

Chapters 5.6-5.10 & 6.4 of *Probability, Statistics, and Random Processes* by A. Leon-Garcia

1. Let X and Y be independent random variables that are uniformly distributed in the interval $[0, 1]$.

(a) Find the pdf of $A = X + Y$.

Solution: Let $F_U(u)$ and $f_U(u)$ be the cdf and pdf of the uniform random variable. Let $F_A(a)$ and $f_A(a)$ be the cdf and pdf of A . As such,

$$\begin{aligned} F_A(a) &= P(X + Y \leq a) \\ &= \int_{-\infty}^{\infty} P(X + y \leq a | Y = y) f_U(y) dy \\ &= \int_{-\infty}^{\infty} P(X \leq a - y | Y = y) f_U(y) dy \\ &= \int_{-\infty}^{\infty} F_U(a - y) f_U(y) dy \end{aligned}$$

By taking the derivative with respect to a , we get

$$\begin{aligned} f_A(a) &= \frac{dF_A(a)}{da} \\ &= \int_{-\infty}^{\infty} f_U(a - y) f_U(y) dy \end{aligned}$$

which is the convolution of the two probability distributions. We note that

$$f_U(u) = \begin{cases} 1 & 0 \leq u \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

and

$$f_U(a - u) = \begin{cases} 1 & 0 \leq a - u \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

Therefore, the only region that $f_U(a - y)f_U(y)$ is non-zero is $\max(a - 1, 0) \leq y \leq \min(a, 1)$. From here, we get

$$f_A(a) = \begin{cases} a & 0 \leq a \leq 1 \\ 2 - a & 1 \leq a \leq 2 \\ 0 & \text{otherwise.} \end{cases}$$

(b) Find the pdf of $B = X - Y$.

Solution: We get a similar solution to part (a) except that the pdf of B is

$$f_B(b) = \int_{-\infty}^{\infty} f_U(b+y)f_U(y)dy.$$

As such, we get

$$f_B(b) = \begin{cases} 1+b & -1 \leq b \leq 0 \\ 1-b & 0 \leq b \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

(c) Find the pdf of $C = XY$.

Solution: As in part (a), we calculate the cdf of C to find the pdf. Let $F_C(c)$ and $f_C(c)$ be the cdf and pdf of C .

$$\begin{aligned} F_C(c) &= P(XY \leq c) \\ &= \int_{-\infty}^{\infty} P(Xy \leq c|Y=y)f_U(y)dy \\ &= \int_{-\infty}^{\infty} P(X \leq \frac{c}{y}|Y=y)f_U(y)dy \\ &= \int_{-\infty}^{\infty} F_U(\frac{c}{y})f_U(y)dy \end{aligned}$$

We note that the third step is only valid because Y is non-negative so multiplying both sides by y does not change the sign.

By taking the derivative with respect to c , we get

$$\begin{aligned} f_C(c) &= \frac{dF_C(c)}{dc} \\ &= \int_{-\infty}^{\infty} f_U(\frac{c}{y})f_U(y)\frac{1}{y}dy \end{aligned}$$

Consider the case when $0 \leq c \leq 1$ since all other cases must have zero for the pdf. As such,

$$\begin{aligned} f_C(c) &= \int_{-\infty}^{\infty} f_U(\frac{c}{y})f_U(y)\frac{1}{y}dy \\ &= \int_c^1 \frac{1}{y}dy \\ &= \ln(y)|_c^1 = \ln(1) - \ln(c) = -\ln(c) \end{aligned}$$

Hence, we get

$$f_C(c) = \begin{cases} -\ln(c) & 0 < c \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

Another way to get this result is to calculate the cdf first and then differentiate. Again, consider the case when $0 \leq c \leq 1$ which results in

$$\begin{aligned} F_C(c) &= \int_{-\infty}^{\infty} F_U\left(\frac{c}{y}\right) f_U(y) dy \\ &= \int_0^c dy + \int_c^1 \frac{c}{y} dy \\ &= c - c \ln(c). \end{aligned}$$

By taking the derivative with respect to c , we get the same result.

- (d) Find the covariance of A and B .

Solution:

$$\begin{aligned} \text{Cov}(A, B) &= \mathbb{E}[AB] - \mathbb{E}[A]\mathbb{E}[B] \\ &= \mathbb{E}[(X + Y)(X - Y)] - \mathbb{E}[(X + Y)]\mathbb{E}[(X - Y)] \\ &= \mathbb{E}[X^2 - Y^2] - \mathbb{E}[(X + Y)]\mathbb{E}[(X - Y)] \\ &= 0 \end{aligned}$$

- (e) Find the covariance of A and C .

Solution:

$$\begin{aligned} \text{Cov}(A, C) &= \mathbb{E}[AC] - \mathbb{E}[A]\mathbb{E}[C] \\ &= \mathbb{E}[(X + Y)XY] - \mathbb{E}[(X + Y)]\mathbb{E}[XY] \\ &= \mathbb{E}[X^2Y + XY^2] - \mathbb{E}[(X + Y)]\mathbb{E}[XY] \\ &= \mathbb{E}[X^2Y] + \mathbb{E}[XY^2] - \mathbb{E}[(X + Y)]\mathbb{E}[XY] \\ &= \mathbb{E}[X^2]\mathbb{E}[Y] + \mathbb{E}[X]\mathbb{E}[Y^2] - \mathbb{E}[X]^2\mathbb{E}[Y] - \mathbb{E}[Y]^2\mathbb{E}[X] \\ &= \mathbb{E}[Y] \cdot (\mathbb{E}[X^2] - \mathbb{E}[X]^2) + \mathbb{E}[X] \cdot (\mathbb{E}[Y^2] - \mathbb{E}[Y]^2) \\ &= \mathbb{E}[Y] \cdot \text{Var}(X) + \mathbb{E}[X] \cdot \text{Var}(Y) \\ &= \frac{1}{2} \cdot \frac{1}{12} + \frac{1}{2} \cdot \frac{1}{12} = \frac{1}{12} \end{aligned}$$

2. Let X be a Gaussian random variable with mean 0 and variance $\sigma^2 = 1$. We define three new random variables as $Y = aX + b$, $Z = X^2$, and $W = X^3$. We require that $a \neq 0$. For reference, $\mathbb{E}[X^3] = 0$, $\mathbb{E}[X^4] = 3\sigma^4$, $\mathbb{E}[X^5] = 0$, and $\mathbb{E}[X^6] = 15\sigma^6$.

- (a) Find the correlation coefficients for each pair (X, Y) , (X, Z) , and (X, W) .

Solution: First, we calculate the expectation of each random variable which are $\mathbb{E}[X] = 0 = \mathbb{E}[Y] = \mathbb{E}[W]$ and $\mathbb{E}[Z] = \sigma^2$.

Next, we calculate the variances of each random variable:

$$\begin{aligned}
 Var(X) &= \sigma^2 = 1 \\
 Var(Y) &= Var(aX + b) = a^2 Var(X) = a^2 \\
 Var(Z) &= Var(X^2) = \mathbb{E}[X^4] - \mathbb{E}[X^2]^2 \\
 &= 3\sigma^4 - \sigma^4 = 2\sigma^4 = 2 \\
 Var(W) &= Var(X^3) = \mathbb{E}[X^6] - \mathbb{E}[X^3]^2 \\
 &= 15\sigma^6 = 15.
 \end{aligned}$$

Now, we calculate the covariance of X to all the other terms.

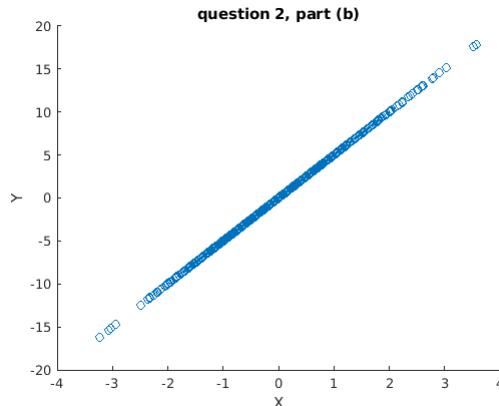
$$\begin{aligned}
 Cov(X, Y) &= \mathbb{E}[X(aX + b)] - \mathbb{E}[aX + b]\mathbb{E}[X] \\
 &= a\mathbb{E}[X^2] - a\mathbb{E}[X]^2 = aVar(X) = a \\
 Cov(X, Z) &= \mathbb{E}[XX^2] - \mathbb{E}[X]\mathbb{E}[X^2] \\
 &= \mathbb{E}[X^3] - \mathbb{E}[X]\mathbb{E}[X^2] = 0 \\
 Cov(X, W) &= \mathbb{E}[XX^3] - \mathbb{E}[X]\mathbb{E}[X^3] \\
 &= \mathbb{E}[X^4] - \mathbb{E}[X]\mathbb{E}[X^3] = 3\sigma^4 = 3.
 \end{aligned}$$

Finally, we can calculate the correlation coefficients:

$$\begin{aligned}
 \rho_{X,Y} &= \frac{Cov(X, Y)}{\sqrt{Var(X)}\sqrt{Var(Y)}} \\
 &= \frac{a}{\sqrt{a^2}} = \frac{a}{|a|} = \begin{cases} 1 & a > 0 \\ -1 & a < 0 \end{cases} \\
 \rho_{X,Z} &= \frac{0}{\sqrt{2\sigma^4}} = 0 \\
 \rho_{X,W} &= \frac{3}{\sqrt{15}}
 \end{aligned}$$

- (b) For the case of $a = 5$ and $b = 0$, randomly sample X 1000 times, use each X to get Y , and use a scatter plot to plot (X, Y) where the values of X is along the x-axis and the values of Y are along the y-axis. Describe how the correlation coefficient relates to what you see in the scatter plot.

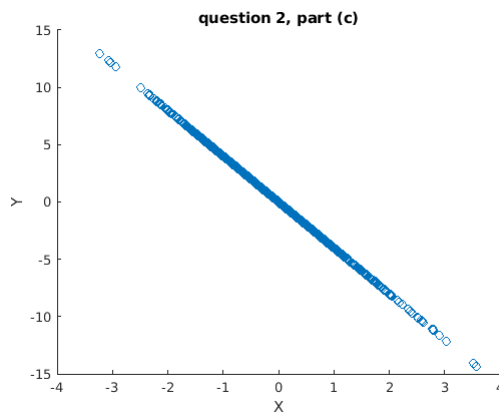
Solution:



We can see that having a positive correlation coefficient implies that X and Y are positively correlated which means that when one increases, the other one is also likely to increase. Even more strongly, when the coefficient is 1, it means that X and Y actually have a linear relationship to each other. That is why you see a straight line in the figure.

- (c) For the case of $a = -4$ and $b = 0$, randomly sample X 1000 times, use each X to get Y , and use a scatter plot to plot (X, Y) where the values of X is along the x-axis and the values of Y are along the y-axis. Describe how the correlation coefficient relates to what you see in the scatter plot and explain the difference to part (b).

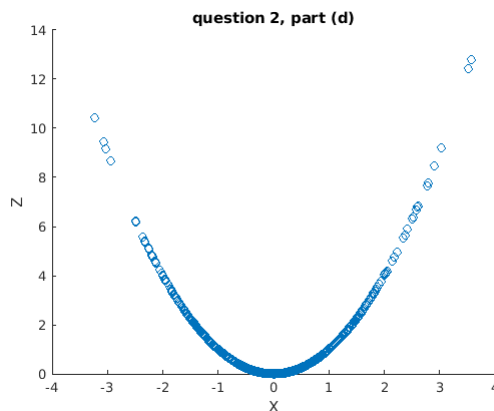
Solution:



Similarly, we can see that having a negative correlation coefficient implies that X and Y are negatively correlated which means that when one increases, the other one is likely to decrease. Additionally, we see that when the coefficient is -1 , X and Y also follow a linear relationship to each other only negatively this time.

- (d) Now, randomly sample X 1000 times, use each X to get Z , and use a scatter plot to plot (X, Z) where the values of X is along the x-axis and the values of Z are along the y-axis. Describe how the correlation coefficient relates to what you see in the scatter plot and explain the difference to parts (b) and (c).

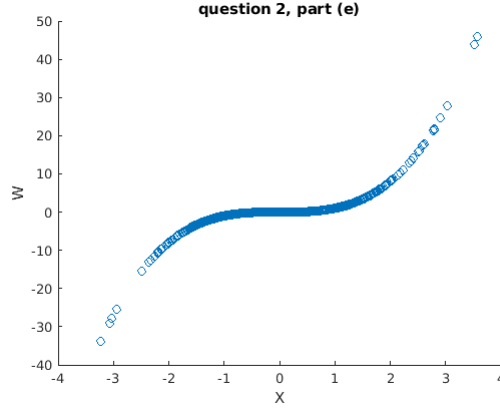
Solution:



Now, we can see that the plot has Z both decreasing and increasing as X increases. This is what is implied by the correlation coefficient being 0. As such, when the coefficient is 0, Z can, on average, either decrease or increase as X increases.

- (e) Now, randomly sample X 1000 times, use each X to get W , and use a scatter plot to plot (X, W) where the values of X is along the x-axis and the values of W are along the y-axis. Describe how the correlation coefficient relates to what you see in the scatter plot.

Solution:



Here we see that the correlation coefficient is positive but not 1. As such, we can clearly see a positive trend since W increases as X increases. Yet, the trend is clearly not linear which is why the correlation coefficient is not 1.

3. Let X and Y be two random variables with identical distributions. These two random variables are not necessarily independent. Answer the following questions given that $C = aX + bY$ and $D = aX - bY$.

- (a) Find $COV[C, D]$ in terms of the variances and covariances of X and Y

Solution:

$$\begin{aligned}
 COV[C, D] &= COV[aX + bY, aX - bY] \\
 &= E[(aX + bY)(aX - bY)] - E[aX + bY]E[aX - bY] \\
 &= E[a^2X^2 - b^2Y^2] - (aE[X] + bE[Y])(aE[X] - bE[Y]) \\
 &= a^2E[X^2] - b^2E[Y^2] - a^2E[X]^2 + b^2E[Y]^2 \\
 &= a^2VAR[X] - b^2VAR[Y] \\
 &= (a^2 - b^2) * VAR[X]
 \end{aligned}$$

- (b) Find the relation between a and b if the random variables C and D are independent

Solution:

Since it is given that C and D are independent, therefore, $COV[C, D] = 0$.

$$\begin{aligned}
 COV[C, D] &= (a^2 - b^2) * VAR[X] \\
 0 &= (a^2 - b^2) * VAR[X]
 \end{aligned}$$

Since $\text{Var}(X) > 0$ (since X is a non-constant random variable), we must have either $a + b = 0$ or $a - b = 0$. If $a + b = 0$, then $a = -b$, and if $a - b = 0$, then $a = b$. Therefore, $a = \pm b$

4. Answer the following. **Show all your work.**

- (a) Consider two random variables X and Y . Prove that the correlation coefficient $\rho_{X,Y}$ satisfies $-1 \leq \rho_{X,Y} \leq 1$.

Solution:

$$\begin{aligned} 0 &\leq \mathbb{E}\left[\left(\frac{X - \mathbb{E}[X]}{\sigma_X} \pm \frac{Y - \mathbb{E}[Y]}{\sigma_Y}\right)^2\right] \\ &= \mathbb{E}\left[\left(\frac{X - \mathbb{E}[X]}{\sigma_X}\right)^2\right] \pm 2\mathbb{E}\left[\frac{(X - \mathbb{E}[X])(Y - \mathbb{E}[Y])}{\sigma_X \sigma_Y}\right] + \mathbb{E}\left[\left(\frac{Y - \mathbb{E}[Y]}{\sigma_Y}\right)^2\right] \\ &= 1 \pm 2\rho_{X,Y} + 1 = 2(1 \pm \rho_{X,Y}) \\ &\implies -1 \leq \rho_{X,Y} \leq 1 \end{aligned}$$

- (b) Let X be a random variable, and Y be another random variable given by $Y = aX + b$. What is the correlation coefficient between X and Y . Does the answer depend on the sign of a ?

Solution:

The correlation coefficient is given by:

$$\rho_{X,Y} = \frac{E[XY] - E[X]E[Y]}{\sigma_X \sigma_Y}.$$

We have that,

$$\begin{aligned} E[XY] &= E[aX^2 + bX] = aE[X^2] + bE[X], \\ E[X]E[Y] &= E[X](aE[X] + b) = aE[X]^2 + bE[X], \\ \text{COV}[X, Y] &= E[XY] - E[X]E[Y] = aE[X^2] + bE[X] - aE[X]^2 - bE[X] = a\text{VAR}[X], \end{aligned}$$

Next, $\sigma_Y = \sqrt{\text{VAR}[aX + b]} = \sqrt{a^2 \text{VAR}[X]} = a\sigma_X$, if $a > 0$ and $-a\sigma_X$, if $a < 0$. and,

$$\sigma_X \sigma_Y = \pm a \sigma_X^2 = \pm a \text{VAR}[X].$$

Thus, if $\text{VAR}[X] \neq 0$,

$$\rho_{X,Y} = \begin{cases} \frac{a\text{Var}[X]}{a\text{Var}[X]} = 1 & \text{if } a > 0 \\ \frac{a\text{Var}[X]}{-a\text{Var}[X]} = -1 & \text{if } a < 0 \end{cases}.$$

Clearly, the sign of a does matter.

5. Let X and Y be jointly Gaussian random variables with $\mathbb{E}[Y] = 0$, $\sigma_X = 4$, $\sigma_Y = 3$ and $\mathbb{E}[X|Y] = \frac{4Y}{9} + 2$. Find the joint pdf of X and Y .

Solution:

Let m_x and m_y be the means of X and Y . By the definition of marginal distributions for jointly Gaussian random variables, we know that

$$\mathbb{E}[X|Y] = \rho_{X,Y} \frac{\sigma_X}{\sigma_Y} (y - m_y) + m_x.$$

As such,

$$\mathbb{E}[X|Y] = \frac{4Y}{9} + 2 \implies m_x = 2, \rho_{X,Y} = \frac{1}{3}.$$

Now we have all the parameters to find the joint pdf of X and Y which is

$$\begin{aligned} f_{X,Y}(x,y) &= \frac{\exp\left[\frac{1}{2(1-\rho_{X,Y}^2)}\left(\left(\frac{x-m_x}{\sigma_X}\right)^2 - 2\rho_{X,Y}\left(\frac{x-m_x}{\sigma_X}\right)\left(\frac{y-m_y}{\sigma_Y}\right) + \left(\frac{y-m_y}{\sigma_Y}\right)^2\right)\right]}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho_{X,Y}^2}} \\ &= \frac{\exp\left[\frac{1}{2(\frac{8}{9})}\left(\left(\frac{x-2}{4}\right)^2 - \frac{(x-2)(y)}{18} + \left(\frac{y}{3}\right)^2\right)\right]}{2\pi(12)\sqrt{\frac{8}{9}}} \\ &= \frac{\exp\left[\frac{9}{16}\left(\left(\frac{x-2}{4}\right)^2 - \frac{(x-2)y}{18} + \left(\frac{y}{3}\right)^2\right)\right]}{16\pi\sqrt{2}} \end{aligned}$$

6. Q3 from HW6

Let X and Y be two jointly continuous random variables with joint pdf

$$f_{XY}(x,y) = \begin{cases} 6xy, & 0 \leq x \leq 1, 0 \leq y \leq \sqrt{x}, \\ 0, & \text{otherwise,} \end{cases}$$

(a) Find $f_X(x)$.

Solution:

For $0 \leq x \leq 1$,

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{XY}(x,y) dy \\ &= \int_0^{\sqrt{x}} 6xy dy \\ &= 3x^2 \end{aligned}$$

Thus

$$f_X(x) = \begin{cases} 3x^2, & 0 \leq x \leq 1 \\ 0, & \text{otherwise.} \end{cases}$$

(b) Find the conditional pdf of X given $Y = y$, $f_{X|Y}(x|y)$.

Solution:

For $0 \leq y \leq 1$,

$$\begin{aligned} f_Y(y) &= \int_{-\infty}^{\infty} f_{XY}(x, y) dx \\ &= \int_{y^2}^1 6xy dx \\ &= 3y(1 - y^4) \end{aligned}$$

Thus

$$\begin{aligned} f_Y(y) &= \begin{cases} 3y(1 - y^4), & 0 \leq y \leq 1 \\ 0, & \text{otherwise,} \end{cases} \\ f_{X|Y}(x|y) &= \frac{f_{XY}(x, y)}{f_Y(y)} \\ &= \begin{cases} \frac{2x}{1-y^4} & y^2 \leq x \leq 1 \\ 0, & \text{otherwise.} \end{cases} \end{aligned}$$

(c) Find $E[X|Y = y]$, for $0 \leq y < 1$. What is $E[X|Y]$?

Solution:

For $0 \leq y < 1$

$$\begin{aligned} E[X|Y = y] &= \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx \\ &= \int_{y^2}^1 x \frac{2x}{1-y^4} dx \\ &= \frac{2(1-y^6)}{3(1-y^4)} \end{aligned}$$

$$E[X|Y] = \frac{2(1-Y^6)}{3(1-Y^4)}$$

(d) Let A be the event $\{X \geq \frac{1}{2}\}$. Find $P[A]$, $f_{X|A}(x)$, and $E[X|A]$.

Solution:

$$P(A) = \int_A f_X(x) dx = \int_{\frac{1}{2}}^1 3x^2 dx = \frac{7}{8}$$

$$f_{X|A}(x|A) = \begin{cases} \frac{f_X(x)}{P[A]} & \text{if } x \in A \\ 0 & \text{otherwise} \end{cases}$$

$$f_{X|A}(x|A) = \begin{cases} \frac{24x^2}{7} & \text{if } \frac{1}{2} \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Therefore,

$$E[X|A] = \int_{-\infty}^{\infty} x f_{X|A}(x|A) dx = \int_{\frac{1}{2}}^1 x \frac{24x^2}{7} dx = \left[\frac{6x^4}{7} \right]_{\frac{1}{2}}^1 = \frac{45}{56}$$

Appendix: Reference code for Q2

```
(a) %% Part (b)
2 X = randn(1000,1);
3 a = 5;
4 Y = a*X;
5 figure()
6 scatter(X,Y)
7 xlabel("X")
8 ylabel("Y")
9 title("question 2, part (b)")
10 %% Part (c)
11 X = randn(1000,1);
12 a = -4;
13 Y = a*X;
14 figure()
15 scatter(X,Y)
16 xlabel("X")
17 ylabel("Y")
18 title("question 2, part (c)")
19
20 %% Part (d)
21 X = randn(10000,1);
22 Z = X.^2;
23 figure()
24 scatter(X,Z)
25 xlabel("X")
26 ylabel("Z")
27 title("question 2, part (d)")
28
29 %% Part (e)
30 X = randn(100,1);
31 W = X.^3;
32 figure()
33 scatter(X,W)
34 xlabel("X")
35 ylabel("W")
36 title("question 2, part (e)")
37
38 %%%
```