

$$\text{Binomial} \sim (n, p) \quad E X = n p \quad V X = n p (1-p)$$

$$P(X=x) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & x=0, 1, \dots, n \\ 0 & \text{else} \end{cases}$$

$$\text{GEOM} \begin{cases} \mu = \frac{1-p}{p} & (1-p)^x p = P(X=x) \quad x \in [0, 1, \dots] \\ & V X = \frac{1-p}{p^2} \\ \mu = \frac{1}{p} & (1-p)^{x-1} p = P(X=x) \quad x \in [1, 2, \dots, n] \end{cases}$$

$$\text{HIPERGEOM} \sim (n, k, N) \quad \mu = \frac{n k}{N}$$

$$f_X(x) = P(X=x) = \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}}$$

$$\text{Multinomial} (n, p_1, p_2, \dots, p_r)$$

$$P(Y_1=y_1, Y_2=y_2, \dots, Y_r=y_r) = \begin{cases} \frac{n!}{y_1! y_2! \dots y_r!} p_1^{y_1} \dots p_r^{y_r} & \text{si } y_j \in \{0, 1, \dots, n\} \forall j=1, 2, \dots, r, \\ & y_1 + y_2 + \dots + y_r = n \\ 0 & \text{en caso contrario} \end{cases}$$

$$\boxed{\text{UNIFORME} \sim U([a, b])}$$

$$\mu = \frac{1}{2}(a+b) \quad \text{var} = \frac{1}{12}(b-a)^2$$

$$f(x) = \begin{cases} \frac{1}{b-a} & x \in [a, b] \\ 0 & \text{c.c.} \end{cases}$$

$$F_X(x) = P(X \leq x) = \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & x \in [a, b] \\ 1 & x > b \end{cases}$$

$$\boxed{\text{BINOM} \sim \text{BN}(r, p)} \quad \mu = r \frac{(1-p)}{p} \quad \text{var} = r \frac{(1-p)}{p^2}$$

$$P(X=x) = \binom{x+r-1}{x} (1-p)^x p^r$$

$$\boxed{\text{POISSON} \sim P(\lambda)} \quad \mu = \lambda \quad \text{var} = \lambda$$

$$P(X=x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

EXPONENCIAL

Sea X v.a., $X \sim E(\lambda)$ ($\lambda > 0$).

$$\mu = \frac{1}{\lambda} \quad \text{Var} = \frac{1}{\lambda^2}$$

Entonces

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & x > 0 \\ 0 & \text{c.c.} \end{cases}$$

Proposición

Sea $X \sim E(\lambda)$. Entonces $P(X > a+b | X > a) = P(X > b) \quad \forall a, b \geq 0$

Proposición (Caracterización de la Distribución exponencial)

Sea X v.a. continua tal que $\forall a, b \geq 0$,

$$P(X > a+b) = P(X > a) \cdot P(X > b)$$

Entonces: $P(X > 0) = 0$ ó $X \sim E(\lambda)$ para algún $\lambda > 0$

NORMAL ESTÁNDAR $N(0,1)$

$f: \mathbb{R} \rightarrow (0, +\infty)$, dada por:

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

NORMAL GRAL $N(\mu, \sigma^2)$

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

GAMMA

$$\Gamma(\alpha) = \int_0^{\infty} u^{\alpha-1} e^{-u} du$$

$$\cdot) \Gamma(\alpha+1) = \alpha \Gamma(\alpha)$$

$$\cdot) \Gamma(n+1) = n!$$

$$\Gamma(1) = 1$$

Sea X v.a. absolutamente continua con función de densidad dada por:

$$f(x) = \begin{cases} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} & x > 0 \\ 0 & \text{c.c.} \end{cases}$$

donde $\lambda > 0, \alpha > 0$.

$$EX = \frac{\alpha}{\lambda} \quad Var = \frac{\alpha}{\lambda^2}$$

Entonces decimos que X tiene distribución Gamma con parámetros α y λ . Notación $X \sim \Gamma(\alpha, \lambda)$

$$X \sim \Gamma(1, \lambda) \Leftrightarrow X \sim E(\lambda)$$

Sea X v.a. tal que $X \sim N(0, \sigma^2)$; entonces $X^2 \sim \Gamma(\frac{1}{2}, \frac{1}{2\sigma^2})$

CAUCHY

Sea X v.a. Se dice que X tiene distribución de Cauchy si su función de densidad es de la forma:

$$f_X(x) = \frac{1}{\pi(1+x^2)} \quad \forall x \in \mathbb{R}$$

Proposición

Sea X r.a. tal que $X \sim U(0,1)$. Sea $Y = g(X) = \tan\left(\pi X - \frac{\pi}{2}\right)$

Entonces $Y \sim$ de Cauchy

$\vdots \quad \uparrow g(t) \quad \vdots$

Cauchy es simétrica

NORMAL BIVARIADA

Caso Particular: $A = \text{id}$ y $\mu = (0,0)$

Distribución Normal Bivariada Estándar

$$f_{X,Y}(x,y) = \frac{1}{2\pi} e^{-\frac{x^2+y^2}{2}}$$

Sumas

$$X_i \sim B(n_i, p)$$

$$\sum_{i=0}^h X_i \sim B(\sum n_i, p)$$

$$X_i \sim P(\lambda_i)$$

$$\sum_{i=0}^h X_i \sim P(\sum_{i=0}^h \lambda_i)$$

$$X_i \sim BN(n_i, p)$$

$$\sum_{i=0}^h X_i \sim BN(\sum_{i=0}^h n_i, p)$$

Proposición

Sean X e Y v.a independientes de finides sobre (Ω, \mathcal{F}, P) tal que $X \sim \Gamma(\alpha_1, \lambda)$ e $Y \sim \Gamma(\alpha_2, \lambda)$. Entonces $(X+Y) \sim \Gamma(\alpha_1 + \alpha_2, \lambda)$

Proposición

Sean X e Y v.a independientes de finides sobre (Ω, \mathcal{F}, P) tal que $X \sim N(\mu_1, \sigma_1^2)$ e $Y \sim N(\mu_2, \sigma_2^2)$. Entonces $(X+Y) \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$

Definición

Sea X v.a. definida sobre (Ω, \mathcal{F}, P) . Entonces X se dice simétrica si X y $-X$ tienen exactamente la misma distribución. Esto es:

$$P(X \leq t) = P(-X \leq t) \quad \forall t \in \mathbb{R}$$

$$P\left(\bigcup_{i=1}^n A_i\right) = \sum_{i=1}^n (-1)^{i-1} S_i \quad 1 \leq i_1 < i_2 < \dots < i_n \leq n$$
$$S_n = \sum P(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_n})$$

$$F_{X+Y}(z) = \int_{-\infty}^z \int_0^z f_{XY}(x, z-x) dx dz$$
$$= \int_{-\infty}^z \int_0^z f_{XY}(u-y, y) du dy$$

$$f_{X+Y}(z) = \int_{-\infty}^z f_X(z-y) f_Y(y) dy$$

$$= \int_{-\infty}^z f_X(x) f_Y(z-x) dx$$

Discreta

$$f_{X+Y}(t) = \sum_{i=0}^t f_X(i) f_Y(t-i)$$
$$P(X+Y=t) = P(X=t) P(X=t-i)$$

Entonces definimos la función $f_{Y/X} : \mathbb{R} \times \mathbb{R} \rightarrow [0,1]$ como:

$$f_{Y/X}(y/x) = \begin{cases} \frac{f(x,y)}{f_X(x)} & \text{si } f_X(x) > 0 \\ 0 & \text{si } f_X(x) = 0 \end{cases}$$

$$Z = XY$$

$$F_Z(z) = \int_{-\infty}^{\infty} \int_{-\infty}^z \frac{f_{XY}(x, \frac{z}{x})}{|x|} dx dy = f_Z(z)$$

Esperanzas

$$U \sim ([a, b])$$

$$EU = \frac{b+a}{2}$$

$$X \sim P(\alpha, \lambda)$$

$$EX = \frac{\alpha}{\lambda}$$

PROPERTIES

$$1) X \text{ et } Y \text{ indépendantes} \Rightarrow E(XY) = EX \cdot EY \quad \text{✗}$$

$$2) E(aX + Y) = aEX + EY$$

$$3) P(X \geq Y) = 1 \Rightarrow EX \geq EY$$

$$P(X \geq t) \leq \frac{EX}{t} \quad (\text{MARKOV})$$

$$P(|X - \mu| \geq t) \leq \frac{\sigma^2}{t^2} \quad (\text{CHEVYCHEV})$$

LEY DE BIL $P(X_n \leq X) \approx \Phi\left(\frac{X_n - EX}{\sqrt{\text{var } X_n}}\right)$

$$P\left(\left|\frac{X_n}{n} - \mu\right| \geq \varepsilon\right) \xrightarrow{n \rightarrow \infty} 0$$

$$\text{Cov}(X, Y) = EXY - EXEY = \text{var}(X)$$

$$\text{car } Y = X$$

$$= 0$$

si X et Y
indépendantes

$$4) \text{Var}(X) = EX^2 - (EX)^2$$

$$5) (EXY)^2 \leq E(X^2)E(Y^2)$$

$$a) |E(XY)| \leq \sqrt{E X^2} \sqrt{E Y^2}$$

$$b) \text{Var}(X+Y) = \text{Var} X + \text{Var} Y + 2 \text{Cov}(X, Y)$$

$$c) \text{Var}(X+c) = \text{Var} X$$

$$d) \text{Var}(cX) = c^2 \text{Var}(X)$$

Lema 3:

Si X_1, X_2, \dots, X_n son r.v.a tal que $E(X_i^2) < \infty \forall i=1, 2, \dots, n$

entonces:

$$\text{Var}\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n \text{Var}(X_i) + \sum_{1 \leq i \neq j \leq n} \text{Cov}(X_i, X_j)$$

$$= \sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{1 \leq i < j \leq n} \text{Cov}(X_i, X_j)$$