

Intercambio de masa de probabilidad conjunta

$$P_{X,Y}(x,y) = P(X=x, Y=y)$$

$$P_X(x) = \sum_y P_{X,Y}(x,y) \text{ y } P_Y(y) = \sum_x P_{X,Y}(x,y)$$

↓
marginales
a partir de la conjunta

$$\textcircled{a} P_Z(z) = \sum_{\{(x,y) | g(x,y)=z\}} P_{X,Y}(x,y) \text{ dado } z = g(X,Y)$$

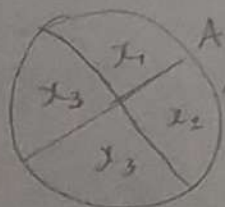
$$\textcircled{a} E[g(X,Y)] = \sum_x \sum_y g(x,y) \cdot P_{X,Y}(x,y)$$

$$\textcircled{a} E(ax + by + c) \\ = a \cdot E(X) + b \cdot E(Y) + c$$

Condicionamiento de
Varios Aleatorios

Condicionar a eventos normalista:

$$\bullet P(X=x | A) = \frac{P(\{X=x\} \cap A)}{P(A)}$$



Teorema:

$$P(A) = \sum_x \frac{P(\{X=x\} \cap A)}{P(A)} = \frac{P(A)}{P(A)} = 1$$

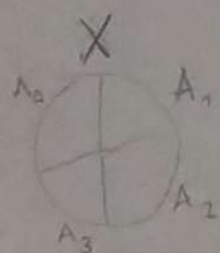
$$\textcircled{a} P_{X|Y}(x|y) = \frac{P(\{X=x\} \cap \{Y=y\})}{P(\{Y=y\})} = \frac{P_{X,Y}(x,y)}{P_Y(y)}$$

$$\Rightarrow P_{X,Y}(x,y) = P_{X|Y}(x|y) \cdot P_Y(y)$$

$$\textcircled{1} P_x(x) = \sum_y P_y(y) \cdot P(x|y)$$

$$\textcircled{2} P_{(x,y)|z} = \frac{P_{x,y,z}(x,y,z)}{P(z)} \quad \textcircled{3} P_{x|y,z} = \frac{P_{x,y,z}(x,y,z)}{P_{y,z}(y,z)}$$

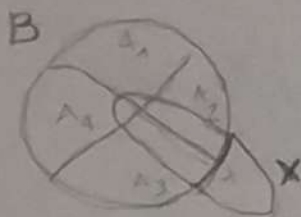
Teorema



Si X tiene A_n particiones con $P(A_i) > 0, i=1, \dots, n$ entonces

$$P_x(x) = \sum_{i=1}^n P(A_i) \cdot P_{x|A_i}(x)$$

Teorema



Sea B un evento t.q. $P(A_i \cap B) > 0, \forall i$

$$\begin{aligned} P_{x|B}(x) &= \sum_{i=1}^n P(A_i|B) \cdot P_{x|(A_i \cap B)}(x) \\ &= \frac{P(A_i \cap B)}{P(B)} \cdot \frac{P(\{x=x\} \cap A_i \cap B)}{P(A_i \cap B)} \end{aligned}$$

Valor esperado condicional

- $E(X|A) = \sum_x x \cdot P_{x|A}(x)$ es un evento
- $E(X|y=y) = \sum_x x \cdot P_{x|y}(x|y=y)$

Teorema

$$E(g(x)|A) = \sum_x g(x) \cdot P_{x|A}(x)$$

Teorema

Sean A_n eventos que forman una partición de $\Omega, P(A_i) > 0, \forall i \in \mathbb{N}$.

$$\textcircled{1} E(x) = \sum_{i=1}^n P(A_i) \cdot E(x|A_i) \quad \textcircled{2} P(A_i \cap B) > 0 \Rightarrow E(x|B) = \sum_{i=1}^n P(A_i|B) \cdot E(x|A_i)$$

$$\textcircled{3} E(x) = \sum_y P_y(y) \cdot E(x|y=y)$$

Independencia

• si $P(\{X=x\} \cap A) = P(\{X=x\}) \cdot P(A) = P_x(x) \cdot P(A)$
 $\forall x \in X$.

Teorema

• sea A un evento $P(A) > 0$, y X indep de A
entonces $P_{X|A}(x) = P_x(x)$

$$\bullet P_{xy}(x, y) = P_x(x) \cdot P_y(y) \quad \forall x \in X, \forall y \in Y$$

Teorema

• dado x e y independientes:

$$P_{xy}(x, y) = P_x(x) \cdot P_y(y)$$

Independencia condicional

$$\bullet P_{(x,y)|A}(x, y) = P_{x|A}(x) \cdot P_{y|A}(y)$$

• indep. condicional \nRightarrow indep. X, Y

• indep. de $X, Y \nRightarrow$ indep. condicional

• dado X, Y y cond. indep y $P_{y|A}(y) > 0$

$$\text{entonces } P_{x|y,A}(x) = P_{x|A}(x)$$

Teorema

• X, Y v.a.'s indep y discretas

$$E(XY) = E(X)E(Y)$$

Teorema

• X, Y v.a.'s indep. $f, g: \mathbb{R} \rightarrow \mathbb{R} \Rightarrow f(x)$ y $g(y)$ son indep.

Corolario

• si X, Y son indep. $f, g: \mathbb{R} \rightarrow \mathbb{R}$

$$E(f(x) \cdot g(y)) = E(f(x)) \cdot E(g(y))$$

Distribuciones de probabilidad

• Normal:

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

$$E[X] = \mu \quad \text{Var}(X) = \sigma^2$$

• Bernoulli:

$$P_X(k) = \begin{cases} p, & \text{si } k=1 \\ 1-p, & \text{si } k=0 \end{cases}$$

$$E(X) = p \quad \text{Var}(X) = p(1-p)$$

• Binomial

$$P_X(k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k=0, 1, \dots, n$$

$$E(X) = np$$

• Geométrica

$$P_X(k) = (1-p)^{k-1} p, \quad k=1, 2, 3, \dots$$