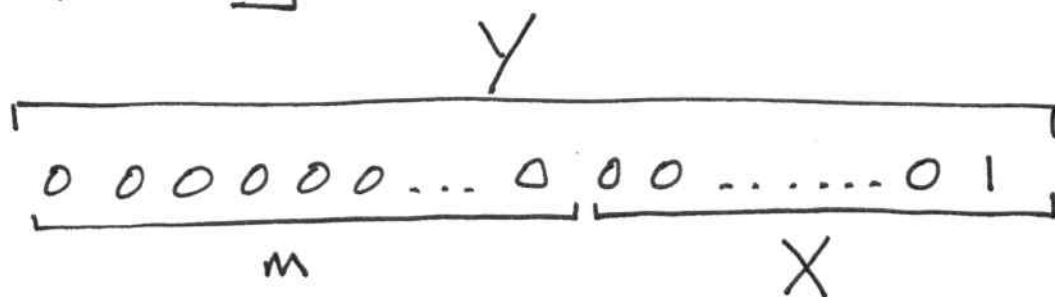


Supplemental Lecture

Ex.Let Y be geometric with parameter p .

so

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

let

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

we must show: $P_Y(k) = P_{X|A}(k) \quad (k=1, 2, 3, \dots)$ We know: $P_Y(k) = p(1-p)^{k-1} \quad (k=1, 2, 3, \dots)$

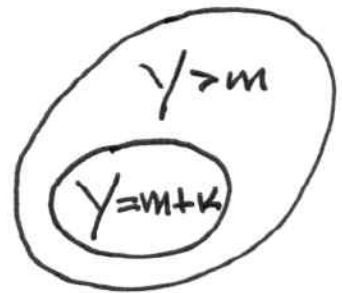
Also (for $k \geq 1$)

$$P_{X|A}^{(k)} = P(X=k|A)$$

$$= P(Y-m=k|Y>m)$$

$$= \frac{P(Y=m+k, Y>m)}{P(Y>m)}$$

$$= \frac{P(Y=m+k)}{P(Y>m)}$$



$$= \frac{p \cdot (1-p)^{m+k-1}}{(1-p)^m}$$

$$= \frac{p \cdot \cancel{(1-p)^m} \cdot (1-p)^{k-1}}{\cancel{(1-p)^m}}$$

$$= p \cdot (1-p)^{k-1} = P_Y(k) \quad (k=1, 2, 3, \dots)$$

~~12~~

Let $A_1, A_2, \dots, A_n \subseteq \Omega$ form a partition of Ω !



- $A_i \cap A_j = \emptyset$ for $i \neq j$
- $A_1 \cup A_2 \cup \dots \cup A_n = \Omega$,

then the total Probability theorem gives

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

Thus

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

diff. w.r.t. x to get

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

As in the discrete case we can condition X on another r.v. Y

Defn

If X, Y are jointly continuous r.v.s, and $f_Y(y) > 0$ for all $y \in \mathbb{R}$, the conditional PDF of X given Y is

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

observe

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

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

$$= \frac{f_Y(y)}{f_Y(y)} = 1. \quad \checkmark$$

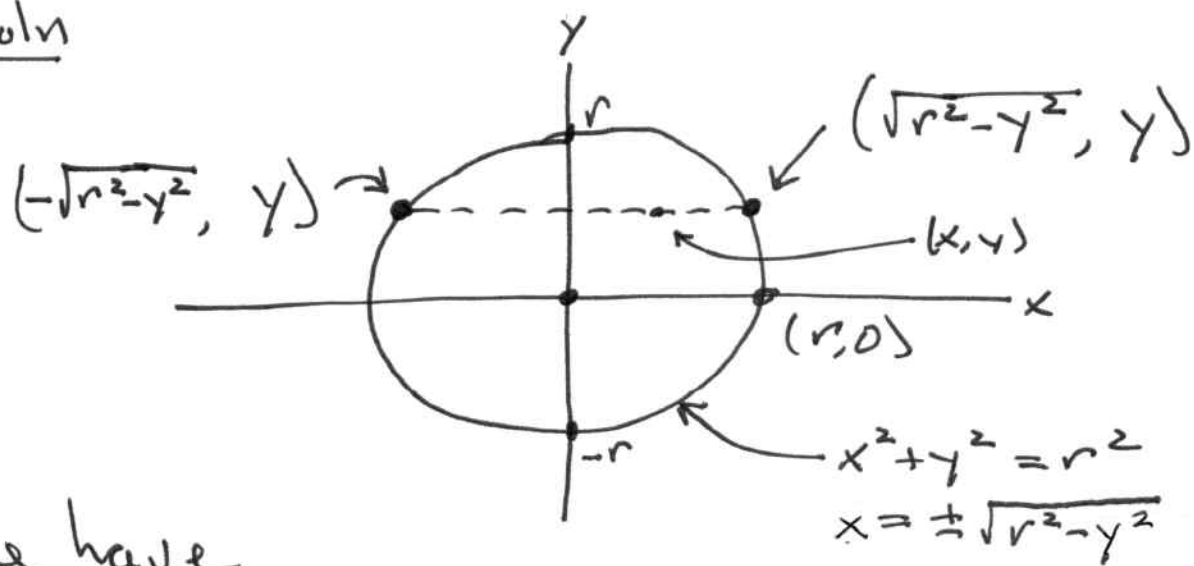
so $f_{X|Y}(x|y)$ is a valid PDF,

for any $y \in \mathbb{R}$ for which $f_Y(y) > 0$.

Ex throw a dart at a circular target, radius r , centered at $(0,0)$. Assume target is hit, but otherwise the impact point (X,Y) is uniformly distributed.

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

Soln



we have

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

so

□

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

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

$$= \frac{1}{\pi r^2} \times \left| \sqrt{r^2-y^2} \right|_{-\sqrt{r^2-y^2}}$$

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

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

Finally,

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

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



Exercise: find $f_X(x)$ and $f_{Y|X}(y|x)$.

All the same identities that hold for PMFs also hold for PDFs, such as

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

and multiplication rule:

19

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

Conditional Expectation

Let X, Y be jointly cont. r.v.s and $A \subseteq \Omega$ with $P(A) > 0$.

Defns

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

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

Expected value rule

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

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

Total expectation theorem

- if $A_1, A_2, \dots, A_n$ partition Ω , and $P(A_i) > 0$ for all i , then

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

- similarly

$$(2) E[X] = \int_{-\infty}^{\infty} f_Y(y) \cdot E[X|Y=y] dy$$

Proof of (1)

start with total Prob. thm.

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

multiply by x and integrate w.r.t. x
from $-\infty$ to ∞ :

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

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

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



Proof of (2)

start with R.H.S

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

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

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

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

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

$$= \int_{-\infty}^{\infty} x \cdot f_X(x) dx = E[X] = \text{L.H.S.}$$

