

## Conditional expectation & conditional variance

(part 1)

Lecture 8a (STAT 24400 F24)

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## Conditional expectation and conditional distribution

The meaning of **conditional expectation**  $\mathbb{E}(X|Y)$ :

Expected value of a r.v.  $X$  conditional on observing the value of another r.v.  $Y$ , i.e. expected value from a conditional distribution.

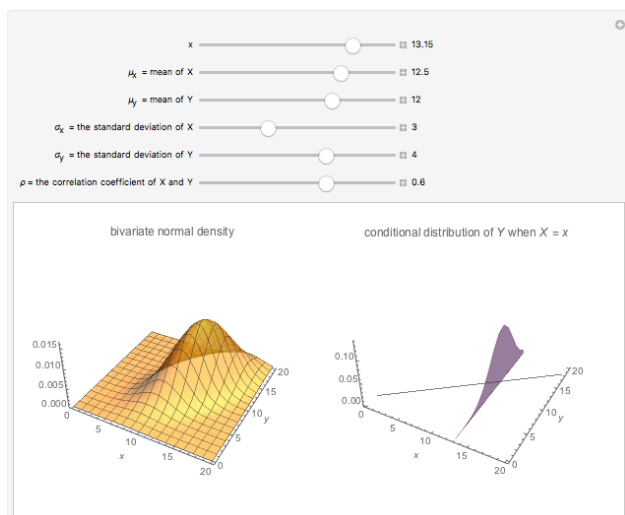
For a joint distribution on  $(X, Y)$ , we can ask about the distribution of  $X$  conditional on observing  $Y = y$ .

As for any distribution, we can calculate its expected value:

e.g. from conditional PMF  $p_{X|Y}(x|y)$  or conditional density  $f_{X|Y}(x|y)$ .

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## Conditional distribution (demo)



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## Conditional expectation (frequentist interpretation)

Intuitively, we can still think of  $\mathbb{E}(X | Y = y)$  as a long-run average:

- Imagine drawing  $(X, Y)$  from its joint distribution many times
- Now throw out all trials except those for which we got  $Y = y$
- Among those trials, what is the average  $X$  value?

These  $X$  are from the conditional distribution given  $Y = y$ .

(This intuition works for discrete  $Y$ , where  $\mathbb{P}(Y = y) > 0$ .

For the continuous case, we can imagine taking limits as it's done before.)

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## Conditional expectation (discrete case)

Discrete case:

$$\mathbb{E}(X \mid Y = y) = \sum_x x \cdot p_{X|Y}(x \mid y)$$

Similarly,

$$\mathbb{E}(Y \mid X = x) = \sum_y y \cdot p_{Y|X}(y \mid x)$$

More generally

$$\mathbb{E}(g(X) \mid Y = y) = \sum_x g(x) \cdot p_{X|Y}(x \mid y)$$

Recall  $p_{X|Y}(x \mid y) = \mathbb{P}(X = x \mid Y = y) = \frac{\mathbb{P}(X=x, Y=y)}{\mathbb{P}(Y=y)}$ , etc.

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## Conditional expectation (continuous case)

Continuous case:

$$\mathbb{E}(X \mid Y = y) = \int_x x \cdot f_{X|Y}(x \mid y) dx$$

More generally

$$\mathbb{E}(g(X) \mid Y = y) = \int_x g(x) \cdot f_{X|Y}(x \mid y) dx$$

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

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## Conditional variance

We can also define conditional variance,

$$\text{Var}(X \mid Y = y) = \mathbb{E}[(X - \mathbb{E}[X \mid Y = y])^2 \mid Y = y]$$

or equivalently,

$$\text{Var}(X \mid Y = y) = \mathbb{E}(X^2 \mid Y = y) - (\mathbb{E}(X \mid Y = y))^2$$

Intuition: Among all trials of  $(X, Y)$ , consider the ones with  $Y = y$ , what is the variability among the corresponding  $X$  values?

Note:  $\mathbb{E}(X \mid Y = y)^2 = (\mathbb{E}(X \mid Y = y))^2$ .

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## Conditional expectation and conditional variance as r.v.'s

- $\mathbb{E}(Y \mid X = x)$  is a real function of  $x \in \mathbb{R}$ .
- When  $\mathbb{E}(Y \mid X = x)$  is well defined for all  $x$  in the support of  $X$ ,  $\mathbb{E}(Y \mid X)$  is a random variable — a function of random variable  $X$ .
- $\text{Var}(Y \mid X = x) \geq 0$  is a real function of  $x \in \mathbb{R}$ .
- When  $\text{Var}(Y \mid X = x)$  is well defined for all  $x$  in the support of  $X$ ,  $\text{Var}(Y \mid X) \geq 0$  is a random variable — a function of random variable  $X$ .

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## Tower law for expectations (Law of Total Expectation)

For jointly distributed  $(X, Y)$ ,

$$\mathbb{E}(Y) = \mathbb{E}[\mathbb{E}(Y | X)]$$

What do we mean by these successive expectations?

We can write this more explicitly as

$$\mathbb{E}(Y) = \mathbb{E}_X \left[ \underbrace{\mathbb{E}_{Y|X}(Y | X)}_{\text{taking expectation over conditional distrib. of } Y \text{ given } X \text{ (the answer is a function of } X\text{)}} \right]$$

taking expectation over marginal distrib. of  $X$   
(i.e., expectation of  $\mathbb{E}(Y|X)$ , which is a function of  $X$ )

Similarly, we may write  $\mathbb{E}(X) = \mathbb{E}[\mathbb{E}(X | Y)]$ .

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## Examples (Tower law for expectations)

1. Let's play the following game: at each round, you toss a coin.

- If it's Heads, you roll a die and win \$1 if you rolled a 6
- If it's Tails, the game ends

What is the expected amount of money you win?

$X$  = total \$ won,  $Y$  = # rounds played. Tower law:  $\mathbb{E}(X) = \mathbb{E}[\mathbb{E}(X | Y)]$

$$X | Y = y \sim \text{Binomial}(y - 1, \frac{1}{6}) \Rightarrow \mathbb{E}(X | Y = y) = \frac{y - 1}{6}.$$

$$\Rightarrow \mathbb{E}(X | Y) = \frac{Y - 1}{6}.$$

$$\mathbb{E}(X) = \mathbb{E}(\mathbb{E}(X | Y)) = \mathbb{E}\left(\frac{Y - 1}{6}\right) = \frac{\mathbb{E}(Y) - 1}{6} = \frac{2 - 1}{6} = \frac{1}{6}.$$

since marginally  $Y \sim \text{Geometric}(\frac{1}{2})$

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## Examples (Tower law used for covariance)

We can also use the tower law to calculate  $\text{Cov}(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y)$ :

$$\mathbb{E}(XY | Y = y) = \mathbb{E}(X | Y = y) \cdot y = \frac{y(y - 1)}{6}$$

$$\Rightarrow \mathbb{E}(XY | Y) = \frac{Y(Y - 1)}{6}$$

$$\mathbb{E}(XY) = \mathbb{E}(\mathbb{E}(XY | Y)) = \mathbb{E}\left(\frac{Y(Y - 1)}{6}\right) = \frac{\mathbb{E}(Y^2)}{6} - \frac{\mathbb{E}(Y)}{6} = \frac{6}{6} - \frac{2}{6} = \frac{2}{3}$$

calculate using  $Y \sim \text{Geometric}(\frac{1}{2})$

$$\text{Cov}(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y) = \frac{2}{3} - \frac{1}{6} \cdot 2 = \frac{1}{3}.$$

Exercise: derive  $\mathbb{E}[Y^2]$  when  $Y$  is geometric( $p$ ).

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## Examples (conditional density & conditional expectation)

2. Let  $(X, Y)$  be supported on the unit square  $[0, 1]^2$  with density

$$f(x, y) = x + y, \quad x \in [0, 1], y \in [0, 1].$$

What is the conditional expectation of  $X | Y$ ?

Marginal density:

$$f_X(x) = \int_{y=0}^1 (x + y) dy = x + 1/2, \quad f_Y(y) = y + 1/2$$

Conditional density: for  $(x, y) \in [0, 1]^2$ ,

$$f_{X|Y}(x | y) = \frac{f_{X,Y}(x, y)}{f_Y(y)} = \frac{x + y}{y + 1/2}$$

Conditional expectation:

$$\mathbb{E}(X | Y = y) = \int_{x=0}^1 x \cdot f_{X|Y}(x | y) dx = \int_{x=0}^1 x \cdot \frac{x + y}{y + 1/2} dx = \frac{y/2 + 1/3}{y + 1/2}$$

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## Tower law for probability

For any event  $A$  and any random variable  $X$ ,

$$\mathbb{P}(A) = \mathbb{E}[\mathbb{P}(A | X)].$$

### Proof:

Let  $\mathbb{1}_A$  be the indicator variable for the event  $A$ .

$$\mathbb{1}_A \sim \text{Bernoulli}(\mathbb{P}(A)) \implies \mathbb{E}(\mathbb{1}_A) = \mathbb{P}(A)$$

Similarly, if we condition on  $X$ ,

$$\mathbb{1}_A | X \sim \text{Bernoulli}(\mathbb{P}(A | X)) \implies \mathbb{E}(\mathbb{1}_A | X) = \mathbb{P}(A | X)$$

By the tower law,

$$\mathbb{P}(A) = \mathbb{E}(\mathbb{1}_A) = \mathbb{E}[\mathbb{E}(\mathbb{1}_A | X)] = \mathbb{E}[\mathbb{P}(A | X)].$$

  
by tower law for expectations,

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## Examples (apply the tower law for probability)

1. You play the following game: at each round, you toss a coin.

- If it's Heads, you roll a die and win \$1 if you rolled a 6
- If it's Tails, the game ends

What is the probability that you win nothing?

$X$  = total \$ won,  $Y$  = # rounds played.

$$\mathbb{P}(X = 0) = \mathbb{E}(\mathbb{P}(X = 0 | Y)) = \mathbb{E}\left(\left(\frac{5}{6}\right)^{Y-1}\right)$$

$$= \sum_{y=1}^{\infty} p_Y(y) \cdot \left(\frac{5}{6}\right)^{y-1} = \sum_{y=1}^{\infty} \left(\frac{1}{2}\right)^y \left(\frac{5}{6}\right)^{y-1} = \frac{6}{7}.$$

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## Examples (apply the tower law for probability)

2. Suppose we build a toy tower of height  $H$ , where  $H \sim \text{Uniform}[100, 200]$ . However, a larger tower is more likely to break; a tower of height  $h$  breaks with probability  $\frac{h}{300}$ , in which case the final height is zero.

What is the probability that the tower doesn't break?

Let  $A$  be the event that the tower doesn't break.

$$\mathbb{P}(A | H = h) = \left(1 - \frac{h}{300}\right)$$

$$\mathbb{P}(A) = \mathbb{E}(\mathbb{P}(A | H)) = \mathbb{E}\left(1 - \frac{H}{300}\right) = 1 - \frac{\mathbb{E}(H)}{300} = 1 - \frac{150}{300} = \frac{1}{2}$$

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## Examples (apply the tower law for probability)

What is the expected height of the tower?

Let  $A$  be the event that the tower doesn't break. Then the final height is  $X = \mathbb{1}_A \cdot H$ .

$$\mathbb{E}(X | H = h) = \mathbb{E}(\mathbb{1}_A \cdot H | H = h) = \mathbb{E}(\mathbb{1}_A | H = h) \cdot h = \left(1 - \frac{h}{300}\right) \cdot h$$

$$\begin{aligned} \mathbb{E}(X) &= \mathbb{E}[\mathbb{E}(X | H)] = \mathbb{E}\left[\left(1 - \frac{H}{300}\right) \cdot H\right] \\ &= \mathbb{E}(H) - \frac{\mathbb{E}(H^2)}{300} = 150 - \frac{23333.33}{300} = 72.22 \end{aligned}$$

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