

Let X and Y be two r.v. with joint pmf or pdf.

$$f(x, y), x \in S_X, y \in S_Y.$$

$E[X|Y]$ is a r.v. which can be written as a function of Y . That is $\exists g$ such that:

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

The set of all possible values of $E[X|Y]$ is given by $\{g(y), y \in S_Y\}$, where:

$$g(y) = E[X|Y=y] = \begin{cases} \sum_{x \in S_X} x f(x, y) \\ \int_{S_X} x f(x, y) dx. \end{cases}$$

$$f(x|y) = \frac{f(x, y)}{f(y)}, \quad f(y) = \int_{S_X} f(x, y) dx \text{ or } \sum_{x \in S_X} f(x, y).$$

$$E(h(Y)|X) = \varphi(X), \text{ where:}$$

$$\varphi(x) = E[h(Y)|X=x] = \begin{cases} \sum_{y \in S_Y} h(y) f(y|x) \\ \int_{S_Y} h(y) f(y|x) dy. \end{cases}$$

So If we know joint pdf we can find the conditional expectation of X given Y .

Example: Roll a die until we get a 6.

let Y be the number of rolls and X the number of 1's we get.

Find $E[X|Y]$?

$$S_Y = \{1, 2, \dots\} = \mathbb{N}^*. \quad Y \hookrightarrow G\left(\frac{1}{6}\right).$$

$$n \geq 1 \quad P(Y=n) = P(\text{the } (n-1) \text{ rolls were not 6 and the } n^{\text{th}} \text{ roll we get 6}).$$

$$= P((n-1) \text{ rolls without 6}) P(n^{\text{th}} \text{ 6}).$$

$$= \frac{1}{6} \cdot \left(\frac{5}{6}\right)^{n-1} = \underbrace{\frac{5}{6} \cdot \frac{5}{6} \cdot \frac{5}{6} \cdots \frac{5}{6}}_{n-1} \cdot \frac{1}{6}.$$

$E[X|Y]$ is a r.v. with values: $E[X|Y=n]$ $n \geq 1$.

Now we need to compute $E[X|Y=n]$.

We have

$$E[X|Y=n] = \sum_{i=0}^{n-1} i P(X=i|Y=n).$$

$$P(X=i|Y=n) = \binom{n-1}{i} \left(\frac{1}{6}\right)^i \left(\frac{5}{6}\right)^{n-1-i}.$$

$$E[X|Y=n] = \sum_{i=0}^{n-1} i \binom{n-1}{i} \left(\frac{1}{6}\right)^i \left(\frac{5}{6}\right)^{n-1-i}.$$

$$= \sum_{i=1}^{n-1} i \frac{(n-1)!}{(n-1-i)! i!} \frac{1}{6^i} \left(\frac{5}{6}\right)^{n-1-i}.$$

$$\begin{aligned}
 E[X|Y=n] &= \sum_{k=0}^{n-1} k \binom{n-1}{k} \frac{1}{5^k} \left(\frac{4}{5}\right)^{n-1-k} \\
 &= \sum_{k=0}^{n-1} \frac{n(n-1)!}{n k! (n-1-k)!} \frac{1}{5^k} \left(\frac{4}{5}\right)^{n-1-k} \\
 &= \frac{1}{n} \sum_{k=0}^{n-1} \frac{n!}{(n-(k+1))! k!} \frac{1}{5^k} \left(\frac{4}{5}\right)^{n-1-k} \\
 \text{(\textcircled{n=5})} \quad E[X|Y=5] &= (n-1) \cdot \frac{1}{5} = \frac{\textcircled{n-1}}{5}
 \end{aligned}$$

$$E[X|Y] = \frac{Y-1}{5} = g(Y), \quad g(n) = \frac{n-1}{5}.$$

Properties of conditional expectation:

- $E[\alpha_1 X_1 + \alpha_2 X_2 | Y] = \alpha_1 E[X_1 | Y] + \alpha_2 E[X_2 | Y]$
- $E[X | Y] = E[X]$ when $X \perp Y$.
- $E[E[X | Y]] = E[X]$.
- $E[E[Y | X]] = E[Y]$.
- $\text{Var}(X | Y=y) = \text{Var}(X)$ if $X \perp Y$.
- $\text{Var}(X | Y) = E[X^2 | Y] - (E[X | Y])^2$
 $\left| \begin{array}{l} \text{If } X \perp Y \\ \text{then } X^2 \perp Y \end{array} \right.$

$$\begin{aligned}
 &= E[X^2] - (E[X])^2 \\
 &= \text{Var}(X).
 \end{aligned}$$

Let X_1, X_2, \dots, X_n be a family of r.v.

$$E\left[\sum_{k=1}^n X_k\right] = \sum_{k=1}^n E[X_k]$$

$$\text{Var}\left(\sum_{k=1}^n X_k\right) = \text{Cov}\left(\sum_{k=1}^n X_k, \sum_{k=1}^n X_k\right) \text{ because.}$$

$$\boxed{\text{Var}(X) = \text{Cov}(X, X)}$$

$$\begin{aligned} \text{Then } \text{Var}\left(\sum_{k=1}^n X_k\right) &= \text{Cov}\left(\sum_{k=1}^n X_k, \sum_{j=1}^n X_j\right) \\ &= \sum_{k=1}^n \sum_{j=1}^n \text{Cov}(X_k, X_j) \end{aligned}$$

$$\begin{aligned} \text{Because } \text{Cov}(X, Y+Z) &= E[X(Y+Z)] - E[X]E[Y+Z] \\ &= E[XY] + E[XZ] - E[X]E[Y] - E[X]E[Z] \end{aligned}$$

$$\boxed{\text{Cov}(X, Y+Z) = \text{Cov}(X, Y) + \text{Cov}(X, Z)}$$

Similarly we have.

$$\text{Cov}(X+Y, Z) = \text{Cov}(X, Z) + \text{Cov}(Y, Z)$$

$$\text{Therefore: } \text{Cov}\left(\underbrace{\sum_{k=1}^n X_k}_X, \underbrace{\sum_{k=1}^n X_k}_{X_1+Z}\right) = \text{Cov}\left(\underbrace{\sum_{k=1}^n X_k}_X, \underbrace{X_1}_Y + \underbrace{\sum_{k=2}^n X_k}_Z\right)$$

$$= \text{Cov}\left(\sum_{k=1}^n X_k, X_1\right) + \text{Cov}\left(\sum_{k=1}^n X_k, \sum_{k=2}^n X_k\right)$$

$$= \sum_{j=1}^n \text{Cov}\left(\sum_{k=1}^n X_k, X_j\right) = \sum_{j=1}^n \sum_{k=1}^n \text{Cov}(X_k, X_j)$$

If $(X_k)_{k \geq 1}$ are independent then

$$\text{Cov}\left(\sum_{k=1}^n X_k, \sum_{j=1}^n X_j\right) = \sum_{k=1}^n \text{Cov}(X_k, X_k).$$

because if $X_k \perp X_j$ $\text{Cov}(X_k, X_j) = 0 \forall k \neq j$.

$$\text{That is } \text{Var}\left(\sum_{k=1}^n X_k\right) = \sum_{k=1}^n \text{Var}(X_k) \cdot (X_k \perp X_j)_{k \neq j}$$

Let X and Y be two random variables.

$$\begin{aligned} \text{Var}(X+Y) &= \text{Cov}(X+Y, X+Y) \\ &= \text{Cov}(X, X) + \text{Cov}(X, Y) + \text{Cov}(Y, X) + \text{Cov}(Y, Y) \\ &= \text{Var}(X) + 2\text{Cov}(X, Y) + \text{Var}(Y). \end{aligned}$$

Remark If $\text{Cov}(X, Y) = 0 \Rightarrow$ does not imply
in general that $X \perp Y$.

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

But $\text{Cov}(X, Y) = 0 \not\Rightarrow X \perp Y$ (in general).

Let X_1, X_2, \dots , be a sequence of i.i.d.
random variables. Then $E[X_1] = E[X_2] = \dots = E[X_n] = \mu$
and $\text{Var}(X_1) = \text{Var}(X_2) = \dots = \text{Var}(X_n) = \sigma^2 \forall n \geq 1$

Consider N a Poisson r.v. with parameter $\lambda > 0$

such that N is $\perp (X_1, X_2, X_3, \dots, X_n, \dots)$. WB-263.

Set $X = \sum_{k=1}^N X_k = X_1 + X_2 + \dots + X_N$. (fog)

This r.v. X is used as a model loss payment of an insurance company. The insurance company wants to know the expected loss for some period and also its standard deviation:

$E[X]$? and $\sqrt{\text{Var}(X)} = \text{sd}(X) = \sigma_X$?

The distribution of X is not known.

First $E[X|N]$? and $\text{Var}(X|N)$?

$E[X|N] = h(N)$ where $h(n) = E[X|N=n]$.

Then $h(n) = E\left[\sum_{k=1}^N X_k | N=n\right] = E\left[\sum_{k=1}^n X_k | N=n\right]$.

$= \sum_{k=1}^n E[X_k | N=n] = \sum_{k=1}^n E[X_k]$

$= \sum_{k=1}^n \mu = \mu n$

Therefore $E[X|N] = h(N) = \mu N$.

$E[X] = E[E[X|N]] = E[\mu N] = \mu E[N]$
 $= \mu d$.

Wednesday. 2 or 3 exercises.