $\beta$  only.

$$\begin{aligned} f_{X,Y}(X=2, Y=A) &= f_{X|Y}(X=2|Y=A) P(Y=A) \\ &= \frac{2}{7} \alpha \end{aligned}$$

4. What is the joint probability  $f_{X,Y}(X=2, Y=B)$ ? The answer must be given in terms of  $\alpha$  and  $\beta$  only.

$$f_{X,Y}(X=2, Y=B) = \frac{5}{7} \beta$$

5. What is  $f_X(X=2)$ ? The answer must be given in terms of  $\alpha$  and  $\beta$  only.

$$f_X(X=2) = \frac{2}{7} \alpha + \frac{5}{7} \beta$$

6. What is the conditional probability  $\Pr(Y=A|X=2)$ ?

$$\begin{aligned} \Pr(Y=A|X=2) &= \frac{f_{X,Y}(X=2, Y=A)}{f_X(2)} \\ &= \frac{\frac{2}{7} \alpha}{\frac{2}{7} \alpha + \frac{5}{7} \beta} = \frac{1}{1 + \frac{5}{2} \times \beta / \alpha} \end{aligned}$$

7. Consider the figure 1. If you trained a new Bayes classifier on this data, what class would you predicted for the query location indicated with “?” ? Explain carefully your answer.

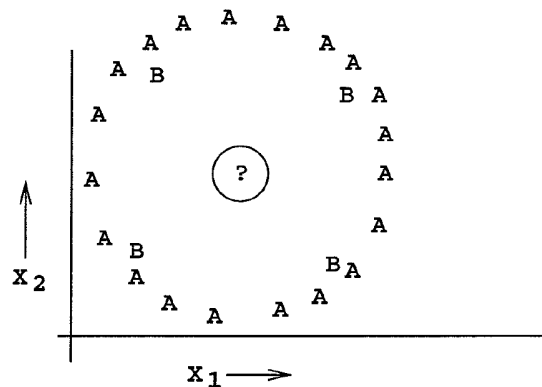


FIGURE 1 – Training data set and query location indicated with “?”.

The A's and B's have almost the same mean and the same covariance.  
But the number of A is larger than the number of B. So the test should choose A.

### Exercise 2 (Test and $p$ -value, $\approx 12$ pts)

Assume that  $x$  is a sample of a random variable  $X$  following an exponential distribution with the unknown parameter  $\theta$ . The exponential probability density function with parameter  $\theta$  is

$$f_{\theta}(x) = \begin{cases} \frac{1}{\theta} e^{-\frac{x}{\theta}} & \text{for } x > 0, \\ 0 & \text{elsewhere.} \end{cases}$$

We want to test  $H_0 : \{\theta = \theta_0\}$  versus  $H_1 : \{\theta = \theta_1\}$  with  $0 < \theta_1 < \theta_0$ .

1. Calculate the cumulative distribution function  $F_{\theta}(x)$  associated to  $f_{\theta}(x)$ .

$$F_{\theta}(x) = \int_{-\infty}^x f_{\theta}(t) dt = \begin{cases} 0 & \text{if } x \leq 0 \\ \int_0^x \frac{1}{\theta} e^{-t/\theta} dt & \text{if } x > 0 \end{cases}$$

$$= \begin{cases} 0 & \text{if } x \leq 0 \\ \frac{1}{\theta} [-\theta e^{-t/\theta}]_0^x = 1 - e^{-x/\theta} & \text{if } x > 0 \end{cases}$$

2. Calculate the decision function  $d(x)$  of the Neyman-Pearson test of size  $\alpha$ . Simplify it such that the final decision function consists in comparing  $x$  to a threshold  $h_\alpha$ .

Neyman-Pearson  $x > 0$

$$\frac{f_{\theta_1}(x)}{f_{\theta_0}(x)} \stackrel{H_1}{>} \lambda \Leftrightarrow \frac{\theta_0}{\theta_1} e^{-\frac{x}{\theta_1} + \frac{x}{\theta_0}} \stackrel{H_1}{>} \lambda$$

$$\Leftrightarrow \ln \frac{\theta_0}{\theta_1} + x \left( \frac{1}{\theta_0} - \frac{1}{\theta_1} \right) \stackrel{H_1}{>} \lambda$$

$$\Leftrightarrow x \stackrel{H_1}{<} h_\alpha = \frac{\left( \lambda - \ln \frac{\theta_0}{\theta_1} \right)}{\frac{1}{\theta_0} - \frac{1}{\theta_1}}$$

since  $\frac{1}{\theta_0} - \frac{1}{\theta_1} < 0$

3. Calculate the threshold  $h_\alpha$  of the test. The threshold must be given in closed-form.

$h_\alpha$  satisfies  $P_x(x < h_\alpha) = \alpha$  when  $x \sim f_{\theta_0}(x)$

So  $F_{\theta_0}(h_\alpha) = \alpha$

$$1 - e^{-\frac{h_\alpha}{\theta_0}} = \alpha \quad \text{since } \alpha > 0$$

$$h_\alpha = -\theta_0 \ln[1 - \alpha] > 0$$

$$= F_{\theta_0}^{-1}(\alpha)$$

4. Describe carefully the critical region  $C_\alpha$  of the test.

$$C_\alpha = (0; h_\alpha)$$

5. Calculate the power of the test, i.e., the probability  $\gamma$  to reject  $H_0$  when  $H_1$  is true.

$$\begin{aligned}\gamma &= P_\tau(x < h_\alpha) \text{ when } x \sim P_{\theta_1}(x) \\ \gamma &= 1 - e^{-h_\alpha/\theta_1} = 1 - e^{\frac{\theta_0}{\theta_1} \ln(1-\alpha)} \\ &= 1 - (1-\alpha)^{\theta_0/\theta_1}\end{aligned}$$

6. Show that  $C_\alpha \subset C_{\alpha'}$  if  $\alpha < \alpha'$ .

$$\begin{aligned}\text{If } \alpha < \alpha' \text{ then } & 1-\alpha > 1-\alpha' \\ & \ln(1-\alpha) > \ln(1-\alpha') \\ & -\theta_0 \ln(1-\alpha) < -\theta_0 \ln(1-\alpha') \\ & h_\alpha < h_{\alpha'} \\ \Rightarrow & C_\alpha \subset C_{\alpha'}\end{aligned}$$

7. Calculate the  $p$ -value  $\hat{p}(x)$  of the sample  $x$  from the definition of the  $p$ -value.

$$\begin{aligned}\hat{p}(x) &= \inf \{ \alpha : x \in C_\alpha \} \\ \text{We get that } \hat{p} &= \hat{p}(x) \text{ is the solution of} \\ h_{\hat{p}} = x &\Rightarrow -\theta_0 \ln(1-\hat{p}) = x \\ \Rightarrow \hat{p} &= 1 - e^{-x/\theta_0} = F_{\theta_0}(x)\end{aligned}$$

8. Show that  $\hat{p}(x)$  is uniformly distributed over  $[0, 1]$  when  $x$  follows the exponential distribution with pdf  $f_{\theta_0}(\cdot)$ .

Let  $t \in (0, 1)$

$$\begin{aligned} P_n(\hat{p}(x) \leq t) &= P_n(F_{\theta_0}(x) \leq t) \\ &= P_n(x \leq F_{\theta_0}^{-1}(t)) \quad \text{because } F_{\theta_0} \text{ is bijective.} \\ &= F_{\theta_0}[F_{\theta_0}^{-1}(t)] = t \end{aligned}$$

we also get  $P_n(\hat{p}(x) \leq 0) = 0$  and  $P_n(\hat{p}(x) \leq 1) = 1$

Hence the cdf of  $\hat{p}(x)$  coincides with the cdf of the uniform distribution over  $[0, 1]$

9. Propose a test equivalent to the Neyman-Pearson test of question 2 whose decision function is  $\hat{p}(x)$ . Precise clearly the threshold of the test.

The test is  $x \stackrel{H_1}{<} h_\alpha$

so it is equivalent to

$$F_{\theta_0}(x) \stackrel{H_1}{<} F_{\theta_0}(h_\alpha) \quad \text{since } F_{\theta_0} \text{ is increasing}$$

$$\Leftrightarrow F_{\theta_0}(x) \stackrel{H_1}{<} F_{\theta_0}(F_{\theta_0}^{-1}(\alpha)) = \alpha.$$

Hence  $\boxed{\hat{p}(x) \stackrel{H_1}{<} \alpha}$