## **Jointly Distributed Random Variables**

## Sums of Independent Random Variables [Ross S6.3]

Say X and Y are independent continuous random variables. What is the pdf of Z = X + Y?

$$F_{Z}(z) = P[X + Y \le z]$$

$$= \iint_{x+y \le z} f_{XY}(x, y) \, dxdy$$

$$= \iint_{x \le z - y} f_{X}(x) f_{Y}(y) \, dxdy$$

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

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

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

Hence:

$$f_Z(z) = \frac{d}{dz} F_Z(z) = \frac{d}{dz} \int_{-\infty}^{\infty} f_Y(y) F_X(z - y) dy$$
$$= \int_{-\infty}^{\infty} f_Y(y) \frac{d}{dz} F_X(z - y) dy$$
$$= \int_{-\infty}^{\infty} f_Y(y) f_X(z - y) dy$$

The pdf of Z = X + Y is the convolution of  $f_X(x)$  and  $f_Y(y)$ !

**Example 26.1:**  $X \sim U(0,1)$  and  $Y \sim U(0,1)$  are independent. What is the pdf of Z = X + Y?

Solution:

## Sum of Normal (Gaussian) Random Variables

**Proposition 26.1** Let  $X_1, X_2, ..., X_n$  be independent random variables with  $X_i \sim \mathcal{N}(\mu_i, \sigma_i^2)$ .

Let 
$$Z = X_1 + X_2 + \cdots + X_n$$
.

Then  $Z \sim \mathcal{N}(\mu_Z, \sigma_Z^2)$  where

$$\mu_Z = \mu_1 + \mu_2 + \dots + \mu_N$$
 $\sigma_Z^2 = \sigma_1^2 + \sigma_2^2 + \dots + \sigma_N^2$ 

Why?

We prove the result for the sum  $Z = X_1 + X_2$ . The general case follows by repeatedly applying the 2 variables case.

First determine the pdf of U = X + Y where  $X \sim \mathcal{N}(0, \sigma^2)$  and  $Y \sim \mathcal{N}(0, 1)$ .

$$f_X(u-y)f_Y(y) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(u-y)^2}{2\sigma^2}\right\} \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{y^2}{2}\right\}$$

$$= \frac{1}{2\pi\sigma} \exp\left\{-\frac{u^2}{2(1+\sigma^2)} - c\left(y - \frac{u}{1+\sigma^2}\right)^2\right\}$$

$$\left[\text{where } c = \frac{1+\sigma^2}{2\sigma^2}\right]$$

$$= \exp\left\{\frac{-u^2}{2(1+\sigma^2)}\right\} \frac{1}{2\pi\sigma} \exp\left\{-c\left(y - \frac{u}{1+\sigma^2}\right)^2\right\}$$

$$\begin{split} f_U(u) &= \int_{-\infty}^{\infty} f_X(u - y) f_Y(y) dy \\ &= \exp\left\{\frac{-u^2}{2(1 + \sigma^2)}\right\} \underbrace{\frac{1}{2\pi\sigma} \int_{-\infty}^{\infty} \exp\left\{-c\left(y - \frac{u}{1 + \sigma^2}\right)^2\right\} dy}_{\text{constant } K} \\ &= K \exp\left\{\frac{-u^2}{2(1 + \sigma^2)}\right\} \end{split}$$

But then  $U \sim \mathcal{N}(0, 1 + \sigma^2)$ .

Now, let  $X_1 \sim \mathcal{N}(\mu_1, \sigma_1^2)$  and  $X_2 \sim \mathcal{N}(\mu_2, \sigma_2^2)$ .

$$Z = X_1 + X_2 = \sigma_2 \left( \underbrace{\frac{X_1 - \mu_1}{\sigma_2}}_{X} + \underbrace{\frac{X_2 - \mu_2}{\sigma_2}}_{Y} \right) + \mu_1 + \mu_2$$
 where 
$$X \sim \mathcal{N}(0, \sigma_1^2/\sigma_2^2)$$
 
$$Y \sim \mathcal{N}(0, 1)$$

So 
$$U = X + Y \sim \mathcal{N}(0, 1 + \frac{\sigma_1^2}{\sigma_2^2})$$

and 
$$Z = \sigma_2 U + (\mu_1 + \mu_2)$$
 
$$\sim \mathcal{N}(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$$

**Definition 26.1:** A random variable Y is called **lognormal** with parameters  $\mu$  and  $\sigma$  if  $\log Y$  is normal with parameter  $\mu$  and  $\sigma^2$ , i.e., if

$$Y = e^X$$

where  $X \sim \mathcal{N}(\mu, \sigma^2)$ .

**Definition 26.2:** If the random variables  $X_1, X_2, \dots, X_n$  are **independent** and **identically distributed**, we say that they are **i.i.d.**, or **iid**.

**Example 26.2:** Let S(n) be the value of an investment at the end of week n.

A model for the evolution of S(n) is that

$$\frac{S(n)}{S(n-1)}$$

are iid lognormal random variables with parameters  $\mu$  and  $\sigma$ .

What is the probability that

- a) the value increases in each of the next two weeks?
- b) the value at the end of two weeks is higher than it is today?

Solution:

