

# Lesson 1.4

## Bivariate Transformation

### Introduction

So far in this module, we have talked about transformations involving a single random variable  $Y$ . It is sometimes of interest to consider **bivariate transformation** such as

$$\begin{aligned}U_1 &= g_1(Y_1, Y_2) \\ U_2 &= g_2(Y_1, Y_2)\end{aligned}$$

To discuss such transformations, we will assume that  $Y_1$  and  $Y_2$  are jointly **continuous** random variables. Furthermore, for the following methods to apply, the transformation needs to be **one-to-one**. We start with the joint distribution of  $\mathbf{Y} = (Y_1, Y_2)$ . Our first goal is to find the joint distribution of  $\mathbf{U} = (U_1, U_2)$ .

### Bivariate Transformation

Suppose that  $\mathbf{Y} = (Y_1, Y_2)$  is a continuous random vector with joint PDF  $f_{Y_1, Y_2}(y_1, y_2)$ . Let  $g : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be a continuous one-to-one vector-valued mapping from  $R_{Y_1, Y_2}$  to  $R_{U_1, U_2}$ , where  $U_1 = g_1(Y_1, Y_2)$  and  $U_2 = g_2(Y_1, Y_2)$ , and where  $R_{Y_1, Y_2}$  and  $R_{U_1, U_2}$  denote two-dimensional supports of  $\mathbf{Y} = (Y_1, Y_2)$  and  $\mathbf{U} = (U_1, U_2)$ , respectively.

If  $g_1^{-1}(u_1, u_2)$  and  $g_2^{-1}(u_1, u_2)$  have continuous partial derivatives with respect to both  $u_1$  and  $u_2$  and the Jacobian,  $J$ , where “det” denote “determinant”,

$$J = \det \begin{bmatrix} \frac{\partial g_1^{-1}(u_1, u_2)}{\partial u_1} & \frac{\partial g_1^{-1}(u_1, u_2)}{\partial u_2} \\ \frac{\partial g_2^{-1}(u_1, u_2)}{\partial u_1} & \frac{\partial g_2^{-1}(u_1, u_2)}{\partial u_2} \end{bmatrix} \neq 0$$

then

$$f_{U_1, U_2}(u_1, u_2) = \begin{cases} f_{Y_1, Y_2}[g_1^{-1}(u_1, u_2), g_2^{-1}(u_1, u_2)] \times |J|, & (u_1, u_2) \in R_{U_1, U_2} \\ 0, & \text{elsewhere} \end{cases} \quad (1)$$

where  $|J|$  denotes the absolute value of  $J$ .

To summarize, the steps involved in bivariate transformation are:

1. Find the joint distribution  $f_{Y_1, Y_2}(y_1, y_2)$  of  $Y_1$  and  $Y_2$ .
2. Find the  $R_{U_1, U_2}$  the support of  $\mathbf{U} = (U_1, U_2)$ .
3. Find the inverse transformations  $y_1 = g_1^{-1}(u_1, u_2)$  and  $y_2 = g_2^{-1}(u_1, u_2)$ .
4. Find the Jacobian,  $J$ , of the inverse transformation.
5. Use (1) to find  $f_{U_1, U_2}(u_1, u_2)$  the joint distribution of  $U_1$  and  $U_2$ .