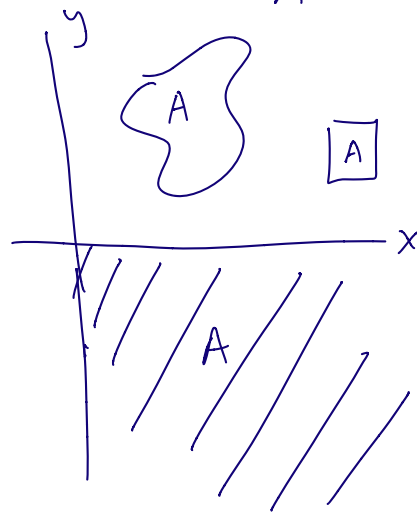


# Theory of Probability

Nov 9, 2020

Joint continuous distributions :

$$P[(X, Y) \in A] = \iint_A f(x, y) dx dy$$

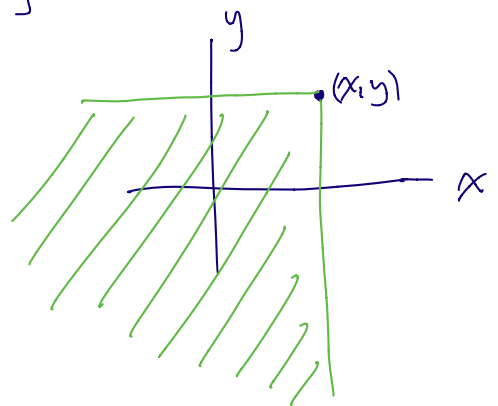


↑ joint probability density function

The distribution function is defined similarly:

$$F(x, y) = P[X \leq x, Y \leq y]$$

$$= \int_{-\infty}^y \int_{-\infty}^x f(u, v) du dv$$



$$\Rightarrow f(x, y) = \frac{\partial^2}{\partial x \partial y} F(x, y)$$

$$\underbrace{f_x(x)}_{\text{density for } X} = \int_{-\infty}^{\infty} f(x, y) dy$$

$$f_y(x) = \int_{-\infty}^{\infty} f(x, y) dx$$

## Independence of Random Variables.

Recall: Events  $A, B$  are independent if

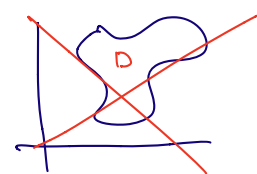
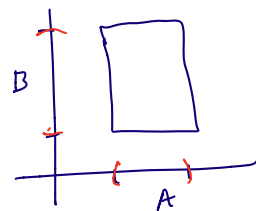
$$P[A|B] = P[A] \quad \Leftrightarrow \quad P[AB] = P[A] P[B].$$

Two random variables  $X, Y$  are independent if

$$P[X \in A, Y \in B] = P[X \in A] P[Y \in B].$$

$$\hookrightarrow = \int_A \int_B f(x,y) dy dx$$

$$= \int_A \left( \int_B f(x,y) dy \right) dx$$



If  $X, Y$  are independent:

$$(1) f(x,y) = f_x(x) f_y(y)$$

$$(2) F(x,y) = F_x(x) F_y(y).$$

$$P[X, Y \in D]$$

$$= \iint_{D(x,y)} f_x(x) f_y(y) dx dy$$

## Sums of Independent Random Variables

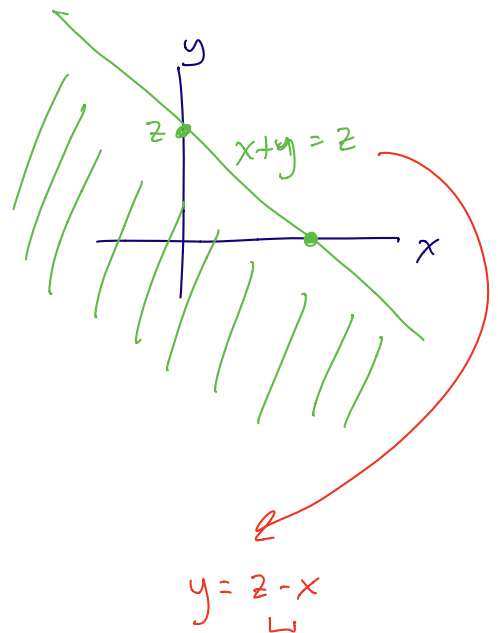
Let  $X, Y$  be independent random variables.

$$\Rightarrow f(x,y) = f_x(x) f_y(y).$$

Consider  $Z = X + Y$

$$\text{What is } F_z(z) = P[Z \leq z] ?$$

$$\begin{aligned}
 P[Z \leq z] &= P[X+Y \leq z] \\
 &= \iint_{x+y \leq z} f(x,y) \, dx \, dy \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{z-y} f(x,y) \, dx \, dy \\
 &= \int_{-\infty}^{\infty} \underbrace{\int_{-\infty}^{z-y} f_X(x) f_Y(y) \, dx}_{\text{does not depend on } y} \, dy
 \end{aligned}$$



$$P[Z \leq z] = \int_{-\infty}^{\infty} F_X(z-y) f_Y(y) \, dy = F_Z(z)$$

We know that

$$\begin{aligned}
 f_Z(z) &= \frac{d}{dz} F_Z(z) \\
 &= \frac{d}{dz} \int_{-\infty}^{\infty} F_X(z-y) f_Y(y) \, dy \\
 &= \underbrace{\int_{-\infty}^{\infty} f_X(z-y) f_Y(y) \, dy}_{\text{convolution of } f_X \text{ and } f_Y}
 \end{aligned}$$

If instead we wanted the density for  $\underbrace{W+X+Y}_U$

$$\Rightarrow P[U \leq u] = P[W+Z \leq u]$$

$$= \int \bar{F}_w(u-z) f_z(z) dz$$

$$\Rightarrow f_u = \int f_w(u-z) f_z(z) dz$$

$$= \int_{-\infty}^{\infty} f_w(u-z) \int_{-\infty}^{\infty} f_x(z-y) f_y(y) dy dz$$

iterated convolution

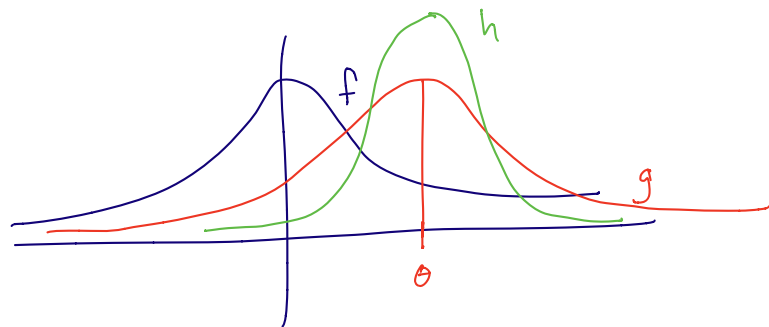
Ex:

$$\text{Gamma}(s, \lambda) + \text{Gamma}(t, \lambda) \sim \text{Gamma}(s+t, \lambda)$$

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

Cauchy

$$f(x) = \frac{1}{\pi} \frac{1}{1+x^2}$$



$$g(x) = f(x-\theta)$$

$$= \frac{1}{\pi} \frac{1}{1+(x-\theta)^2}$$

$$h(x) = f\left(\frac{x-\theta}{\tau}\right)$$

$$= \frac{1}{\pi} \frac{1}{1 + \frac{(x-\theta)^2}{\tau^2}} = \frac{\tau^2}{\pi} \frac{1}{\tau^2 + (x-\theta)^2}$$