

### 6.3. Sum of independent R.V.s

1. The case that both  $X$  and  $Y$  are continuous.

Let  $X, Y$  have densities  $f_X(x), f_Y(y)$  resp.

Since  $X$  and  $Y$  are assumed to be independent,

$X, Y$  have a joint density

$$f(x, y) = f_X(x) f_Y(y).$$

Now let  $a \in \mathbb{R}$ , then

$$F_{X+Y}(a) = P\{X+Y \leq a\}$$

$$= \iint_{\substack{(x,y) \in \mathbb{R}^2: \\ x+y \leq a}} f(x, y) \, dx \, dy$$

$$= \iint_{(x,y) \in \mathbb{R}^2: x+y \leq a} f_X(x) f_Y(y) \, dx \, dy$$

$$= \int_{-\infty}^{\infty} \left( \int_{-\infty}^{a-y} f_X(x) f_Y(y) \, dx \right) dy$$

$$= \int_{-\infty}^{\infty} f_Y(y) \cdot \left( \int_{-\infty}^{a-y} f_X(x) \, dx \right) dy$$

$$= \int_{-\infty}^{\infty} f_Y(y) F_X(a-y) dy$$

$$=: F_X * f_Y(a)$$

└────────── Convolution

( For  $g, h: \mathbb{R} \rightarrow \mathbb{R}$ , we let

$$g * h(a) = \int_{-\infty}^{\infty} g(a-y) h(y) dy )$$

$$f_{X+Y}(a) = \frac{d}{da} F_{X+Y}(a)$$

$$= \frac{d}{da} \int_{-\infty}^{\infty} F_X(a-y) f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} \left( \frac{d}{da} F_X(a-y) \right) f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy$$

$$= f_X * f_Y(a).$$

Thm A. If  $X, Y$  are independent, and have densities  $f_X(x), f_Y(y)$ , then

$$\left\{ \begin{array}{l} F_{X+Y}(a) = F_X * f_Y(a) \\ f_{X+Y}(a) = f_X * f_Y(a) \end{array} \right.$$

Example 1. Let  $X, Y$  be independent, both unif. dist. on  $[0, 1]$ . Calculate the density of  $X+Y$ .

Solution: Let  $a \in \mathbb{R}$ . Then

$$f_{X+Y}(a) = f_X * f_Y(a)$$

$$= \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy$$

$$= \int_0^1 f_X(a-y) dy \quad (\text{since } f_Y(y) = \begin{cases} 1 & \text{if } y \in [0, 1] \\ 0 & \text{otherwise} \end{cases})$$

$$\stackrel{\text{letting } z=a-y}{=} \int_{a-1}^a f_X(z) dz$$

If  $0 < a \leq 1$ ,

$$\int_{a-1}^a f_X(z) dz = \int_0^a 1 dz = a.$$

If  $1 < a \leq 2$ ,

$$\int_{a-1}^a f_X(z) dz = \int_{a-1}^1 1 dz = 2-a.$$

If  $a > 2$  or  $a < 0$ ,

$$\int_{a-1}^a f_X(z) dz = 0.$$

Hence

$$f_{X+Y}(a) = \begin{cases} a & \text{if } 0 < a \leq 1 \\ 2-a & \text{if } 1 < a \leq 2 \\ 0 & \text{otherwise,} \end{cases}$$





Example 2. Let  $X, Y$  be independent normal r.v.'s with parameters  $(0, 1)$  and  $(0, \sigma^2)$ . Find out the distribution of  $X + Y$ .

Solution: Recall that

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad x \in \mathbb{R}$$

$$f_Y(y) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{y^2}{2\sigma^2}}, \quad y \in \mathbb{R}.$$

Hence for  $a \in \mathbb{R}$ ,

$$\begin{aligned} f_X * f_Y(a) &= \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy \\ &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(a-y)^2}{2}} \cdot \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{y^2}{2\sigma^2}} dy \\ &= \frac{1}{2\pi \sigma} \int_{-\infty}^{\infty} e^{-\frac{(a-y)^2}{2} - \frac{y^2}{2\sigma^2}} dy. \end{aligned}$$

Notice that

$$\frac{(a-y)^2}{2} + \frac{y^2}{2\sigma^2} = \frac{\left(\sqrt{\sigma^2+1} y - \frac{a\sigma^2}{\sqrt{\sigma^2+1}}\right)^2}{2\sigma^2} + \frac{a^2}{2(\sigma^2+1)}$$

( verify it )

Hence

$$f_{X+Y}(a) = \frac{1}{2\pi\sigma} e^{-\frac{a^2}{2(\sigma^2+1)}} \int_{-\infty}^{\infty} e^{-\frac{\left(\sqrt{\sigma^2+1} y - \frac{a\sigma^2}{\sqrt{\sigma^2+1}}\right)^2}{2\sigma^2}} dy$$

$$\left( \text{letting } z = \frac{\sqrt{\sigma^2+1} y - \frac{a\sigma^2}{\sqrt{\sigma^2+1}}}{\sigma} \right)$$

$$= \frac{1}{2\pi\sigma} \cdot \frac{\sigma}{\sqrt{\sigma^2+1}} e^{-\frac{a^2}{2(\sigma^2+1)}} \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} dz$$

$$\left( \text{using the fact } \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} dz = 1 \right)$$

$$= \frac{1}{\sqrt{2\pi} \sqrt{\sigma^2+1}} e^{-\frac{a^2}{2(\sigma^2+1)}}.$$

Hence  $X+Y$  is a normal r.v. with parameters  $(0, \sigma^2+1)$ .

Remark: In general, if  $X, Y$  are independent,  
normal r.v.'s with parameters  
 $(\mu_1, \sigma_1^2)$ , and  $(\mu_2, \sigma_2^2)$ , then  
 $X + Y$  has a normal distribution  
with parameters  $(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$ .

Example 1. Let  $X, Y$  be independent Poisson r.v.'s  
with parameters  $\lambda_1, \lambda_2$ , resp.  
Calculate the distribution of  $X+Y$ .

Solution:  $P\{X=k\} = e^{-\lambda_1} \cdot \lambda_1^k / k!$ ,  $k=0, 1, 2, \dots$   
 $P\{Y=k\} = e^{-\lambda_2} \lambda_2^k / k!$ ,  $k=0, 1, 2, \dots$

For  $n=0, 1, 2, \dots$ ,

$$P\{X+Y=n\} = \sum_{k=0}^{\infty} P\{X=k\} P\{Y=n-k\}$$

(but  $P\{Y=n-k\} = 0$  if  $k > n$ )

$$= \sum_{k=0}^n P\{X=k\} P\{Y=n-k\}$$

$$= \sum_{k=0}^n e^{-\lambda_1} \frac{\lambda_1^k}{k!} e^{-\lambda_2} \frac{\lambda_2^{n-k}}{(n-k)!}$$

$$= e^{-\lambda_1 - \lambda_2} \cdot \frac{1}{n!} \sum_{k=0}^n \binom{n}{k} \lambda_1^k \lambda_2^{n-k}$$

By the binomial Thm

$$= e^{-\lambda_1 - \lambda_2} \frac{(\lambda_1 + \lambda_2)^n}{n!}$$

(Recall  $(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$ )

Hence  $X+Y$  has the Poisson distribution with parameter  $\lambda_1 + \lambda_2$ .

## § 6.4 Conditional distribution.

### 1. Discrete case.

Def. Let  $X, Y$  be two discrete r.v.'s.

Then

$$P\{X=x \mid Y=y\} = \frac{P\{X=x, Y=y\}}{P\{Y=y\}},$$

provided that  $P\{Y=y\} > 0$ .

2. Suppose that  $X$  and  $Y$  are jointly cts with density  $f(x, y)$ .

Def. The conditional density function of  $X$  given  $Y=y$ , is given by

$$f_{X|Y}(x|y) := \frac{f(x, y)}{f_Y(y)},$$

provided that  $f_Y(y) > 0$ .

Def. For  $A \subset \mathbb{R}$ , the conditional prob. of  $X$  taking values in  $A$  given  $Y=y$  is given by

$$P\{X \in A \mid Y=y\} = \int_A f_{X|Y}(x|y) dx$$

In particular,

$$\begin{aligned} F_{X|Y}(a|y) &:= P\{X \leq a \mid Y=y\} \\ &= \int_{-\infty}^a f_{X|Y}(x|y) dx. \end{aligned}$$

Remark: If  $X$  and  $Y$  are independent,

$$\text{then } f_{X|Y}(x|y) = f_X(x).$$

(since in such case  $f(x, y) = f_X(x) f_Y(y)$ )

Remark: One may view

$$P\{X \in A \mid Y = y\}$$

$$= \lim_{\varepsilon \rightarrow 0} P\{X \in A \mid y - \varepsilon < Y < y + \varepsilon\}$$

$$= \lim_{\varepsilon \rightarrow 0} \frac{P\{X \in A, y - \varepsilon < Y < y + \varepsilon\}}{P\{y - \varepsilon < Y < y + \varepsilon\}}$$

Example 2. Suppose the joint density of  $X$  and  $Y$  is given by

$$f(x, y) = \begin{cases} e^{-x/y} e^{-y} / y & \text{if } x > 0, y > 0 \\ 0 & \text{otherwise} \end{cases}$$

Find  $P\{X > 1 \mid Y = y\}$ .

$$\begin{aligned} \text{Solution: } f_Y(y) &= \int_{-\infty}^{\infty} f(x, y) dx \\ &= \int_0^{\infty} e^{-x/y} e^{-y} / y dx \quad (\text{if } y > 0) \\ &= -e^{-x/y} e^{-y} \Big|_{x=0}^{\infty} \\ &= e^{-y} \quad \text{if } y > 0. \end{aligned}$$

Hence for  $y > 0$ ,

$$f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = e^{-x/y} / y$$

if  $x > 0$ .



Therefore

$$P\{X > 1 | Y = y\} = \int_1^{\infty} e^{-x/y} / y \, dx$$

$$= -e^{-x/y} \Big|_{x=1}^{\infty}$$

$$= e^{-1/y} \quad \text{if } y > 0.$$



- Distribution of functions of r.v.'s.

Setup:  $X_1, X_2$  are joint cts with density  $f_{X_1, X_2}(x_1, x_2)$ . Let  $g_1, g_2: \mathbb{R}^2 \rightarrow \mathbb{R}$ .

Let  $Y_1 = g_1(X_1, X_2)$ ,  $Y_2 = g_2(X_1, X_2)$ .

Find out the joint distribution of  $Y_1, Y_2$

Thm. Assumptions: ①  $x_1, x_2$  can be solved in terms of  $y_1, y_2$ .

②  $g_1, g_2$  have cts partial derivatives

and

$$J(x_1, x_2) := \begin{vmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} \end{vmatrix}$$

$$= \frac{\partial g_1}{\partial x_1} \frac{\partial g_2}{\partial x_2} - \frac{\partial g_1}{\partial x_2} \frac{\partial g_2}{\partial x_1} \neq 0$$

Then  $Y_1, Y_2$  have a joint density

$$f_{Y_1, Y_2}(y_1, y_2) = f_{X_1, X_2}(x_1, x_2) \cdot |J(x_1, x_2)|^{-1}$$

Exer 1.  $X, Y$  have joint density

$$f(x, y) = \begin{cases} \frac{1}{x^2 y^2}, & \text{if } x > 1, y > 1 \\ 0, & \text{otherwise.} \end{cases}$$

Let  $U = XY$ ,  $V = X/Y$ . Find out the joint density of  $U$  and  $V$ .

Solution: Let  $g_1(x, y) = xy$  and  $g_2(x, y) = x/y$ .

Then

$$J(x, y) = \begin{vmatrix} \frac{\partial g_1}{\partial x} & \frac{\partial g_1}{\partial y} \\ \frac{\partial g_2}{\partial x} & \frac{\partial g_2}{\partial y} \end{vmatrix} = \begin{vmatrix} y & x \\ \frac{1}{y} & -\frac{x}{y^2} \end{vmatrix} \\ = -\frac{2x}{y}$$

By the thm,

$$f_{U,V}(u, v) = f(x, y) \cdot |J(x, y)|^{-1} \\ = \begin{cases} \frac{1}{x^2 y^2} \cdot \frac{y}{2x} = \frac{1}{2x^3 y} & \text{if } x > 1, y > 1 \\ 0 & \text{otherwise} \end{cases}$$

Notice that  $x = \sqrt{uv}$   
 $y = \sqrt{u/v}$   
 $u, v > 0$ .

$$\begin{cases} \sqrt{uv} > 1 \Rightarrow v > \frac{1}{u} \\ \sqrt{u/v} > 1 \Rightarrow v < u \end{cases}$$

Also notice that  $x, y > 1 \Leftrightarrow u > 1, \frac{1}{u} < v < u$ .

$$\text{Hence } f_{U,V}(u, v) = \begin{cases} \frac{1}{2u^2 v}, & \text{if } u > 1, \frac{1}{u} < v < u \\ 0, & \text{otherwise} \end{cases} \quad \square$$