

## Transformation

Thm:  $X \sim F_X(x)$ ,  $Y = g(X)$ , Supports  $\mathcal{X}$  and  $\mathcal{Y}$  are defined as

$$\mathcal{X} = \{x: f_X(x) > 0\}, \quad \mathcal{Y} = \{y: y = g(x), x \in \mathcal{X}\}$$

1° If  $g(\cdot)$  is increasing on  $\mathcal{X}$ , then  $Y \sim F_X(g^{-1}(y))$  for  $y \in \mathcal{Y}$

2° If  $g(\cdot)$  is decreasing on  $\mathcal{X}$ , and  $X$  is a continuous r.v., then

$$Y \sim 1 - F_X(g^{-1}(y)) \text{ for } y \in \mathcal{Y}$$

Thm: (Probability integral transformation)

1°  $X \sim F_X(x)$ , which is continuous, and  $Y = F_X(X)$ ,  
then  $Y \sim U(0,1)$ .

2°  $U \sim U[0,1]$ ,  $\forall$  cdf  $F(\cdot)$ , let  $X = F^{-1}(U)$ ,  
then  $X \sim F(x)$

Main proof of 1°: For  $0 < y < 1$ ,

$$P(Y \leq y) = P(F_X(X) \leq y)$$

$$= P(F_X^{-1}(F_X(X)) \leq F_X^{-1}(y)) \quad (\because F_X^{-1} \text{ increasing})$$

$$= P(X \leq F_X^{-1}(y))$$

$$= F_X(F_X^{-1}(y))$$

$$= y$$

$$\text{and } P(Y \leq y) = \begin{cases} 1, & y \geq 1 \\ 0, & y \leq 0 \end{cases}$$

Special examples:

1° If  $X_1 \sim \Gamma(\alpha_1, \beta)$ ,  $X_2 \sim \Gamma(\alpha_2, \beta)$  and  $X_1 \perp X_2$ .

then  $X_1 + X_2 \sim \Gamma(\alpha_1 + \alpha_2, \beta)$

2° If  $Z \sim N(0, 1)$ , then  $Z^2 \sim \Gamma(\frac{1}{2}, 2) = \chi^2(1)$ ;

If  $Z_1, \dots, Z_n$  iid.  $N(0, 1)$ , then  $Z_1^2 + \dots + Z_n^2 \sim \Gamma(\frac{n}{2}, 2) = \chi^2(n)$ .

3° If  $X_1, \dots, X_{n+1}$  iid. Exponential (1),  $S_k = \sum_{i=1}^k X_i$  ( $k=1, \dots, n+1$ ),  
then  $(\frac{S_1}{S_{n+1}}, \dots, \frac{S_n}{S_{n+1}}) \sim (U_{(1)}, \dots, U_{(n)})$ , where  $U_i \sim U[0, 1]$ .

proof:  $(X_1, \dots, X_{n+1}) \sim e^{-x_1} \dots e^{-x_{n+1}} \mathbb{I}(x_1 > 0, \dots, x_{n+1} > 0)$   
 $\underline{S}$  is a linear, nonsingular transformation of  $\underline{X}$ , and

$$|J_{\underline{X}}(\underline{S})| = 1$$

then  $(S_1, \dots, S_{n+1}) \sim e^{-S_{n+1}} \mathbb{I}(0 < S_1 < S_2 < \dots < S_{n+1})$

Let  $(W_1, \dots, W_n, W_{n+1}) = (\frac{S_1}{S_{n+1}}, \dots, \frac{S_n}{S_{n+1}}, S_{n+1})$ ,

then  $\underline{W}$  is a 1-1 transformation with

$$|J_{\underline{S}}(\underline{W})| = \frac{1}{S_{n+1}^n}$$

$\therefore (W_1, \dots, W_{n+1}) \sim W_{n+1}^n e^{-W_{n+1}} \mathbb{I}(0 < W_1 < \dots < W_n < 1, W_{n+1} > 0)$

$$= [n! \mathbb{I}(0 < W_1 < \dots < W_n < 1)] \cdot [\frac{W_{n+1}^n}{n!} e^{-W_{n+1}} \mathbb{I}(W_{n+1} > 0)]$$

$\therefore (W_1, \dots, W_n) \perp W_{n+1}$ , and  $(W_1, \dots, W_n) \sim n! \mathbb{I}(0 < W_1 < \dots < W_n < 1)$ ,  
 $\sim (U_{(1)}, \dots, U_{(n)})$   $\square$ .

4° Given  $U_{(r)} \sim \text{Beta}(r, n+1-r)$ , we get  $U_{(b)} - U_{(a)} \sim U_{(b-a)}$  ( $b > a$ )

proof:  $U_{(b)} - U_{(a)} \stackrel{d}{=} \frac{S_b - S_a}{S_{n+1}} = \frac{X_{a+1} + \dots + X_b}{S_{n+1}}$   
 $\stackrel{d}{=} \frac{X_1 + \dots + X_{b-a}}{S_{n+1}}$  (exchangeable)  
 $\sim U_{(b-a)}$   
 $\sim \text{Beta}(b-a, n+1-b+a)$   $\square$ .

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Function of two random variables  $Z = g(X, Y)$

1<sup>o</sup> Direct method:  ~~$P_Z(z) = \int P_{X,Y}(x,y) \delta(z - g(x,y)) dx dy$~~

Transformation  
~~Theory~~

$$F_Z(z) = P\{Z \leq z\}$$

$$= P\{g(X, Y) \leq z\}$$

$$= P\{(X, Y) \in G_Z\}$$

$$= \iint_{G_Z} f_{X,Y}(x,y) dx dy$$

$$f_Z(z) = \frac{d}{dz} F_Z(z)$$

Eg.:  $Z = X + Y$

$$f_Z(z) = \int_{-\infty}^{+\infty} f_{X,Y}(z-y, y) dy$$

If  $X \perp Y$ ,  $f_Z(z) =$   ~~$f_X(z) f_Y(z)$~~

$$\int_{-\infty}^{+\infty} f_X(z-y) f_Y(y) dy$$

$$= (f_X * f_Y)(z)$$

Characteristic functions:

$$\Phi_Z(w) = \Phi_X(w) \cdot \Phi_Y(w)$$

1<sup>o</sup>  ~~$f_X(x) = \lambda_x e^{-\lambda_x x}$~~   
 $X, Y$  exponential ( $\lambda_x > \lambda_y$ )

$$f_Z(z) = \frac{\lambda_x \lambda_y}{\lambda_x - \lambda_y} [e^{-\lambda_y z} - e^{-\lambda_x z}] u(z)$$

2<sup>o</sup>  $X, Y$  Gaussian

$$f_Z(z) = N(z; m_x + m_y, \sigma_x^2 + \sigma_y^2)$$

Eg.:  $Z = \max\{X, Y\}$ ,  $f_Z(z) = \frac{d}{dz} F_{X,Y}(z, z)$

Eg 2:  
 ~~$Y = AX + b$~~

$$\Phi_Y(w) = E e^{i w^T (AX + b)}$$

$$= e^{i w^T b} E e^{i (A^T w)^T X}$$

$$= e^{i w^T b} \Phi_X(A^T w)$$

3<sup>o</sup> If  $X, Y$  Poisson,  
 $Z \sim \text{Poisson}(\theta_x + \theta_y)$

2° Conditional method

1° Find  $f_{Z|X}$  from  $f_{Y|X}$  and  $g_X(Y) = g(X, Y)$

$$2° f_Z(z) = \int_{-\infty}^{+\infty} f_{Z|X}(z|x) f_X(x) dx$$

3° Transformation of r.v.'s:

$$\begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix} = \begin{pmatrix} g_1(X_1, X_2) \\ g_2(X_1, X_2) \end{pmatrix}$$

$$f_Y(y) = \sum_{x_i: g(x_i) = y} \frac{f_X(x_i)}{|J_g(x_i)|}$$

$J_g$  is the Jacobian matrix of  $g(\cdot)$

Eg.:  $Y = AX + b$ ,  $A$  invertible.  
(Affine transformation)

$$f_Y(y) = \frac{f_X(A^{-1}(y-b))}{|\det(A)|}$$

If  $X$  jointly Gaussian, then  $Y$  also jointly Gaussian.

$$\begin{cases} m_Y = A m_X + b \\ K_Y = A K_X A^T \end{cases}$$

If  $X \sim \text{beta}(\alpha, \beta)$ ,  
 $Y \sim \text{beta}(\alpha+\beta, \gamma)$ ,  
and  $X \perp\!\!\!\perp Y$ , then

$$XY \sim \text{beta}(\alpha, \beta+\gamma)$$

Eg.: 2°  $\begin{pmatrix} X \\ Y \end{pmatrix} \rightarrow \begin{pmatrix} R \\ \theta \end{pmatrix}$

$$f_{R, \theta}(r, \theta) = r f_{X, Y}(r \cos \theta, r \sin \theta)$$

If  $X, Y$  are independent,  
standard Gaussian,  
then  $\frac{X}{Y} \sim \text{Cauchy}(1)$