

Conditional Expectation (Chapter 5.7.2)

(a) Given an observed value of x ,
what is $E(Y|x)$?

$$E(Y|x) = \int_{-\infty}^{\infty} y f_y(y|x) dy$$

this answer
depends on
 x

could just as well say

$$E(Y|x) = g(x) \text{ for some } x$$

(b) what is $E(Y|X)$? = $g(X)$

And $g(x)$ is a random variable, since
 x is a random variable

So we can take its expected value too.

$$E(E(Y|X)) = \int_{-\infty}^{\infty} \left[\int_{-\infty}^{\infty} y f_y(y|x) dy \right] f_x(x) dx$$

expectation'
wrt
 x

expectation'
wrt
 y

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

$$= \int_{-\infty}^{\infty} y f_y(y) dy = E(Y)$$

$$\begin{aligned} \text{So } E(Y) &= E(E(Y|X)) && \text{Law of} \\ &= E_x(E_y(Y|X)) && \text{iterated} \\ &&& \text{expectations} \end{aligned}$$

$$\text{Also: } E(h(Y)) = E_x(E_y(h(Y)|X))$$

$$\text{ex: } E(Y^k) = E(E(Y^k|X))$$

This is extremely powerful!

The "traditional" approach to compute $E(Y)$:

$$\left. \begin{array}{l} f_Y(y|x) \\ f_X(x) \end{array} \right\} \xrightarrow{\text{multiply}} f_{XY}(x,y) \xrightarrow{\text{integrate}} f_Y(y) \xrightarrow{\text{integrate}} E(Y)$$

Using iterated expectations to compute $E(Y)$:

$$f_Y(y|x) \xrightarrow{\text{integrate}} E(Y|X)$$

$$f_X(x) \xrightarrow{\text{integrate}} \downarrow E(Y)$$

IF Y conditioned on X is a common RV like, say, exponential, then the top integration becomes trivial.

Example of why iterated expectations are so powerful.

Let X be uniform on $[0, 1]$

Let Y be uniform on $[0, X]$

What is $E(Y)$?

method 1 (the slow way)

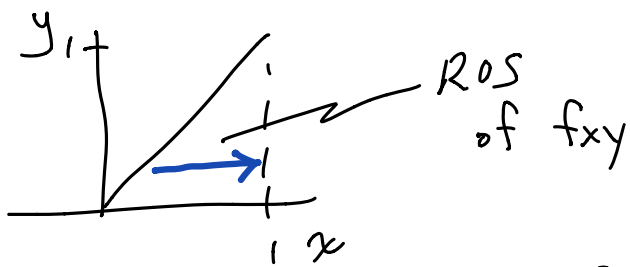
$$f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{else} \end{cases}$$

$$f_Y(y|x) = \begin{cases} 1/x & 0 \leq y < x \\ 0 & \text{else} \end{cases}$$

from the
problem
statement

joint pdf

$$f_{XY}(x, y) = f_Y(y|x)f_X(x) = \begin{cases} 1/x & 0 < y \leq x \leq 1 \\ 0 & \text{else} \end{cases}$$



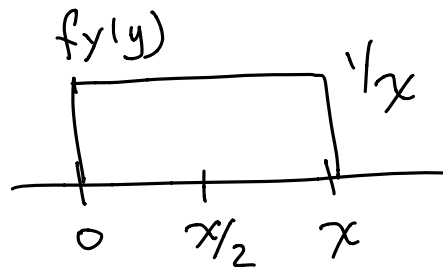
$$\text{marginal } f_Y(y) = \int_{-\infty}^{\infty} f_{XY}(x, y) dx = \int_y^1 \frac{1}{x} dx$$

$$= \ln x \Big|_y^1 = -\ln y \quad \text{for } 0 < y < 1$$

$$\text{expectation } E(Y) = \int_{-\infty}^{\infty} y f_Y(y) dy = - \int_0^1 y \ln y dy = \dots = 1/4 \quad (\text{details omitted})$$

Method 2 using iterated expectation

$$E(Y|X) = \frac{X}{2}$$



because the mean of a uniform RV on $[0, x]$ is $x/2$

$$\begin{aligned} E(E(Y|X)) &= E\left(\frac{X}{2}\right) = \frac{1}{2} E(X) \\ &= \left(\frac{1}{2}\right)\left(\frac{1}{2}\right) = \boxed{\frac{1}{4}} \end{aligned}$$

because the mean of X is $1/2$

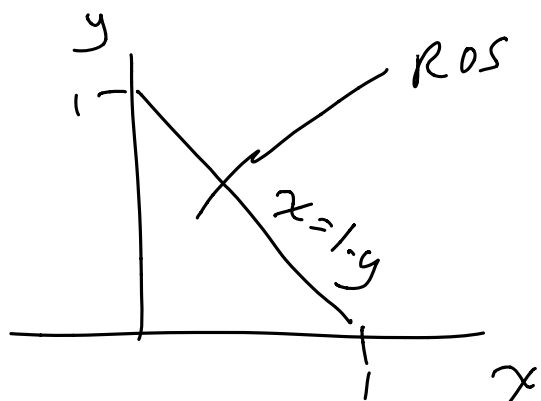
because it's a uniform RV on $[0, 1]$

Another example of iterated expectations

X and Y are uniformly distributed

on triangle formed by $(0,0), (1,0), (0,1)$

what is $E(X)$?



Steps:

- Find $f_x(x|y)$
- Compute $E(X|y)$
- Compute $E(X)$
 $= E(E(X|Y))$

a) use chop and scale

For a given y , X is uniform on $[0, 1-y]$.

$$f_x(x|y) = \begin{cases} \frac{1}{1-y} & 0 \leq x \leq 1-y \\ 0 & \text{else} \end{cases}$$

b) $E(X|y) = \frac{1-y}{2}$ because $X|y$ is uniform

$$c) E(X) = E\left(\frac{1-Y}{2}\right) = \frac{1}{2} - \frac{E(Y)}{2}$$

But by symmetry, we can also say that

$$E(X) = E(Y)$$

Combining, $E(Y) = \frac{1}{2} - \frac{E(Y)}{2} \Rightarrow E(Y) = E(X) = \frac{1}{3}$