

Homework 3 Solutions1. (10 points) *Messenger*

- a. (X, Y) is a random point on the rectangle 13×2 whose center is at the origin, in other words X and Y are independent uniform over $(-6.5, 6.5)$ and $(-1, 1)$, respectively, and the distance is $Z = |X| + |Y|$. Also note that the four quadrants are symmetric.

$$E(Z) = \int_{-1}^1 \int_{-6.5}^{6.5} (|x| + |y|) \frac{1}{13} \frac{1}{2} dx dy \quad (1)$$

$$= 4 \int_0^1 \int_0^{6.5} (x + y) \frac{1}{26} dx dy \quad (2)$$

$$= \frac{4}{26} \int_0^1 (6.5^2/2 + 6.5y) dy = 4/26(6.5^2/2 + 6.5/2) = 3.75 \quad (3)$$

- b. Recall that the Markov inequality for a positive random variable is $P(X > a) \leq E(X)/a$. Also clearly distance is positive:

$$P(Z > 5) \leq 3.75/5 = 75\% \quad (4)$$

2. (10 points) *Pasta and rice*

- a. The constraints are $100 \leq \max\{X, R\} \leq 300$. The joint pdf is constant over that region. Figure 1 contains the diagram.

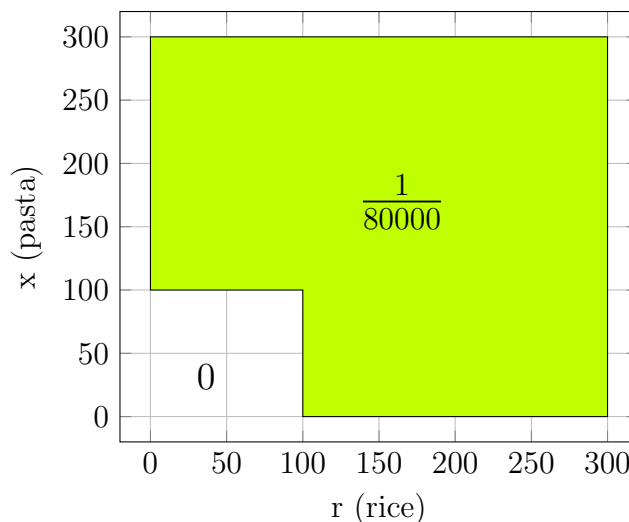


Figure 1: Joint pdf of R and X.

- b. We compute

$$E(RX) = \int_{x=100}^{300} \int_{r=0}^{300} \frac{rx}{80000} dx dr + \int_{x=0}^{100} \int_{r=100}^{300} \frac{rx}{80000} dx dr \quad (5)$$

$$= \frac{1}{80000} \left(\frac{x^2}{2} \Big|_{100}^{300} \frac{r^2}{2} \Big|_0^{300} + \frac{x^2}{2} \Big|_0^{100} \frac{r^2}{2} \Big|_{100}^{300} \right) \quad (6)$$

$$= 25000. \quad (7)$$

$$E(X) = \int_{x=100}^{300} \int_{r=0}^{300} \frac{x}{80000} dx dr + \int_{x=0}^{100} \int_{r=100}^{300} \frac{x}{80000} dx dr \quad (8)$$

$$= \frac{1}{80000} \left(300 \frac{x^2}{2} \Big|_{100}^{300} + 200 \frac{x^2}{2} \Big|_0^{100} \frac{r^2}{2} \Big|_{100}^{300} \right) \quad (9)$$

$$= 162.5. \quad (10)$$

By symmetry $E(R) = 162.5$. The covariance $\text{Cov}(R, X) = E(RX) - E(R)E(X) = 1406.25$ so R and X are negatively correlated.

c. The variables are correlated, which implies they cannot be independent.

3. (10 points) *Restaurant*

a. Let N be the number of customers and C_i the money customer i spends. The money made on a given night is

$$M = \sum_{i=1}^N C_i. \quad (11)$$

We are interested in the mean of this random variable. Conditioned on $N = n$ the mean is

$$E(M|N = n) = \sum_{i=1}^n E(C_i|N = n) = n E(C_1|N = n), \quad (12)$$

as long as all the C_i have the same conditional mean given the number of customers. This assumption is implicit in what the management is saying.

Now, if N and C_i (i.e. the number of customers and what each customer spends) are **independent** then $E(C_i|N = n) = E(C_i) = 30$, so $E(M|N = n) = 30n$. This would imply that by iterated expectation

$$E(M) = E(E(M|N)) = E(N\mu) = E(30N) = 30 E(N) = 1200 \text{ dollars}. \quad (13)$$

In fact, if you want to be very rigorous, we don't need N and C_i to be independent, as long as $E(C_i|N = n) = E(C_i) = 30$, which is a less strong requirement.

b. Under the imagined scenario

$$E(N) = 100 P(N = 100) + 10 P(N = 10) \quad (14)$$

$$= 100 P(\text{good night}) + 10 (1 - P(\text{good night})) \quad (15)$$

$$= 90 P(\text{good night}) + 10 = 40, \quad (16)$$

so $P(\text{good night}) = 1/3$.

c. To compute the expected earned money per night under the assumptions we first compute the conditional expectation given N

$$E(M|N) = \begin{cases} 1000 & \text{if } N = 100 \\ 400 & \text{if } N = 10. \end{cases} \quad (17)$$

Now by iterated expectation

$$E(E(M|N)) = 1000 P(N = 100) + 400 P(N = 10) = 600 \text{ dollars.} \quad (18)$$

You are telling this story to show that the actual expected gain can be very different (in this case half!) from the management's back-of-the-envelope calculation if the money spent by each customer depends on the number of clients.

4. (10 points) *Copper*

- a. Let us define the random variables C (amount of copper), X (price of copper) and V (value of the stored copper). From the problem statement $V \leq 2.5$ million dollars and $E(V) = 2$ million dollars. Since we only know the mean of the random variable, we use Markov's inequality. Since we want to bound the probability that V is *smaller* than a certain quantity, we will apply the inequality to $-V$ or better still to $Y = 2.5 \text{ million} - V$ which is nonnegative:

$$P(V \leq 1 \text{ million}) = P(Y > 1.5 \text{ million}) \quad (19)$$

$$\leq \frac{E(Y)}{1.5 \text{ million}} = \frac{E(2.5 \text{ million} - V)}{1.5 \text{ million}} = \frac{0.5 \text{ million}}{1.5 \text{ million}} = \frac{1}{3}. \quad (20)$$

The probability of the event of interest is at most $1/3$.

- b. Probably not, as the company will buy more copper when it is cheaper and sell when it is more expensive, so it seems plausible for C and X to have negative correlation.
- c. In order to use the additional information we apply Chebyshev's inequality. For this we need the variance of V . First, note that by independence,

$$E(V) = E(XC) = E(X)E(C), \quad (21)$$

so $E(C) = E(V)/E(X) = 4/9$ million. We can now compute

$$E(V^2) = E(X^2C^2) \quad (22)$$

$$= E(X^2)E(C^2) \text{ by independence} \quad (23)$$

$$= (\text{Var}(X) + E^2(X))(\text{Var}(C^2) + E^2(C)) \quad (24)$$

$$= (0.2^2 + 4.5^2) \left(10000^2 + \left(\frac{4000000}{9} \right)^2 \right) \quad (25)$$

$$\approx 4 \cdot 10^{12} + 9.93 \cdot 10^9. \quad (26)$$

$$\text{Var}(V) = E(V^2) - E^2(V) \approx 9.93 \cdot 10^9. \quad (27)$$

Now,

$$P(V \leq 1 \text{ million}) \leq P(|V - E(V)| > 1 \text{ million}) \quad (28)$$

$$\leq \frac{\text{Var}(V)}{10^{12}} \approx 0.993\%. \quad (29)$$

A **much** sharper bound.

5. (10 points) *Law of conditional variance*

- a. Given $X = x$, X is constant, and its value is x , therefore once x is fixed $\text{Var}(Y|X = x)$ is a **number**. It represents the variance of the random variable Y given the information $X = x$. In other words, given two random variables X and Y , look at their joint density, condition X to the value x . This slice gives rise to a new distribution of the random variable $Y|(X = x)$, $\text{Var}(Y|X = x)$ is precisely the variance of this random variable.

$$\text{Var}(Y|X = x) = E((Y - E(Y|X = x))^2|X = x) \quad (30)$$

- b. Now we don't keep x fixed but rather we treat it as a variable, for any x we assign a value $\text{Var}(Y|X = x)$, this is a function from the range of X to real numbers, which we call by h , so $h(x) = \text{Var}(Y|X = x)$. Since this function is defined on a probability space we can regard it as a **random variable** by $h(X) = \text{Var}(Y|X)$.

- c. Using iterated expectations we get:

$$E[\text{Var}(Y|X)] = E[E(Y^2|X)] - E[E(Y|X)^2] \quad (31)$$

$$= E(Y^2) - E[E(Y|X)^2] \quad (32)$$

$$\text{Var}[E(Y|X)] = E[E(Y|X)^2] - E[E(Y|X)]^2 \quad (33)$$

$$= E[E(Y|X)^2] - E(Y)^2 \quad (34)$$

$$\implies E[\text{Var}(Y|X)] + \text{Var}[E(Y|X)] = E(Y^2) - E(Y)^2 = \text{Var}(Y) \quad (35)$$

Average of the variance of Y given X , plus the variance of the average of Y given X .

- d. Let T be the time at which a runner gets injured. And let A be the random variable that describes the age group: $A = 1$ be the group of runners below 30, and $A = 2$ be the group above 30. So that $T|\{A = 1\} \sim \text{exp}(1)$ and $T|\{A = 2\} \sim \text{exp}(2)$. We are also given that $P(A = 2) = 0.2$. Also note that if $X \sim \text{exp}(\lambda)$ then $E(X^2) = \text{Var}(X) + E(X)^2 = 1/\lambda^2 + 1/\lambda^2 = 2/\lambda^2$.

$$E(T) = E(E(T|A)) = E(T|A = 1)P(A = 1) + E(T|A = 2)P(A = 2) \quad (36)$$

$$= 1 \cdot 0.8 + 2 \cdot 0.2 = 1.2 \quad (37)$$

$$E(T^2) = E(E(T^2|A)) = E(T^2|A = 1)P(A = 1) + E(T^2|A = 2)P(A = 2) \quad (38)$$

$$= 2/1^2 \cdot 0.8 + 2/2^2 \cdot 0.2 = 1.7 \quad (39)$$

$$\implies \text{Var}(T) = 0.26 \quad (40)$$