

Conditional Probability

Let the probability set function $P(C)$ be defined on the sample space \mathcal{C} and C_1 be a subset of \mathcal{C} such that $P(C_1) > 0$.

We agree to consider only those outcomes of Random Experiment that are elements of C_1 (basically we are going to take elements of C_1 to be the sample space.)

This probability is denoted by $P(C_2|C_1)$ where $C_2 \subseteq \mathcal{C}$

$$P(C_2|C_1) = \frac{P(C_1 \cap C_2)}{P(C_1)}$$

$$\left. \begin{aligned} P(C_1|C_1) &= 1 \\ P(C_2 \cap C_1|C_1) &= P(C_2|C_1) \end{aligned} \right\}$$

$$\frac{P(C_2 \cap C_1|C_1)}{P(C_1|C_1)} = \frac{P(C_2 \cap C_1)}{P(C_1)}$$

Properties :-

$$P(C_2|C_1) \geq 0$$

$$P(C_2 \cup C_3 \dots \text{if } C_1) = P(C_2|C_1) + P(C_3|C_1)$$

if $C_2, C_3 \dots$ are disjoint

$$P(C_1|C_1) = 1$$

Marginal Distribution

Suppose $f(x_1, x_2)$ be pdf of 2 RV x_1 & x_2

Consider an event $a < x_1 < b$ & $-\infty < x_2 < \infty$

and $a < x_1 < b$ these two events are identical

$$\begin{aligned} \Pr(a < x_1 < b) &= \Pr(a < x_1 < b, -\infty < x_2 < \infty) \\ &= \int_{a=-\infty}^b \int_{-\infty}^{\infty} f(x_1, x_2) dx_2 dx_1 \end{aligned}$$

$$f_1(x_1) = \int_{-\infty}^{\infty} f(x_1, x_2) dx_2$$

↑
marginal probability of pdf x_1

Let x_1 & x_2 be discrete RV
Now define $A_1 = \{(x_1, x_2) : x_1 = x_1'\}$ $A_2 = \{(x_1, x_2) : x_2 = x_2'\}$

$$P(A_2|A_1) = \frac{P(A_1 \cap A_2)}{P(A_1)} = \frac{\Pr(x_1 = x_1', x_2 = x_2')}{P(x_1 = x_1')} = \frac{f(x_1, x_2)}{f(x_1)}$$

same can be done for cts case

$$f(x_2|x_1) = \frac{f(x_1, x_2)}{f(x_1)}$$

The correlation coefficient

Let x, y be two RV with joint distribution $f(x, y)$

$$\sigma_{12} = E[(x - \mu_1)(y - \mu_2)]$$

where $\mu_1 = \text{mean of RV } x = E[x]$

$\mu_2 = \text{mean of RV } y = E[y]$

From Linearity of $E =$

$$\sigma_{12} = E[xy] - \mu_1 \mu_2$$

~~covariance~~ \rightarrow covariance of x & y

$$\text{Correlation} = \frac{\sigma_{12}}{\sqrt{\sigma_{11} \sigma_{22}}}$$

$$-1 < \rho < 1$$

\Rightarrow Let x and y be two RV

$$\text{Var}(ax + y) \geq 0$$

$$\Rightarrow a^2 \text{Var}(x) + 2a \text{Cov}(x, y) + \text{Var}(y) \geq 0 \quad \forall a$$

\therefore discriminant ≤ 0

$$\Rightarrow 4 \text{Cov}(x, y)^2 - 4 \text{Var}(x) \text{Var}(y) \leq 0$$

$$\Rightarrow -1 \leq \rho \leq 1$$

Change of variable

Some time we are given a ~~pdf~~ RV X and we need to know the distribution of $f(x)$.

Notice that as X is an RV $f(x)$ will also be an RV.

There are various methods to find Pdf or Pmf of $f(x)$.

For Continuous distributions

- 1) Using CDF :- We know that CDF of a distribution is unique and we can find the ~~pdf~~ pdf just by differentiating the CDF.

How to approach with ~~pdf~~ CDF.

Suppose X be any RV with pdf $f(x)$

Let $g(x)$ You have to find distribution of $g(x)$

$$\therefore \Pr(g(x) \leq t) = F_{g(x)}(t)$$

||
This represents the CDF of $g(x)$

Now we can ~~exp~~ solve $\Pr(g(x) \leq t)$ in terms of x and get the answer.

Let's take an example.

$$X \sim N(0,1)$$

we have to find distribution of x^2

$$\therefore F_{x^2}(t) = \Pr(x^2 \leq t)$$

$$= \Pr(-\sqrt{t} \leq x \leq \sqrt{t})$$

$$F_{x^2}(t) = \int_{-\sqrt{t}}^{\sqrt{t}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

Now differentiate to get the pdf.

$$\frac{d}{dt} F_{x^2}(t) = \frac{1}{\sqrt{2\pi}} e^{-\frac{t}{2}} \frac{1}{2\sqrt{t}} + \frac{1}{2\sqrt{t}} \frac{e^{-\frac{t}{2}}}{\sqrt{2\pi}} = \frac{e^{-\frac{t}{2}}}{\sqrt{2\pi t}}$$

Interesting fact this is chi square distributed with degree of freedom 1.

This above method is universal and work in every case. Remember it's not the easiest to apply But when applied correctly always give the correct result.

Example suppose x & y are both iid exponential RV with parameter λ .
Find distribution of $x+y$.

$$F_{x+y}(t) = \Pr(x+y \leq t)$$

$$\int_0^t \int_0^{t-y} \lambda e^{-\lambda x} \lambda e^{-\lambda y} dx dy$$

$$f(x, y) = \lambda_2 \lambda_1 e^{-\lambda_1 x - \lambda_2 y}$$

$$Pr(X+Y \leq t) = \int_0^t \int_0^{t-y} \lambda_2 \lambda_1 e^{-\lambda_1 x} e^{-\lambda_2 y} dx dy$$

$$\int_0^{t-y} e^{-\lambda_1 x} dx = \left. \frac{e^{-\lambda_1 x}}{-\lambda_1} \right|_0^{t-y} = \frac{1 - e^{-\lambda_1(t-y)}}{\lambda_1}$$

$$\lambda_2 \int_0^t \frac{1 - e^{-\lambda_1(t-y)}}{\lambda_1} e^{-\lambda_2 y} dy$$

$$\lambda_2 \int_0^t e^{-\lambda_2 y} - e^{-\lambda_1 t + \lambda_1 y - \lambda_2 y} dy$$

$$\lambda_2 \left[\left. \frac{e^{-\lambda_2 y}}{-\lambda_2} \right|_0^t + \left. \frac{e^{-\lambda_1 t + (\lambda_1 - \lambda_2)y}}{\lambda_1 - \lambda_2} \right|_0^t \right]$$

$$1 - e^{-\lambda_2 t} + \frac{\lambda_2}{\lambda_1 - \lambda_2} e^{-\lambda_1 t} - e^{-\lambda_1 t + \lambda_1 t - \lambda_2 t}$$

$$F_{X+Y}(t) = 1 - e^{-\lambda_2 t} + \frac{\lambda_2}{\lambda_1 - \lambda_2} [e^{-\lambda_1 t} - e^{-\lambda_2 t}]$$

$$\frac{d}{dt} F_{X+Y}(t) = \lambda_2 e^{-\lambda_2 t} + \frac{\lambda_2}{\lambda_1 - \lambda_2} [-\lambda_1 e^{-\lambda_1 t} + \lambda_2 e^{-\lambda_2 t}]$$

But if $\lambda_1 = \lambda_2$ then

$$F_{X+Y}(t) = 1 - e^{-\lambda t} - \lambda t e^{-\lambda t}$$

$$f_{X+Y}(t) = \lambda e^{-\lambda t} - \lambda [-\lambda t e^{-\lambda t} + e^{-\lambda t}] = \lambda^2 t e^{-\lambda t}$$

This distribution is Erlang (2, λ)



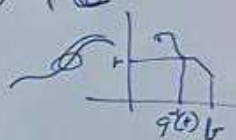
2) Find pdf with help of change of formula technique.

Let X be a RV.

Suppose $g(x)$ be a function st $g'(t)$ is non zero in an interval (a, b) .

$\therefore g(x)$ can be increasing or decreasing.

Then $F_{g(X)}(t) = \Pr(g(X) \leq t) = \sum_{\substack{\text{all such } (a,b) \\ \text{st } g'(t) \text{ is non} \\ \text{zero in } (a,b)}} \Pr(X \in (a,b) \mid g(X) \leq t) + C$



Increasing case

$$\sum_{(a,b)} \Pr(a \leq X \leq g^{-1}(t))$$

$$\sum_{(a,b)} F_X(g^{-1}(t)) - F_X(a)$$

Decreasing case

$$\sum_{(a,b)} \Pr(b \geq X \geq g^{-1}(t))$$

$$\sum_{(a,b)} F_X(b) - F_X(g^{-1}(t))$$

$$f_{g(X)}(t) = \sum_{(a,b)} f_X(g^{-1}(t)) (g^{-1}(t))' \quad \text{or} \quad \sum_{(a,b)} -f_X(g^{-1}(t)) (g^{-1}(t))'$$

$$= \sum_{(a,b)} f_X(g^{-1}(t)) |(g^{-1}(t))'|$$

$$\sum_{(a,b)} f_X(g^{-1}(t)) |(g^{-1}(t))'|$$

$$g(g^{-1}(t)) = t$$

$$g'(g^{-1}(t)) = 1$$

Suppose $x \sim N(0, 1)$ and $y = x^2$ find pdf of y

$$\therefore f_y(t) = \frac{f_x(x)}{|g'(x)|}$$

$$g(x) = x^2$$

$$\therefore g^{-1}(t) = -\sqrt{t} \quad \text{if } x \in (-\infty, 0) \quad g^{-1}(t) = \frac{1}{2\sqrt{t}}$$

$$g^{-1}(t) = \sqrt{t} \quad \text{if } x \in (0, \infty) \quad g^{-1}(t) = \frac{1}{2\sqrt{t}}$$

$$\therefore f_y(t) = f_x(-\sqrt{t}) \frac{1}{2\sqrt{t}} + f_x(\sqrt{t}) \frac{1}{2\sqrt{t}}$$

$$= \frac{1}{\sqrt{t}} f_x(\sqrt{t}) = \frac{1}{\sqrt{2\pi t}} e^{-t/2}$$

In case of ~~two~~ ~~bin~~ multivariate distributions:-

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} A \\ B \end{pmatrix}$$

$$\frac{f(A, B)}{A, B} = f(a, b) = f_{x, y}(T^{-1}(a, b)) || J_{T^{-1}}(a, b) ||$$

Suppose x and y are such that $f(x, y) = 1$ $0 \leq x \leq 1$ and $0 \leq y \leq 1$

Find $x+y, x-y$

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x+y \\ x-y \end{pmatrix} = \begin{pmatrix} A \\ B \end{pmatrix}$$

$$f_{A, B}(a, b) = f_{x, y} \left(\frac{A+B}{2}, \frac{A-B}{2} \right) \frac{1}{2}$$

$$T^{-1} \begin{pmatrix} A \\ B \end{pmatrix} = \begin{pmatrix} \frac{A+B}{2} \\ \frac{A-B}{2} \end{pmatrix}$$

$$f_{A, B}(a, b) = \frac{1}{2} \quad \begin{matrix} 0 \leq \frac{A+B}{2} \leq 1 \\ 0 \leq \frac{A-B}{2} \leq 1 \end{matrix}$$

$$J_{T^{-1}} = \begin{vmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{vmatrix} = \frac{1}{2}$$