

**PROPOSED SOLUTIONS
HOMEWORK 1
GRA6039 ECONOMETRICS WITH PROGRAMMING
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The definitions, propositions, etc. that I refer to are those in the Lecture notes. If you spot mistakes in the proposed solutions, please send me an email.

Solution to Ex. 3. (d) Recall that $B \setminus A = B \cap A^c$. Write $B = (B \cap A) \cup (B \setminus A) =$ (make a Venn diagram!), and notice that $B \cap A$ and $B \cap A^c$ are disjoint. Use Def. 1.2(iii) in the Lecture notes,

$$\Pr(B) = \Pr((B \cap A) \cup (B \setminus A)) = \Pr(B \cap A) + \Pr(B \setminus A).$$

Subtract $\Pr(B \cap A)$ on both sides. **(e)** Write

$$A \cup B = (A \setminus B) \cup (B \setminus A) \cup (A \cap B),$$

and note that all three events on the right are disjoint. Use Def. 1.2(iii) and (d),

$$\begin{aligned} \Pr(A \cup B) &= \Pr(A \setminus B) + \Pr(B \setminus A) + \Pr(A \cap B) \\ &= \{\Pr(A) - \Pr(A \cap B)\} + \{\Pr(B) - \Pr(A \cap B)\} + \Pr(A \cap B) \\ &= \Pr(A) + \Pr(B) - \Pr(A \cap B). \end{aligned}$$

(f) Assume that $A \subset B$. Then $A \cap B = A$, so

$$\Pr(B \setminus A) = \Pr(B) - \Pr(A \cap B) = \Pr(B) - \Pr(A),$$

but $\Pr(B \setminus A) \geq 0$, so $\Pr(A) \leq \Pr(B)$.

Solutions to Ex. 4. (a) From the definition of conditional probability, multiplying both sides by $\Pr(B)$ we get $\Pr(A \cap B) = \Pr(A | B)\Pr(B)$. Here we can think of zero divided by zero as zero, so that if $\Pr(B) = 0$ this relation is still valid. To see that this is ok, note that since $A \cap B \subset B$, so that $\Pr(B) = 0$ entails that $\Pr(A \cap B) = 0$ (see Ex. 3(f)). By the symmetry $\Pr(A \cap B) = \Pr(B \cap A)$, so $\Pr(A \cap B) = \Pr(B | A)\Pr(A)$.

(b) The events A_1, \dots, A_k are pairwise disjoint and their union equals the sample space Ω . Therefore (make a drawing!),

$$B = (B \cap A_1) \cup (B \cap A_2) \cup \dots \cup (B \cap A_{k-1}) \cup (B \cap A_k) = \bigcup_{i=1}^k (B \cap A_i)$$

Since the A_i s are pairwise disjoint, so are the $(B \cap A_i)$ s, that is $(B \cap A_i) \cap (B \cap A_j) = \emptyset$, whenever $i \neq j$. Using Def. 1.2(iii) and (a) we then get

$$\Pr(B) = \Pr\left(\bigcup_{i=1}^k (B \cap A_i)\right) = \sum_{i=1}^k \Pr(B \cap A_i) = \sum_{i=1}^k \Pr(B | A_i) \Pr(A_i).$$

(c) Combining the definition of conditional probability with (a), we get Bayes theorem.

$$\Pr(A | B) = \frac{\Pr(A \cap B)}{\Pr(B)} = \frac{\Pr(B | A) \Pr(A)}{\Pr(B)}.$$

(d) Combine (b) and (c).

(e) Define $R = \{\text{car is red}\}$ and $W = \{\text{witness says car was red}\}$. We are being told that $\Pr(R) = 1/10^3$. The witness can say the car is red when it is (the event R), and when it is not (the event R^c), therefore,

$$\begin{aligned} \Pr(W) &= \Pr(W | R) \Pr(R) + \Pr(W | R^c) \Pr(R^c) \\ &= \Pr(W | R) \Pr(R) + \Pr(W | R^c) \{1 - \Pr(R^c)\}. \end{aligned}$$

We also know from the psychologist that $\Pr(W | R) = 1$, and that $\Pr(W | R^c) = 5/100$. Use Bayes theorem,

$$\begin{aligned} \Pr(R | W) &= \frac{\Pr(W | R) \Pr(R)}{\Pr(W | R) \Pr(R) + \Pr(W | R^c) \Pr(R^c)} \\ &= \frac{1/10^3}{1/10^3 + 5/100 \times 999/1000} = \frac{100}{5095} \approx 0.02. \end{aligned}$$

Solution to Ex. 5. (a)

$$\sum_{n=1}^6 2^n = 2 + 2^2 + 2^3 + 2^4 + 2^5 + 2^6 = 2 + 4 + 8 + 16 + 32 + 64 = 126.$$

(b)

$$\begin{aligned} \sum_{k=1}^5 (3k - 2) &= (3 \times 1 - 2) + (3 \times 2 - 2) + (3 \times 3 - 2) + (3 \times 4 - 2) + (3 \times 5 - 2) \\ &= 1 + 4 + 7 + 10 + 13 = 35. \end{aligned}$$

(c)

$$\begin{aligned} \sum_{n=0}^5 2/(n+1) &= 2/(0+1) + 2/(1+1) + 2/(2+1) + 2/(3+1) + 2/(4+1) + 2/(5+1) \\ &= 2 + 1 + 2/3 + 1/2 + 2/5 + 1/3 = 4 + 9/10 = 4.9. \end{aligned}$$

(d) For the sum $\sum_{i=1}^3 (\sum_{j=1}^3 2^{i+j})$ notice that for $i = 1, 2, 3$, the inner sum in

$$\sum_{j=1}^3 2^{i+j} = 2^{i+1} + 2^{i+2} + 2^{i+3} = 2^i (2 + 2^2 + 2^3) = 2^i (2 + 4 + 8) = 2^i \times 14.$$

Therefore,

$$\begin{aligned} \sum_{i=1}^3 \left(\sum_{j=1}^3 2^{i+j} \right) &= \sum_{i=1}^3 2^i 14 = 2^1 + 2^2 14 + 2^3 14 = \\ &= (2 + 2^2 + 2^3) 14 = (14) 14 = 196. \end{aligned}$$

Solution to Ex. 6. To show that the equality is not true, it suffices to find one counterexample. Take $n = 2$, and suppose that $X_1 = 1$ and $X_2 = -1$. Then

$$\frac{1}{n} \sum_{i=1}^n X_i^2 = \frac{1}{2}(1 + 1) = 1 \neq \left(\frac{1}{n} \sum_{i=1}^n X_i \right)^2 = \left(\frac{1}{2}(1 - 1) \right)^2 = 0.$$

Solution to Ex. 7. (a)

$$\begin{aligned} \sum_{i=1}^n (X_i + Y_i) &= (X_1 + Y_1) + \cdots + (X_n + Y_n) \\ &= X_1 + \cdots + X_n + Y_1 + \cdots + Y_n = \sum_{i=1}^n X_i + \sum_{i=1}^n Y_i. \end{aligned}$$

(b) When a is a constant,

$$\sum_{i=1}^n aX_i = aX_1 + \cdots + aX_n = a(X_1 + \cdots + X_n) = a \sum_{i=1}^n X_i.$$

(c) In (a), set $Y_i = a$ for $i = 1, \dots, n$, then that exercise tells us that

$$\sum_{i=1}^n (X_i + a) = \sum_{i=1}^n X_i + \sum_{i=1}^n a = \sum_{i=1}^n X_i + na.$$

(d) In (b) set $a = 1/b$ for $b \neq 0$, then $\sum_{i=1}^n X_i/b = (1/b) \sum_{i=1}^n X_i$.

Solution to Ex. 8. (a) Combining Ex. 7(a) and (b), we get

$$\sum_{i=1}^n (aX_i + bY_i) \stackrel{\text{Ex. 7(a)}}{=} \sum_{i=1}^n aX_i + \sum_{i=1}^n bY_i \stackrel{\text{Ex. 7(b)}}{=} a \sum_{i=1}^n X_i + b \sum_{i=1}^n Y_i.$$

(b) $A_i = aX_i$ for $i = 1, \dots, n$, and $\bar{A}_n = (1/n) \sum_{i=1}^n A_i$. Then

$$\bar{A}_n = \frac{1}{n} \sum_{i=1}^n A_i = \frac{1}{n} \sum_{i=1}^n aX_i \stackrel{\text{Ex. 7(b)}}{=} \frac{a}{n} \sum_{i=1}^n X_i = a\bar{X}_n,$$

where $\bar{X}_n = (1/n) \sum_{i=1}^n X_i$.

(c) $B_i = X_i + Y_i$ for $i = 1, \dots, n$. Then

$$\bar{B}_n = \frac{1}{n} \sum_{i=1}^n B_i = \frac{1}{n} \sum_{i=1}^n (X_i + Y_i) \stackrel{\text{Ex. 7(a)}}{=} \frac{1}{n} \sum_{i=1}^n X_i + \frac{1}{n} \sum_{i=1}^n Y_i = \bar{X}_n + \bar{Y}_n,$$

where $\bar{X}_n = (1/n) \sum_{i=1}^n X_i$ and $\bar{Y}_n = (1/n) \sum_{i=1}^n Y_i$.

(d) $C_i = aX_i + bY_i$ for $i = 1, \dots, n$. Then

$$\bar{C}_n = \frac{1}{n} \sum_{i=1}^n C_i = \frac{1}{n} \sum_{i=1}^n (aX_i + bY_i) \stackrel{\text{Ex. 8(a)}}{=} \frac{a}{n} \sum_{i=1}^n X_i + \frac{b}{n} \sum_{i=1}^n Y_i = a\bar{X}_n + b\bar{Y}_n.$$

Solution to Ex. 9. (a) Let X_1, \dots, X_n be numbers, obs., rv's.

(a) We define $\tilde{X}_i = X_i - \bar{X}_n$ for $i = 1, \dots, n$, where $\bar{X}_n = (1/n) \sum_{i=1}^n X_i$ as always. The average of $\tilde{X}_1, \dots, \tilde{X}_n$ is

$$\frac{1}{n} \sum_{i=1}^n \tilde{X}_i = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n) \stackrel{\text{Ex. 7(c)}}{=} \frac{1}{n} \sum_{i=1}^n X_i - \frac{n}{n} \bar{X}_n = \bar{X}_n - \bar{X}_n = 0.$$

(b) Now set

$$\tilde{X}_i = \frac{X_i - \bar{X}_n}{s_X}, \quad \text{for } i = 1, \dots, n,$$

where $s_X = \sqrt{s_X^2}$, and

$$s_X^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2,$$

is the empirical variance. The empirical average of $\tilde{X}_1, \dots, \tilde{X}_n$ is zero:

$$\frac{1}{n} \sum_{i=1}^n \tilde{X}_i = \frac{1}{n} \sum_{i=1}^n \frac{X_i - \bar{X}_n}{s_X} \stackrel{\text{Ex. 7(d)}}{=} \frac{1}{s_X} \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n) \stackrel{\text{Ex. 9(a)}}{=} 0,$$

because $1/s_X$ is a constant.

The empirical variance $s_{\tilde{X}}^2$ of $\tilde{X}_1, \dots, \tilde{X}_n$ is one:

$$\begin{aligned} s_{\tilde{X}}^2 &= \frac{1}{n-1} \sum_{i=1}^n \left(\tilde{X}_i - \frac{1}{n} \sum_{i=1}^n \tilde{X}_i \right)^2 = \frac{1}{n-1} \sum_{i=1}^n \tilde{X}_i^2 \\ &= \frac{1}{n-1} \sum_{i=1}^n \left(\frac{X_i - \bar{X}_n}{s_X} \right)^2 = \frac{1}{s_X^2} \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2 = \frac{s_X^2}{s_X^2} = 1, \end{aligned}$$

so the empirical standard deviation $s_{\tilde{X}} = \sqrt{s_{\tilde{X}}^2}$ must also be one.

(c) Now set $\tilde{X}_i = aX_i + b$ for $i = 1, \dots, n$. Then the empirical average of $\tilde{X}_1, \dots, \tilde{X}_n$ is

$$\frac{1}{n} \sum_{i=1}^n \tilde{X}_i = \frac{1}{n} \sum_{i=1}^n (aX_i + b) \stackrel{\text{Ex. 7(c)}}{=} \frac{a}{n} \sum_{i=1}^n X_i + b = a\bar{X}_n + b.$$

And

$$\begin{aligned} s_{\tilde{X}}^2 &= \frac{1}{n-1} \sum_{i=1}^n \{aX_i + b - (a\bar{X}_n + b)\}^2 = \frac{1}{n-1} \sum_{i=1}^n (aX_i - a\bar{X}_n)^2 \\ &\stackrel{\text{Ex. 7(b)}}{=} \frac{a^2}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2 = a^2 s_X^2. \end{aligned}$$

The result in (a) is the special case where $a = 1$ and $b = -\bar{X}_n$. The result in (b) is the special case where $a = 1/s_X$ and $b = -\bar{X}_n/s_X$.

Solution to Ex. 10. We have numbers, observations, rv's X_1, \dots, X_n and Y_1, \dots, Y_n . The empirical covariance is defined as

$$s_{X,Y} = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)(Y_i - \bar{Y}_n).$$

(a)

$$\begin{aligned} s_{X,Y} &= \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)(Y_i - \bar{Y}_n) = \frac{1}{n-1} \sum_{i=1}^n \{(X_i - \bar{X}_n)Y_i - (X_i - \bar{X}_n)\bar{Y}_n\} \\ &\stackrel{\text{Ex. 7(a)}}{=} \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)Y_i - \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)\bar{Y}_n \\ &\stackrel{\text{Ex. 7(b)}}{=} \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)Y_i - \frac{\bar{Y}_n}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n) = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)Y_i. \end{aligned}$$

because $\sum_{i=1}^n (X_i - \bar{X}_n) = 0$ as shown in Ex. 9(a). By the same argument we see that $s_{X,Y} = (1/(n-1)) \sum_{i=1}^n (Y_i - \bar{Y}_n)X_i$.

(b) From (a) and using that $s_{X,X} = s_X^2$, we get

$$\begin{aligned} s_X^2 &= \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)X_i = \frac{1}{n-1} \sum_{i=1}^n (X_i^2 - \bar{X}_n X_i) \\ &\stackrel{\text{Ex. 7(a)}}{=} \frac{1}{n-1} \sum_{i=1}^n X_i^2 - \frac{1}{n-1} \sum_{i=1}^n \bar{X}_n X_i \stackrel{\text{Ex. 7(b)}}{=} \frac{1}{n-1} \sum_{i=1}^n X_i^2 - \bar{X}_n \frac{1}{n-1} \sum_{i=1}^n X_i \\ &= \frac{1}{n-1} \sum_{i=1}^n X_i^2 - \frac{n}{n-1} \bar{X}_n^2, \end{aligned}$$

for the last equality using that since

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i, \quad \text{we have} \quad n\bar{X}_n = \sum_{i=1}^n X_i.$$

Solution to Ex. 11. For $a < b$, define $A = \{X \leq a\}$ and $B = \{a < X \leq b\}$. (a) Then $A \cap B = \emptyset$, so they are disjoint, and (b) $A \cup B = \{X \leq b\}$. (c). Use Def. 1.2(iii)

$$\Pr(X \leq b) = \Pr(A \cup B) = \Pr(A) + \Pr(B) = \Pr(X \leq a) + \Pr(a < X \leq b),$$

and subtract $\Pr(X \leq a)$ on both sides to get $\Pr(a < X \leq b) = \Pr(X \leq b) - \Pr(X \leq a)$.

You could also argue like this: When $a < b$,

$$\{a < X \leq b\} = \{X \leq b\} \setminus \{X \leq a\} = \{X \leq b\} \cap \{X > a\}^c,$$

and $\{X \leq b\} \cap \{X \leq a\} = \{X \leq a\}$, so by Ex. 3(d),

$$\Pr(a < X \leq b) = \Pr(\{X \leq b\} \setminus \{X \leq a\}) = \Pr(X \leq b) - \Pr(X \leq a).$$