

The joint probability mass function (jpmf) of two discrete r.v.s X and Y is

$$P_{X,Y}(x_i, y_j) = \Pr(X=x_i, Y=y_j) = \Pr(\{X=x_i\} \cap \{Y=y_j\})$$

Properties of jpmf:

$$1) 0 \leq P_{X,Y}(x_i, y_j) \leq 1$$

$$2) \sum_{x_i} \sum_{y_j} P_{X,Y}(x_i, y_j) = 1$$

$$3) \Pr((X,Y) \in A) = \sum_{(x_i, y_j) \in A} P_{X,Y}(x_i, y_j)$$

$$4) P_X(x_i) = \sum_{y_j} P_{X,Y}(x_i, y_j)$$

$$P_Y(y_j) = \sum_{x_i} P_{X,Y}(x_i, y_j)$$

The joint cumulative distribution function (jcdf) of two r.v.s X and Y is

$$F_{X,Y}(x, y) = \Pr(X \leq x, Y \leq y)$$

Properties of jcdf:

$$1) 0 \leq F_{X,Y}(x, y) \leq 1$$

$$2) F_{X,Y}(-\infty, y) = F_{X,Y}(x, -\infty) = 0$$

$$F_{X,Y}(\infty, \infty) = 1$$

$$3) F_{X,Y}(x_1, y_1) \leq F_{X,Y}(x_2, y_2)$$

$$\text{if } x_1 \leq x_2 \text{ and } y_1 \leq y_2$$

$$4) F_Y(y) = \lim_{x \rightarrow \infty} F_{X,Y}(x, y)$$

$$F_X(x) = \lim_{y \rightarrow \infty} F_{X,Y}(x, y)$$

The joint probability density function (jpdf)

of two jointly continuous r.v.s X and Y is

$$f_{X,Y}(x, y) = \frac{\partial^2}{\partial x \partial y} F_{X,Y}(x, y)$$

or equivalently

$$F_{X,Y}(x, y) = \int_{-\infty}^y \int_{-\infty}^x f_{X,Y}(x', y') dx' dy'$$

Properties of jpdf:

$$1) f_{X,Y}(x, y) \geq 0$$

$$2) \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx dy = 1$$

$$3) P_r((X, Y) \in A) = \iint_A f_{X, Y}(x, y) dx dy$$

$$4) f_Y(y) = \int_{-\infty}^{\infty} f_{X, Y}(x, y) dx$$

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

The mean or expected value of $g(X, Y)$ is

$$E[g(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X, Y}(x, y) dx dy$$

$$= \sum_{y_j} \sum_{x_i} g(x_i, y_j) P_{X, Y}(x_i, y_j)$$

if X, Y discrete

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

$$E[X] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{X, Y}(x, y) dx dy$$

$$E[Y] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f_{X, Y}(x, y) dx dy$$

Conditional cdf, pdf, pmf

Recall that if X is a r.v. and M is an event with $\Pr(M) > 0$ then

$$F_x(x|M) = \Pr(X \leq x | M)$$

$$f_x(x|M) = \frac{d}{dx} F_x(x|M)$$

Let X and Y be jointly continuous r.v.s.

Let $M = \{Y = y\}$. Since $\Pr(Y = y) = 0$,

$F_x(x|Y=y)$ and $f_x(x|Y=y)$

cannot be defined as above.

Instead the following definitions are used:

The conditional cdf of X given Y is

$$F_{x|Y}(x|y) = \lim_{dy \rightarrow 0} F_x(x | y \leq Y \leq y + dy)$$

The conditional pdf of X given Y is

$$f_{x|Y}(x|y) = \frac{d}{dx} F_{x|Y}(x|y)$$

$$= \frac{f_{x,Y}(x,y)}{f_Y(y)}, \quad f_Y(y) > 0$$

Likewise,

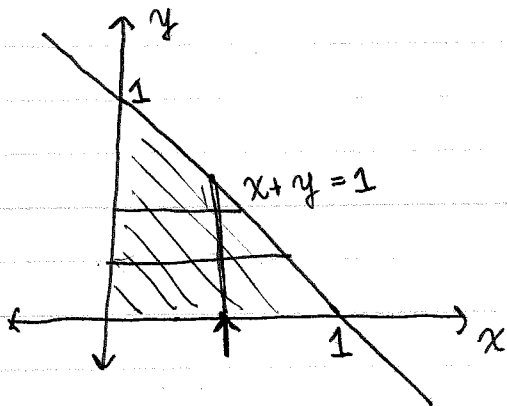
$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)}, \quad f_X(x) > 0$$

Ex: Let X, Y be r.v.s with jpdf

$$f_{X,Y}(x,y) = \begin{cases} kx & , \quad 0 \leq x+y \leq 1, \quad x > 0, \quad y > 0 \\ 0 & , \quad \text{else} \end{cases}$$

k is a constant.

a) Find k



$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy \quad (= 1)$$

$$= \int_0^1 \int_0^{1-y} kx dx dy$$

$$= k \int_0^1 \left. \frac{x^2}{2} \right|_0^{1-y} dy$$

$$= k \int_0^1 \frac{(1-y)^2}{2} dy, \quad u = 1-y, \quad du = -dy$$

$$= k \int_0^1 \frac{u^2}{2} du = \frac{k}{6} = 1$$

$$\Rightarrow \boxed{k=6}$$

b) Find $f_x(x)$, $f_y(y)$, $f_{x|y}(x|y)$.

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

$$= \int_0^{1-x} 6x dy, \quad 0 \leq x \leq 1$$

$$= 6x y \Big|_0^{1-x}$$

$$= 6x(1-x), \quad 0 \leq x \leq 1$$

$$= 0, \quad \text{else}$$

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

$$= \int_0^{1-y} 6x dx, \quad 0 \leq y \leq 1$$

$$= 3(1-y)^2, \quad 0 \leq y \leq 1$$

$$= 0, \quad \text{else}$$

$$f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} \quad \begin{array}{l} \leftarrow 0 \leq x+y \leq 1, x, y > 0 \\ \leftarrow 0 \leq y \leq 1 \end{array}$$

$$= \frac{2x}{(1-y)^2}, \quad \begin{array}{l} 0 \leq y \leq 1 \\ 0 \leq x \leq 1-y \end{array}$$

$$= 0, \quad \text{else}$$