

Theorem: Cov/Cor of Independent RVs

If $X \perp Y$ then $\text{Cov}(X, Y) = \text{Cor}(X, Y) = 0$.

pf:
$$\begin{aligned} \text{Cov}(X, Y) &= E[XY] - EX EY \\ &= (EX)(EY) - (EX)(EY) = 0 \end{aligned}$$

So $\text{Cor}(X, Y) = 0$ b/c cor is just re-scaled cov.

Converse is generally false.

If $\text{Cor}(X, Y) = 0$ they may or may not be independent.

Ex $X \sim N(0, 1)$ and $Y = X^2$

not independent

$$\begin{aligned} \text{Cov}(X, Y) &= E[XY] - EX EY \\ &= E[X^3] - (EX)(E[X^2]) \end{aligned}$$

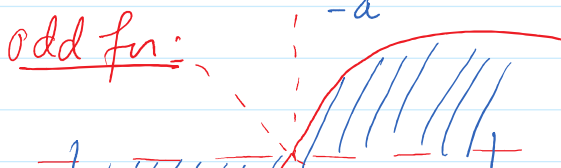
$$= E[X^3]$$

$$= 0$$

$$E[X^3] = \int_{-\infty}^{\infty} x^3 \frac{1}{\sqrt{2\pi}} e^{-x^2} dx = 0$$

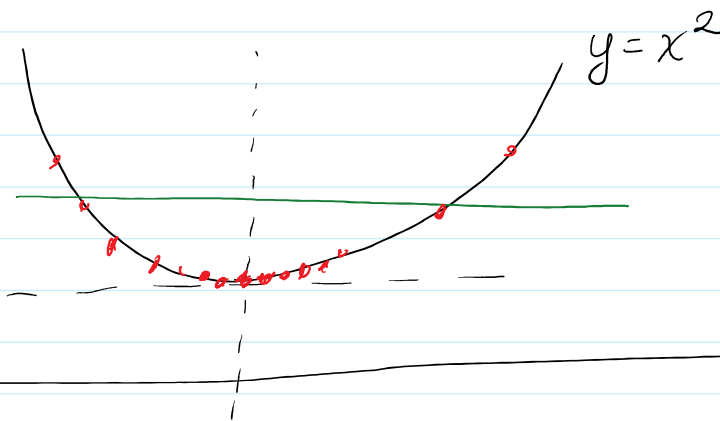
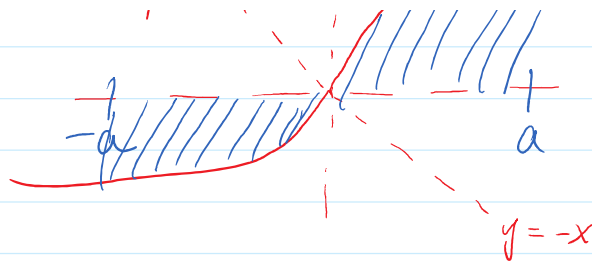
odd

$$\int_{-a}^a \text{odd fn} = 0$$



$$= 0$$

So $\text{Cor}(X, Y) = 0.$



Bayes' Theorem

Events: $P(A|B) = \frac{P(B|A) P(A)}{P(B)}$

RVs: $f(x|y) = \frac{f(y|x) f_x(x)}{f_y(y)}$

Law of Total Probability

What is the dist of Y ?

Law of Total Prob.

$$Z \sim \text{Pois}(\lambda)$$
$$f(z) = \lambda^z e^{-\lambda}$$

$$f(y) = \int_{\mathbb{R}} f(y|x) f(x) dx = \int_0^{\infty} \underbrace{\frac{x^y e^{-x}}{y!}}_{\text{Pois}(x)} \underbrace{\lambda e^{-\lambda x}}_{\text{Exp}(\lambda)} dx$$

$$= \frac{\lambda}{y!} \int_0^{\infty} x^{\overbrace{y}^{a-1}} e^{-\overbrace{(\lambda+1)}^b x} dx$$

PDF Gamma(a, b)

$$\frac{x^{a-1} e^{-bx}}{\Gamma(a)} \frac{b^a}{\Gamma(a)}$$

$$= \frac{\lambda}{y!} \frac{\Gamma(a)}{b^a} \int_0^{\infty} \underbrace{x^{a-1} e^{-bx} \frac{b^a}{\Gamma(a)}}_{\text{integral of a Gamma PDF}} dx$$

integral of a Gamma PDF
 $= 1$

$$= \frac{\lambda}{y!} \frac{\cancel{\Gamma(y+1)}}{\cancel{y!} (\lambda+1)^{y+1}}$$

$$f(y) = \frac{\lambda}{(\lambda+1)^{y+1}} \quad \text{for } y = 0, 1, 2, 3, \dots$$

Ex.

$$Y \sim \text{Pois}(\lambda)$$

$$X|Y=y \sim \text{Bin}(y, p)$$

$$0 \leq X \leq Y$$

$$0 < p < 1$$

$$\binom{y}{x} \frac{1}{y!} = \frac{y!}{x!(y-x)!} \frac{1}{y!}$$

What is the dist of X ?

$$= \frac{1}{x!(y-x)!}$$

$$f(x) = \sum_y \underbrace{f(x|y)}_{\text{Bin}(y,p)} \underbrace{f(y)}_{\text{Pois}(\lambda)} = \sum_{y=x}^{\infty} \binom{y}{x} p^x (1-p)^{y-x} \frac{\lambda^y e^{-\lambda}}{y!}$$

$$= \frac{p^x e^{-\lambda} \lambda^x}{x!} \sum_{y=x}^{\infty} \frac{1}{(y-x)!} (1-p)^{y-x} \lambda^{y-x}$$

$$= \frac{p^x e^{-\lambda} \lambda^x}{x!} \sum_{y=x}^{\infty} \frac{1}{(y-x)!} [(1-p)\lambda]^{y-x}$$

$$= \frac{p^x e^{-\lambda} \lambda^x}{x!} \sum_{y=0}^{\infty} \frac{1}{y!} \underbrace{[(1-p)\lambda]^y}_{e^{(1-p)\lambda}}$$

$$e^z = \sum_{i=0}^{\infty} \frac{z^i}{i!}$$

$$= \frac{p^x e^{-\lambda} \lambda^x e^{(1-p)\lambda}}{x!}$$

$$f(x) = \frac{(p\lambda)^x e^{-(p\lambda)}}{x!}$$

$$Z \sim \text{Pois}(\lambda)$$

$$\frac{\lambda^3 e^{-\lambda}}{3!}$$

$$X \sim \text{Pois}(\lambda)$$

Theorem: Iterated Expectation

If X and Y are RVs then

$$E[X] = E[E[X|Y]]$$

$$E[X|Y=y] = \int x f(x|y) dx = g(y)$$

For each $y \in \mathbb{R}$ this defines some fn

$$g(y) = E[X|Y=y]$$

↑ a real valued fn

e.g. $g(y) = y^2$ or $g(y) = 1 + y$

We can plug Y into g to get $g(Y)$

e.g. $g(Y) = Y^2$ or $g(Y) = 1 + Y$

Might want to write

$$g(Y) = E[X|Y=Y]$$

$$g(y) = E[X|Y=y]$$

$$g(y) = E[X | Y=y]$$

unknow

a number

$$= E[X | Y]$$

notation
a RV

Summary: $E[X | Y=y] = \text{a number}$
 $E[X | Y] = \text{a RV}$

ex, $E[X | Y=y] = y^2$

then $E[X | Y] = Y^2$

Ex. Prev. Ex. $Y \sim \text{Pois}(\lambda)$

$$X|Y=y \sim \text{Bin}(y, p)$$

What is $E[X]$?

Iterated Expectation: $E[X] = E[E[X|Y]]$

$$\textcircled{1} E[X|Y=y] = yp$$

$$\textcircled{2} E[X|Y] = Yp$$

$$\textcircled{3} E[E[X|Y]] = E[Yp] = pE[Y]$$

$= p\lambda$

Ex. $P \sim \text{Beta}(\alpha, \beta)$

$$X|P=p \sim \text{Bin}(n, p)$$

Known

$E[X]$?

$$\textcircled{1} E[X|P=p] = np$$

$$\textcircled{2} E[X|P] = nP$$

$$\textcircled{3} E[E[X|P]] = E[nP] = nE[P]$$

$= n \frac{\alpha}{\alpha + \beta} = E[X]$

pf. (cts case)

$$\rightarrow \textcircled{1} \quad f(x) = \int f(x, y) dy$$

$$\textcircled{2} \quad f(x|y) = \frac{f(x, y)}{f(y)} \Leftrightarrow f(x, y) = f(x|y) f(y)$$

$$\textcircled{3} \quad g(y) = E[X|Y=y] = \int x f(x|y) dx$$

$$E[X] = \int x f(x) dx \stackrel{\textcircled{1}}{=} \int x \int f(x, y) dy dx$$

$$\stackrel{\textcircled{2}}{=} \int x \int f(x|y) f(y) dy dx$$

re arrange

$$= \int \underbrace{\int x f(x|y) dx}_{g(y)} f(y) dy$$

$$= \int g(y) f(y) dy$$

$$= E[g(Y)]$$

$$= E[E[X|Y]]$$

notation for $g(Y)$