

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

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

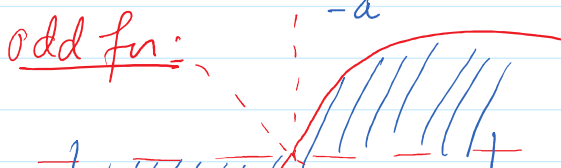
$$= E[X^3]$$

$$= 0$$

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

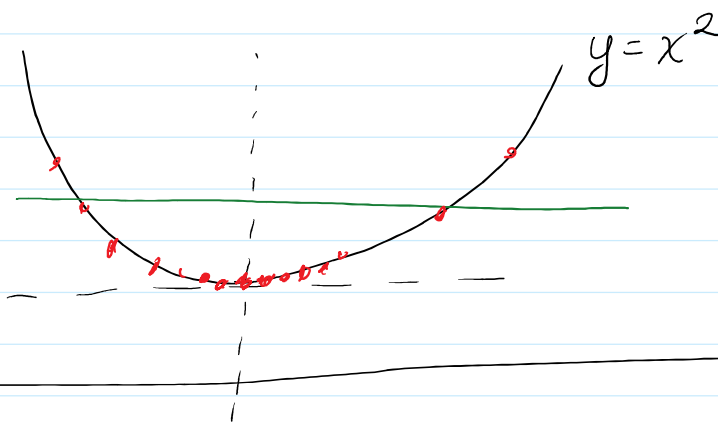
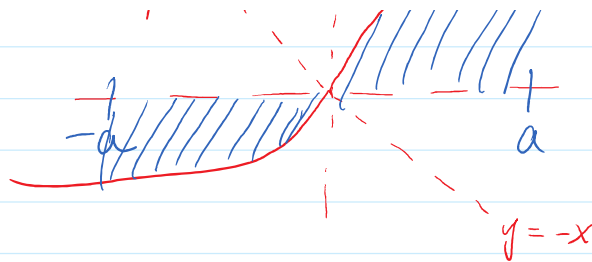
odd

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



$$= 0$$

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

Events: (C_i) partition S then

$$P(A) = \sum_i P(A|C_i) P(C_i)$$

RVs: discrete:

$$f(y) = \sum_x f(y|x) f(x)$$

cts:

$$f(y) = \int_{\mathbb{R}} f(y|x) f(x) dx$$

pf. (cts)

$$(1) f(y|x) = \frac{f(x,y)}{f(x)} \Leftrightarrow \underline{f(x,y) = f(y|x) f(x)}$$

$$(2) \underline{f(y)} = \int_{\mathbb{R}} f(x,y) dx = \int_{\mathbb{R}} \underline{f(y|x) f(x)} dx$$

Ex. $X \sim \text{Exp}(\lambda)$

$Y|X=x \sim \underline{\text{Pois}(x)}$

What is the dist of Y ?

Law of Total Prob.

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

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

Pois(x) Exp(λ)

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

PDF Gamma(a, b)

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

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

integral of a Gamma PDF
= 1

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

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

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

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

$$0 \leq X \leq Y$$

$$0 < p < 1$$

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

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

What is the dist of X ?

$$f(x) = \sum f(x|y) f(y) = \sum_{y=x}^{\infty} \binom{y}{x} p^x (1-p)^{y-x} \lambda e^{-\lambda}$$

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

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

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

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

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

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

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

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

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

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

Theorem: Iterated Expectation

If X and Y are RVs then

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

a RV

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

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

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

awkward

$$= E[X|Y]$$

notation

a RV

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

a number

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

① $E[X|Y=y] = yp$

② $E[X|Y] = Yp$

③ $E[E[X|Y]] = E[Yp] = pEY$
 $\boxed{= p\lambda}$

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

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

$E[X]?$

Known

$$(1) E[X|P=p] = np$$

$$(2) E[X|P] = nP$$

$$(3) E[E[X|P]] = E[nP] = nE[P]$$

$$= n \frac{\alpha}{\alpha + \beta} = EX$$

pf. (cts case)

$$\rightarrow (1) f(x) = \int f(x,y) dy$$

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

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

$$EX = \int x f(x) dx \stackrel{(1)}{=} \int x \int f(x,y) dy dx$$

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

re arrange

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

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

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

notation for $g(Y)$

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