

The random vector $X \equiv (X_1, \dots, X_m) \in \mathbb{R}^m$ is considered random variable as a whole because of the joint PDF $f \equiv f(x_1, \dots, x_m)$ that is used to calculate expectation values with respect to each component X_i . In other words, the components X_1, \dots, X_m of the random vector X are in general not-independent. This is reflected in the Σ as $\sigma_{ij}^2 \neq 0$ in general. Thus,

Let $f(x_1, \dots, x_m)$ be the PDF of a random $X \in \mathbb{R}^m$

\rightarrow If $f(x_1, \dots, x_m) \equiv f_1(x_1) \dots f_m(x_m) \Leftrightarrow X_1, \dots, X_m$ mutually independent $\Leftrightarrow \forall X_i, X_j$ no relation $X_i \neq g(X_j) \Rightarrow X_1, \dots, X_m$ mutually uncorrelated $\Leftrightarrow \forall X_i, X_j$ Σ is diagonal $\sigma_{ij}^2 = 0$

\rightarrow If X_1, \dots, X_m are not mutually independent then X_1, \dots, X_m can still be mutually uncorrelated $\Leftrightarrow \sigma_{ij}^2 = 0 \forall i, j = 1, \dots, m$ but $\forall X_i, X_j$ relation $X_i \equiv g(X_j)$

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(*) The latter is in particular relevant in regression theory $Y = h(X_1, \dots, X_m) + \epsilon \dots$

What about the correlation or independence between two random vectors $X \in \mathbb{R}^{m_x}, Y \in \mathbb{R}^{m_y}$

joint CDF: $F(x, y) \equiv P[X \leq x, Y \leq y] \equiv P[X_1 \leq x_1, \dots, X_{m_x} \leq x_{m_x}, Y_1 \leq y_1, \dots, Y_{m_y} \leq y_{m_y}]$

joint PDF: $f(x, y) \equiv P[X \in [x, x+dx], Y \in [y, y+dy]] \equiv \partial^{m_x+m_y} F(x, y) / \partial x_1 \dots \partial x_{m_x} \partial y_1 \dots \partial y_{m_y} = f(x_1, \dots, x_{m_x}, y_1, \dots, y_{m_y})$

marginal PDF: $f_X(x) \equiv P[X \in [x, x+dx], Y \in \mathbb{R}] \equiv \int_{-\infty}^{\infty} f(x, y) dy_1 \dots dy_{m_y}$

$f_Y(y) \equiv P[Y \in [y, y+dy], X \in \mathbb{R}] \equiv \int_{-\infty}^{\infty} f(x, y) dx_1 \dots dx_{m_x}$

conditional PDF: $f_{X|Y}(x/y) \equiv P[X \in [x, x+dx] / Y = y] \equiv f(x, y) / f_Y(y)$

$f_{Y|X}(y/x) \equiv P[Y \in [y, y+dy] / X = x] \equiv f(x, y) / f_X(x)$

2 random vectors $X \in \mathbb{R}^{m_x}, Y \in \mathbb{R}^{m_y}$ are independent iff $f(x, y) \equiv f_X(x) \cdot f_Y(y)$

(*) the random vectors X, Y do NOT need to have the same dimension to define joint PDF or independence. examine!

(*) It can be that $\rightarrow X_1, \dots, X_{m_x}$ mutually independent, Y_1, \dots, Y_{m_y} mutually independent BUT X, Y dependent

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