## Multivariate Expectation Application Solutions

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24 June, 2020

## Exercises

1. Let X and Y be continuous random variables with joint pmf:

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

where  $0 \le x \le 1$  and  $0 \le y \le 1$ .

a) Find E(X) and E(Y).

$$E(X) = \int_0^1 x \left( x + \frac{1}{2} \right) dx = \frac{x^3}{3} + \frac{x^2}{4} \Big|_0^1 = \frac{1}{3} + \frac{1}{4} = \frac{7}{12} = 0.583$$

$$E(Y) = \int_0^1 y \left( y + \frac{1}{2} \right) dy = 0.583$$

b) Find Var(X) and Var(Y).

$$\operatorname{Var}(X) = \operatorname{E}(X^2) - \operatorname{E}(X)^2$$

$$\operatorname{E}(X^2) = \int_0^1 x^2 \left( x + \frac{1}{2} \right) dx = \frac{x^4}{4} + \frac{x^3}{6} \Big|_0^1 = \frac{1}{4} + \frac{1}{6} = \frac{5}{12} = 0.417$$

So,  $Var(X) = 0.417 - 0.583^2 = 0.076$ .

Similarly, Var(Y) = 0.076.

c) Find Cov(X, Y) and  $\rho$ . Are X and Y independent?

$$\operatorname{Cov}(X,Y) = \operatorname{E}(XY) - \operatorname{E}(X)\operatorname{E}(Y)$$

$$\operatorname{E}(XY) = \int_0^1 \int_0^1 xy(x+y) \, \mathrm{d}y \, \mathrm{d}x = \int_0^1 \frac{x^2y^2}{2} + \frac{xy^3}{3} \Big|_0^1 \, \mathrm{d}x = \int_0^1 \frac{x^2}{2} + \frac{x}{3} \, \mathrm{d}x$$

$$= \frac{x^3}{6} + \frac{x^2}{6} \Big|_0^1 = \frac{1}{3} = 0.333$$

So,

$$Cov(X,Y) = \frac{1}{3} - \left(\frac{7}{12}\right)^2 = -0.007$$

$$\rho = \frac{\text{Cov}(X,Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}} = \frac{-0.007}{\sqrt{0.076 \times 0.076}} = -0.909$$

With a non-zero covariance, X and Y are not independent.

d) Find Var(3X + 2Y).

$$Var(3X + 2Y) = Var(3X) + Var(2Y) + 2Cov(3X, 2Y) = 9Var(X) + 4Var(Y) + 12Cov(X, Y)$$
$$= 9 * 0.076 + 4 * 0.076 + 12 * -0.007 = 0.910$$

2. Let X and Y be continuous random variables with joint pmf:

$$f_{X,Y}(x,y) = 1$$

where  $0 \le x \le 1$  and  $0 \le y \le 2x$ .

a) Find E(X) and E(Y).

$$E(X) = \int_0^1 x \cdot 2x \, dx = \frac{2x^3}{3} \Big|_0^1 = 0.667$$

$$E(Y) = \int_0^2 y \left(1 - \frac{y}{2}\right) dy = \frac{y^2}{2} - \frac{y^3}{6} \Big|_0^2 = 2 - \frac{8}{6} = 0.667$$

b) Find Var(X) and Var(Y).

$$E(X^2) = \int_0^1 x^2 \cdot 2x \, dx = \frac{x^4}{2} \Big|_0^1 = 0.5$$

So,  $Var(X) = 0.5 - \left(\frac{2}{3}\right)^2 = \frac{1}{18} = 0.056$ 

$$E(Y^2) = \int_0^2 y^2 \left(1 - \frac{y}{2}\right) dy = \frac{y^3}{3} - \frac{y^4}{8} \Big|_0^2 = \frac{8}{3} - 2 = 0.667$$

So, 
$$Var(Y) = \frac{2}{3} - \left(\frac{2}{3}\right)^2 = \frac{2}{9} = 0.222$$

c) Find Cov(X, Y) and  $\rho$ . Are X and Y independent?

$$E(XY) = \int_0^1 \int_0^{2x} xy \, dy \, dx = \int_0^1 \frac{xy^2}{2} \Big|_0^{2x} \, dx = \int_0^1 2x^3 \, dx = \frac{x^4}{2} \Big|_0^1 = \frac{1}{2}$$

So,

$$Cov(X,Y) = \frac{1}{2} - \frac{2}{3}\frac{2}{3} = \frac{1}{18} = 0.056$$

$$\rho = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}} = \frac{\frac{1}{18}}{\sqrt{\frac{1}{18}\frac{2}{9}}} = 0.5$$

X and Y appear to be positively correlated (thus not independent).

d) Find Var  $\left(\frac{X}{2} + 2Y\right)$ .

$$Var\left(\frac{X}{2} + 2Y\right) = \frac{1}{4}Var(X) + 4Var(Y) + 2Cov(X, Y) = \frac{1}{72} + \frac{8}{9} + \frac{1}{9} = 1.014$$

3. Suppose X and Y are independent random variables. Show that E(XY) = E(X)E(Y).

If X and Y are independent, then Cov(X,Y) = 0. So,

$$Cov(X, Y) = E(XY) - E(X)E(Y) = 0$$

Thus,

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

- 4. You are playing a game with a friend. Each of you roll a fair sided die and record the result.
- a) Write the joint probability mass function.

Let X be the number on your die and Y be the number on your friend's die.

b) Find the expected value of the product of your score and your friends score.

Find E[XY] so we just go through an find all 36 values of the product and multiply by the probabilities. Since the probabilities are all constant, we will take that out. Now

$$E[XY] = \frac{1}{36}(1+2+3+4+5+6+2+4+$$

$$6+8+10+12+3+6+9+12+15+18+4+8+12+16+20+24+$$

$$5+10+15+20+25+30+6+12+18+24+30+36)$$

$$= 12.25$$

c) Verify the previous part using simulation.

```
set.seed(1012)
(do(100000)*(sample(1:6,size=2,replace=TRUE))) %>%
mutate(prod=V1*V2) %>%
summarize(Expec=mean(prod))
```

```
## Expec
## 1 12.25016
```

d) Using simulation, find the expected value of the maximum number on the two roles. In the next lesson, we will show how to find this using a transformation method.

```
(do(100000)*max(sample(1:6,size=2,replace=TRUE))) %>%
   summarize(Expec=mean(max))
```

```
## Expec
## 1 4.4737
```

5. A miner is trapped in a mine containing three doors. The first door leads to a tunnel that takes him to safety after two hours of travel. The second door leads to a tunnel that returns him to the mine after three hours of travel. The third door leads to a tunnel that returns him to his mine after five hours. Assuming that the miner is at all times equally likely to choose any one of the doors, yes a bad assumption but it makes for a nice problem, what is the expect length of time until the miner reaches safety?

Simulating this is a little more challenging because we need a conditional but we try it first before going to the mathematical solution.

Let's write a function that takes a vector and returns the sum of the values up to the first time 2 appears. Anytime you are repeating something more than 5 times, it might make sense to write a function.

```
miner_time <- function(x){
  index <- which(x==2)[1]
  total<-cumsum(x)
  return(total[index])
}</pre>
```

```
set.seed(113)
(do(10000)*miner_time(sample(c(2,3,5),size=20,replace=TRUE))) %>%
summarise(Exp=mean(miner_time))
```

```
## Exp
## 1 10.0092
```

Now let's find it mathematically.

Let X be the time it takes and Y the door. Then we have

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

Now if door 2 is selected

$$E[X|Y = 2] = E[X] + 3$$

since the miner will travel for 3 hours and then be back at the starting point.

Likewise if door 3 is select

$$E[X|Y=2] = E[X] + 5$$

So

$$E[x] = \frac{1}{3}2 + \frac{1}{3}(E[X] + 3) + \frac{1}{3}(E[X] + 5)$$
$$E[x] - \frac{2}{3}E[X] = \frac{2}{3} + \frac{3}{3} + \frac{5}{3}$$
$$\frac{1}{3}E[X] = \frac{10}{3}$$

6. ADVANCED: Let  $X_1, X_2, ..., X_n$  be independent, identically distributed random variables. (This is often abbreviated as "iid"). Each  $X_i$  has mean  $\mu$  and variance  $\sigma^2$  (i.e., for all i,  $E(X_i) = \mu$  and  $Var(X_i) = \sigma^2$ ).

Let 
$$S = X_1 + X_2 + ... + X_n = \sum_{i=1}^n X_i$$
. And let  $\bar{X} = \sum_{i=1}^n \frac{X_i}{n}$ .

Find E(S), Var(S),  $E(\bar{X})$  and  $Var(\bar{X})$ .

$$E(S) = E(X_1 + X_2 + ... + X_n) = E(X_1) + E(X_2) + ... + E(X_n) = \mu + \mu + ... + \mu = n\mu$$

Since the  $X_i$ s are all independent:

$$Var(S) = Var(X_1 + X_2 + ... + X_n) = Var(X_1) + Var(X_2) + ... + Var(X_n) = n\sigma^2$$

$$E(\bar{X}) = \frac{1}{n}E(X_1 + X_2 + \dots + X_n) = \frac{1}{n}n\mu = \mu$$
$$Var(\bar{X}) = \frac{1}{n^2}Var(X_1 + X_2 + \dots + X_n) = \frac{1}{n^2}n\sigma^2 = \frac{\sigma^2}{n}$$

## File Creation Information

• File creation date: 2020-06-24

• Windows version: Windows 10 x64 (build 17763)

• R version 3.6.3 (2020-02-29)

• mosaic package version: 1.6.0

• tidyverse package version: 1.3.0