

P11

(a) Para ver que $C = 1$ notemos que la densidad

$f_X(x)$ debe cumplir que $\int_{-\infty}^{+\infty} f_X(x) dx = 1$

$$1 = \int_{-\infty}^{+\infty} f_X(x) dx = \int_{-1}^0 -x dx + \int_0^1 (C-x) dx$$

$$= -\frac{x^2}{2} \Big|_{-1}^0 + \left(Cx - \frac{x^2}{2} \right) \Big|_0^1$$

$$= \frac{1}{2} + C - \frac{1}{2} = C \Rightarrow C = 1$$

(b) Queremos calcular $F_X(x) \quad \forall x \in \mathbb{R}$

Por la definición de $f_X(x)$ vemos que

$$\forall x < -1 \quad F_X(x) = 0$$

$$\forall x \geq 1 \quad F_X(x) = 1$$

Si $x \in (-1, 0)$

$$P(X \leq x) = F_X(x) = \int_{-\infty}^x -t dt = \int_{-1}^x -t dt$$

$$= -\frac{t^2}{2} \Big|_{-1}^x = \frac{1}{2} - \frac{x^2}{2}$$

ANTES DE -1
VALE CERO

$$S: x \in (0, 1]$$

$$P(X \leq x) = F_X(x) = \int_{-\infty}^x f_X(t) dt$$

$$= \int_{-1}^0 f_X(t) dt + \int_0^x f_X(t) dt$$

$$= \int_{-1}^0 -t dt + \int_0^x (c - t) dt$$

$$= -\frac{t^2}{2} \Big|_{-1}^0 + \left(t - \frac{t^2}{2} \right) \Big|_0^x = \frac{1}{2} + x - \frac{x^2}{2}$$

$$c=1$$

(c) Analicemos la distribución de Y .

$$F_Y(y) = P(Y \leq y) = P(|X| \leq y)$$

$$= P(-y \leq X \leq y)$$

Notar que si $y > 1$

$$F_Y(y) = 1$$

Si $y < 0$ no tiene sentido, $F_Y(y) = 0$

$$\begin{aligned} |X| \leq y \\ \Leftrightarrow -y \leq X \leq y \end{aligned}$$

$$= P(X \leq y) - P(X \leq -y) = F_X(y) - F_X(-y)$$

$$= \frac{1}{2} + y - \frac{y^2}{2} - \left(\frac{1}{2} - \frac{y^2}{2} \right)$$

Por (b)

$$= y$$

$$\Rightarrow \forall y \in [0, 1]$$

$$f_Y(y) = 1$$

(Recordar que $f_Y(y) = \frac{d}{dy} (F_Y(y))$)

$\overline{P2}$ $X_1 \sim \exp(\lambda_1)$
 $X_2 \sim \exp(\lambda_2)$ indep.

Veamos la distribución de Y

$$\begin{aligned} F_Y(y) &= P(\min\{X_1, X_2\} \leq y) \\ &= 1 - P(\min\{X_1, X_2\} \geq y) \\ &= 1 - P(X_1 \geq y, X_2 \geq y) \\ &= 1 - P(X_1 \geq y) P(X_2 \geq y) \\ &= 1 - (1 - P(X_1 \leq y))(1 - P(X_2 \leq y)) \quad (*) \end{aligned}$$

Recordar que si $Z \sim \exp(\mu)$

$$F_Z(t) = 1 - e^{-\mu t}$$

Por lo tanto en (*)

$$\begin{aligned} (*) &= 1 - (e^{-\lambda_1 y})(e^{-\lambda_2 y}) \\ &= 1 - e^{-(\lambda_1 + \lambda_2)y} \end{aligned}$$

$\therefore$ la densidad es

$$f_Y(y) = \frac{d}{dy} F_Y(y) = (\lambda_1 + \lambda_2) e^{-(\lambda_1 + \lambda_2)y}$$

$$\Rightarrow \min\{X_1, X_2\} \sim \exp(\lambda_1 + \lambda_2)$$

[P3]

$$X \sim \exp(\lambda)$$

con $k \in \mathbb{N}$

$$(a) P(\lfloor X \rfloor = k) = P(k \leq X < k+1)$$

$$= \int_k^{k+1} f_X(x) dx = \int_k^{k+1} \lambda e^{-\lambda x} dx = \left(-e^{-\lambda x} \right) \Big|_k^{k+1} \\ = e^{-\lambda k} - e^{-\lambda(k+1)}$$

(b) Notemos que para una exponencial se tiene que

$$P(X > y) = 1 - P(X \leq y) = 1 - \int_{-\infty}^y \lambda e^{-\lambda t} dt = 1 - \int_0^y \lambda e^{-\lambda t} dt$$

$$= 1 - \left(-e^{-\lambda t} \right) \Big|_0^y = 1 - (1 - e^{-\lambda y}) =$$

$$= e^{-\lambda y}$$

$$\{X > t+s\} \supseteq \{X > s\}$$

$$\therefore P(X > t+s | X > s) = \frac{P(X > t+s, X > s)}{P(X > s)} = \frac{P(X > t+s)}{P(X > s)}$$

$$= \frac{e^{-(t+s)\lambda}}{e^{-s\lambda}} = e^{-t\lambda} = P(X > t)$$

[P4]

$$\begin{aligned} E\left(\frac{1}{1+X}\right) &= \sum_{x \in R_X} g(x) P(X=x) \\ &= \sum_{k=0}^m g(k) P(X=k) = \sum_{k=0}^m \frac{1}{1+k} \underbrace{\binom{m}{k} p^k (1-p)^{m-k}}_{P(X=k)} \\ &\quad \uparrow \\ &\quad g(x) = \frac{1}{1+x} \quad X \sim \text{Bin}(m, p) \\ &= \sum_{k=0}^m \frac{m!}{(m-k)! k!} \cdot \frac{1}{(k+1)} p^k (1-p)^{m-k} \\ &= \frac{1}{m+1} \sum_{k=0}^m \frac{(m+1)!}{((m+1)-(k+1))!} \cdot \frac{1}{(k+1)!} p^k (1-p)^{m-k} \\ &= \frac{1}{m+1} \sum_{k=0}^m \binom{m+1}{k+1} p^k (1-p)^{m-k} \\ &= \frac{1}{m+1} \sum_{k=1}^{m+1} \binom{m+1}{k} p^{k-1} (1-p)^{m-(k-1)} \\ &= \frac{1}{m+1} \left(\sum_{k=0}^{m+1} \binom{m+1}{k} \frac{p^k}{p} (1-p)^{(m+1)-k} - \frac{(1-p)^{m+1}}{p} \right) \\ &= \frac{1}{m+1} \left(\frac{1}{p} \left(\sum_{k=0}^{m+1} \binom{m+1}{k} p^k (1-p)^{(m+1)-k} - (1-p)^{m+1} \right) \right) \\ &= \frac{1}{p(m+1)} \left((p+1-p)^{m+1} - (1-p)^{m+1} \right) \\ &= \frac{1 - (1-p)^{m+1}}{p(m+1)} \end{aligned}$$

P5. Si X es una v.a. discreta

$$E(X) = \sum_{k \in R_X} k P(X=k)$$

Entonces podemos tomar una variable que

$$R_X = \{1, 2, \dots\} \quad P(X=k) = \frac{6}{\pi^2 k^2}$$

→ Esto vale de que sabemos $\sum_{k=1}^{+\infty} \frac{1}{k^2} = \frac{\pi^2}{6}$

$$\Rightarrow \sum_{k=1}^{+\infty} \frac{6}{\pi^2 k^2} = 1$$

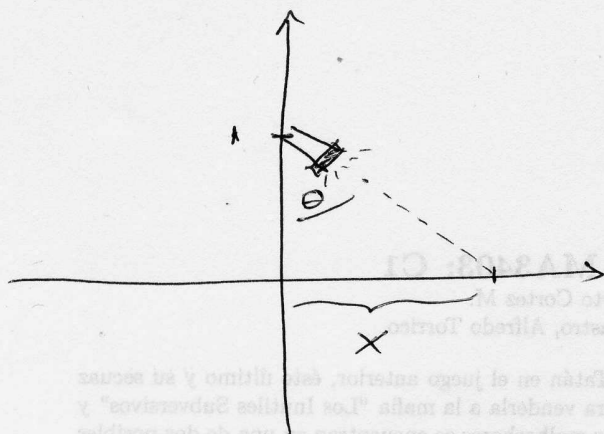
Dado lo anterior

$$E(X) = \sum_{k \in R_X} k P(X=k) = \sum_{k=1}^{+\infty} k \cdot \frac{6}{\pi^2 k^2}$$

$$= \frac{6}{\pi^2} \sum_{k=1}^{+\infty} \frac{1}{k}$$

Suma que sabemos que no converge.

P6



Notemos que $\tan(\theta) = X$ y sabemos que

$$\theta \sim U(-\pi/2, \pi/2)$$

Buscamos $F_X(t)$ y $f_X(t)$

$$F_X(t) = P(X \leq t) = P(\tan(\theta) \leq t)$$

$\tan(\cdot)$ biyectiva en $(-\pi/2, \pi/2)$

$$= P(\theta \leq \arctan(t)) = \int_{-\infty}^{\arctan(t)} f_{\theta}(u) du$$

con $-\pi/2 < \arctan(t) \leq \pi/2$

$$= \int_{-\pi/2}^{\arctan(t)} \frac{1}{\frac{\pi}{2} - (-\frac{\pi}{2})} du = \frac{1}{\pi} (\arctan(t) - (-\pi/2))$$

$$= \frac{\arctan(t)}{\pi} + \frac{1}{2} \quad \forall t \in (-\infty, +\infty)$$

$$\Rightarrow f_X(t) = \frac{d}{dt} F_X(t) = \frac{d}{dt} \left(\frac{\arctan(t)}{\pi} + \frac{1}{2} \right) = \frac{1}{\pi(1+t^2)}$$

veamos ahora

$$E(X) = \int_{-\infty}^{+\infty} t f_X(t) dt = \int_{-\infty}^{+\infty} \frac{t}{\pi(1+t^2)} dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{2t}{1+t^2} dt$$

$$= \frac{1}{2\pi} \int_{-\infty}^{+\infty} (\ln(1+t^2))' dt = \frac{1}{2\pi} (\ln(1+t^2) \Big|_{-\infty}^{+\infty}) (*)$$

(*) ES INDEFINIDO PORQUE

$$(*) = \lim_{a \rightarrow +\infty} \frac{1}{2\pi} (\ln(1+t^2) \Big|_{-2a}^a)$$

$$= \lim_{a \rightarrow +\infty} \frac{1}{2\pi} (\ln(1+a^2) - \ln(1+4a^2))$$

$$= \frac{1}{2\pi} \lim_{a \rightarrow +\infty} \ln\left(\frac{1+a^2}{1+4a^2}\right) = \frac{1}{2\pi} \ln\left(\frac{1}{4}\right)$$

y también

$$(*) = \lim_{a \rightarrow +\infty} \frac{1}{2\pi} (\ln(1+t^2) \Big|_{-a}^a)$$

$$= \lim_{a \rightarrow +\infty} \frac{1}{2\pi} (\ln(1+a^2) - \ln(1+a^2)) = 0$$

[P7]. Definamos

$$X_i = \begin{cases} 1 & \text{la bolita } i \text{ es blanca} \\ 0 & \text{no} \end{cases}$$

$$P(X_i = 1) = p = \frac{m}{N}$$

$$X = \sum_{i=1}^m X_i = \text{cantidad de bolitas blancas (al sacar } m)$$

$$E(X) = E\left(\sum_{i=1}^m X_i\right) \stackrel{\text{linealidad}}{=} \sum_{i=1}^m E(X_i) = \sum_{i=1}^m \frac{m}{N} = \boxed{\frac{m}{N} \cdot m}$$
$$E(X_i) = 1 \cdot P(X_i = 1) + 0 \cdot P(X_i = 0) = 1 \cdot p = \frac{m}{N}$$