

Recall :-

- Conditional Probability

- Independence

$$x_1, x_2, \dots, x_n \text{ are i.i.d.} \equiv \begin{cases} i - \text{independent} \\ i - \text{identically} \\ d - \text{distributed} \end{cases}$$

[Discrete r.v.s]

$$P(x_i \leq x) = P(x_i \leq x) \quad \forall i \geq 2$$

$$\text{and } P(x_{i_1} = x_1, x_{i_2} = x_2, \dots, x_{i_n} = x_n) \\ \{x_i \in \mathbb{R}, i \geq 1\} = \prod_{j=1}^n P(x_{i_j} = x_j)$$

[Discrete r.v.s] - Joint distribution (x_1, \dots, x_n)

$$\{P(x_1 = x_1, \dots, x_n = x_n)\}_{x_i \in \text{Range}(x_i)}$$

Sums of Random Variables

Recall :

Let $X, Y \sim \text{Bernoulli}(p)$ be two independent random variables.

$Z = X + Y, Z \sim ?$. $Z \sim \text{Binomial}(2, p)$

Inductively :-

X_1, X_2, \dots, X_n are i.i.d Bernoulli (p)

$\Rightarrow X = \sum_{i=1}^n X_i \sim \text{Binomial}(n, p)$

Sums of Random Variables

Let $X \sim \text{Poisson}(\lambda_1)$ and $Y \sim \text{Poisson}(\lambda_2)$ be independent random variables.

(a) Let $Z = X + Y$. Find the distribution of Z .

(b) Find the conditional distribution of $X | Z$.

(a) $X \sim \text{Poisson}(\lambda_1) \Rightarrow \text{Range}(X) \in \{0, 1, 2, \dots\}$
 $Y \sim \text{Poisson}(\lambda_2) \Rightarrow \text{Range}(Y) \in \{0, 1, 2, \dots\}$
 $\Rightarrow Z = X + Y$ has $\text{Range}(Z) = \{0, 1, 2, \dots\}$

Take $z \in \text{Range}(Z)$

$$P(Z=z) = P(X+Y=z) = P\left(\bigcup_{i=0}^z (X=i, Y=z-i)\right)$$

$$= \sum_{i=0}^z P(X=i, Y=z-i)$$

Mutually
Exclusive
Events

$$= \sum_{i=0}^z P(X=i) P(Y=z-i)$$

X and Y are
independent

$X \sim \text{Poisson}(\lambda_1)$

$Y \sim \text{Poisson}(\lambda_2)$

$$= \sum_{i=0}^z \frac{e^{-\lambda_1} \lambda_1^i}{i!} \frac{e^{-\lambda_2} (\lambda_2)^{z-i}}{(z-i)!}$$

$$= \frac{e^{-(\lambda_1 + \lambda_2)}}{z!} \sum_{i=0}^z \frac{\lambda_1^i (\lambda_2)^{z-i}}{i! (z-i)!} \cdot z!$$

$$= \frac{e^{-(\lambda_1 + \lambda_2)}}{z!} (\lambda_1 + \lambda_2)^z$$

If $X \sim \text{Poisson}(\lambda_1)$ and $Y \sim \text{Poisson}(\lambda_2)$ then

$$Z \sim \text{Poisson}(\lambda_1 + \lambda_2)$$

Inductively : $X_i \sim \text{Poisson}(\lambda_i)$ $i=1, 2, \dots, n$
and independent

$$Z = \sum_{i=1}^n X_i \quad \text{then} \quad Z \sim \text{Poisson}\left(\sum_{i=1}^n \lambda_i\right)$$

(b) Find distribution of $X | Z$

Fix $z \in \text{Range}(Z)$

Step 1: $X | Z=z$ - Find its distribution

$$\mathbb{P}(X=x | Z=z) \equiv ? \quad \left\{ \begin{array}{ll} 0 & x > z \\ ? & x \leq z \end{array} \right.$$

$x \in \text{Range}(X)$

$$\text{For } 0 \leq x \leq z \Rightarrow = \frac{\mathbb{P}(X=x, Z=z)}{\mathbb{P}(Z=z)}$$

$$= \frac{\mathbb{P}(X=x, Y=z-x)}{\mathbb{P}(Z=z)}$$

independence of X and Y \leftarrow

$$= \frac{\mathbb{P}(X=x) \mathbb{P}(Y=z-x)}{\mathbb{P}(Z=z)}$$

$$= \frac{\frac{e^{-\lambda_1} \lambda_1^x}{x!} \frac{e^{-\lambda_2} \lambda_2^{z-x}}{(z-x)!}}{\frac{e^{-(\lambda_1+\lambda_2)} (\lambda_1+\lambda_2)^z}{z!}}$$

$$= \frac{z!}{x! (z-x)!} \frac{\lambda_1^x \lambda_2^{z-x}}{(\lambda_1+\lambda_2)^z}$$

$$P(X=x | Z=z) = \binom{z}{x} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^x \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{z-x}$$

$$x = 0, 1, 2, \dots, z$$

independent-

$$X \sim \text{Poisson}(\lambda_1) \quad Y \sim \text{Poisson}(\lambda_2) \quad Z \sim \text{Poisson}(\lambda_1 + \lambda_2)$$

$$Z = X + Y$$

$$\Rightarrow X | Z=z \sim \text{Binomial}(z, \frac{\lambda_1}{\lambda_1 + \lambda_2})$$

$$X | Z \sim \text{Binomial}(Z, \frac{\lambda_1}{\lambda_1 + \lambda_2})$$

Expectation :- X, Y are random variables

$$E[X+Y] = E[X] + E[Y]$$

Q: $E[XY] = ?$ $E[X]$ and $E[Y]$

Expectation of product

Ex:- Construct X, Y random variables
 $E[XY] \neq E[X]E[Y]$

In general $E[XY] \neq E[X]E[Y]$ but

If X and Y are independent, then $E[XY] = E[X]E[Y]$

$$\begin{aligned} E[XY] &= \sum_{t \in \text{Range}(XY)} t \cdot \mathbb{P}(XY=t) \\ &= \sum_{\substack{u \in \text{Range}(X) \\ v \in \text{Range}(Y)}} uv \mathbb{P}(X=u, Y=v) \end{aligned}$$

independence of X and Y $\leftarrow = \sum_{\substack{u \in \text{Range}(X) \\ v \in \text{Range}(Y)}} P(X=u) P(Y=v)$

$$= \sum_{u \in \text{Range}(X)} P(X=u) \sum_{v \in \text{Range}(Y)} P(Y=v)$$

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

D

X, Y
Independent $\Rightarrow E[XY] - E[X]E[Y] = 0$

"Dependence"

measure

$$E[(X - E[X])(Y - E[Y])] = ?$$

$$\begin{aligned} E[(X - E[X])(Y - E[Y])] &= E[XY] - E[X]E[Y] - E[Y]E[X] \\ &\quad + E[X]E[Y] \\ &= E[XY] - E[X]E[Y] \end{aligned}$$

Ex 1 Construct random variables

(i) $E[XY] - E[X]E[Y] = 0$

(ii) X and Y are NOT independent

Covariance - measure of dependence between X and Y

If X and Y are independent $\Rightarrow \text{Cov}[X, Y] = 0$

The "covariance of X and Y " is defined as

$$\text{Cov}[X, Y] = E[(X - E[X])(Y - E[Y])].$$

$\text{Cov}[X, Y] > 0 \equiv$ $\left\{ \begin{array}{l} X - \text{larger than its average} \\ \text{at the same time} \\ Y - \text{larger than its average} \end{array} \right. \left| \begin{array}{l} \text{smaller than} \\ \text{its average} \end{array} \right.$

Similar interpretation one can do for $\text{Cov}[X, Y] < 0$.

Covariance-Facts

$$\text{Cov}[X, Y] = \text{Cov}[Y, X]; \quad - \text{Symmetric}$$

$$\text{Cov}[X, aY + bZ] = a \cdot \text{Cov}[X, Y] + b \cdot \text{Cov}[X, Z]; \quad - \text{linear}$$

$$\text{Cov}[aX + bY, Z] = a \cdot \text{Cov}[X, Z] + b \cdot \text{Cov}[Y, Z]; \quad \text{and}$$

If X and Y are independent with a finite covariance, then
 $\text{Cov}[X, Y] = 0$.

$$\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y] + 2\text{Cov}[X, Y].$$

$$\equiv E[(X+Y) - E[X+Y]]^2$$

\equiv expand & do algebra

Conditional Variance

Theorem 4.4.9

Let X and $Y : S \rightarrow T$ be two discrete random variables on a sample space S . Let $g(y) = E[X|Y = y]$. Let $h(y) = \text{Var}[X|Y = y]$. Denoting $g(Y)$ by $E[X|Y]$ and denoting $h(Y)$ by $\text{Var}[X|Y]$, then

$$\text{Var}[X] = E[\text{Var}[X|Y]] + \text{Var}[E[X|Y]].$$

$$\text{and: } E[X] = E[E[X|Y]]$$

Conditional Variance

Question: The number of eggs N found in nests of a certain species of turtles has a Poisson distribution with mean λ . Each egg has probability p of being viable and this event is independent from egg to egg. Find the mean and variance of the number of viable eggs per nest.

$N =$ number of eggs

$X =$ number of viable ones.

$N = n$ for some $n \geq 1$

$$X \mid N=n \sim \text{Binomial}(n, p)$$

$$g(n) := E[X \mid N=n] = np$$

\Rightarrow

$$h(n) := \text{Var}[X \mid N=n] = np(1-p)$$

$$\Rightarrow \quad \overset{E[X|N]}{\parallel} g(N) = Np \quad \& \quad \overset{\text{Var}[X|N]}{\parallel} h(N) = Np(1-p)$$

$$\begin{aligned} \cdot E[X] &= E[E[X|N]] = E[pN] = p E[N] \\ &= p\lambda \end{aligned}$$

• Analysis of Variance

$$\text{Var}[X] = E[\text{Var}[X|Y]] + \text{Var}[E[X|Y]].$$

$$\begin{aligned} \cdot \text{Var}[X] &= E[h(N)] + \text{Var}[g(N)] \\ &= E[Np(1-p)] + \text{Var}[Np] \\ &= p(1-p) E[N] + p^2 \text{Var}[N] \end{aligned}$$

$N \sim \text{Poisson}(\lambda)$

$$= p(1-p)\lambda + p^2\lambda = p\lambda$$