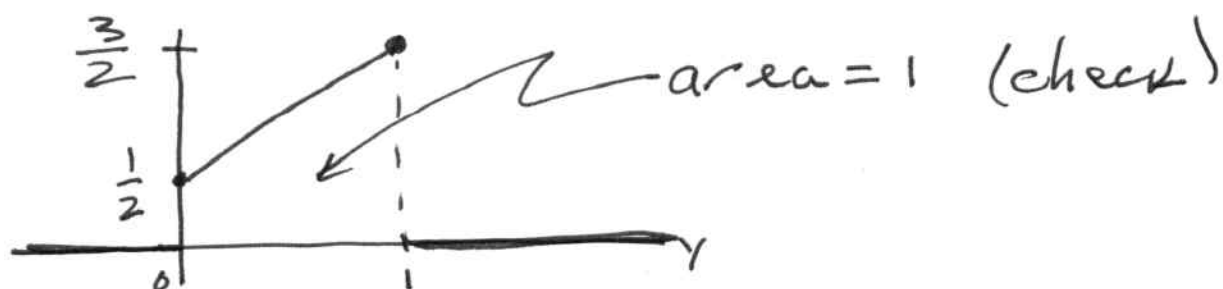


Supplemental Lecture

Ex.

Let X, Y be jointly continuous, with

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



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 $E[X]$.

Soln.

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

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

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

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

and

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

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

$$= \frac{1}{2} \frac{1}{2} y^2 + \frac{1}{3} y \Big|_0^1 = \frac{1}{4} + \frac{1}{3} = \boxed{\frac{7}{12}}$$

another way:

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

$$= \begin{cases} x+y & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

so

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

$$= \int_0^1 (x+y) dy = xy + \frac{1}{2}y^2 \Big|_0^1$$

$$= \begin{cases} x + \frac{1}{2} & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

so

$$\begin{aligned}
 E[X] &= \int_{-\infty}^{\infty} x f_X(x) dx \\
 &= \int_0^1 x \left(x + \frac{1}{2}\right) dx = \int_0^1 \left(x^2 + \frac{1}{2}x\right) dx \\
 &= \left. \frac{1}{3}x^3 + \frac{1}{4}x^2 \right|_0^1 \\
 &= \frac{1}{3} + \frac{1}{4} = \boxed{\frac{7}{12}}
 \end{aligned}$$

Back up a little, note $E[X]$ is a number, but

$$E[X|Y=y] = \begin{cases} \frac{\frac{1}{2}y + \frac{1}{3}}{y + \frac{1}{2}} & 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

is a function of y .

we can compose this fn. with a r.v. to get another r.v.

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

compose with Y to get

$$g(Y) = E[X | Y] = \frac{\frac{1}{2}Y + \frac{1}{3}}{Y + \frac{1}{2}}$$

we did this already:

$$E[E[X | Y]] = \frac{7}{12}$$

In general

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

↑ this is Part (2) of total expectation theorem, called law of iterated expectations.

Independence

Let X, Y be jointly continuous r.v.s

Defn

We say X and Y are independent iff

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

for all $x \in \text{range}(X)$, $y \in \text{range}(Y)$.

or

$$f_{X|Y}(x|y) = f_X(x) \text{ for } y \text{ s.t. } f_Y(y) > 0$$

or

$$f_{Y|X}(y|x) = f_Y(y) \text{ for } x \text{ s.t. } f_X(x) > 0.$$

Generalize to 3 or more r.v.s

□

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

If X, Y are independent, then any events of the form

$$\{X \in I\} \text{ and } \{Y \in J\}$$

would be independent. In particular if $I = (-\infty, x]$, and $J = (-\infty, y]$ then

$$\begin{aligned} F_{X,Y}(x,y) &= \int_{-\infty}^y \int_{-\infty}^x f_{X,Y}(s,t) ds dt \\ &= \int_{-\infty}^y \int_{-\infty}^x f_X(s) \cdot f_Y(t) ds dt \end{aligned}$$

$$= \int_{-\infty}^x f_X(s) ds \cdot \int_{-\infty}^y f_Y(t) dt$$

$$= P(X \leq x) \cdot P(Y \leq y)$$

$$= \bar{F}_X(x) \cdot \bar{F}_Y(y)$$

Note: converse is also true, i.e.

this eqn. $\bar{F}_{X,Y}(x,y) = \bar{F}_X(x) \cdot \bar{F}_Y(y)$

imply independence. (exercise:

Prove this.)

As in discrete case, we have
for independent continuous r.v.s

$$(1) \quad E[X \cdot Y] = E[X] \cdot E[Y] \quad -$$

$$(2) \quad E[g(X) \cdot h(Y)] = E[g(X)] \cdot E[h(Y)] \quad -$$

and

$$(3) \quad \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) \quad .$$

Exercise

- Prove (1)
- Prove X, Y indep. $\Rightarrow g(X), h(Y)$ indep.
adapt Proof in Prob. #44 (ch 2, p. 134)
- Prove (3). mimic Proof from discrete case.

3.6 Bayes Rule

Recall Bayes Rule

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$

If $A_1, A_2, \dots, A_n$ form a Partition of Ω ,
we may use the total Prob. thm

$$P(B) = \sum_{i=1}^n P(B \cap A_i)$$

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

so for any k in $1 \leq k \leq n$:

$$P(A_k|B) = \frac{P(B|A_k) \cdot P(A_k)}{\sum_{i=1}^n P(B|A_i) \cdot P(A_i)}$$

$\Rightarrow$ If X, Y are discrete r.v.,
this becomes

$$P_{X|Y}(x|y) = \frac{P_{Y|X}(y|x) P_X(x)}{\sum_t P_{Y|X}(y|t) \cdot P_X(t)}$$

In the continuous case, we have

$$f_{X|Y}(x|y) = \frac{f_{Y|X}(y|x) \cdot f_X(x)}{\int_{-\infty}^{\infty} f_{Y|X}(y|t) \cdot f_X(t) dt}$$

Ex Let Y be a normal r.v. with s.d. 1 and mean X , another r.v.

Suppose X is continuous uniform on $[1, 3]$.

a.) find $f_{Y|X}$

Soln

We're given

$$f_X(x) = \begin{cases} \frac{1}{2} & \text{if } 1 \leq x \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

and

$$f_{Y|X}(y|x) = \begin{cases} \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{(y-x)^2}{2}} & \text{if } 1 \leq x \leq 3 \\ & y \in \mathbb{R} \\ 0 & \text{otherwise} \end{cases}$$

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

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

$$= \frac{1}{2} \int_1^3 \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{(x-y)^2}{2}} dx$$

u-sub. $u = x - y$ $x=1 \Rightarrow u=1-y$
 $du = dx$ $x=3 \Rightarrow u=3-y$

$$= \frac{1}{2} \int_{1-y}^{3-y} \frac{e^{-u^2/2}}{\sqrt{2\pi}} du$$

$$= \frac{1}{2} \left(\Phi(3-y) - \Phi(1-y) \right)$$

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

Soln

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

$$= \frac{\frac{1}{\sqrt{2\pi}} \cdot e^{-(y-x)^2/2}}{\Phi(3-y) - \Phi(1-y)}$$

Exercises:

c.) Suppose we sample Y and get $Y=3$. What is the that $X \leq 2$, i.e. $P(X \leq 2)$?

d.) find $E[Y]$.