### Université de Genève

# MULTIMEDIA SECURITY AND PRIVACY 14x016

## TP 3: Elements of Detection Theory

Author: Deniz Sungurtekin

E-mail: Deniz.Sungurtekin@etu.unige.ch

April 2020



#### 1 Exercices

#### 1.1

Let  $X \sim \mathcal{N}(0,1)$  and  $Y \sim \mathcal{N}(0,3)$ :

$$P[-2 < Y \le 1]: \tag{1}$$

$$= > \{(-2 - (0))/3 < N \le (1 - 0)/3\}:$$
(2)

$$=> P[-2/3 < N \le 1/3] = \phi(1/3) - \phi(-2/3) = \phi(1/3) - (1 - \phi(2/3)) = 0.63 - (1 - 0.74) = 0.37$$
 (3)

$$P[Y > 5.5] = 1 - P[Y \le 5.5]: \tag{4}$$

$$P[Y \le 5.5] = \phi(5.5/3) = 0.96 \tag{5}$$

$$=>1-0.96=0.04$$
 (6)

$$P[-2 < X \le 2]: (7)$$

$$=> P[-2 < N \le 2] = \phi(2) - (1 - \phi(2)) = 0.97 - 0.03 = 0.94$$
(8)

$$P[X > 1.5] = 1 - P[X \le 1.5]: \tag{9}$$

$$=> P[X \le 1.5] = \phi(1.5) = 0.93 \tag{10}$$

$$=>1-0.93=0.07\tag{11}$$

#### 1.2

Let X denote the peak temperature in Geneva, in June, as measured in Celsius, for which holds:  $X \sim N(27, 9)$ . What is:

$$P[X > 35] = 1 - P[X \le 35]: \tag{12}$$

$$P[X \le 35] = \phi((35 - 27)/9) = 0.81 \tag{13}$$

$$=>1-0.81=0.19$$
 (14)

$$P[X \le 5] = \phi((5 - 27)/9) = \phi(-2.4) = 1 - \phi(2.4) = 0.01 \tag{15}$$

$$P[20 < X \le 40] = P[((20 - 27)/9) < N \le ((40 - 27)/9)] = P[-7/9 < N \le 13/9]$$
(16)

$$= \phi(13/9) - \phi(-7/9) = \phi(13/9) - (1 - \phi(7/9)) = 0.91 - (1 - 0.77) = 0.68$$
(17)

#### 1.3

Let X be a Gaussian random variable, for which E[X] = 0 and  $P[|X| \le 10] = 0.3$ . What is  $\sigma_X$ ?

$$P[|X| \le 10] = P[-10 \le X \le 10] = P[-10/\sigma \le N \le 10/\sigma] \tag{18}$$

$$= \phi(10/\sigma) - (1 - \phi(10/\sigma)) = 0.3 \tag{19}$$

$$=> \phi(10/\sigma) = 0.65$$
 (20)

$$=>10/\sigma\cong0.4\tag{21}$$

$$=> \sigma = 25 \tag{22}$$

#### 1.4

Prove that:

$$Q(n) = \frac{1}{2} erfc(\frac{n}{\sqrt{2}}) :$$

$$=> Q(n) = \frac{1}{\sqrt{2\pi}} \int_{n}^{\infty} e^{\frac{-x^2}{2}} dx = \frac{1}{2} \left(\frac{2}{\sqrt{\pi}} \int_{n/\sqrt{2}}^{\infty} exp(-x^2) dx\right)$$
 (23)

$$= \frac{1}{2} - \frac{1}{2}erf(\frac{n}{\sqrt{2}}) = \frac{1}{2}erfc(\frac{n}{\sqrt{2}})$$
 (24)

#### 1.5

Let there be two hypothesizes,  $H_0$  and  $H_1$ :

$$H_0: X = Z$$

$$H_1: \mathbf{X} = \mu \mathbf{1} + \mathbf{Z}$$

where  $Z \sim N (0, 1)$  and  $\mu 1 = 1$ .

Determine the separation threshold  $\tau$  following the MAP hypothesis, or likelihood ratio test:

We have that  $X \sim \mathcal{N}(0,1)$  with hypothesis 0 and  $X \sim \mathcal{N}(1,1)$  with hypothesis 1:

$$x \in A_0, ifx \le \frac{e^{\frac{-x^2}{2\sigma^2}}}{e^{\frac{-(x-1)^2}{2\sigma^2}}} \ge \frac{P[H_1]}{P[H_0]}$$
$$x \in A_0, ifx \le \ln(\frac{P[H_1]}{P[H_0]})^{\frac{2\sigma^2 + 1}{-2}}$$
$$x \in A_0, ifx \le \tau$$

Determine the probability of correct detection  $p_d = 1 - p_m$ :

$$p_d = 1 - P(A_0|H_1) = 1 - \int_{x < \tau} f_{x|H_1}(x)dx = 1 - \phi(\tau - 1)$$

#### 1.6

Derive the general formula's for the Probability of Miss,  $p_m$ , the Probability of correct detection,  $p_d$ , and the Probability of False Alarm,  $p_f$ :

$$p_d = 1 - P(A_0|H_1) = 1 - \int_{x < \tau} f_{x|H_1}(x)dx = 1 - \phi(\tau - \mu 1)$$

$$p_m = P(A_0|H_1) = \phi(\tau - \mu 1)$$

$$p_f = P(A_1|H_0) = 1 - \phi(\tau)$$

Receiver Operating Characteristic (ROC) curve for each different value of  $\mu 1 = 0, 1, 2$ :

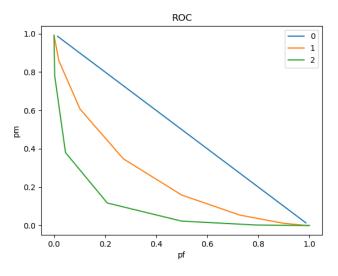


Figure 1: ROC

We clearly observe that for  $\mu 1 = 0$ , pf = 1 - pm. The more  $\mu 1$  is big the more the pm decrease quickly over the pf. It can be explained by the fact that we centralize  $\tau$  by  $\mu 1$  so the obtained pm by using  $\phi$  is much smaller. Furthermore, the bigger is  $\mu 1$  the bigger is the  $p_d$ .

#### 1.7

Formulate hypothesises  $H_0$  and  $H_1$ :

$$H_0$$
: Y = W

$$H_1$$
:  $Y = V + W$ 

Use the binary hypothesis likelihood ratio test to determine the rule that minimises  $p_{err}$ , the probability of a decoding error:

$$x \in A_0, ify \le \frac{f_{Y|H_0}}{f_{Y|H_1}} \tag{25}$$

$$x \in A_0, ify \le \frac{e^{-y}}{ye^{-y}} \tag{26}$$

$$=> y \le \frac{1}{y} \tag{27}$$

$$=>\tau=1\tag{28}$$

Determine  $p_{err}$  for the optimum decision rule:

$$p_{err} = p_m p(H_1) + p_f p(H_0) = p_m 0.5 + p_f 0.5$$
(29)

$$<=> p_f = \int_{y>\tau} f_{Y|H_0} = 0.36 <=> p_m = \int_{y<\tau} f_{Y|H_1} = 0.26 => p_{err} = 0.5 * 0.26 + 0.5 * 0.36 = 0.31$$
 (30)