

#### Exercise 4

**Exercise 4.** Let  $X_1, X_2$  and  $X_3$  be three discrete random variables with

$$\mathbb{P}[X_1 = 0] = \mathbb{P}[X_1 = 1] = \mathbb{P}[X_2 = 0] = \mathbb{P}[X_2 = 1] = \frac{1}{2}$$

and

$$\mathbb{P}[X_3 = 0] = 1.$$

(i) Characterize all possible coupling between  $X_1$  and  $X_2$ .

$X_2 \backslash X_1$	0	1	
0	a	b	$1/2$
1	c	d	$1/2$
	$1/2$	$1/2$	

}  $\mathbb{P}[X_2 = x]$

}  $\mathbb{P}[X_1 = x]$

with  $a := \mathbb{P}[X_1=0, X_2=0]$ ;  $b := \mathbb{P}[X_1=1, X_2=0]$   
 $c := \mathbb{P}[X_1=0, X_2=1]$ ;  $d := \mathbb{P}[X_1=1, X_2=1]$

The parameters  $a, b, c, d$  have to fulfill some properties such that they define a valid probability measure:

$$(I) \quad a + b = c + d = a + c = b + d = \frac{1}{2}$$

which also implies:  $a + b + c + d = 1$

$$(II) \quad a, b, c, d \geq 0$$

which combined with (I) implies  $a, b, c, d \in [0, \frac{1}{2}]$

Examples:

$$1) \quad a = 0 \Rightarrow b = c = 1 \Rightarrow d = 0$$

$$2) \quad a = \frac{1}{4} \Rightarrow b = c = \frac{1}{4} \Rightarrow d = \frac{1}{4}$$

These examples show that it is sufficient to fix one parameter to describe the coupling:

For instance: pick  $a = P[X_1=0, X_2=0] \in [0, \frac{1}{2}]$ , then you can compute:

$$b = \frac{1}{2} - a$$

$$c = \frac{1}{2} - a$$

$$d = \frac{1}{2} - b = \frac{1}{2} - c = \frac{1}{2} - (\frac{1}{2} - a) = a$$

→ In the following we will express a specific coupling by parameter  $a$ .

- (ii) Which coupling maximizes the correlation? Which coupling minimizes the correlation? Do you have an intuitive explanation why these couplings are the ones that minimize/maximize the correlation?

$$\text{corr}(X_1, X_2) = \frac{\text{cov}(X_1, X_2)}{\sqrt{\text{Var}(X_1) \text{Var}(X_2)}}$$

$$a^* = \underset{a \in [0, \frac{1}{2}]}{\text{argmax}} \frac{\text{cov}(X_1, X_2)}{\sqrt{\text{Var}(X_1) \text{Var}(X_2)}} = \underset{a \in [0, \frac{1}{2}]}{\text{argmax}} \text{cov}(X_1, X_2)$$

$$= \underset{a \in [0, \frac{1}{2}]}{\text{argmax}} \left( E[X_1 X_2] - E[X_1] E[X_2] \right)$$

$$= \underset{a \in [0, \frac{1}{2}]}{\text{argmax}} E[X_1 X_2]$$

get rid of terms not depending on  $a$

$$E[X_1 X_2] = 0 \cdot (P[X_1=0, X_2=0] + P[X_1=0, X_2=1] + P[X_1=1, X_2=0]) + 1 \cdot P[X_1=1, X_2=1]$$

$$= 0 + d = d$$

In (i) we have seen that for a valid coupling  $a=d$ .

$$\rightarrow a^* = \underset{a \in [0, \frac{1}{2}]}{\text{argmax}} E[X_1 X_2] = \underset{a \in [0, \frac{1}{2}]}{\text{argmax}} d = \underset{a}{\text{argmax}} a = \frac{1}{2}$$

If  $a=d=\frac{1}{2}$ , we have  $b=c=0$ .

$$\text{corr}(X_1, X_2) = \frac{E[X_1 X_2] - E[X_1] E[X_2]}{\sqrt{\text{Var}(X_1) \text{Var}(X_2)}} = \frac{\frac{1}{2} - \frac{1}{2} \cdot \frac{1}{2}}{\sqrt{\frac{1}{4} \cdot \frac{1}{4}}} = 1$$

$$\Gamma E[X_1] = E[X_2] = \frac{1}{2}$$

$$\text{Var}(X_1) = \text{Var}(X_2) = E[(X_2 - \frac{1}{2})^2] = (0 - \frac{1}{2})^2 \cdot \frac{1}{2} + (1 - \frac{1}{2})^2 \cdot \frac{1}{2} = \frac{1}{4}$$

L

same as before

$$\begin{aligned} a^* &= \operatorname{argmin}_{a \in [0, \frac{1}{2}]} \frac{\operatorname{cov}(X_1, X_2)}{\sqrt{\operatorname{Var}(X_1) \operatorname{Var}(X_2)}} = \dots = \operatorname{argmin}_{a \in [0, \frac{1}{2}]} \mathbb{E}[X_1 X_2] \\ &= \operatorname{argmin}_{a \in [0, \frac{1}{2}]} d = \operatorname{argmin}_{a \in [0, \frac{1}{2}]} a = 0 \end{aligned}$$

If  $a = d = 0$ , we have  $b = c = \frac{1}{2}$ .

$$\operatorname{corr}(X_1, X_2) = \frac{\mathbb{E}[X_1 X_2] - \mathbb{E}[X_1] \mathbb{E}[X_2]}{\sqrt{\operatorname{Var}(X_1) \operatorname{Var}(X_2)}} = \frac{0 - \frac{1}{2} \cdot \frac{1}{2}}{\sqrt{\frac{1}{4} \cdot \frac{1}{4}}} = -1$$

In these two scenarios ( $a = \frac{1}{2}$  and  $a = 0$ ) we have  $|\operatorname{corr}(X_1, X_2)| = 1$ . In both cases it is sufficient to know the value of  $X_1$  or  $X_2$  to infer the value of the respective other RV.

In the case  $a = \frac{1}{2}$ ,  $X_1$  and  $X_2$  always take on the same value, in the case  $a = 0$ , they always take on a different value.

This means we have perfect correlation, with the difference that when  $a = 0$ , we have perfect negative correlation.

(iii) Which coupling makes the two random variables uncorrelated?

$$\operatorname{corr}(X_1, X_2) \stackrel{!}{=} 0$$

$$\Rightarrow \operatorname{cov}(X_1, X_2) = 0$$

$$\Leftrightarrow \mathbb{E}[X_1 X_2] - \mathbb{E}[X_1] \mathbb{E}[X_2] = 0$$

$$\Leftrightarrow \mathbb{E}[X_1 X_2] = \mathbb{E}[X_1] \mathbb{E}[X_2]$$

$$d = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} //$$

$$\Rightarrow a = d = \frac{1}{4}, \quad b = c = \frac{1}{4}$$

Here, we have the opposite to the previous cases. If we know  $X_1$  (or  $X_2$ ), we cannot make any assumptions about  $X_2$  (or respectively  $X_1$ ). The two RVs are independent:

$$\mathbb{P}[X_1 = x_1, X_2 = x_2] = \frac{1}{4} = \mathbb{P}[X_1 = x_1] \cdot \mathbb{P}[X_2 = x_2] \quad \forall x_1, x_2 \in \{0, 1\}$$

(iv) Do the tasks (i) – (iii) but for  $X_1$  and  $X_3$ .

$X_3 \backslash X_1$	0	1	
0	$\frac{1}{2}$	$\frac{1}{2}$	1
1	0	0	0
	$\frac{1}{2}$	$\frac{1}{2}$	

}  $P[X_3 = x]$

}  $P[X_1 = x]$

→ There is only one possible coupling

$$\text{Corr}(X_1, X_3) = \frac{E[X_1 X_3] - E[X_1]E[X_3]}{\sqrt{\text{Var}(X_1) \text{Var}(X_3)}} = 0$$

$\underbrace{\quad}_{=0}$

$$E[X_1 X_3] = 1 \cdot P[X_1=1, X_3=1] = 0$$

$$E[X_1] = \frac{1}{2}, \quad E[X_3] = 1$$

$$\text{Var}(X_1) = \frac{1}{4}, \quad \text{Var}(X_3) = 0$$

→  $X_1$  and  $X_3$  are uncorrelated