W10 Notes

Review: True/False

TRUE or FALSE:

- (a) Suppose Cov[X,Y] = 0.05. It is possible that the correlation coefficient $\rho_{X,Y} = 0.05$.
- (b) Suppose Cov[X, Y] = 0.05. It is possible that X and Y have a strong linear relationship.
- (c) Suppose Cov[X, Y] = 0. It is possible that X and Y are *not* independent.
- (d) Suppose X and Y are *not* independent. It is possible that Cov[X, Y] is equal to 0.
- (e) Suppose X and Y *are* independent. Cov[X, Y] must be equal to 0.

Review: Conditional probability

Recall some items related to conditional probability.

Conditioning definition:

$$P[-\mid A] = \frac{P[-\cap A]}{P[A]}$$

Multiplication rule:

$$P[AB] = P[B \mid A] P[A]$$

Division into Cases / Total Probability:

$$P[B] = P[B \mid A_1] P[A_1] + \cdots + P[B \mid A_n] P[A_n]$$

Conditional distribution

01 Theory

⊞ Conditional distribution - fixed event

Suppose X is a random variable, and suppose $A \subset \mathbb{R}$. The distribution of X conditioned on A describes the probabilities of values of X given knowledge that $X \in A$.

Discrete case:

$$P_{X|A}(k) \; = \; egin{cases} rac{1}{P[A]}P_X(k) & k \in A \ 0 & k
otin A \end{cases}$$

Continuous case:

There is also a conditional CDF, of which this conditional PDF is the derivative:

$$F_{X|A}(x) \ = \ P[X \le x \mid A], \qquad f_{X|A}(x) \ = \ rac{d}{dx} F_{X|A}(x)$$

The Law of Total Probability has versions for distributions:

$$P_X(k) = P_{X|A_1}(k) P[A_1] + \cdots + P_{X|A_n}(k) P[A_n]$$

$$f_X(x) = f_{X|A_1}(x)\,P[A_1] + \dots + f_{X|A_n}(x)\,P[A_n]$$

⊞ Conditional distribution - variable event

Suppose X and Y are any two random variables. The **distribution** of X conditioned on Y describes the probabilities of values of X in terms of y, given knowledge that Y = y.

Discrete case:

$$egin{aligned} P_{X|Y}(k|\ell) &= P[X=k \mid Y=\ell] \ \\ &= rac{P_{X,Y}(k,\ell)}{P_Y(\ell)} \qquad ext{(assuming $P_Y(\ell)
eq 0$)} \end{aligned}$$

Continuous case:

$$f_{X|Y}(x|y) \; = \; rac{f_{X,Y}(x,y)}{f_Y(y)} \qquad ext{(assuming } f_Y(y)
eq 0)$$

Remember: $P_{X,Y}(k,\ell)$ is the probability that "X=k and $Y=\ell$."

Sometimes it is useful to have the formulas rewritten like this:

$$P_{X,Y}(k,\ell) = P_{X|Y}(k|\ell) P_Y(\ell)$$

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

\blacksquare Extra - Deriving $f_{X|Y}(x|y)$

The density $f_{X|Y}$ ought to be such that $f_{X|Y}(x|y) dx$ gives the probability of $X \in [x, x + dx]$, given knowledge that $Y \in [y, y + dy]$. Calculate this probability:

$$egin{aligned} Pigg[x \leq X \leq x + dx \ igg| \ y \leq Y \leq y + dy igg] \ \gg \gg & rac{Pigg[x \leq X \leq x + dx, \ y \leq Y \leq y + dy igg]}{Pigg[y \leq Y \leq y + dy igg]} \ \gg \gg & rac{f_{X,Y}(x,y) \ dx \ dy}{f_Y(y) \ dy} \ \gg \gg & rac{f_{X,Y}(x,y)}{f_Y(y)} \ dx \end{aligned}$$

02 Illustration

≡ Example - Conditional PMF, variable event, via joint density

Suppose X and Y have joint PMF given by:

$$P_{X,Y}(k,\ell) \quad = \quad egin{cases} rac{k+\ell}{21} & k=1,2,3; \ \ell=1,2 \ 0 & ext{otherwise} \end{cases}$$

Find $P_{X|Y}(k|\ell)$ and $P_{Y|X}(\ell,k)$.

Solution

Marginal PMFs:

$$P_X(k) = rac{2k+3}{21}, \qquad k=1,2,3$$
 $P_Y(\ell) = rac{\ell+2}{7}, \qquad \ell=1,2$

Assuming $\ell = 1$ or 2, for each k = 1, 2, 3 we have:

$$P_{X|Y}(k|\ell) = rac{P_{X,Y}(k,\ell)}{P_{Y}(\ell)} \gg rac{k+\ell}{3\ell+6}$$

Assuming $k = 1, 2, \text{ or } 3, \text{ for each } \ell = 1, 2 \text{ we have:}$

$$P_{Y|X}(\ell|k) = rac{P_{Y,X}(\ell,k)}{P_{X}(k)} \gg rac{k+\ell}{2k+3}$$

Conditional expectation

03 Theory

⊞ Expectation conditioned by a fixed event

Suppose X is a random variable and $A \subset \mathbb{R}$. The **expectation of** X **conditioned on** A describes the typical value of X given the hypothesis that $X \in A$ is known.

Discrete case:

$$E[\,X\mid A\,] \quad = \quad \sum_k k\, P_{X\mid A}(k)$$

$$E[\,g(X)\mid A\,] \quad = \quad \sum_k g(k)\,P_{X\mid A}(k)$$

Continuous case:

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

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

Conditional variance:

$$\operatorname{Var}[X \mid A] = E[(X - \mu_{X|A})^2 \mid A] = E[X^2 \mid A] - \mu_{X|A}^2$$

Division into Cases / Total Probability applied to expectation:

$$E[X] = E[X \mid A_1] P[A_1] + \cdots + E[X \mid A_n] P[A_n]$$

Linearity of conditional expectation:

$$E[\,aX_1 + bX_2 + c\mid Y = y\,] = a\,E[\,X_1\mid Y = y\,] + b\,E[\,X_2\mid Y = y\,] + c$$

Extra - Proof: Division of Expectation into Cases

We prove the discrete case only.

1. Expectation formula:

$$E[X] = \sum_k k P_X(k)$$

2. Division into Cases for the PMF:

$$P_X(k) = \sum_{i=1}^n P_{X|A_i}(k) \, P[A_i]$$

3. Substitute in the formula for E[X]:

$$egin{aligned} \sum_k k \, P_X(k) &\gg\gg \sum_k k \, \sum_{i=1}^n P_{X|A_i}(k) \, P[A_i] \ &\gg\gg \sum_{i=1}^n P[A_i] \sum_k k \, P_{X|A_i}(k) \ &\gg\gg \sum_{i=1}^n P[A_i] \, E[\, X \mid A_i\,] \end{aligned}$$

⊞ Expectation conditioned by a variable event

Suppose X and Y are any two random variables. The **expectation of** X **conditioned on** Y = y describes the typical of value of X in terms of y, given the hypothesis that Y = y is known.

Discrete case:

$$E[\,X\mid Y=y\,] \quad = \quad \sum_k k\, P_{X\mid Y}(k\mid \! y) \quad (k ext{ over all poss. vals.})$$

$$E[\,g(X,Y)\mid Y=y\,] \quad = \quad \sum_k g(k,y)\,P_{X|Y}(k|y)$$

Continuous case:

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

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

05 Illustration

≔ Example - Conditional PMF, fixed event, expectation

Suppose *X* measures the lengths of some items and has the following PMF:

$$P_X(k) = egin{cases} 0.15 & k=1,2,3,4 \ 0.1 & k=5,6,7,8 \ 0 & ext{otherwise} \end{cases}$$

Let *L* be the event that $X \geq 5$.

- (a) Find the conditional PMF of X given that L is known.
- (b) Find the conditional expected value and variance of X given L.

Solution

(a)

Conditional PMF formula with $x \in L$ plugged in:

$$P_{X|L}(x) = egin{cases} rac{P_X(x)}{P[L]} & x=5,6,7,8 \ 0 & ext{otherwise} \end{cases}$$

Compute P[L] by adding cases:

$$P[L] \ = \ \sum_{k=-5}^{8} P_X(k) \gg \gg 0.4$$

Divide nonzero PMF entries by 0.1:

$$P_{X|L}(k) = egin{cases} 0.25 & k=5,6,7,8 \ 0 & ext{otherwise} \end{cases}$$

(b)

Find $E[X \mid L]$:

$$E[\,X\mid L\,] \;=\; \sum_{k=5}^8 k\, P_{X\mid L}(k)$$

$$\gg \gg 5 \cdot (0.25) + 6 \cdot (0.25) + 7 \cdot (0.25) + 8 \cdot (0.25)$$

$$\gg\gg$$
 6.5 min

Find $E[X^2 \mid L]$:

$$E[\,X^2\mid L\,] \;=\; \sum_{k=5}^8 k^2\, P_{X\mid L}(k)$$

$$\gg \gg 5^2 \cdot (0.25) + 6^2 \cdot (0.25) + 7^2 \cdot (0.25) + 8^2 \cdot (0.25)$$

$$\gg \gg 43.5 \, \mathrm{min}^2$$

Find $Var[X \mid L]$ using "short form" with conditioning:

$$Var[X | L] = E[X^2 | L] - E[X | L]^2 \gg 1.25 min^2$$

04 Theory - extra

⊞ Expectation conditioned by a random variable

Suppose X and Y are any two random variables. The **expectation of** X **conditioned on** Y is a random variable giving the typical value of X on the assumption that Y has value determined by an outcome of the experiment.

$$E[X \mid Y] = g(Y)$$
 where $g(y) = E[X \mid Y = y]$

In other words, start by defining a function g(y):

$$g: \mathbb{R}
ightarrow \mathbb{R} \ y \mapsto E[\, X \mid Y = y\,]$$

Now $E[X \mid Y]$ is defined as the composite random variable g(Y).

Considered as a random variable, E[X | Y] takes an outcome $s \in S$, computes Y(s), sets y = Y(s), then returns the expectation of X conditioned on Y = y.

Notice that X is *not* evaluated at s, only Y is.

Because the value of $E[X \mid Y]$ depends only on Y(s), and not on any additional information about s, it is common to *represent* a conditional expectation $E[X \mid Y]$ using only the function g.

☐ Iterated Expectation

$$E[\; E[X \mid Y] \;] = E[X]$$

Proof of Iterated Expectation, discrete case

$$\begin{split} E[\,E[\,X\mid Y\,]\,] &= &\sum_{\ell} E[X\mid Y=\ell]\,P_Y(\ell) \\ \\ &= &\sum_{\ell} \sum_{k} k\,P_{X\mid Y}(k|\ell)\,P_Y(\ell) \\ \\ &= &\sum_{k} k\sum_{\ell} P_{X,Y}(k,\ell) \\ \\ &= &\sum_{k} k\,P_X(k) = E[X] \end{split}$$

05 Illustration - extra

≔ Example - Conditional expectations from joint density

Suppose *X* and *Y* are random variables with joint density given by:

$$f_{X,Y}(x,y) = egin{cases} rac{1}{y}e^{-x/y}e^{-y} & x,y \in (0,\infty) \ 0 & ext{otherwise} \end{cases}$$

Find $E[X \mid Y = y]$. Use this to compute E[X].

Solution

(1) Derive the marginal density $f_Y(y)$:

$$f_Y(y) \quad \gg \gg \quad \int_0^{+\infty} rac{1}{y} e^{-x/y} e^{-y} \, dx$$

$$\gg\gg -e^{-x/y}e^{-y}\Big|_{x=0}^{\infty}\gg\gg e^{-y}$$

(2) Use $f_Y(y)$ to compute $f_{X|Y}(x|y)$:

$$f_{X|Y}(x|y) \quad \gg \gg \quad rac{f_{X,Y}(x,y)}{f_{Y}(y)}$$

$$\gg \gg \frac{1}{y} e^{-x/y} e^{-y} \cdot (e^{-y})^{-1} \gg \gg \frac{1}{y} e^{-x/y}$$

(3) Use $f_{X|Y}(x|y)$ to calculate expectation conditioned on the variable event:

$$E[X\mid Y=y] \quad \gg \gg \quad \int_{-\infty}^{+\infty} x\, f_{X\mid Y}(x\mid y)\, dx$$

$$\gg\gg\int_0^\inftyrac{x}{y}e^{-x/y}\,dx \gg\gg y$$

(4) Apply Iterated Expectation:

Set g(y) = y. By Iterated Expectation, we know that E[X] = E[g(Y)]. Therefore:

$$E[X] = E[g(Y)] \quad = \quad \int_{-\infty}^{+\infty} g(y) \, f_Y(y) \, dy$$

$$>>> \int_0^{+\infty} y \, e^{-y} \, dy >>> 1$$

Notice that g(Y) = Y, so $E[X \mid Y] = Y$, and Iterated Expectation says that E[X] = E[Y].

¡ Example - Flip coin, choose RV

Suppose $X \sim \text{Ber}(1/3)$ and $Y \sim \text{Ber}(1/4)$ represent two biased coins, giving 1 for heads and 0 for tails.

Here is the experiment:

- 1. Flip a fair coin.
- 2. If heads, flip the *X* coin; if tails, flip the *Y* coin.
- 3. Record the outcome as Z.

What is E[Z]?

Solution

Let $G \sim \text{Ber}(1/2)$ describe the fair coin. Then:

$$egin{aligned} E[Z] &= E[\,E[Z \,|\, G\,]\,] \ \gg \gg & E[\,Z \,|\, G=0\,]\, P_G(0) + E[\,Z \,|\, G=1\,]\, P_G(1) \ &\gg \& E[Y]\, P_G(0) + E[X]\, P_G(1) \ &\gg \& rac{1}{4} \cdot rac{1}{2} + rac{1}{3} \cdot rac{1}{2} &\gg \gg rac{7}{24} \end{aligned}$$

≔ Example - Sum of random number of RVs

Let N denote the number of customers that enter a store on a given day.

Let X_i denote the amount spent by the i^{th} customer.

Assume that E[N] = 50 and $E[X_i] = \$8$ for each *i*.

What is the expected total spend of all customers in a day?

Solution

A formula for the total spend is $X = \sum_{i=1}^{N} X_i$.

By Iterated Expectation, we know $E[X] = E[E[X \mid N]]$.

Now compute $E[X \mid N]$ as a function of N:

$$egin{aligned} E[X \mid N=n] &\gg\gg & E\left[\left(\sum_{i=1}^{N}X_i
ight) \mid N=n
ight] \ &\gg\gg & E\left[\left(\sum_{i=1}^{n}X_i
ight) \mid N=n
ight] \ &\gg\gg & \sum_{i=1}^{n}E[X_i \mid N=n] \ &\gg\gg & \sum_{i=1}^{n}E[X_i] &\gg\gg & 8n \end{aligned}$$

Therefore g(n) = 8n and g(N) = 8N and $E[X \mid N] = 8N$.

Then by Iterated Expectation, E[X] = E[8N] = 8E[N] = \$400.