

# Data Analysis 2014-2015

## Home Assignment 4

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## 1 Classification

## 2 Probability theory refreshment

We have an urn that contains five red, three orange, and one blue ball. We now select two balls at random

1. **What is the sample space of this experiment**

$\{\{R, R\}, \{R, O\}, \{R, B\}, \{O, O\}, \{O, R\}, \{O, B\}, \{B, R\}, \{B, O\}\}$

2. **What is the probability of each point in the sample space**

If we use the same ordering as above,  $\{5/18, 5/24, 5/72, 1/12, 5/24, 1/24, 5/72, 1/24\}$ , the sum of these probabilities sum up to one, meaning that we have a complete sample space.

3. **Let  $X$  represent the number of orange balls selected. What are the possible values of  $X$ ?**

The possible values of  $X$  are  $\{0, 1, 2\}$ .

4. **Calculate  $\mathbb{P}\{X = 0\}$**

$5/18 + 5/72 + 5/72 = 5/12$

5. **Calculate  $\mathbb{E}[X]$**

$5/24 + 1/12 \cdot 2 + 5/24 + 1/24 + 1/24 = 2/3$

## 3 Probability theory refreshment

From probability theory we have the following definitions and properties:

$$(a) p_X(x) = \sum_{y \in \mathcal{Y}} p_{XY}(x, y)$$

$$(b) \text{ If } X \text{ and } Y \text{ are independent, then } P_{XY}(x, y) = p_X(x)p_Y(y)$$

$$(c) \mathbb{E}[X] = \sum_{x \in \mathcal{X}} xp_X(x)$$

$X$  and  $Y$  are discrete random variables that take values from in  $\mathcal{X}$  and  $\mathcal{Y}$ .  $p_X$  is the distribution of  $X$ ,  $p_Y$  the distribution of  $Y$  and  $p_{XY}$  the distribution of  $X$  and  $Y$ .

### 1.

We prove the following identity:

$$\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$$

By (c), the expected value of  $X$  is given by:

$$\mathbb{E}[X] = \sum_{x \in \mathcal{X}} xp_X(x)$$

Therefore the expected value of  $X + Y$  would be

$$\begin{aligned} \mathbb{E}[X + Y] &= \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} (x + y)p_{XY}(x, y) \\ &= \sum_{x \in \mathcal{X}} x \sum_{y \in \mathcal{Y}} p_{XY}(x, y) + \sum_{y \in \mathcal{Y}} y \sum_{x \in \mathcal{X}} p_{XY}(x, y) \\ &= \sum_{x \in \mathcal{X}} xp(x) + \sum_{y \in \mathcal{Y}} yp(y) \\ &= \mathbb{E}[X] + \mathbb{E}[Y] \end{aligned}$$

In the last step we use the definition for the expected value of a random variable. We have now shown that  $\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$ .

## 2.

To prove the following identity, we use that the random variables  $X$  and  $Y$  are independent.

$$\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$$

We can write  $\mathbb{E}[XY]$  as

$$\mathbb{E}[XY] = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} xyp_{XY}(x, y)$$

This is where we use that  $X$  and  $Y$  are independent - using property (b):

$$\sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} xyp_X(x)p_Y(y)$$

This can be reduced to prove our identity

$$\begin{aligned} \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} xyp_X(x)p_Y(y) &= \sum_{x \in \mathcal{X}} xp_X(x) \sum_{y \in \mathcal{Y}} yp_Y(y) \\ &= \mathbb{E}[X]\mathbb{E}[Y] \end{aligned}$$

This proves the identity  $\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$ .

## 3.

A bag has 2 red apples and 2 green apples. There is taken 2 apples from the bag without putting them back into the bag. Let  $X$  be the first apple and let  $Y$  be the second apple. The joint distribution table of  $X$  and  $Y$  is seen below:

X / Y	Red	Green
Red	$\frac{1}{6}$	$\frac{2}{6}$
Green	$\frac{2}{6}$	$\frac{1}{6}$

The probability of apple  $X$  being red is:

$$\mathbb{E}[X = \text{Red}] = \frac{1}{2}$$

Which is the same probability for apple  $Y$  being red. We have that

$$\mathbb{E}[X = \text{Red} \wedge Y = \text{Red}] = \frac{1}{6}$$

Since  $\frac{1}{2} \frac{1}{2} = \frac{1}{4} \neq \frac{1}{6}$  then

$$\mathbb{E}[XY] \neq \mathbb{E}[X]\mathbb{E}[Y]$$

in this example.

## 4.

The identity to be proved:

$$\mathbb{E}[\mathbb{E}[X]] = \mathbb{E}[X]$$

We know that  $\mathbb{E}[X] = k$  and that  $\mathbb{E}[k] = k$ . That means taking the expected value of an expected value will just return the constant you already found. This can be done more than 2 times and it will always be the constant  $k$  that is your result.

**5.**

We want to show that  $\mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$ .

$$\begin{aligned}\mathbb{E}[(X - \mathbb{E}[X])^2] &= \mathbb{E}[X^2 - 2 \cdot X \cdot \mathbb{E}[X] + (\mathbb{E}[X])^2] \\ &= \mathbb{E}[X^2] - 2 \cdot \mathbb{E}[X] \cdot \mathbb{E}[X] + (\mathbb{E}[X])^2 \\ &= \mathbb{E}[X^2] - (\mathbb{E}[X])^2\end{aligned}$$

**4 Markov's inequality vs. Hoeffding's inequality vs. binomial bound**

**5 Hoeffding's inequality**