

## Lecture 4: Random Vectors

On a probability space  $(\Omega, \mathcal{F}, P)$  we can have  $n$  random variables  $X_1, \dots, X_n$  such that  $X_i: \Omega \rightarrow \mathbb{R}$ . This is called a random vector and can also be written  $\vec{X}: \Omega \rightarrow \mathbb{R}^n$ .

The distribution of  $\vec{X}$  is defined by

$$G_{\vec{X}}(A) = P(\vec{X} \in A) \text{ for } A \subseteq \mathbb{R}^n.$$

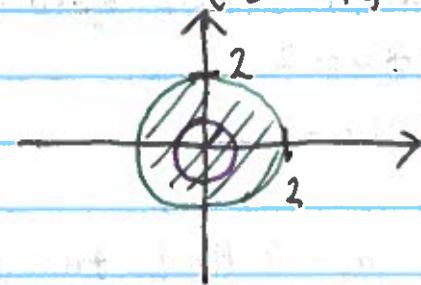
The joint density function (PDF) is defined by

$$P(\vec{X} \in A) = \int_A f(x_1, \dots, x_n) dx_1 \cdots dx_n$$

### Example:

$\vec{X} = (X, Y)$  where  $\vec{X}: \Omega \rightarrow \mathbb{R}^2$  is the  $(x, y)$  coordinate of a point chosen uniformly in a circle of radius 2.

$$f(x, y) = \begin{cases} \frac{1}{4\pi} & \text{if } x^2 + y^2 \leq 4 \\ 0 & \text{if } x^2 + y^2 > 4 \end{cases}$$

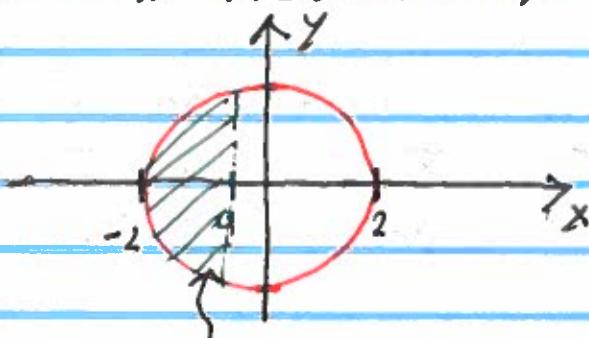


What is  $P(\vec{X} \in \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq 1\})$ ?

$$P(\vec{X} \in \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq 1\}) = \iint_{x^2 + y^2 \leq 1} \frac{1}{4\pi} dx dy = \frac{1}{4\pi} \int_0^{2\pi} \int_0^1 r dr dt = \frac{1}{2}.$$

$X, Y$  are also random variables. What is the PDF of  $\vec{X}$ ?

$$P(\bar{X} \leq a) = \int_{\{(x,y) \in \mathbb{R}^2 : x \leq a\}} f(x,y) dx dy = F_{\bar{X}}(a)$$



integral over this region.

$$\Rightarrow P(\bar{X} \leq a) = \int_{-2}^a \int_{-\sqrt{4-x^2}}^{0} \frac{1}{4\pi} dy dx$$

$$= \int_{-2}^a \frac{1}{2\pi} \sqrt{4-x^2} dx.$$

$$= F_{\bar{X}}(a)$$

$$\Rightarrow \frac{dF_{\bar{X}}}{da} = f(a) = \frac{1}{2\pi} (\sqrt{4-a^2})$$

The PDF is therefore,

$$f(x) = \begin{cases} \frac{1}{2\pi} \sqrt{4-x^2}, & -2 \leq x \leq 2 \\ 0, & |x| > 2 \end{cases}$$

Similarly, the PDF for  $\bar{Y}$  is given by

$$g(y) = \begin{cases} \frac{1}{2\pi} \sqrt{4-y^2}, & -2 \leq y \leq 2 \\ 0, & |y| > 2 \end{cases}$$

The densities of  $\bar{X}, \bar{Y}$  are called the marginal densities.

Random variables  $X_1, \dots, X_n$  are independent if for all  $A \in \mathcal{F}$ :

$$P(X_1 \in A_1, \dots, X_n \in A_n) = P(X_1 \in A_1) \cdots P(X_n \in A_n)$$

$$\Rightarrow f(x_1, \dots, x_n) = f(x_1) \cdots f(x_n) \leftarrow \text{product of marginals}$$

$$= P(\bar{X} \in A_1 \cap \dots \cap A_n) = P(X_1 \in A_1 \text{ and } X_2 \in A_2 \text{ and } \dots \text{ and } X_n \in A_n).$$

Example:

↑ IID

Let  $(\bar{X}, \bar{Y})$  be independent, identically distributed with standard Gaussian distributions.

1. What is  $P(\bar{X}^2 + \bar{Y}^2 \leq 1)$ ?

$$\begin{aligned} P(\bar{X}^2 + \bar{Y}^2 \leq 1) &= \iint_{x^2+y^2 \leq 1} f(x, y) dx dy \\ &= \iint_{x^2+y^2 \leq 1} \frac{1}{2\pi} e^{-x^2/2 - y^2/2} dx dy \quad \text{I.I.D.} \\ &= \iint_{x^2+y^2 \leq 1} \frac{1}{2\pi} e^{-(x^2+y^2)/2} dx dy \\ &= \int_0^{2\pi} \int_0^1 \frac{1}{2\pi} e^{-r^2/2} r dr dt \\ &\quad \text{density in polar coordinates.} \\ &= \int_0^{2\pi} -e^{-r^2/2} \Big|_0^1 dt \\ &= \frac{1}{2\pi} (1 - e^{-1/2}). \end{aligned}$$

2. Let  $R = (\bar{X}^2 + \bar{Y}^2)^{1/2}$ . Then

$$\begin{aligned} E[R] &= \int_0^{2\pi} \int_0^\infty \frac{1}{2\pi} e^{-r^2/2} r^2 dr \\ &= \frac{1}{2\pi} \int_0^{2\pi} \int_0^\infty \frac{1}{2\pi} e^{-r^2/2} r^2 dr dt \\ &= \frac{2\pi}{\sqrt{2\pi}} \cdot \frac{1}{2} \\ &= \sqrt{\frac{\pi}{2}}. \end{aligned}$$

3. What is the PDF of  $R$ ?

$$\begin{aligned} P(R \leq r) &= \int_0^r \int_0^{2\pi} \frac{1}{2\pi} e^{-s^2/2} s dr dt \\ &= \int_0^r s e^{-s^2/2} ds \\ \Rightarrow f(r) &= r e^{-r^2/2}. \end{aligned}$$

Definition - If  $(X, Y)$  is a random vector, then the covariance is given by

$$\begin{aligned}\text{Cov}(X, Y) &= \mathbb{E}[(X - \mathbb{E}[X])\mathbb{E}[(Y - \mathbb{E}[Y])]] \\ &= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]\end{aligned}$$

For a vector  $\mathbf{X} = (X_1, \dots, X_n)$  the covariance matrix is given by

$$C_{ij} = \text{Cov}(X_i, X_j).$$

\* Covariance is the expected change of one random variable with respect to another.

$\Rightarrow$  If  $C_{ij} < 0$  then if  $X_i$  increases  $X_j$  decreases.