

## Review.

- Joint distribution of two r.v.'s
- Independence of two r.v.'s.

Let  $X$  and  $Y$  be two r.v.'s. The joint cumulative distribution function (CDF) of  $X$  and  $Y$  is defined as

$$F(a, b) = P\{X \leq a, Y \leq b\}, \quad a, b \in \mathbb{R}.$$

Def. We say that  $X$  and  $Y$  are independent if

$$P\{X \in A, Y \in B\} = P\{X \in A\} P\{Y \in B\}$$

for all  $A, B \subset \mathbb{R}$ .

Theoretically, one can prove that

$$X \text{ and } Y \text{ are independent} \iff F(a, b) = F_X(a) F_Y(b) \quad \forall a, b \in \mathbb{R}.$$

- Moreover, when  $X, Y$  are both discrete,

$$X, Y \text{ are independent} \Leftrightarrow p(x, y) = p_X(x) p_Y(y), \forall x, y \in \mathbb{R}.$$

- When  $X$  and  $Y$  are jointly cts,

$$X \text{ and } Y \text{ are independent} \Leftrightarrow f(x, y) = f_X(x) f_Y(y) \quad \forall x, y \in \mathbb{R}.$$

### § 6.3 Sums of independent r.v.'s.

Question: Let  $X, Y$  be independent r.v.'s.

How to calculate the distribution of  $X+Y$ ?

Suppose  $X, Y$  are jointly cts and independent. Then  
the density of  $X+Y$  is given by

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

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

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

$$\begin{aligned}
 f_{X+Y}(a) &= f_X * f_Y(a) \\
 &= \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy \\
 &= \int_0^1 f_X(a-y) dy \quad (\text{since } f_Y(y) = \begin{cases} 1 & \text{if } y \in [0, 1] \\ 0 & \text{otherwise} \end{cases}) \\
 &\stackrel{\text{letting } z=a-y}{=} \int_{a-1}^a f_X(z) dz
 \end{aligned}$$

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

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

( verify it )

Hence

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

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

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

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

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

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  $(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$ .