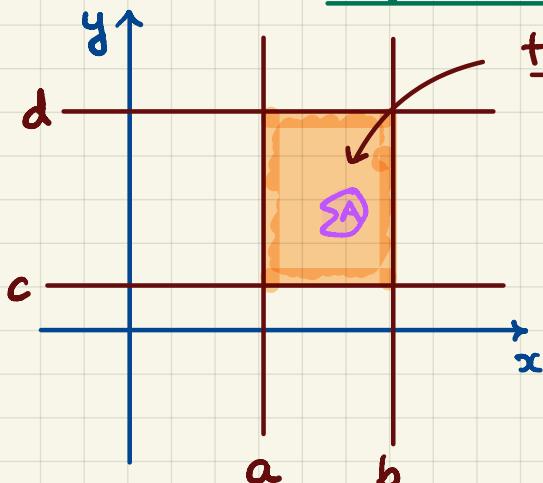


Chapter 5.

Section 5.1. Uniform Densities.



the rectangular region

$$B = [a, b] \times [c, d]$$

"Throwing fair (unbiased) darts @ the region B."

Position of the dart is determined by a pair of coordinates:

(X, Y) a RANDOM PAIR

We introduce the joint density function of (X, Y) as:

$$f_{X,Y}(x, y) = \begin{cases} \text{const} & \text{for } (x, y) \in B \\ 0 & \text{for } (x, y) \notin B \end{cases}$$

for $\text{const} = \frac{1}{\text{area}(B)} = \frac{1}{(b-a)(d-c)}$

Let $A \subseteq B$:

$$\text{TP}[(X, Y) \in A] = \frac{\text{area}(A)}{\text{area}(B)} = \iint_A f_{X,Y}(x, y) dx dy$$

Example. A bird will land @ random in a square area w/ each side of length 2.

What is the probability that the bird lands in a trap which is circular w/ radius $3/4$?

→: $\frac{\text{area(trap)}}{\text{area(field)}} = \frac{\left(\frac{3}{4}\right)^2 \pi}{2^2} = \frac{9}{64} \pi$



Def'n. We say that (X, Y) are jointly continuous w/ pdf

$$f_{X,Y}: \mathbb{R}^2 \rightarrow \mathbb{R}$$

If

$$\mathbb{P}[(X, Y) \in A] = \iint_A f_{X,Y}(x, y) dx dy$$

Def'n. Random variables X and Y are **independent**, if

$$\mathbb{P}[X \in A_1, Y \in A_2] = \mathbb{P}[X \in A_1] \cdot \mathbb{P}[Y \in A_2]$$

for all $A_1, A_2 \subseteq \mathbb{R}$

Recall: For X and Y jointly discrete w/ joint pmf $p_{X,Y}$, they are **independent** iff

$$p_{X,Y}(x, y) = p_X(x) \cdot p_Y(y)$$

Def'n. Marginal Densities.

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy \quad \text{for all } x \in \mathbb{R}$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx \quad \text{for all } y \in \mathbb{R}$$

Then, jointly continuous X and Y are **independent** iff

$$f_{X,Y}(x, y) = f_X(x) \cdot f_Y(y) \quad \text{for all } (x, y) \in \mathbb{R}^2$$

Note: X and Y independent

$$\mathbb{E}[XY] = \iint_{-\infty}^{+\infty} xy f_{X,Y}(x, y) dx dy$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_X(x) f_Y(y) dx dy$$

$$= \int_{-\infty}^{\infty} x f_X(x) dx \int_{-\infty}^{\infty} y f_Y(y) dy = \mathbb{E}[X] \cdot \mathbb{E}[Y]$$

Independent Normals.

Imagine repeated measurements:

$$X_1, X_2, \dots, X_n \sim N(\mu, \sigma^2)$$

μ
the actual
value we're
measuring

We define the sample mean:

$$\bar{X} = \frac{1}{n}(X_1 + X_2 + \dots + X_n) \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$