

CSE 107 2-14-26

11

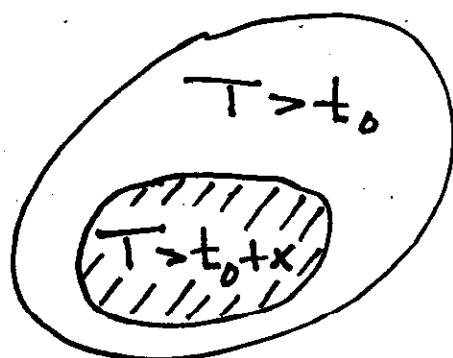
Supplemental Lecture

Recall (3.1) that

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

so for $x \geq 0$

$$\begin{aligned} P(X > x | A) &= P(T > t_0 + x | T > t_0) \\ &= \frac{P(T > t_0 + x, T > t_0)}{P(T > t_0)} \end{aligned}$$



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

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

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

$$= e^{-\lambda x}$$

so

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

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

$$= 1 - e^{-\lambda x}$$

$$F_{X|A}(x) = \begin{cases} 1 - e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

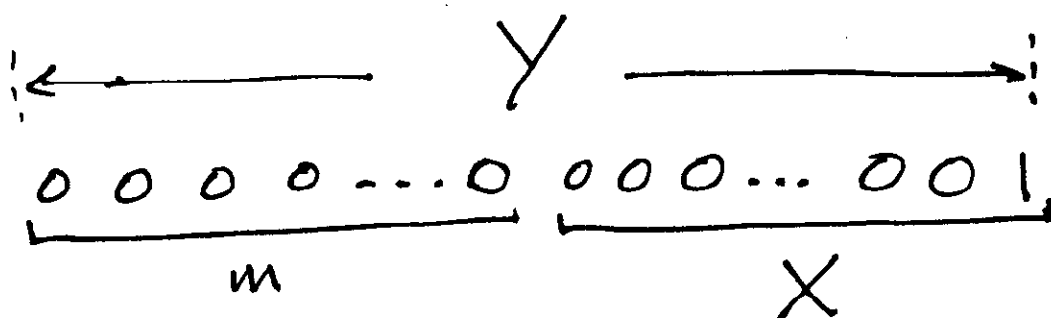
now differentiate w.r.t. x

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

Thus additional time to burn-out is also exponential, and Exponential r.v. has no memory.

Exercise

Do same thing for Geometric(p).



Let $Y = \# \text{ trials to } 1^{\text{st}} \text{ success}$

$X = \# \text{ trials after } m^{\text{th}}, \text{ to } 1^{\text{st}} \text{ success}$

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

show that

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

If $A_1, A_2, \dots, A_n \subseteq \Omega$ form a partition of Ω , Total-Prob. then gives

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

Thus

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

differentiate w.r.t x !

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

As in discrete case, we can condition on another r.v.

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

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

note

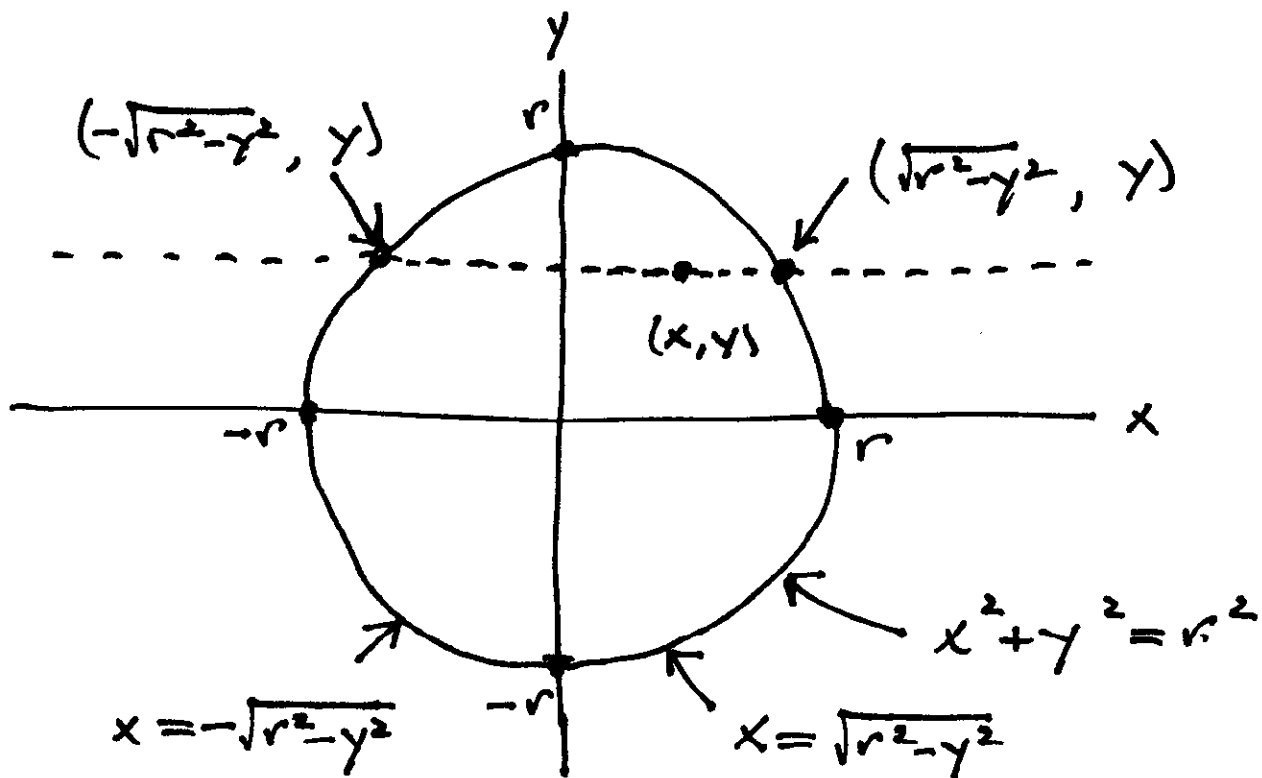
$$\int_{-\infty}^{\infty} f_{X|Y}(x|y) dx = \int_{-\infty}^{\infty} \frac{f_{X,Y}(x,y)}{f_Y(y)} dx$$

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

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

Ex throw a dart at a circular target centered at $(0,0)$. Assume the impact Point (X,Y) to be jointly uniform over target.

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



We are 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

$$\begin{aligned} f_Y(y) &= \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx \\ &= \int_{-\sqrt{r^2 - y^2}}^{\sqrt{r^2 - y^2}} \frac{1}{\pi r^2} dx \end{aligned}$$

$$= \frac{1}{\pi r^2} \cdot x \Big|_{-\sqrt{r^2-y^2}}^{\sqrt{r^2-y^2}}$$

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

i.e.

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

note: though $f_{X,Y}(x,y)$ is uniform,
the marginal $f_Y(y)$ is not

Exercise: find marginal $f_X(x)$.

Finally

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

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

Exercise

find $f_{Y|X}(y|x)$

note: Same identities hold for PDFs as for PMFs

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

conditional Expectation

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

Defn

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

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

Expected value rules

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

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

Total Expectation Theorem

- if $A_1, A_2, \dots, A_n$ form a Partition of Ω , $P(A_i) > 0$ for all i ,

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

- similarly

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

Proof of (1)

Total Prob. thm for P.D.F.s :

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

mult. by x and integrate

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

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

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

Proof of (2)

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

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

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \underbrace{f_{X|Y}(x|y) f_Y(y)}_{\text{joint density}} dx dy$$

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

$$= E[X] = LHS$$



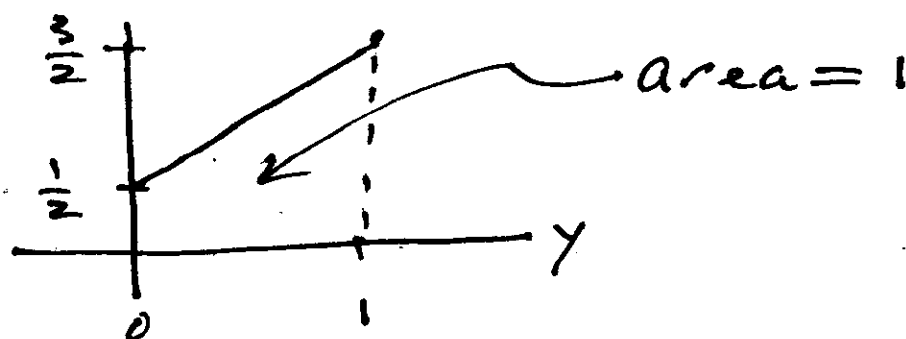
note: Partition in (2) is

$$A_y = \{Y=y\} \text{ for } y \in \text{range}(Y).$$

Ex.

Let X, Y be jointly cont. with

$$f_Y(y) = \begin{cases} y + \frac{1}{2} & \text{if } 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



and

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

Find $E[X]$.

Solution

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

$$= \int_0^1 x \left(\frac{x+y}{y+\frac{1}{2}} \right) dx = \frac{1}{y+\frac{1}{2}} \int_0^1 (x^2 + xy) dx$$

$$= \frac{1}{y+\frac{1}{2}} \left(\frac{1}{3} x^3 + \frac{1}{2} x^2 y \right) \Big|_0^1$$

$$= \frac{\frac{1}{2}y + \frac{1}{3}}{y + \frac{1}{2}} \quad (0 \leq y \leq 1)$$

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

$$= \int_0^1 \left(\frac{\frac{1}{2}y + \frac{1}{3}}{\cancel{y + \frac{1}{2}}} \right) (\cancel{y + \frac{1}{2}}) dy$$

$$= \frac{1}{2} \cdot \frac{1}{2} y^2 + \frac{1}{3} y \Big|_0^1$$

$$= \frac{1}{4} + \frac{1}{3} = \boxed{\frac{7}{12}}$$

check using!

$$E[X] = \int_{-\infty}^{\infty} x \underbrace{f_X(x)}_{\text{first find this}} dx$$

first find this

note 2nd version of Cond.
expectation thm.

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

can be written as

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

↑
Law of iterated Expectation