

This leads us to the joint density function of (Y_1, Y_2, \dots, Y_m) .

Consequence

There is a general version of theorem.

The transformation/mapping $g = (g_1, g_2, \dots, g_n)$
 $g: \mathbb{R}^n \rightarrow \mathbb{R}^n$ does not need to be one-to-one on the whole domain \mathbb{R}^n .

In that case, we partition (divide it into disjoint subsets) \mathbb{R}^n such that in each partition g is one-to-one. The partition must be finite.

Example $g: \mathbb{R} \rightarrow \mathbb{R}$ such that $g(x) = x^2$.

Then g is not one-to-one. But if we divide \mathbb{R} into two disjoint halves $(-\infty, 0]$ and $(0, \infty)$.

Then $g: (-\infty, 0] \rightarrow \mathbb{R}$ and $g: (0, \infty) \rightarrow \mathbb{R}$ is one-to-one and ~~is~~ ~~increasing~~ so invertible.

Let A_1, A_2, \dots, A_k be a partition of \mathbb{R}^n that is,

$A_i \cap A_j = \emptyset$ for $i \neq j$ and $\bigcup_{i=1}^k A_i = \mathbb{R}^n$. Further

assume that the restriction of g on A_i is one-to-one and so invertible for every $i=1, 2, \dots, k$. Let

~~denote the~~ $g^{(i)}$ denote the restriction of g on A_i .

Recall $\underline{g} = (g_1, g_2, \dots, g_n)$.

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$$\underline{g}^{(i)} = (g_1^{(i)}, g_2^{(i)}, \dots, g_n^{(i)}) \text{ for } 1 \leq i \leq n.$$

Define $\underline{h}^{(i)}$ to be the inverse of $\underline{g}^{(i)}$. Then

for each $i = 1, 2, \dots, n$, we can compute

$$J^{(i)} = \begin{vmatrix} \frac{\partial h_1^{(i)}}{\partial y_1} & \frac{\partial h_1^{(i)}}{\partial y_2} & \dots & \frac{\partial h_1^{(i)}}{\partial y_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial h_n^{(i)}}{\partial y_1} & \frac{\partial h_n^{(i)}}{\partial y_2} & \dots & \frac{\partial h_n^{(i)}}{\partial y_n} \end{vmatrix}$$

It is important to have $\det(J^{(i)}) \neq 0$ for all $i = 1, 2, \dots, n$.

~~Then~~ Suppose that the random vector $\underline{X} = (X_1, X_2, \dots, X_n)$ has joint density $f(x_1, \dots, x_n)$. Then the joint density of $(Y_1, Y_2, \dots, Y_n) = \underline{g}(X_1, \dots, X_n)$

$$= (g_1(X_1, X_2, \dots, X_n), g_2(X_1, X_2, \dots, X_n), \dots, g_n(X_1, X_2, \dots, X_n))$$

is given by

$$w(y_1, y_2, \dots, y_n) = \sum_{i=1}^n |J_i| f(h_1^{(i)}(y_1, y_2, \dots, y_n), h_2^{(i)}(y_1, y_2, \dots, y_n), \dots, h_n^{(i)}(y_1, y_2, \dots, y_n))$$

$$1[(y_1, y_2, \dots, y_n) \in \underline{g}(A_i)]$$