

Section 4.4 Bivariate Distribution of the Continuous Type

- Extend the idea of joint dist. to two RVs of continuous type.

□ Joint PDF

DEF $f_{X,Y}(x,y)$ is called a **joint PDF** of two RVs X & Y if:

(a) $f_{X,Y}(x,y) \geq 0$ for any x, y .

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

(c) $P((X,Y) \in A) = \iint_A f_{X,Y}(x,y) dx dy$ for any region A in the plane.

- As before, PDF of each X and Y is called **marginal PDF** and can be computed by

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

- **CAUTION** Unlike the discrete-type case, two continuous RVs need not have a joint PDF! One such class of examples are RVs X & Y related by a function, say $Y = u(X)$ for some $u(x)$. Then even if both X and Y are continuous, the joint dist. of X and Y is not "jointly continuous".

Ex Let X & Y have the joint PDF

$$f(x,y) = \frac{4}{3}(1-xy), \quad 0 < x < 1, \quad 0 < y < 1.$$

Then

$$(1) P(X < Y) = P((X,Y) \in A), \quad \text{where}$$

$$A = \{(x,y) : x < y\},$$

and so,

$$\begin{aligned}
 P(X < Y) &= \iint_A f(x, y) dx dy = \iint_{x < y} f(x, y) dx dy \\
 &= \int_0^1 \int_0^y \frac{4}{3} (1 - xy) dx dy = \int_0^1 \left[\frac{4}{3} x - \frac{2}{3} x^2 y \right]_{x=0}^{x=y} dy \\
 &= \int_0^1 \left(\frac{4}{3} y - \frac{2}{3} y^3 \right) dy = \left[\frac{2}{3} y^2 - \frac{1}{6} y^4 \right]_0^1 = \frac{1}{2}.
 \end{aligned}$$

(2) The marginal PDF of X is, for $0 < x < 1$,

$$\begin{aligned}
 f_X(x) &= \int_{-\infty}^{\infty} f(x, y) dy = \int_0^1 \frac{4}{3} (1 - xy) dy \\
 &= \left[\frac{4}{3} y - \frac{2}{3} xy^2 \right]_{y=0}^{y=1} = \frac{4}{3} - \frac{2}{3} x, \quad \underline{0 < x < 1}
 \end{aligned}$$

(3) The expectation of X may be computed in two ways:

▷ Using the marginal PDF:

$$E(X) = \int_{-\infty}^{\infty} x f_X(x) dx = \int_0^1 x \cdot \left(\frac{4}{3} - \frac{2}{3} x \right) dx = \left[\frac{2}{3} x^2 - \frac{2}{9} x^3 \right]_0^1 = \frac{4}{9}$$

▷ Using the joint PDF:

$$\begin{aligned}
 E(X) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x, y) dx dy = \int_0^1 \int_0^1 x \cdot \frac{4}{3} (1 - xy) dx dy \\
 &= \int_0^1 \left[\frac{2}{3} x^2 - \frac{4}{9} x^3 y \right]_{x=0}^{x=1} dy = \int_0^1 \left(\frac{2}{3} - \frac{4}{9} y \right) dy \\
 &= \left[\frac{2}{3} y - \frac{2}{9} y^2 \right]_0^1 = \frac{4}{9}.
 \end{aligned}$$

□

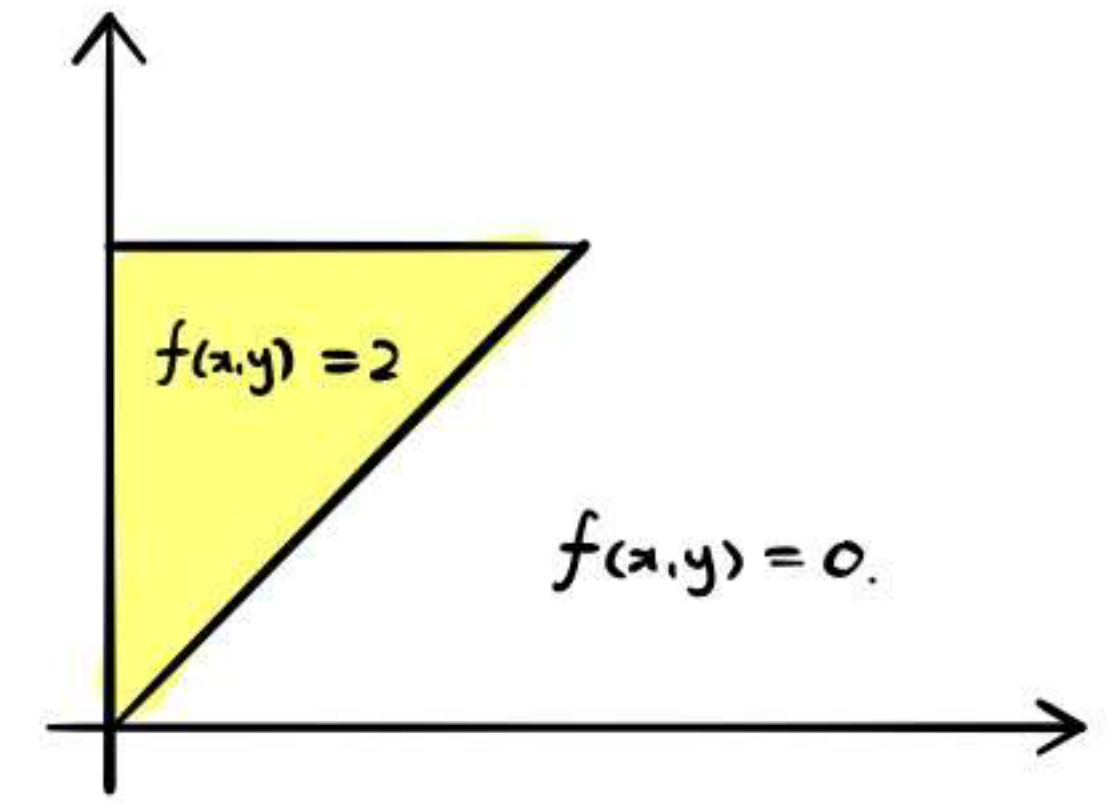
2 Independence

DEF Let X and Y have the joint PDF $f_{X,Y}(x, y)$. X and Y are independent if

$$f_{X,Y}(x, y) = f_X(x) f_Y(y) \quad \text{for any } x, y.$$

Ex Let X and Y have the joint PDF

$$f(x,y) = 2, \quad 0 < x < y < 1.$$



Then the marginal PDFs are

▷ If $0 < x < 1$, then $f_X(x) = \int_{-\infty}^{\infty} f(x,y) dy = \int_x^1 2 dy = 2(1-x)$,

otherwise, $f_X(x) = 0$. (why?)

▷ If $0 < y < 1$, then $f_Y(y) = \int_{-\infty}^{\infty} f(x,y) dx = \int_0^y 2 dx = 2y$.

otherwise, $f_Y(y) = 0$.

▷ Then

$$f_{X,Y}(x,y) = \begin{cases} 2, & 0 < x < y < 1, \\ 0, & \text{elsewhere,} \end{cases}$$

but

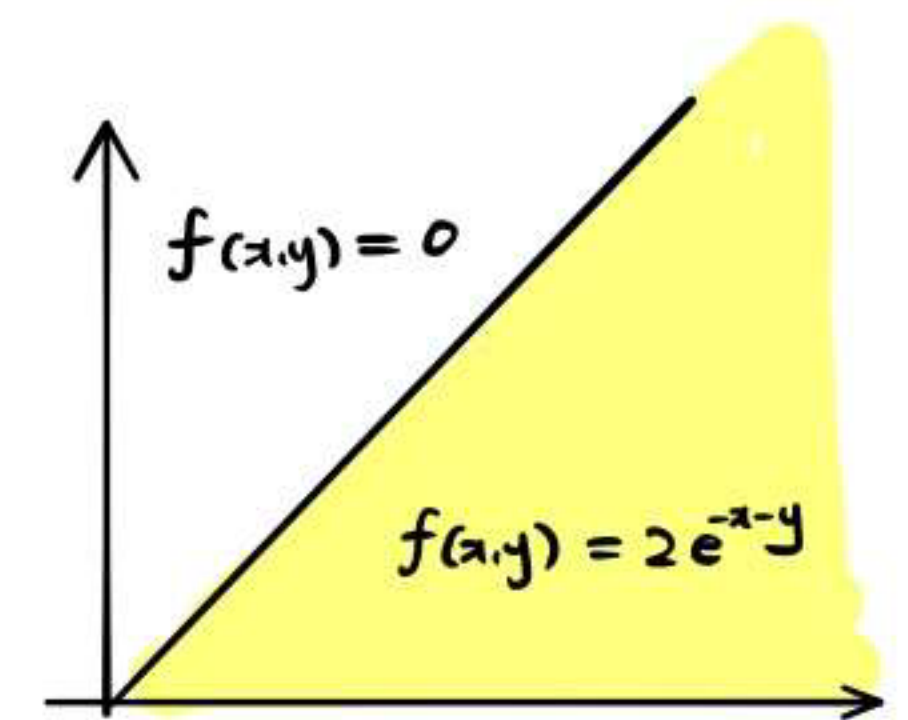
$$f_X(x)f_Y(y) = \begin{cases} 4(1-x)y, & 0 < x < 1, 0 < y < 1, \\ 0, & \text{elsewhere} \end{cases}$$

so X and Y are not independent. □

[3] Correlation Coefficient

Ex Let X and Y have the joint PDF:

$$f(x,y) = 2e^{-x-y}, \quad 0 < y < x < +\infty$$



Then

$$\mu_X = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x,y) dx dy = \int_0^{\infty} \int_0^x x \cdot 2e^{-x-y} dy dx = \int_0^{\infty} 2x e^{-x} (1 - e^{-x}) dx = \frac{3}{2}.$$

$$\mu_Y = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x,y) dx dy = \int_0^{\infty} \int_y^{\infty} y \cdot 2e^{-x-y} dx dy = \int_0^{\infty} 2y e^{-2y} dy = \frac{1}{2}.$$

$$\begin{aligned}
 E(XY) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x,y) dx dy = \int_0^{\infty} \int_y^{\infty} 2xy e^{-x-y} dx dy \\
 &= \int_0^{\infty} 2ye^{-y} \left[-xe^{-x} - e^{-x} \right]_{x=y}^{x=+\infty} dy = \int_0^{\infty} 2y(y+1) e^{-2y} dy = 1.
 \end{aligned}$$

So we get

$$Cov(X,Y) = E(XY) - \mu_X \mu_Y = 1 - \left(\frac{1}{2}\right) \left(\frac{3}{2}\right) = \frac{1}{4}.$$

In particular, X and Y are not independent. (Recall: indep. $\Rightarrow$ zero cov.) $\square$

4 Conditional Distributions

DEF Let X and Y have the joint PDF $f(x,y)$.

▷ The **conditional PDF** of Y given $X=x$ is defined by

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

▷ The **conditional expectation** of $u(Y)$ given $X=x$ is defined by

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

▷ $E(Y|X)$ and $Var(Y|X)$ are defined as in the discrete case.

Rmk) Both the law of total exp/var hold in this case.

Ex Let X and Y be as in the previous ex:

$$f(x,y) = 2e^{-x-y}, \quad 0 < y < x < +\infty$$

Then we can check that

$$f_X(x) = 2e^{-x}(1-e^{-x}), \quad 0 < x < +\infty,$$

and so,

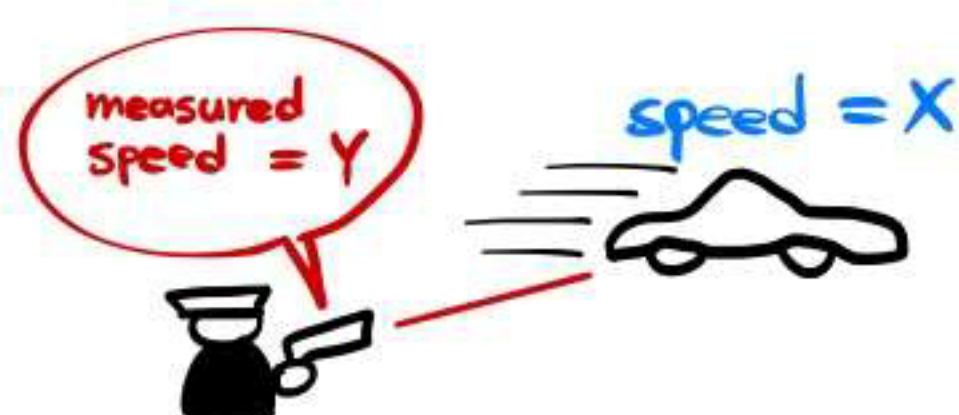
$$f_{Y|X}(y|x) = \frac{2e^{-x-y}}{2e^{-x}(1-e^{-x})} = \frac{e^{-y}}{1-e^{-x}}, \quad 0 < y < x.$$

Then for $0 < x < +\infty$,

$$\begin{aligned} E(Y|X=x) &= \int_{-\infty}^{\infty} y \cdot f_{Y|X}(y|x) dy = \int_0^x \frac{y e^{-y}}{1-e^{-x}} dx \\ &= \frac{1}{1-e^{-x}} \left[-(y+1)e^{-y} \right]_{y=0}^{y=x} = \frac{1}{1-e^{-x}} (1 - (x+1)e^{-x}) \\ &= 1 - \frac{x}{e^x - 1}. \end{aligned}$$

□

Ex A police officer is measuring the speed of cars on a highway.



- ▷ The actual speed X of a car is $\mathcal{U}(80, 120)$,
- ▷ Due to the inaccuracy of the speed gun, the speed Y measured, given $X=x$, is $\mathcal{N}(x, \frac{1}{100}x)$.

Find the mean/var of Y .

Sol) By the law of total E:

$$E(Y) = E[E(Y|X)] = E[X] = 100.$$

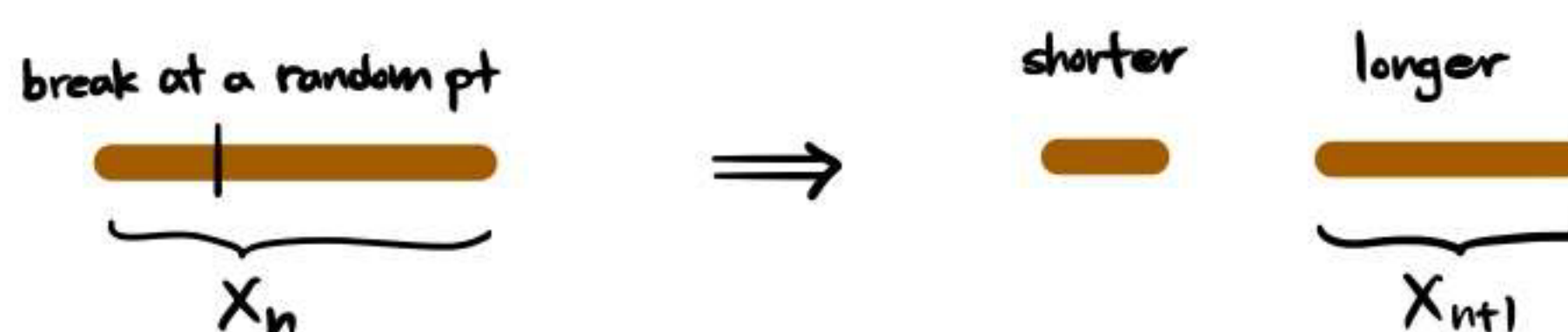
Likewise, by the law of total Var:

$$\begin{aligned} \text{Var}(Y) &= E(\text{Var}(Y|X)) + \text{Var}(E(Y|X)) \\ &= E\left(\frac{1}{100}X\right) + \text{Var}(X) \\ &= \frac{1}{100} \cdot 100 + \frac{1}{12} \cdot (120 - 80)^2 \\ &= 1 + \frac{400}{3}. \end{aligned}$$

□

Ex Let $X_0, X_1, \dots$ be defined as follows:

- (1) Start with a stick of length X_0 .
- (2) If we have a stick of length X_n , break it at a uniformly randomly chosen point and X_{n+1} be the length of the longer piece:



▷ Given $X_n = x$, let Y be the breaking position. Then Y is $U(0, x)$, and $X_{n+1} = \max\{Y, x - Y\}$:

So we get

$$\begin{aligned} E(X_{n+1} | X_n = x) &= E(\max\{Y, x - Y\} | X_n = x) \\ &= \int_0^x \max\{y, x - y\} \cdot \frac{1}{x} dy = \frac{3}{4}x, \end{aligned}$$

and hence $E(X_{n+1} | X_n) = \frac{3}{4} X_n$. Then

$$E(X_n) \stackrel{\substack{\uparrow \\ \text{Law of Total Exp.}}}{=} E[E(X_n | X_{n-1})] = E\left(\frac{3}{4} X_{n-1}\right) = \dots = \left(\frac{3}{4}\right)^n E(X_0).$$

