

# MASD Final Exam

MASD 2018  
Department of Computer Science  
University of Copenhagen

Eksamensnummer: 35

Version 1  
**Due:** November 4th, 2018, 16:00

## Contents

<b>1. Exercise</b>	<b>3</b>
<b>2. Exercise</b>	<b>3</b>
<b>3. Exercise</b>	<b>3</b>
(a) . . . . .	3
(b) . . . . .	3
<b>4. Exercise</b>	<b>3</b>
(a) . . . . .	3
(b) . . . . .	4
(c) . . . . .	4

## 1. Exercise

## 2. Exercise

## 3. Exercise

### (a)

The expression *pmf* stands for the "probability mass function".

The *pmf* of  $X \sim \text{Poiss}(\lambda)$  can be written as follows, as per table 1.1 in the "MASD\_essentials\_ch1-updated.pdf" on Absalon:

$$p(x) = e^{-\lambda} \frac{\lambda^x}{x!}$$

This expression is used to calculate the probability of  $x$  events happening, when there is an average of  $\lambda$  events happening.

### (b)

The expectation of  $Y$ ,  $\mathbb{E}Y$  can be written as follows, as per definition 1.8 in the "MASD\_essentials\_ch1-updated.pdf" on Absalon:

$$\mathbb{E}Y = \sum_y yp(y)$$

The variance of  $Y$ ,  $\text{var}(Y)$ , can be written as follows, as per definition 1.10 in the "MASD\_essentials\_ch1-updated.pdf" on Absalon:

$$\text{var}(Y) = \mathbb{E}(Y - \mathbb{E}Y)^2 = \sum_y y^2 p(y) - (\sum_y yp(y))^2$$

To find  $\mathbb{P}(a < V \leq b)$  we first have to make the observation that the cdf  $F_V(x)$  is defined as being the probability that the variable  $V$  will be *less than or equal to*  $x$ , thus we can say that:

$$F_V(x) = \mathbb{P}(V \leq x)$$

This means that to calculate the probability of it being inside the interval  $a < V \leq b$ , we calculate  $\mathbb{P}(V \leq b)$ , and subtract  $\mathbb{P}(V \leq a)$ .

$$\mathbb{P}(a < V \leq b) = F_V(b) - F_V(a)$$

## 4. Exercise

### (a)

For this, I assume that the given values for  $x$  and  $y$  are all the values possible. If that assumption is true, then the final joint pmf will look as follows:

$p(x, y)$	$y=-1$	$y=1$	$y=3$
$x=-1$	0.1	0.1	0.4
$x=1$	0.3	0.05	0.05

The reason for this is because the joint pmf must be equal to 1. So we simply calculate:

$$1 - (0.1 + 0.1 + 0.3 + 2(0.05)) = 0.4$$

**(b)**

To compute  $\mathbb{E}(X + Y)$  we have to make the following observation:

$$\mathbb{E}(X + Y) = \mathbb{E}X + \mathbb{E}Y$$

This means that we simply have to compute:

$$\mathbb{E}X + \mathbb{E}Y = \sum_x x p_X(x) + \sum_y y p_Y(y)$$

To find the marginal pmfs  $p_X(x)$  and  $p_Y(y)$  we use the following calculations:

$$p_X(x) = \sum_y p_{X,Y}(x, y)$$

$$p_Y(y) = \sum_x p_{X,Y}(x, y)$$

First we will find  $\mathbb{E}X$ , then  $\mathbb{E}Y$ :

$$\mathbb{E}X = -1(0.1 + 0.1 + 0.4) + 1(0.3 + 0.05 + 0.05) = -0.2$$

$$\mathbb{E}Y = -1(0.1 + 0.3) + 1(0.1 + 0.05) + 3(0.4 + 0.05) = 1.1$$

Then we simply write:

$$\mathbb{E}(X + Y) = -0.2 + 1.1 = 0.9$$

This should be interpreted as expecting to make back 90% of the money put in.

**(c)**

First we will compute the pmf  $p_Z$  of  $Z = X * Y$ .  
To do this, first we will define what  $p_Z$  looks like:

$$p_Z(z) = \mathbb{P}(Z = z)$$

We can see from the table from Exercise 4a that there are six possible values for  $Z$ , since there are 2 values for  $X$  and 3 for  $Y$ . These values are the following:

$X * Y$	$y = -1$	$y = 1$	$y = 3$
$x = -1$	1	-1	-3
$x = 1$	-1	1	3

Now we simply compute  $\mathbb{P}(X * Y)$  for each value of  $x$  and  $y$ , and get the following table:

$p_Z(z)$	$\mathbb{P}(X * Y = z)$
$z = -3$	0.4
$z = -1$	0.4
$z = 1$	0.15
$z = 3$	0.05

To figure out if  $X$  and  $Y$  are independent, we have to look at their covariance, which is defined as thus:

$$\text{cov}(X, Y) = \mathbb{E}XY - \mathbb{E}X\mathbb{E}Y$$

We start off by finding  $\mathbb{E}XY$ :

$$\mathbb{E}XY = \sum_{x,y} xyp(x, y)$$

I have done this calculation in the attached Jupyter Notebook named: "Calculations" since it would be tedious to do by hand. It has given me the following:

$$\mathbb{E}XY = -1.3$$

We have already calculated  $\mathbb{E}X$  and  $\mathbb{E}Y$  further up, in exercise 4b. So we can write the following:

$$\text{cov}(X, Y) = (-1.3) - (-0.2) * 1.1 = -1.08$$

Since we now know that the covariance  $\text{cov}(X, Y) \neq 0$ , we also know that  $X$  and  $Y$  are dependant on each other.