

# Joint PDF of Continuous Random Variables

## 1. Two Continuous R.V.

Let  $X$  and  $Y$  be cont. r.v. Then  $f(x, y)$  is their joint probability density function or joint PDF for  $X$  and  $Y$  if for any two-dimensional set  $A$ ,

$$P[(X, Y) \in A] = \int \int_A f(x, y) dx dy$$

In particular, if  $A$  is the 2-D rectangle with  $\{a \leq X \leq b, c \leq Y \leq d\}$ , then we can bound the integral as  $\int_c^d \int_a^b f(x, y) dx dy$ . Conditions for a joint PDF: (1) it must be non-negative, i.e.  $f(x, y) \geq 0$  for all  $x$  and  $y$  and (2)  $\int \int f(x, y) dx dy = 1$ .

## 2. Obtaining Marginal PDF's from Joint PDF's

Given the joint PDF  $f(x, y)$  of two cont. r.v., the marginal PDF, or the marginal density, of  $X$  and  $Y$ , can be obtained by integrating the joint PDF over the other variable; recall the marginal PMF's of discrete r.v. are obtained by summing the joint PMF over values of the other variable.

## Joint CDF (for discrete and cont. r.v.)

The joint cumulative distribution function of the  $k$  r.v.  $X_1, X_2, \dots, X_k$  is the function defined by

$$F(x_1, \dots, x_k) = P(X_1 \leq x_1, \dots, X_k \leq x_k).$$

The random variables can be discrete or cont., or some be discrete and other be continuous.

## 3. Joint PDF & CDF for Cont. R.V.

If  $f(x, y)$  is the joint PDF for  $X$  and  $Y$ , their joint CDF is

$$F(x, y) = \int_{-\infty}^x \int_{-\infty}^y f(u, v) du dv.$$

Conversely, if  $F(x, y)$  is the joint CDF for  $X$  and  $Y$ , their joint PDF is

$$f(x, y) = \frac{\partial^2}{\partial x \partial y} F(x, y).$$

## 4. Independent Random Variables

Recall that two events  $A$  and  $B$  are independent if  $P(A \cap B) = P(A)P(B)$  and similarly two random variables  $X$  and  $Y$  are independent if  $P(X \in A, Y \in B) = P(X \in A)P(Y \in B)$  for any sets  $A$  and  $B$ .

Two discrete random variables  $X$  and  $Y$  are independent iff

$$p(x, y) = p_X(x)p_Y(y)$$

for all  $x$  and  $y$ , i.e. the joint PMF is the product of their marginal PMF's.

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Their joint PDF is

$$f(x, y) = \frac{\partial^2}{\partial x \partial y} F(x, y) = \frac{\partial^2}{\partial x \partial y} F_X(x)F_Y(y) = F'_X(x)F'_Y(y) = f_X(x)f_Y(y)$$

Conversely, if  $f(x, y) = f_X(x)f_Y(y)$ , their joint PDF is

$$\begin{aligned} F(x, y) &= \int_{-\infty}^x \int_{-\infty}^y f(u, v) du dv = \int_{-\infty}^x \int_{-\infty}^y f_X(u)f_Y(v) du dv \\ &= \int_{-\infty}^x f_X(u) du \int_{-\infty}^y f_Y(v) dv = F_X(x)F_Y(y) \end{aligned}$$

i.e. independent iff joint PDF = the product of marginal PDF's.

## 5. A Simple Method

It seems like we have to find marginal distributions (a lot of work) to check independence. However, there's a simple criterion:  $X$  and  $Y$  are independent if the joint PMF/PDF can be written as the product of a function of  $x$  and a function of  $y$  (which ought to be fairly intuitive), i.e.

$$f(x, y) = g(x)h(y), \text{ for all } x, y.$$

Here  $g(x) \geq 0$  and  $h(y) \geq 0$  are not necessarily PMFs/PDFs.