

Ex.

The time T until a new light bulb burns out is an exponential r.v. with parameter λ . Alice turns on the light at time $t=0$. She returns t_0 hours later, finding light still on. i.e. the event

$$A = \{T > t_0\}$$

has occurred. note

$$f_T(t) = \begin{cases} \lambda e^{-\lambda t} & t \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

Let X be the additional time to failure, i.e. $X = T - t_0$ or $T = X + t_0$. Find $f_{X|A}^{(x)}$.

Solution

first find cond. CDF

$$\begin{aligned} F_{X|A}^{(x)} &= P(X \leq x | A) \\ &= \int_{-\infty}^x f_{X|A}^{(s)} ds \end{aligned}$$

then differentiate:

$$\frac{d}{dx} F_{X|A}^{(x)} = f_{X|A}^{(x)}$$

Recall (11-29-24, p.14)

$$P(T > a) = \begin{cases} e^{-\lambda a} & \text{if } a > 0 \\ 1 & \text{if } a \leq 0 \end{cases}$$

Thus, for $x \geq 0$:

$$P(X > x | A) = P(T > t_0 + x | T > t_0)$$

$$= \frac{P(T > t_0 + x, T > t_0)}{P(T > t_0)}$$

$$= \frac{P(T > t_0 + x)}{P(T > t_0)}$$

$$= \frac{e^{-\lambda(t_0 + x)}}{e^{-\lambda t_0}} = \frac{\cancel{e^{-\lambda t_0}} \cdot e^{-\lambda x}}{\cancel{e^{-\lambda t_0}}} = e^{-\lambda x}$$

Thus

$$F_{X|A}(x) = P(X \leq x | A)$$

$$= 1 - P(X > x | A)$$

$$= \begin{cases} 1 - e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

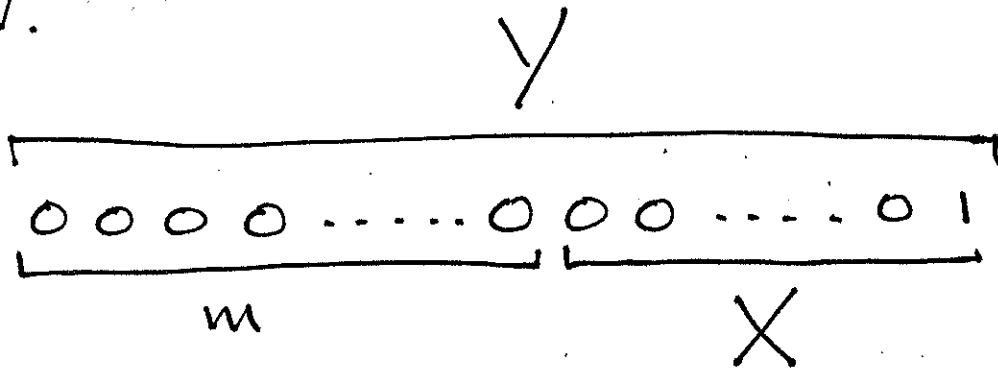
Hence

$$f_{X|A}(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

so cond. PDF is again exponential with same Param. i.e. exponential r.v.s have no memory.

Exercise

Do same thing for the geometric
r.v.



let

$Y = \# \text{ flips til } 1^{\text{st}} \text{ head}$

$X = \# \text{ flips after } m^{\text{th}} \text{ to } 1^{\text{st}} \text{ head}$

$$A = \{ Y > m \}.$$

show : $P_{X|A}(k) = P_Y(k) \quad (k=1, 2, 3, \dots)$

Recall : $P_Y(k) = P(1-P)^{k-1} \quad (k=1, 2, 3, \dots)$

If events $A_1, A_2, \dots, A_n$ form a Partition of Ω , then the Total Prob. Thm. says

$$P(X \leq x) = \sum_{i=1}^n P(A_i) \cdot P(X \leq x | A_i)$$

Thus

$$F_X(x) = \sum_{i=1}^n P(A_i) \cdot F_{X|A_i}(x)$$

diff. w.r.t. x to get

$$f_X(x) = \sum_{i=1}^n P(A_i) \cdot f_{X|A_i}(x)$$

Defn

jointly

If X, Y are $\wedge$ continuous r.v.s
and $f_Y(y) > 0$ for all $y \in \mathbb{R}$,

the conditional PDF of X given Y is

$$f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$$

note:

$$\int_{-\infty}^{\infty} f_{X|Y}(x|y) dx = \int_{-\infty}^{\infty} \frac{f_{X,Y}(x,y)}{f_Y(y)} dx$$

$$= \frac{\int_{-\infty}^{\infty} f_{X,Y}(x,y) dx}{f_Y(y)} = \frac{f_Y(y)}{f_Y(y)} = 1 \quad \checkmark$$

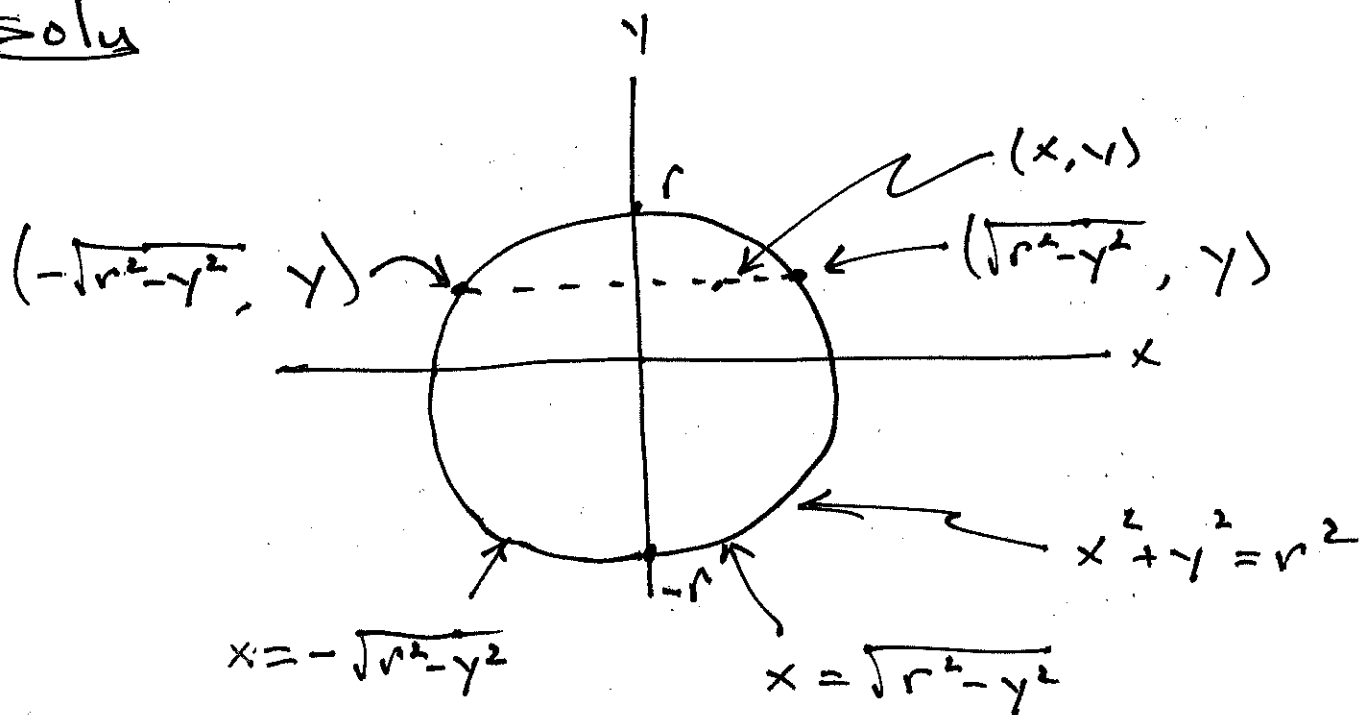
Ex.

Throw a dart at circular target of radius r , centered at $(0,0)$.

Assume target is always hit, and otherwise the landing point (x,y) is uniformly distributed. Let

X, Y be the coordinates hit.

Find $f_{X,Y}(x,y)$, $f_Y(y)$ and $f_{X|Y}(x|y)$.

Solu

we're given

$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{\pi r^2} & \text{if } x^2 + y^2 \leq r^2 \\ 0 & \text{otherwise} \end{cases}$$

Thus

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx$$

$$= \int_{-\sqrt{r^2 - y^2}}^{\sqrt{r^2 - y^2}} \frac{1}{\pi r^2} dx$$

$$= \frac{1}{\pi r^2} \cdot 2\sqrt{r^2 - y^2}$$

$$= \begin{cases} \frac{2\sqrt{r^2 - y^2}}{\pi r^2} & \text{if } -r \leq y \leq r \\ 0 & \text{otherwise} \end{cases}$$

Finally

$$f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$$

$$= \begin{cases} \frac{1}{2\sqrt{r^2 - y^2}} & \text{if } x^2 + y^2 \leq r \\ 0 & \text{otherwise} \end{cases}$$



The PDF identities all hold for PDFs. for instance

$$f_{X|Y}(x|y) \cdot f_Y(y) = f_{X,Y}(x,y) = f_{Y|X}(y|x) \cdot f_X(x)$$

also multiplication rule:

$$f_{X,Y,Z}(x,y,z) = f_{X|Y,Z}(x|y,z) \cdot f_{Y|Z}(y|z) \cdot f_Z(z)$$

Conditional Expectation

let X, Y be jointly cont., and
 $A \subseteq \Omega$ with $P(A) > 0$.

Definitions

$$\bullet E[X|A] = \int_{-\infty}^{\infty} x f_{X|A}(x) dx$$

$$\bullet E[X|Y=y] = \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx$$

expected value rules:

$$\bullet E[g(X)|A] = \int_{-\infty}^{\infty} g(x) f_{X|A}(x) dx$$

$$\bullet E[g(X)|Y=y] = \int_{-\infty}^{\infty} g(x) f_{X|Y}(x|y) dx$$

Total expectation thm

- (1) let $A_1, \dots, A_n$ be a Partition of Ω .
and $P(A_i) > 0$ for $i=1, \dots, n$.

$$E[X] = \sum_{i=1}^n P(A_i) \cdot E[X|A_i]$$

- (2) let Y satisfy $f_Y(y) > 0$ for all $y \in \mathbb{R}$
note events $\{Y=y\}$ form a Partition
of Ω , where $y \in \text{range}(Y)$.

$$E[X] = \int_{-\infty}^{\infty} f_Y(y) \cdot E[X|Y=y] dy$$

Proof of (1)

start with total. Prob. thm.

$$f_X(x) = \sum_{i=1}^n P(A_i) f_{X|A_i}(x)$$

so

$$\int_{-\infty}^{\infty} x f_X(x) dx = \int_{-\infty}^{\infty} x \cdot \sum_{i=1}^n P(A_i) f_{X|A_i}(x) dx$$

$$= \sum_{i=1}^n P(A_i) \cdot \int_{-\infty}^{\infty} x f_{X|A_i}(x) dx$$

thus

$$E[X] = \sum_{i=1}^n P(A_i) \cdot E[X|A_i]$$



Proof of (2)

Start with RHS

$$\int_{-\infty}^{\infty} E[X|Y=y] \cdot f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx \right) \cdot f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{X|Y}(x|y) \cdot f_Y(y) dx dy$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \cdot f_{X,Y}(x,y) dy dx$$

$$= \int_{-\infty}^{\infty} x \cdot \left(\int_{-\infty}^{\infty} f_{X,Y}(x,y) dy \right) dx$$

$$= \int_{-\infty}^{\infty} x f_X(x) dx$$

$$= E[X]$$

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