Independent random variables

• Definition. Random variables  $X_1, \ldots, X_k$  are independent means

 $P(X_1 \in A_1 \text{ and } X_2 \in A_2 \text{ and } \cdots \text{ and } X_k \in A_k) = P(X_1 \in A_1)P(X_2 \in A_2)\cdots P(X_k \in A_k)$ 

for all  $A_1, A_2, \ldots, A_k \in \mathcal{B}(R)$ .

- We say  $X_1, \ldots, X_k$  are dependent if  $X_1, \ldots, X_k$  are not independent.
- Fact 1  $X_1, \ldots, X_k$  are independent if and only if

 $P(X_1 \leq x_1 \text{ and } X_2 \leq x_2 \text{ and } \cdots \text{ and } X_k \leq x_k) = P(X_1 \leq x_1)P(X_2 \leq x_2)\cdots P(X_k \leq x_k)$ 

for all  $x_1, \ldots, x_k \in R$ .

• Fact 2 Suppose that  $(X_1, \ldots, X_k)$  is a discrete random vector,  $X_1, \ldots, X_k$  are independent if and only if

$$p_{X_1,\ldots,X_k}(x_1,\ldots,x_k) = p_{X_1}(x_1)\cdots p_{X_k}(x_k)$$

for all  $x_1, \ldots, x_k \in R$ .

• Fact 3 Suppose that  $(X_1, \ldots, X_k)$  is has a joint PDF  $f_{X_1, \ldots, X_k}$ . Define

$$g(x_1,\ldots,x_k) = f_{X_1}(x_1)\cdots f_{X_k}(x_k)$$

for all  $x_1, \ldots, x_k \in (-\infty, \infty)$ , where  $f_{X_1}, \ldots, f_{X_k}$  are PDFs of  $X_1, \ldots, X_k$  respectively. Then g can be a PDF of  $(X_1, \ldots, X_k)$  if and only if  $X_1, \ldots, X_k$  are independent.

- Remark. Suppose that X, Y are random variables and (X, Y) has joint PDF. Let  $f_X$  be a PDF of X and  $f_Y$  is a PDF of Y. Let  $R_Y = \{y : f_Y(y) > 0\}$ . If X and Y are independent, then  $\{f_X : y \in R_Y\}$  is a version of the conditional PDF of X given Y.
- The following result can be used together with Fact 3 to show that random variables are not independent.

Fact 4 Suppose that g and  $f_{X_1,\ldots,X_k}$  are both PDFs of  $(X_1,\ldots,X_k)$ , and g and  $f_{X_1,\ldots,X_k}$  are continuous on an open set A, then  $g(x_1,\ldots,x_k) = f_{X_1,\ldots,X_k}(x_1,\ldots,x_k)$  for every  $(x_1,\ldots,x_k) \in A$ .

• Example 1. Suppose that (X, Y) has a joint PDF  $f_{X,Y}$ , where

$$f_{X,Y}(x,y) = \begin{cases} 1.5x + 0.5y & \text{if } (x,y) \in (0,1) \times (0,1); \\ 0 & \text{otherwise.} \end{cases}$$

Show that X and Y are not independent.

Sol. Compute  $f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy$  and  $f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx$ , then we have

$$f_X(x) = \begin{cases} 1.5x + 0.25 & \text{if } x \in (0,1); \\ 0 & \text{otherwise} \end{cases}$$

and

$$f_Y(y) = \begin{cases} 0.75 + 0.5y & \text{if } y \in (0,1); \\ 0 & \text{otherwise} \end{cases}$$

Let  $g(x,y) = f_X(x)f_Y(y)$  for  $(x,y) \in R^2$ . Then g and  $f_{X,Y}$  are both continuous on  $(0,1) \times (0,1)$ . If g is a PDF of (X,Y), then by Fact 4,  $g = f_{X,Y}$  on  $(0,1) \times (0,1)$ . It is clear that on  $(0,1) \times (0,1)$ ,

$$g(x,y) = (1.5x + 0.25)(0.75 + 0.5y) \neq 1.5x + 0.5y = f_{X,Y}(x,y)$$

for some  $(x, y) \in (0, 1) \times (0, 1)$ , so g cannot be a PDF of (X, Y), which implies that X and Y are not independent by Fact 3.

Below is another way to deduce that X and Y are not independent based on the  $f_{X,Y}$ ,  $f_X$  and  $f_Y$  computed above. Suppose that X and Y are independent, then

$$P(X \le s \text{ and } Y \le t) = P(X \le s)P(X \le t) \text{ for } (s,t) \in (0,1) \times (0,1),$$
(1)

which implies that for  $(s, t) \in (0, 1) \times (0, 1)$ ,

$$\int_0^s \int_0^t (1.5x + 0.5y) dy dx = \int_0^s (1.5x + 0.25) dx \int_0^t (0.5y + 0.75) dy.$$

Take the partial derivative  $\partial^2/\partial t\partial s$  at both sides of the above equation and we have

$$(1.5s + 0.5t) = (1.5s + 0.25)(0.5t + 0.75)$$

for all  $(s,t) \in (0,1) \times (0,1)$ , which is clearly not true. Therefore, (1) does not hold and X and Y are not independent.

• Example 2. Suppose that (X, Y) has a joint PDF  $f_{X,Y}$ , where  $f_{X,Y}(x, y) = g(x)h(y)$  for  $(x, y) \in \mathbb{R}^2$  for some functions g and h. Show that X and Y are independent.

The proof is left as an exercise.

• Example 3. Suppose that (X, Y) is a discrete random vector with joint PMF  $p_{X,Y}$ , where

$$p_{X,Y}(x,y) = \begin{cases} 0.5 & \text{if } (x,y) = (1,2); \\ 0.3 & \text{if } (x,y) = (2,3); \\ 0.2 & \text{if } (x,y) = (1,5); \\ 0 & \text{otherwise.} \end{cases}$$

Determine whether X and Y are independent.

Ans. No since  $p_{X,Y}(2,5) = 0 \neq p_X(2)p_Y(5) = 0.3 \cdot 0.2$ .

• Fact 5 Suppose that X and Y are independent and E(u(X)) and E(v(Y)) are finite, then

$$E(u(X)v(Y)) = E(u(X))E(v(Y))$$

• Fact 6 Suppose that (X, Y) has joint MGF  $M_{X,Y}$ , where  $M_{X,Y}(t_1, t_2)$  is finite for  $|t_1| < \delta$  and  $|t_2| < \delta$  for some  $\delta > 0$ . Then X and Y are independent if and only if

$$M_{X,Y}(t_1, t_2) = M_{X,Y}(t_1, 0)M_{X,Y}(0, t_2)$$

for  $|t_1| < \delta$  and  $|t_2| < \delta$ .

• Example 4. For  $\mu \in (-\infty, \infty)$  and  $\sigma > 0$ , define

$$f_{\mu,\sigma}(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/(2\sigma^2)}$$

for  $x \in (-\infty, \infty)$ . It can be shown that  $\int_{-\infty}^{\infty} f_{\mu,\sigma}(x) dx = 1$  (see the remark after Homework problem 35 for example). Suppose that Z is a random variable with PDF  $f_{\mu,\sigma}$ . Suppose that X and Y are independent random variables with PDFs  $f_{1,3}$  and  $f_{2,4}$  respectively.

- (a) Find the MGF of Z.
- (b) Find the joint MGF of (X, Y).
- (c) Find the MGF of X Y.
- (d) Find a PDF of X Y.

A sketch of solution.

(a) Let  $M_Z$  be the MGF of Z, then from the solution to Problem 30(b), we have

$$M_Z(t) = E(e^{tZ}) = \int_{-\infty}^{\infty} e^{tz} f_{\mu,\sigma}(z) dz = e^{\mu t + 0.5\sigma^2 t^2}$$

for  $t \in (-\infty, \infty)$ .

(b) Let  $M_{X,Y}$  be the MGF of (X, Y), then by independence,

$$M_{X,Y}(t_1, t_2) = M_X(t_1)M_Y(t_2) = e^{t_1 + 4.5t_1^2}e^{2t_2 + 8t_2^2}$$

for  $(t_1, t_2) \in R^2$ .

(c) Let M be the MGF of X - Y, then

$$M(t) = M_{X,Y}(t, -t) = e^{t+4.5t^2} e^{-2t+8(-t)^2} = e^{-t-12.5t^2}$$

for  $t \in (-\infty, \infty)$ .

- (d) Since the MGF of X Y is the same as the MGF of Z when  $\mu = -1$ and  $\sigma = 5$  and  $f_{-1,5}$  is a PDF of Z when  $\mu = -1$  and  $\sigma = 5$ ,  $f_{-1,5}$  is a PDF of X - Y.
- The distribution of Z in Example 4 is called the normal distribution with mean  $\mu$  and variance  $\sigma^2$ , denoted by  $N(\mu, \sigma^2)$ .