

Part 1 : multivariate random vectors and
the multivariate normal distribution

[L1]
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Random vector, $f(x)$, $F(x)$

Random vector = vector of random variables (RV)

$$\underset{p \times 1}{\mathbf{X}} = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_p \end{bmatrix} \quad \text{EKS: height, weight, phys. act, diet}$$

We will focus on continuous random variables.

- 1) $\mathbf{X}$ has a joint distribution (density) function (pdf)

$$f(x) = f(x_1, x_2, \dots, x_p)$$

Remember $f(x) \geq 0 \quad \forall x$

$$\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_p) dx_1 \dots dx_p = 1$$

- 2) $\mathbf{X}$ has a joint cumulative distribution function

(cdf) $F(x) = P(\mathbf{X} \leq x) = P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_p \leq x_p)$

$$f(x) = \frac{\partial^p F(x_1, \dots, x_p)}{\partial x_1 \partial x_2 \dots \partial x_p}$$

Diagram illustrating the relationship between the joint probability density function (pdf) $f(x)$ and the joint cumulative distribution function (cdf) $F(x)$:

- The pdf $f(x)$ is shown as a function of $x_1, \dots, x_p$.
- The cdf $F(x)$ is shown as a function of $x_1, \dots, x_p$.
- The relationship is given by the partial derivative: $f(x) = \frac{\partial^p F(x_1, \dots, x_p)}{\partial x_1 \partial x_2 \dots \partial x_p}$.
- The integral relationship is also shown: $F(x) = \int_{-\infty}^{x_1} \dots \int_{-\infty}^{x_p} f(t_1, t_2, \dots, t_p) dt_1 \dots dt_p$.

$$\text{Let } \mathbf{X} = \begin{bmatrix} \mathbf{X}_A \\ \mathbf{X}_B \end{bmatrix}$$

$$\begin{aligned} \mathbf{X}_A &\in \mathbb{R}^k \\ \mathbf{X}_B &\in \mathbb{R}^{p-k} \end{aligned}$$

$$\text{eg. } \mathbf{X}_A = \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} \quad \mathbf{X}_B = \begin{bmatrix} X_4 \\ \vdots \\ X_p \end{bmatrix}$$

3) $\mathbf{X}_A$ has marginal distribution function

$$f_A(\mathbf{x}_A) = \underbrace{\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty}}_{p-k} f(\mathbf{x}) d\mathbf{x}_B$$

"integrating out the variables not of interest"

4) and conditional distribution of $\mathbf{X}_B$ given $\mathbf{X}_A$

$$f(\mathbf{x}_B | \mathbf{x}_A) = \frac{f(\mathbf{x}_A, \mathbf{x}_B)}{f_A(\mathbf{x}_A)} \quad \text{for } f_A(\mathbf{x}_A) > 0$$

$$\text{Ex: } f(x_1, x_2) = \frac{1}{2\pi} e^{-\frac{1}{2}(x_1^2 + x_2^2)}$$

$$\text{Hint: } \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2} dx_1 = 1 \quad -\infty < x_1, x_2 < \infty$$

$$\begin{aligned} 1) \text{ Find } f_1(x_1) &= \int_{-\infty}^{\infty} f(x_1, x_2) dx_2 \\ &= \int_{-\infty}^{\infty} \frac{1}{2\pi} e^{-\frac{1}{2}(x_1^2 + x_2^2)} dx_2 \end{aligned}$$

$$\begin{aligned}
 &= \int_{-\infty}^{\infty} \underbrace{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2}}_1 \cdot \underbrace{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_2^2}}_1 dx_2 \\
 &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2} \cdot \underbrace{\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_2^2} dx_2}_1 \\
 &= \underline{\underline{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2}}}
 \end{aligned}$$

$$f_2(x_2) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_2^2}$$

$$\begin{aligned}
 2) \quad f(x_2 | x_1) &= \frac{f(x_1, x_2)}{f_1(x_1)} = \frac{\cancel{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_2^2}}{\cancel{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2}}} \\
 &= \underline{\underline{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_2^2}}} = f_2(x_2)
 \end{aligned}$$

$$3) \quad F(x) = \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} f(t_1, t_2) dt_1 dt_2 = \dots$$

$f(x_1, x_2)$

not on closed form, use
R software

Independence

DEF: X_A and X_B are independent



$$f(x_A, x_B) = f_A(x_A) \cdot f_B(x_B)$$

Also: $f(x_A | x_B) = f_A(x_A)$, $f(x_B | x_A) = f_B(x_B)$

EX: $f(x_1, x_2) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_2^2} = f_1(x_1) \cdot f_2(x_2)$

What about $F(x)$, $F_A(x_A)$, $F_B(x_B)$ when X_A and X_B are independent?

$$F(x) = \int_{-\infty}^{x_A} \int_{-\infty}^{x_B} \underset{\substack{\uparrow \\ f_A(t_A) f_B(t_B)}}{f(t)} dt_A dt_B$$

$$= \int_{-\infty}^{x_A} f_A(t_A) dt_A \cdot \int_{-\infty}^{x_B} f_B(t_B) dt_B = F_A(x_A) \cdot F_B(x_B)$$

From $f(x_1, x_2) \rightarrow \begin{matrix} f_2(x_2) \\ f_1(x_1) \end{matrix}$ by integrating

How can we get from $f_1(x_1)$ and $f_2(x_2)$ to $f(x_1, x_2)$?

$f_1(x_1)$ and $f_2(x_2) \Rightarrow f(x_1, x_2) = f_1(x_1) \cdot f_2(x_2)$
 x_1 and x_2 independent 

But what if x_1 and x_2 are not independent?



solution is Copula $\Rightarrow$ see slides!