

We have introduced two rvs X and Y and discussed events of the type $\{(X, Y) \in D\}$ for $D \subset \mathbb{R}^2$

Can write probabilities for this type of event in terms of the measure P :

$$P((X, Y) \in D)$$

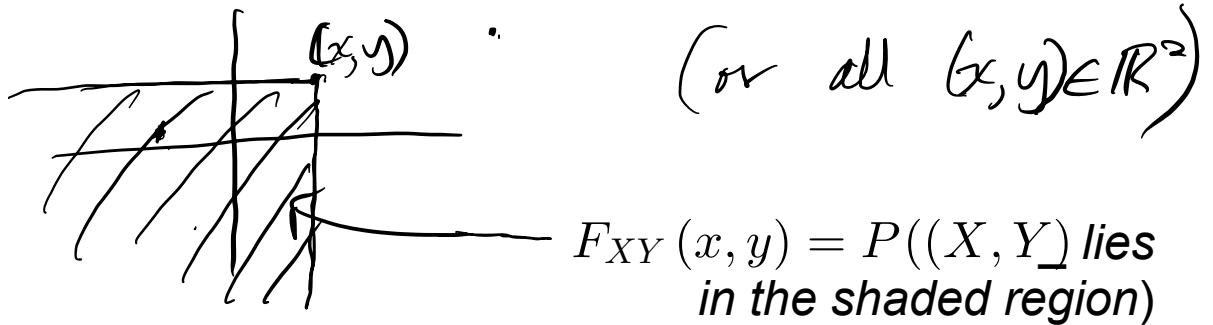
Now consider more practical representations of the distribution describing X, Y .

The Joint Cumulative Distribution Function

Defn. The joint cumulative distribution function, or joint distribution function, or joint cdf, of rvs X and Y is

$$F_{XY}(x, y) = P(\{X \leq x\} \cap \{Y \leq y\})$$

$$= P(X \leq x, Y \leq y) \quad \forall x, y \in \mathbb{R}$$



Some properties of F_{XY} :

①

$$\lim_{x \rightarrow -\infty} F_{XY}(x, y)$$

$$= \lim_{y \rightarrow -\infty} F_{XY}(x, y) = 0$$

②

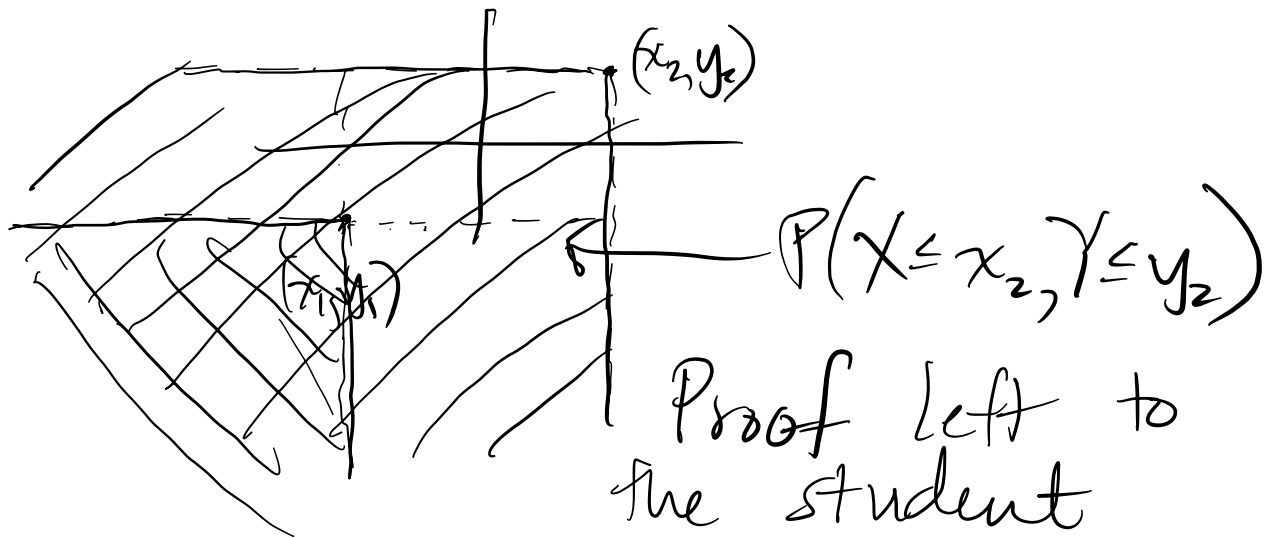
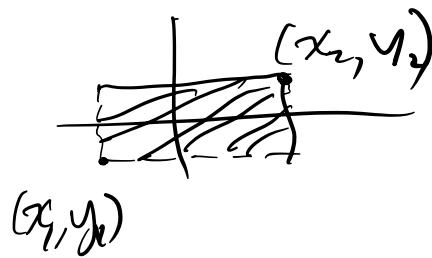
$$\lim_{x \rightarrow \infty} F_{XY}(x, y) = F_Y(y), \forall y \in \mathbb{R}$$

$$\lim_{y \rightarrow \infty} F_{XY}(x, y) = F_X(x), \forall x \in \mathbb{R}$$

F_X and F_Y are called the marginal cdfs of X and Y

③ for $x_1 < x_2, y_1 < y_2,$

$$P(x_1 < X \leq x_2, y_1 < Y \leq y_2) = \\ F_{XY}(x_2, y_2) - F_{XY}(x_2, y_1) - F_{XY}(x_1, y_2) \\ + F_{XY}(x_1, y_1)$$



It is difficult to write $P((X, Y) \in D)$ for some D , where D is a circle

for example, in terms of the joint cdf, so the joint pdf is more useful in practice, often.

The Joint Probability Density Function

Defn. The joint probability density function, or joint pdf, or joint density function, of X and Y is

$$f_{XY}(x, y) = \frac{\partial}{\partial x} \frac{\partial}{\partial y} F_{XY}(x, y)$$

for all x, y where these partials exist.

Using calculus, can show that for $D \subset \mathbb{R}^2$,

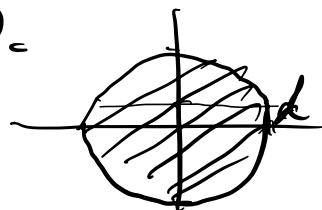
$$P((X, Y) \in D) = \iint_D f_{XY}(x, y) dx dy$$

Important! ↗

Ex. let $D_d = \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq d^2\}$

for some $d > 0$.

Then $P((X, Y) \in D_d)$



$$= \int_{-d}^d \int_{-\sqrt{d^2-x^2}}^{\sqrt{d^2-x^2}} f_{XY}(r, s) ds dr$$

(would probably want to do a change of variables to polar coordinates to evaluate this)

Some properties of the joint pdf

$$\textcircled{1} \underbrace{\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy}_{P((X, Y) \in \mathbb{R}^2)} = 1$$

$$\textcircled{2} F_{XY}(x, y) =$$

$$\underbrace{\int_{-\infty}^x \int_{-\infty}^y f_{XY}(r, s) ds dr}_{P(X \leq x, Y \leq y)}, \text{ for all } x, y \in \mathbb{R}$$

③ The marginal pds of X and Y can be found using

$$f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, y) dy$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{XY}(x, y) dx$$

If X and Y are both discrete, we use the joint pmf:

The Joint Probability Mass Function

Defn. The joint probability mass function, or joint pmf, of two discrete rvs X, Y is

$$\begin{aligned} p_{XY}(x, y) &= P(\{X = x\} \cap \{Y = y\}) \\ &= P(X = x, Y = y) \end{aligned}$$

$$\forall x \in \mathcal{R}_X, y \in \mathcal{R}_Y$$

Some properties

$$\textcircled{1} \sum_{x \in \mathcal{R}_X} \sum_{y \in \mathcal{R}_Y} p_{XY}(x, y) = 1$$

② Marginal pmfs:

$$p_X(x) = \sum_{y \in \mathcal{R}_Y} p_{XY}(x, y), \quad \forall x \in \mathcal{R}_X$$

$$p_Y(y) = \sum_{x \in \mathcal{R}_X} p_{XY}(x, y), \\ \forall y \in \mathcal{R}_Y$$