

## Machine Learning Worksheet 01

Shang-Hsin Yu – 03681048 – shanghsin.yu@tum.de

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### Problem 1

From the problem statements we know that:

$$\begin{aligned}p(Terrorist) &= 0.01, p(Scanned | Terrorist) = 0.95 \\p(NotTerrorist) &= 0.99, p(Scanned | NotTerrorist) = 0.05\end{aligned}$$

And we can further calculate the probability of a random person being scanned is:

$$p(Scanned) = p(S | T) \cdot p(T) + p(S | NT) \cdot p(NT) = 0.059$$

After that we can get the solution by applying Bayes rule

$$p(Terrorist | Scanned) = \frac{p(S | T)p(T)}{p(S)} = \frac{0.95 \cdot 0.01}{0.059} \approx 16.1\%$$

The result is about 16.1%

### Problem 2

The chances of there being 0, 1 or 2 red balls are tied to the result of coin toss which is a binomial distribution as follows

$$\begin{aligned}p(R = 0) &= \binom{2}{0} \cdot 0.5 \cdot 0.5 = 0.25 \\p(R = 1) &= \binom{2}{1} \cdot 0.5 \cdot 0.5 = 0.5 \\p(R = 2) &= \binom{2}{2} \cdot 0.5 \cdot 0.5 = 0.25\end{aligned}$$

Then we need to know the probability of 3 red balls are being drawn from the box which is

$$p(ThreeRed) = 0 \cdot 0 \cdot 0 \cdot p(R = 0) + 0.5 \cdot 0.5 \cdot 0.5 \cdot p(R = 1) + 1 \cdot 1 \cdot 1 \cdot p(R = 2) = 0.3125$$

Finally, we can once again use Bayes rule to get the result

$$p(R = 2 | ThreeRed) = \frac{p(ThreeRed | R = 2)p(R = 2)}{p(ThreeRed)} = \frac{1 \cdot 0.25}{0.3125} = 0.8$$

The solution is 0.8

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**Problem 3**

This problem is just getting the expected value of two random variables.

$$E[Tail] = \sum_{x=1}^{\infty} x \cdot \left(\frac{1}{2}\right)^x \cdot \frac{1}{2} = \frac{1}{2} \cdot \left(\frac{1}{2} + 2 \cdot \frac{1}{4} + 3 \cdot \frac{1}{8} + \dots\right)$$

Assume  $y = \frac{1}{2} + 2 \cdot \frac{1}{4} + 3 \cdot \frac{1}{8} + \dots$

$$y - \frac{1}{2}y = \frac{1}{2}y = \frac{1}{2} + \frac{1}{4} + \frac{1}{8} = \frac{1}{2} \cdot 2 = 1$$

So then we get  $y = 2$  and thus

$$E[Tail] = \frac{1}{2} \cdot y = 1$$

Similarly, the expected value of head is

$$E[Head] = \sum_{x=0}^{\infty} 1 \cdot \left(\frac{1}{2}\right)^x \cdot \frac{1}{2} = \frac{1}{2} \cdot \left(1 + \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots\right) = \frac{1}{2} \cdot 1 \cdot 2 = 1$$

**Problem 4**

$$\begin{aligned} \text{mean} = E[X] &= \int_a^b \frac{1}{b-a} x dx = \frac{1}{b-a} \left(\frac{1}{2}b^2 - \frac{1}{2}a^2\right) = \frac{(b-a)(b+a)}{2(b-a)} = \frac{a+b}{2} \\ \text{variance} = E[X^2] - E[X]^2 &= \int_a^b \frac{1}{b-a} x^2 dx - \frac{(a+b)^2}{4} = \frac{1}{b-a} \cdot \frac{b^3 - a^3}{3} - \frac{(a+b)^2}{4} \\ &= \frac{(b-a)(a^2 + ab + b^2)}{3(b-a)} - \frac{(a+b)^2}{4} = \frac{a^2 - 2ab + b^2}{12} = \frac{(a-b)^2}{12} \end{aligned}$$

**Problem 5**

$$\begin{aligned} E_Y[E_{X|Y}[X]] &= E_Y\left[\int p(x|y)x dx\right] = \int \int xp(x|y)dxp(y)dy = \int \int xp(x|y)p(y)dx dy \\ &= \int \int xp(y|x)p(x)dx dy = \int xp(x)dx \int p(y|x)dy = E[X] \cdot 1 = E[X] \end{aligned}$$

$$\begin{aligned} E_Y[Var_{x|Y}[X]] + Var[E_{X|Y}[X]] &= E_Y[E_{X|Y}[X^2] - E_{X|Y}[X]^2] + E_Y[E_{X|Y}[X]^2] - E_Y[E_{X|Y}[X]]^2 \\ &= E_Y[E_{X|Y}[X^2]] - E_Y[E_{X|Y}[X]]^2 = E[X^2] - E[X]^2 = Var[X] \end{aligned}$$

**Problem 6**

We create a new random variable  $Y$  which is the sum of  $X_1$  through  $X_n$ .

Since  $X_1$  to  $X_n$  are i.i.d. random variables,  $Y$  has the mean of  $nE[X_1]$  and the variance of  $nVar[X_1]$ .

Then let us look at the equation

$$p\left(\left|\frac{1}{n}\sum_{i=1}^n X_i - E[X_i]\right| > \epsilon\right) = p\left(\left|\sum_{i=1}^n X_i - E[X_i]\right| > n\epsilon\right) = p\left(\left|Y - E[Y]\right| > n\epsilon\right)$$

according to Chebyshevs inequality

$$p\left(\left|Y - E[Y]\right| > n\epsilon\right) \leq \frac{Var(Y)}{n^2\epsilon^2}$$

we already know that the variance of  $X$  is finite, so the variance of  $Y$  will also be finite. As  $n$  approaches  $\infty$ , the term on the right will approach 0. So we have

$$p\left(\left|Y - E[Y]\right| > n\epsilon\right) \leq 0 \quad \text{as } n \rightarrow \infty$$