

Chapter four  
The three or more random variables joint probability distribution

1).The joint probability distribution is three or more random variables which values and the joint probability density function is consist of the probability distribution. The joint probability distribution can derived to the joint distribution function, coefficient, joint conditional probability distribution in a special value region, the conditional probability distribution and the marginal probability distribution.

This chapter is about the two random variables joint probability distribution.

Suppose  $X_1 \sim U(0,1)$ ,  $X_2 \sim U(-1,1)$ ,  $X_3 \sim U(-2,2)$ ,  $X_1, X_2$  and  $X_3$  are independent random variables.

1.1)the joint probability density function is

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = \begin{cases} \frac{1}{8}, & 0 < x_1 < 1, -1 < x_2 < 1, -2 < x_3 < 2, \\ 0, & \text{otherwise} \end{cases}$$

1.2)the joint distribution function is

$$F_{X_1, X_2, X_3}(x_1, x_2, x_3) = \begin{cases} 0, & x_1 \leq 0 \quad \text{or} \quad x_2 \leq -1 \quad \text{or} \quad x_3 \leq -2 \\ \frac{x_1 \times (x_2 + 1) \times (x_2 + 2)}{8}, & 0 < x_1 < 1, -1 < x_2 < 1, -2 < x_3 < 2, \\ \frac{(x_2 + 1) \times (x_2 + 2)}{8}, & x_1 \geq 1, -1 < x_2 < 1, -2 < x_3 < 2, \\ \frac{x_1 \times (x_2 + 2)}{8}, & 0 < x_1 < 1, x_2 \geq 1, -2 < x_3 < 2, \\ \frac{x_1 \times (x_2 + 1)}{8}, & 0 < x_1 < 1, -1 < x_2 < 1, x_3 \geq 2, \\ x_1, & 0 < x_1 < 1, x_2 \geq 1 \\ 1, & x_1 \geq 1, x_2 \geq 1, x_3 \geq 2 \end{cases}$$

1.3)The marginal moment or joint monet,

$$E[H(X_1, X_2, X_3)] = \int \int \int H(x_1, x_2, x_3) \times f_{X_1, X_2, X_3}(x_1, x_2, x_3) \times dx_1 \times dx_2 \times dx_3,$$

$$H(X_1, X_2, X_3) \text{ can be } X_1, X_1 + X_2, X_1 + X_2 + X_3, (X_1 + X_2 + X_3)^3, \\ (X_1 + X_2 + X_3) \times \cos(X_2 \times \pi) \times \exp(-X_3), \dots \text{ etc.}$$

1.4) The joint conditional probability distribution in a special value region,

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3 | -1 \leq x_1 + x_2 + x_3 \leq 1), F_{X_1, X_2, X_3}(x_1, x_2, x_3 | -1 \leq x_1 + x_2 + x_3 \leq 1),$$

1.5) The conditional probability distribution,

$$f_{X_1|X_2=x_2}(x_1|X_2=x_2) = \frac{f_{X_1, X_2}(x_1, x_2)}{f_{X_2}(x_2)}, \quad f_{X_2|X_1=x_1}(x_2|X_1=x_1) = \frac{f_{X_1, X_2}(x_1, x_2)}{f_{X_1}(x_1)},$$

$$F_{X_1|X_2=x_2}(x_1|X_2=x_2) = \int_0^{x_1} f_{X_1|X_2=x_2}(x_1|X_2=x_2) dx_1,$$

$$F_{X_2|X_1=x_1}(x_2|X_1=x_1) = \int_0^{x_2} f_{X_2|X_1=x_1}(x_2|X_1=x_1) dx_2,$$

$$E[H(X_1)|X_2=x_2] = \int H(x_1) \times f_{X_1|X_2=x_2}(x_1|X_2=x_2) dx_1,$$

$$E[H(X_2)|X_1=x_1] = \int H(x_2) \times f_{X_2|X_1=x_1}(x_2|X_1=x_1) dx_2,$$

$$f_{X_1, X_2 | X_3 = x_3}(x_1, x_2 | X_3 = x_3) = \frac{f_{X_1, X_2, X_3}(x_1, x_2, x_3)}{f_{X_3}(x_3)},$$

$$E[H(X_1, X_2) | X_3 = x_3] = \int \int H(x_1, x_2) \times f_{X_1, X_2 | X_3 = x_3}(x_1, x_2 | X_3 = x_3) \times dx_1 \times dx_2, ,$$

1.6) The marginal probability distribution,

$$f_{X_1}(x_1) = \int \int f_{X_1, X_2, X_3}(x_1, x_2, x_3) \times dx_2 \times dx_3,$$

$$f_{X_2}(x_2) = \int \int f_{X_1, X_2, X_3}(x_1, x_2, x_3) \times dx_1 \times dx_3,$$

$$f_{X_3}(x_3) = \int \int f_{X_1, X_2, X_3}(x_1, x_2, x_3) \times dx_1 \times dx_2,$$

This software P\_S\_CCC.exe can support their functions, but the can show the one variable, two random variables and three random variables probability distribution only.

1.7) Three or more random variables joint probability distribution will be transferred to the other joint probability distribution or marginal probability distribution. The transformation need 1) the transformation equation, 2) new random variable range. The process of transformation in marginal probability distribution.

step 1. The  $X_1, X_2, X_3$  joint pdf or df or their special distribution name. There are the joint pdf of  $X_1, X_2, X_3, , f_{X_1, X_2, X_3}(x_1, x_2, x_3)$ .

step 2. The transferred mathematical equation,

$$Y_1 = H_1(X_1, X_2, X_3), Y_2 = H_2(X_1, X_2, X_3), Y_3 = H_3(X_1, X_2, X_3),$$

step 3. The jacobian and  $X_1 = G_1(Y_1, Y_2, Y_3), X_2 = G_2(Y_1, Y_2, Y_3), X_3 = G_3(Y_1, Y_2, Y_3),$

are existed,  $J = \frac{\partial(x_1, x_2, x_3)}{\partial(y_1, y_2, y_3)} = \begin{vmatrix} \frac{\partial G_1(y_1, y_2, y_3)}{\partial y_1} & \frac{\partial G_1(y_1, y_2, y_3)}{\partial y_2} & \frac{\partial G_1(y_1, y_2, y_3)}{\partial y_3} \\ \frac{\partial G_2(y_1, y_2, y_3)}{\partial y_1} & \frac{\partial G_2(y_1, y_2, y_3)}{\partial y_2} & \frac{\partial G_2(y_1, y_2, y_3)}{\partial y_3} \\ \frac{\partial G_3(y_1, y_2, y_3)}{\partial y_1} & \frac{\partial G_3(y_1, y_2, y_3)}{\partial y_2} & \frac{\partial G_3(y_1, y_2, y_3)}{\partial y_3} \end{vmatrix},$

step 4. The new random variables joint pdf

$$\int \int \int f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) dy_3 dy_2 dy_1 = \int \int \int f_{X_1, X_2, X_3}(x_1, x_2, x_3) \times J \times dx_3 \times dx_2 \times dx_1.$$

$$f_{Y_1}(y_1) = \int \int f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) dy_2 dy_3, f_{Y_2}(y_2) = \int \int f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) dy_1 dy_3,$$

$$f_{Y_3}(y_3) = \int \int f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) dy_1 dy_2,$$

Sometimes, the transformation is a big problem when the transferred mathematical equation is not matched to the random variables.

1.8) How to find the joint probability distribution in the software P\_S\_CCC.exe.

1.8.1)  $X_1, X_2, X_3$  marginal probability distribution are known and  $X_1, X_2$  and  $X_3$  are independent random variables.

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = f_{X_1}(x_1) \times f_{X_2}(x_2) \times f_{X_3}(x_3).$$

1.8.2)  $X_1$  marginal probability distribution are known and  $X_2$  conditional probability distribution are known, but the parameters are function of  $X_1$  and the  $X_3$  conditional probability distribution of  $X_1$  and  $X_2$ .

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = f_{X_1}(x_1) \times f_{X_2|X_1=x_1}(x_2|X_1=x_1) \times f_{X_3|X_1=x_1, X_2=x_2}(x_3|X_1=x_1, X_2=x_2)$$

1.8.3)  $X_1, X_2, X_3$  marginal probability distribution are known and  $X_1, X_2$  and  $X_3$  are independent random variables. But the values range of  $X_1, X_2$  and  $X_3$  are the special range constrains.

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = f_{X_1}(x_1) \times f_{X_2}(x_2) \times f_{X_3}(x_3).$$

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3 | a \leq X_1 + X_2 + X_3 \leq b) = \frac{f_{X_1, X_2, X_3}(x_1, x_2, x_3)}{P(a \leq X_1 + X_2 + X_3 \leq b)},$$

1.8.4)  $X_1$  marginal probability distribution are known and  $X_2$  conditional probability distribution are known, but the parameters are function of  $X_1$  and the  $X_3$  conditional probability distribution of  $X_1$  and  $X_2$ .

The values range of  $X_1, X_2$  and  $X_3$  are the special constrains.

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = f_{X_1}(x_1) \times f_{X_2|X_1=x_1}(x_2|X_1=x_1) \times f_{X_3|X_1=x_1, X_2=x_2}(x_3|X_1=x_1, X_2=x_2),$$

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3 | a \leq X_1 + X_2 + X_3 \leq b) = \frac{f_{X_1, X_2, X_3}(x_1, x_2, x_3)}{P(a \leq X_1 + X_2 + X_3 \leq b)},$$

Note: i.i.d. is independently and identically distributed.