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?

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

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

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

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

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

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

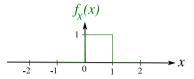
Hence:

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

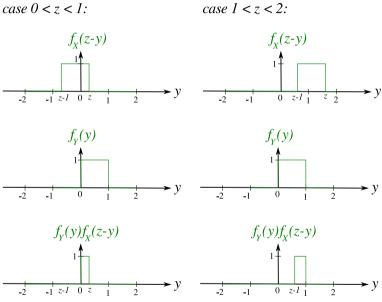
The pdf of Z = X + Y is the convolution of $f_X(x)$ and $f_Y(y)$!

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:



case 0 < z < 1:



Calculating the area of these rectangles:

$$f_Z(z) = \begin{cases} (z - 0) \times 1 & 0 \le z \le 1\\ (1 - (z - 1)) \times 1 & 1 \le z \le 2\\ 0 & \text{else} \end{cases}$$

$$= \begin{cases} z & 0 \le z \le 1 \\ 2-z & 1 \le z \le 2 \\ 0 & \text{else} \end{cases}$$

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 μ and σ if $\log Y$ is normal with parameter μ and σ^2 , i.e., if

$$Y = e^X$$

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

Definition 26.2: If the random variables $X_1, X_2, ..., X_n$ 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 μ and σ .

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: Let $U_1 \sim \mathcal{N}(\mu, \sigma^2)$, $U_2 \sim \mathcal{N}(\mu, \sigma^2)$, $Z_1 \sim \mathcal{N}(0, 1)$ and $Z_2 \sim \mathcal{N}(0, 1)$ be independent.

$$P[S(1) > S(0), S(2) > S(1)] = P\left[\frac{S(1)}{S(0)} > 1, \frac{S(2)}{S(1)} > 1\right]$$

$$= P\left[\ln \frac{S(1)}{S(0)} > 0, \ln \frac{S(2)}{S(1)} > 0\right]$$

$$= P[U_1 > 0] P[U_2 > 0]$$

$$= P\left[\frac{U_1 - \mu}{\sigma} > \frac{-\mu}{\sigma}\right] P\left[\frac{U_2 - \mu}{\sigma} > \frac{-\mu}{\sigma}\right]$$

$$= (1 - \Phi(-\mu/\sigma))^2$$

b)
$$P[S(2) > S(0)] = P\left[\frac{S(2)}{S(0)} > 1\right]$$

$$= P\left[\frac{S(2)}{S(1)} \frac{S(1)}{S(0)} > 1\right]$$

$$= P\left[\ln \frac{S(2)}{S(1)} + \ln \frac{S(1)}{S(0)} > 0\right]$$

$$= P\left[\underbrace{U_2 + U_1}_{\sim \mathcal{N}(\mu + \mu; \sigma^2 + \sigma^2)} > 0\right]$$

$$= P\left[\frac{U_2 + U_1 - 2\mu}{\sqrt{2\sigma^2}} > \frac{0 - 2\mu}{\sqrt{2\sigma^2}}\right]$$

$$= P\left[Z > -\frac{2\mu}{\sqrt{2\sigma^2}}\right]$$

$$= 1 - \Phi\left(-\frac{2\mu}{\sqrt{2\sigma^2}}\right)$$

 $P[X + Y = n] = P[\bigcup_{k=-\infty}^{\infty} \{X = k, Y = n - k\}]$

Example 26.3: Let $X \sim \mathsf{Poisson}(\lambda_1)$ and $Y \sim \mathsf{Poisson}(\lambda_2)$ be independent. What is the pmf of Z = X + Y?

Solution:

$$= \sum_{k=-\infty}^{\infty} P[X = k, Y = n - k]$$

$$= \sum_{k=-\infty}^{\infty} P[X = k]P[Y = n - k]$$

$$= \sum_{k=-\infty}^{\infty} P[X = k]P[Y = n - k]$$
 [since X and Y are ≥ 0]

$$\begin{split} &= \sum_{k=0}^{n} \frac{\lambda_{1}^{k}}{k!} e^{-\lambda_{1}} \frac{\lambda_{2}^{n-k}}{(n-k)!} e^{-\lambda_{2}} \\ &= e^{-(\lambda_{1}+\lambda_{2})} \sum_{k=0}^{n} \frac{\lambda_{1}^{k}}{k!} \frac{\lambda_{2}^{n-k}}{(n-k)!} \\ &= \frac{e^{-(\lambda_{1}+\lambda_{2})}}{n!} \sum_{k=0}^{n} \frac{n!}{k!(n-k)!} \lambda_{1}^{k} \lambda_{2}^{n-k} \\ &= \frac{e^{-(\lambda_{1}+\lambda_{2})}}{n!} \sum_{k=0}^{n} \binom{n}{k} \lambda_{1}^{k} \lambda_{2}^{n-k} \\ &= \frac{e^{-(\lambda_{1}+\lambda_{2})}}{n!} (\lambda_{1}+\lambda_{2})^{n} \qquad \text{[by Binomal Thm]} \end{split}$$

So $Z \sim \mathsf{Poisson}(\lambda_1 + \lambda_2)$.