

## 2.4 Independence of random variables

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### Definition

Suppose that the joint distribution sequence of a discrete random vector  $(\xi, \eta)$  satisfies

$$P(\xi = x_i, \eta = y_j) = P(\xi = x_i)P(\eta = y_j), \\ i, j = 1, 2, \dots,$$

then we call  $\xi$  and  $\eta$  mutually independent.

$$p_{ij} = p_{i\cdot} \cdot p_{\cdot j}, \quad i, j = 1, 2, \dots.$$

For any  $x$  and  $y$ ,

$$\begin{aligned}P(\xi \leq x, \eta \leq y) &= \sum_{x_i \leq x} \sum_{y_j \leq y} P(\xi = x_i, \eta = y_j) \\&= \sum_{x_i \leq x} P(\xi = x_i) \sum_{y_j \leq y} P(\eta = y_j) \\&= P(\xi \leq x)P(\eta \leq y).\end{aligned}$$

That is,

$$F(x, y) = F_\xi(x)F_\eta(y). \quad (2.62)$$

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On the contrary,

$$(2.62) \implies P(\xi = x_i, \eta = y_j) = P(\xi = x_i)P(\eta = y_j)$$

## Definition

Suppose that  $F(x, y)$ ,  $F_\xi(x)$  and  $F_\eta(y)$  are the joint distribution function and marginal distribution functions of  $(\xi, \eta)$  respectively. If

$$F(x, y) = F_\xi(x)F_\eta(y), \quad \forall x, y,$$

$$(i.e., \quad P(\xi \leq x, \eta \leq y) = P(\xi \leq x)P(\eta \leq y), \quad \forall x, y)$$

then we say  $\xi$  and  $\eta$  are independent.

## Theorem

*Suppose that  $p(x, y)$ ,  $p_\xi(x)$  and  $p_\eta(y)$  are the joint density function and marginal density functions of  $(\xi, \eta)$  respectively. Then  $\xi$  and  $\eta$  are independent if and only if*

$$p(x, y) = p_\xi(x)p_\eta(y).$$

**Proof.** For any  $x, y$ , it follows

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$$\begin{aligned} F(x, y) &= F_{\xi}(x)F_{\eta}(y) \\ \Leftrightarrow \int_{-\infty}^x \int_{-\infty}^y p(u, v)dudv &= \int_{-\infty}^x p_{\xi}(u)du \int_{-\infty}^y p_{\eta}(v)dv \end{aligned}$$

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This is the desired conclusion.

### Example

Suppose  $(\xi, \eta) \sim N(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2, r)$ . Find out the necessary and sufficient condition for  $\xi, \eta$  to be independent.

**Solution.** Note that  $\xi \sim N(\mu_1, \sigma_1^2)$  and  $\eta \sim N(\mu_2, \sigma_2^2)$ . By definition,

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**Solution.** Note that  $\xi \sim N(\mu_1, \sigma_1^2)$  and  $\eta \sim N(\mu_2, \sigma_2^2)$ . By definition,

$$\begin{aligned} \xi, \eta \text{ are independent} &\Leftrightarrow p(x, y) = p_\xi(x)p_\eta(y) \\ \Leftrightarrow &\frac{1}{\sqrt{2\pi}\sigma_1} \exp\left\{-\frac{(x - \mu_1)^2}{2\sigma_1^2}\right\} \\ &\times \frac{1}{\sqrt{2\pi}\sigma_2\sqrt{1-r^2}} \exp\left\{-\frac{\left[y - \mu_2 - \frac{r\sigma_2}{\sigma_1}(x - \mu_1)\right]^2}{2\sigma_2^2(1-r^2)}\right\} \\ &= \frac{1}{2\pi\sigma_1\sigma_2} \exp\left\{-\frac{1}{2}\left[\frac{(x - \mu_1)^2}{\sigma_1^2} + \frac{(y - \mu_2)^2}{\sigma_2^2}\right]\right\} \\ \Leftrightarrow &r = 0. \end{aligned}$$

$n$  random variables:

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$$F(x_1, \cdots, x_n) = F_1(x_1) \cdots F_n(x_n).$$

$$\begin{aligned} & \left( i.e., P(\xi_1 \leq x_1, \cdots, \xi_n \leq x_n) \right. \\ & \quad \left. = P(\xi_1 \leq x_1) \cdots P(\xi_n \leq x_n), \quad \forall x_1, \cdots, x_n \right) \end{aligned}$$

## Corollary

*If  $\xi_1, \dots, \xi_n$  are mutually independent, then so are any  $r$  random variables ( $2 \leq r < n$ ).*

**Proof.**



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**Proof.** By the definition of the independence of  $\xi_1, \dots, \xi_n$ , we have for all  $x_1, \dots, x_n$ ,

$$P(\xi_1 \leq x_1, \dots, \xi_n \leq x_n) = P(\xi_1 \leq x_1) \cdots P(\xi_n \leq x_n).$$

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It follows that

$$\begin{aligned} & P(\xi_{i_1} \leq x_{i_1}, \dots, \xi_{i_r} \leq x_{i_r}) \\ &= P(\xi_{i_1} \leq x_{i_1}) \cdots P(\xi_{i_r} \leq x_{i_r}), \quad \forall x_{i_1}, \dots, x_{i_r}. \end{aligned}$$

**思考题:** 1. Find  $n$  random variables (or vectors)  $\xi_1, \dots, \xi_n$ , such that they are not independent, but every  $r$  ( $2 \leq r < n$ ) of them are independent.

2. Suppose  $F$  is a distribution function. Find  $n$  random variables (or vectors)  $\xi_1, \dots, \xi_n$ , such that  $\xi_i \sim F$ ,  $i = 1, \dots, n$ ,  $\xi_1, \dots, \xi_n$  are not independent, but every  $r$  ( $2 \leq r < n$ ) of them are independent.

- $\xi_1, \dots, \xi_n$  are indept. iff (if and only if)

$$P(\xi_1 \in B_1, \dots, \xi_n \in B_n) = P(\xi_1 \in B_1) \cdots P(\xi_n \in B_n)$$

for any  $B_1, \dots, B_n \in \mathcal{B}$ .

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- An  $n$ -dimensional  $\boldsymbol{\xi}$  and an  $m$ -dimensional  $\boldsymbol{\eta}$  are indept. iff

$$P(\boldsymbol{\xi} \in A, \boldsymbol{\eta} \in B) = P(\boldsymbol{\xi} \in A)P(\boldsymbol{\eta} \in B),$$

for all  $A \in \mathcal{B}^n, B \in \mathcal{B}^m$ , equivalently,

$$P(\boldsymbol{\xi} \leq \boldsymbol{x}, \boldsymbol{\eta} \leq \boldsymbol{y}) = P(\boldsymbol{\xi} \leq \boldsymbol{x})P(\boldsymbol{\eta} \leq \boldsymbol{y}), \quad \forall \boldsymbol{x} \in \mathcal{R}^n, \boldsymbol{y} \in \mathcal{R}^m.$$

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- If two random vectors are independent, then so are their sub-vectors.

### Example

Suppose that  $\xi$  is a constant  $a$ , show  $\xi$  and  $\eta$  are independent for any random variable  $\eta$ .

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**Proof** Let  $B_1$  and  $B_2$  be two Borel sets. We want to prove

$$P(\xi \in B_1, \eta \in B_2) = P(\xi \in B_1)P(\eta \in B_2). \quad (*)$$



If  $a \notin B_1$ , then  $P(\xi \in B_1) = 0$  and

$$P(\xi \in B_1, \eta \in B_2) \leq P(\xi \in B_1) = 0.$$

(\*) is true.

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(\*) is true.

If  $a \in B_1$ , then  $P(\xi \in B_1) = 1$  and

$$\begin{aligned} P(\xi \in B_1, \eta \in B_2) &= P(\eta \in B_2) - P(\xi \notin B_1, \eta \in B_2) \\ &= P(\eta \in B_2). \end{aligned}$$

(\*) is also true.

思考题: Suppose that  $X$  and  $Y$  are independent,  $X + Y$  and  $X$  have the same distribution. Prove  $P(Y = 0) = 1$ .

## Infinite many of random variables:

### Definition

Suppose that  $\{\xi_n; n \geq 1\}$  is a sequence of random variables.

Then we call them mutually independent if for each  $n$ ,  $\xi_1, \dots, \xi_n$  are independent.

思考题: Find a sequence  $\{\xi_i; i \geq 1\}$  of random variables such that every two of them are independent, but every three of them are not independent.