

ERASMUS, B4 Course

Winter semester 2020/2021, December 17.

Subject : A concept of the joint distribution of the random vector

### Introduction

So far we assumed, that whenever we were observing some RE, we only need one random variable.

Now we are going to use more. First we assume that for given  $(\Omega, \mathcal{F}, P)$  we have two random variables, namely

$$X, Y: \Omega \rightarrow \mathbb{R} \text{ and}$$

$$\forall t, s \in \mathbb{R}, \{ \omega \in \Omega: X(\omega) \leq t \}, \{ \omega \in \Omega: Y(\omega) \leq s \} \in \mathcal{F}.$$

It means, that the result of our observation is as follows

$$\Omega \ni \omega \longrightarrow (X, Y)(\omega) = (X(\omega), Y(\omega)) \in \mathbb{R} \times \mathbb{R}$$

The object  $(X, Y)$  is called a random vector of rank 2 with coordinates:  $X$ , and  $Y$  respectively.

And generally, for  $n \geq 2$ ,  $X_1, X_2, \dots, X_n$  random variables by the random vector of rank  $n$  we understand the object  $(X_1, X_2, \dots, X_n)$ .



Then

$$\Omega \ni \omega \longrightarrow (X_1, X_2, \dots, X_n)(\omega) = (X_1(\omega), X_2(\omega), \dots, X_n(\omega))$$

and  $X_j$  is called the  $j^{\text{th}}$  coordinate  $\in \mathbb{R}^n$

Assumption.

In this course we will only deal with 2-rank vectors  
variables  $(X, Y)$ .

Def 1

By the joint distribution of the v.v.  $(X, Y)$

we understand the multi function

$$\mathbb{R} \times \mathbb{R} \ni (t, s) \longrightarrow F_{(X, Y)}(t, s) \stackrel{\text{def}}{=} P(\{\omega \in \Omega : X(\omega) \leq t \text{ \& } Y(\omega) \leq s\})$$

Then  $F_{(X, Y)}$  is called the cumulative probability function

Remarks.

$$1^{\circ} \text{ Let } A_t = \{\omega \in \Omega : X(\omega) \leq t\}, \quad t, s \in \mathbb{R}$$

$$B_s = \{\omega \in \Omega : Y(\omega) \leq s\}.$$

Then  $F_{(X, Y)}(t, s) = P(A_t \cap B_s)$ . So if in particular

$X$  &  $Y$  are statistically independent, then



$$F_{(X,Y)}(t,s) = P(A_t | B_s) = F_X(t) F_Y(s)$$

and we can say that the joint distribution is equal to the product of distributions.

2<sup>nd</sup> let's take  $F_{(X,Y)}$  for  $(X,Y)$ , and fix  $t_0, s_0$ .

Then we can consider respectively:

$$\mathbb{R} \rightarrow t \longrightarrow F_{(X,Y)}(t, s_0)$$

$$\mathbb{R} \rightarrow s \longrightarrow F_{(X,Y)}(t_0, s).$$

Now we have

$$F_{(X,Y)}(t, s_0) = P(\{\omega \in \Omega : X(\omega) < t \text{ and } Y(\omega) < s_0\})$$

$$\xrightarrow{s_0 \rightarrow +\infty} P(\{\omega \in \Omega : X(\omega) < t \text{ and } Y(\omega) \in \mathbb{R}\}) =$$

$$= P(\{\omega \in \Omega : X(\omega) < t\} \cap \Omega) = P(\{\omega \in \Omega : X(\omega) < t\}) = F_X(t).$$

Similarly,

$$F_{(X,Y)}(t_0, s) \xrightarrow{t_0 \rightarrow +\infty} F_Y(s), \quad t, s \in \mathbb{R}.$$



This gives us the following theorem

Th 1. (on marginal distributions)

For any r.v.  $(X, Y)$  and C.P.F.  $F_{(X, Y)}$ ,

$$F_X(t) = \lim_{s \rightarrow +\infty} F_{(X, Y)}(t, s), \quad t \in \mathbb{R}$$

$$F_Y(s) = \lim_{t \rightarrow +\infty} F_{(X, Y)}(t, s), \quad s \in \mathbb{R}.$$

So  $F_X$  and  $F_Y$  are determined by  $F_{(X, Y)}$ .

If in addition  $X$  and  $Y$  are S.I., then

$$F_{(X, Y)} = F_X F_Y.$$

Remark In general,  $F_{(X, Y)} \neq F_X F_Y$  //

The case of discrete r.v.

We will limit our considerations further to the case where  $X$  and  $Y$  are discrete and finite, so we have:

$$X(\Omega) = \{x_1, x_2, \dots, x_n\}, \quad n \geq 2$$

$$Y(\Omega) = \{y_1, y_2, \dots, y_m\}, \quad m \geq 2$$



$$\forall_{1 \leq i \leq n} p_i := P(\{\omega \in \Omega: X(\omega) = x_i\}) ,$$

$$\forall_{1 \leq j \leq m} q_j := P(\{\omega \in \Omega: Y(\omega) = y_j\}) , \text{ and}$$

$$\text{of course } \sum_{1 \leq i \leq n} p_i = \sum_{1 \leq j \leq m} q_j = 1.$$

Then we have

$$(X, Y)(\Omega) = \{(x_i, y_j), i=1, \dots, n, j=1, \dots, m\}$$

For every  $i \in \{1, \dots, n\}$ ,  $j \in \{1, \dots, m\}$  we define

$$A_{ij} = \{\omega \in \Omega: (X, Y)(\omega) = (x_i, y_j)\} .$$

$$\text{Let } p_{ij} = P(A_{ij}) \text{ and } \mathcal{A} = \{A_{ij}, i=1, \dots, n, j=1, \dots, m\} .$$

Proposition 1.

For discrete finite vector  $(X, Y)$  as above we have

1. The family  $\mathcal{A}$  is a partition of  $\Omega$

$$2. \quad \forall_{i,j} \quad p_{ij} \in (0, 1)$$

$$3. \quad \sum_j \sum_i p_{ij} = 1$$



$$\forall \quad \sum_{j=1}^m p_{ij} = p_i$$

$$\forall \quad \sum_{i=1}^n p_{ij} = q_j$$

Proof. It is a task for YOU.

Now let's define the matrix  $P \in M_{n \times m}$ , where

$$P = [p_{ij}]_{n \times m} = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1m} \\ p_{21} & p_{22} & \dots & p_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ p_{i1} & p_{i2} & \dots & p_{im} \\ \vdots & \vdots & \ddots & \vdots \\ p_{n1} & p_{n2} & \dots & p_{nm} \end{bmatrix}$$

By Prop 1 the sum of all elements from the  $i$ th rows of  $P$ ,  $\sum_{j=1}^m p_{ij}$  defines the distribution of  $X$ ,  
for  $1 \leq i \leq n$

and similarly, the sum of all elements for the  $j$ th columns,  $\sum_{i=1}^n p_{ij}$  define the distribution of  $Y$ ,  
for  $1 \leq j \leq m$ .



Further, we denote

$$P_{i.} \stackrel{\text{def}}{=} \sum_{j=1}^m P_{ij} (= p_{i.})$$

$$P_{.j} \stackrel{\text{def}}{=} \sum_{i=1}^n P_{ij} (= q_j).$$

Th2 (on the distribution of finite discrete r.v.)

Every finite discrete r.v.  $(X, Y)$  can be uniquely represented by the matrix  $P = [P_{ij}]_{n \times m}$ , where

$$(i) \quad n = |X(\Omega)|, \quad m = |Y(\Omega)|$$

$$(i') \quad P_{ij} = P(\{\omega \in \Omega : (X, Y)(\omega) = (x_i, y_j)\}),$$

$$\text{where } X(\Omega) = \{x_1, \dots, x_n\}$$

$$Y(\Omega) = \{y_1, \dots, y_m\},$$

(ii)  $P_{i.}$ ,  $i=1, \dots, n$  and  $P_{.j}$ ,  $j=1, \dots, m$  determine the distributions of  $X$  &  $Y$ , so are marginals to  $(X, Y)$ .

In particular,  $X$  and  $Y$  are stoch. ind. iff

$$\forall_{i,j} \quad P_{i.} P_{.j} = P_{ij}$$

And vice versa, whenever we have the matrix  $[P_{ij}]_{n \times m}$  with the properties (i)-(iii), then there exist at least one R.E. and two r.v.  $X, Y$  for which (i) and (ii) hold.

Remark.

The matrix  $P$  is called the stochastic representation of the r.v.  $(X, Y)$ , so we write

$$(X, Y) \longrightarrow P_{(X, Y)}.$$

Example 1

Prove that for some  $a, b, c \in \mathbb{Z}$ , the matrix

$$A = \begin{bmatrix} 1/6 & 1/6 & 1/6 \\ 1/12 & 1/12 & 1/12 \\ a & b & c \end{bmatrix} \text{ represents the joint}$$

distribution for some r.v.  $(X, Y)$ .



We are going to apply Th. 2.

The  $a, b, c$  must satisfy the following conditions:

1)  $a, b, c \in (0, 1)$

2)  $3 \cdot \frac{1}{6} + 3 \cdot \frac{1}{12} + (a+b+c) = 1$

3)  $(\frac{1}{6} + \frac{1}{12} + a) + (\frac{1}{6} + \frac{1}{12} + b) + (\frac{1}{6} + \frac{1}{12} + c) = 1$

It means that

(\*)  $a+b+c = \frac{1}{4}$  and  $a, b, c \in (0, 1)$

It is (almost) clear that there exists an infinity many solutions of (\*).

Let  $a_0, b_0, c_0$  be one of them. Then

the marginal distributions are the following

$X: P_{1.0} = \frac{1}{2}, P_{2.0} = \frac{1}{4}, P_{3.0} = a_0 + b_0 + c_0 = \frac{1}{4}$

$Y: P_{.1} = a_0 + \frac{1}{4}, P_{.2} = b_0 + \frac{1}{4}, P_{.3} = c_0 + \frac{1}{4}.$

TASK 1

When  $X$  and  $Y$  from the ex 1 are st. ind.?



The problem of distribution of  $X+Y$  - as an application of the joint distribution.

We will show how to establish the distribution of  $Z = X+Y$  if the joint distribution is known.

By assumption we have the matrix  $P = [p_{ij}]_{n \times m}$

For  $Z = X+Y$  we can write

$$Z(\omega) = X(\omega) \oplus Y(\omega) \stackrel{\text{def}}{=} \{c \in \mathbb{R} : c = a+b, a \in X(\omega), b \in Y(\omega)\}$$

$\uparrow$   
 the complex union

Then  $Z$  has a discrete distribution, and for  $c \in Z(\Omega)$  we have

$$P(\{\omega \in \Omega : Z(\omega) = c\}) = P(\{\omega \in \Omega : \exists_{a,b} (X(\omega), Y(\omega)) = (a,b) \text{ \& } c = a+b\})$$

For given  $a \in X(\Omega)$  and  $b \in Y(\Omega)$ , let

$$A_{a,b} = \{\omega \in \Omega : (X(\omega), Y(\omega)) = (a,b)\}$$



then by additivity property of  $P$ ,

$$(\#) P(\{\omega \in \Omega: Z(\omega) = c\}) = \sum_{a+b=c} P(A_{a,b}) = \sum_{a+b=c} p_{a,b}$$

In particular, if  $X$  and  $Y$  are S.I., then

$$p_{a,b} = p_a \cdot p_b \quad \text{and hence}$$

$$(\#\#) P(\{\omega \in \Omega: Z(\omega) = c\}) = \sum_{a+b=c} p_a \cdot p_b.$$

The operations  $(\#)$  &  $(\#\#)$  are called the convolution operation of the distributions.

Example 2

Let  $X(\omega) = \{1, 2\}$ ,  $Y(\omega) = \{-1, 1\}$  and

$$(X, Y) \rightarrow P = \begin{matrix} & \begin{matrix} Y=-1 & Y=1 \end{matrix} \\ \begin{matrix} X=1 \\ X=2 \end{matrix} & \begin{bmatrix} 1/6 & 1/6 \\ 1/3 & 1/3 \end{bmatrix} \end{matrix} \begin{matrix} 1/3 \\ 2/3 \end{matrix} \quad \text{so}$$

$$P(\{\omega \in \Omega: X(\omega) = 1\}) = 1/3, \quad P(\{\omega \in \Omega: X(\omega) = 2\}) = 2/3$$

$$P(\{\omega \in \Omega: Y(\omega) = -1\}) = P(\{\omega \in \Omega: Y(\omega) = 1\}) = 1/2$$



So, we can write

$$X: \begin{array}{c|c} 1 & 2 \\ \hline 1/15 & 2/15 \end{array}$$

$$Y: \begin{array}{c|c} -1 & 1 \\ \hline 1/12 & 1/12 \end{array}$$

} as marginals

We note that  $X$  and  $Y$  are independent (Why?)

and  $Z(1) = \{0, 1, 2, 3\}$ .

We can apply (#1) for  $Z$ .

### TASK 2

When  $P = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$  represents some r.v.  $(X, Y)$ .

When  $X$  and  $Y$  are s.i.

### TASK 3

Compute  $Z = X + Y$  for  $X, Y$  from TASK 2.