

Exercise 1

In the game of blackjack, the player is initially dealt two cards from a deck of ordinary playing cards. Without going into all the game's details, it is enough to know the best possible hand for a player to receive on the initial deal is a combination of an ace of any suit and any face card or ten. What is the probability that a player will be dealt this combination?

The number of ways a single player can be dealt a blackjack (assuming there is only one player) is $\binom{4}{1}\binom{16}{1} = 64$. So the number of events in our event space (which we'll call A) is 64. The total number of ways to deal 2 cards to that player is $\binom{52}{2} = 1326$. So, the number of events in the whole sample space is 1326. the probability of a single player being dealt a blackjack is therefore

$$\frac{N_A}{N} = \frac{\binom{4}{1}}{\binom{16}{1}} = \frac{64}{1326} = 0.048$$

Exercise 2

Discrete and random variables X and Y can each take on integer values 1, 3, and 5. The joint probability table of X and Y is given below.

Table 1: Your first table.

X	Y=1	Y=3	Y=5
1	1/18	1/18	1/18
3	1/18	1/18	1/6
5	1/18	1/6	1/3

Problem (a)

Are the random variables X and Y independent?

No, the table clearly shows that the probability of X taking on certain values changes depending on the value of Y . For instance $P(X = 3|Y = 3) \neq P(X = 3|Y = 5)$ If X and Y were independent then $P(X = x|Y = y)$ would be the same for all values of y .

Problem (b)

Find the unconditional (marginal) probability $P(Y = 5)$.

The marginal probability $P(Y = 5) = P(Y = 5|X = 1) + P(Y = 5|X = 3) + P(Y = 5|X = 5) = (1/18) + (1/6) + (1/3) = (5/9)$.

Problem (c)

What is the conditional probability $P(Y = 5|X = 3)$?

Generally $P(A|B) = \frac{P(A,B)}{P(B)}$. This means we need to calculate the marginal probability $P(X = 3) = (1/18) + (1/18) + (1/6) = (5/18)$. With this information we can find

$$P(Y = 5|X = 3) = \frac{P(Y = 5, X = 3)}{P(X = 3)} = \frac{1/6}{5/18} = \frac{3}{5}$$

Exercise 3

Determine the value of a in the function

$$f_X(x) = \begin{cases} ax(1-x) & x \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

so that $f_X(x)$ is a valid probability density function.

For $f_X(x)$ to be a valid probability density function $\int_{-\infty}^{\infty} f_X(x)dx$ must be equal to one.

$$\begin{aligned} 1 &= \int_{-\infty}^{\infty} f_X(x)dx \\ &= \int_0^1 ax(1-x)dx \\ &= a \int_0^1 (x - x^2)dx \\ &= a \left[\frac{x^2}{2} - \frac{x^3}{3} \right]_0^1 \\ &= a \left(\frac{1}{2} - \frac{1}{3} \right) \\ 1 &= \frac{a}{6} \\ a &= 6 \end{aligned}$$

Exercise 4

The probability density function of an exponentially distributed random variable is defined as follows

$$f_X(x) = \begin{cases} ae^{-ax} & x > 0 \\ 0 & x \leq 0 \end{cases}$$

where $a \geq 0$.

Problem (a)

Find the probability distribution function of an exponentially distributed random variable.

$$\begin{aligned}P(a \leq X \leq b) &= \int_a^b f_X(x) dx \\&= \int_a^b a e^{-ax} dx \\&= -e^{-ax} \Big|_a^b\end{aligned}$$

assuming $0 < a \leq b$.

Problem (b)

Find the mean of an exponentially distributed random variable.

$$\begin{aligned}E[X] &= \int_{-\infty}^{\infty} x f_X(x) dx \\&= a \int_0^{\infty} x e^{-ax} dx \\&= a \left[\left(-\frac{x}{a} - \frac{1}{a^2} \right) e^{-ax} \right]_0^{\infty} \\&= \left[\left(-x - \frac{1}{a} \right) e^{-ax} \right]_0^{\infty} \\&= \left(-\infty - \frac{1}{a} \right) e^{-a\infty} + \frac{1}{a} e^0\end{aligned}$$

Because $e^{-\infty} = 0$, $E[X] = \frac{1}{a}$.

Problem (c)

Find the second moment of an exponentially distributed random variable.

$$\begin{aligned}E[X^2] &= \int_{-\infty}^{\infty} x^2 f_X(x) dx \\&= \int_0^{\infty} x^2 a e^{-ax} dx \\&= \frac{1}{a^2} \int_0^{\infty} t^2 e^{-t} dt \\&= \frac{1}{a^2} \Gamma(2+1) = \frac{2!}{a^2} \\&= \frac{2}{a^2}\end{aligned}$$

Problem (d)

Find the variance of an exponentially distributed random variable.

$$\begin{aligned}\sigma_x^2 &= E[X^2] - (E[X])^2 \\ &= \frac{2}{a^2} - \left(\frac{1}{a}\right)^2 \\ &= \frac{1}{a^2}\end{aligned}$$

Problem (e)

What is the probability that an exponentially distributed random variable takes on a value within one standard deviation of its mean?

The standard deviation is $\sqrt{\frac{1}{a}} = \frac{1}{a}$, so we simply need to evaluate the probability distribution function from 0 to $\frac{2}{a}$:

$$\begin{aligned}P\left(0 < X \leq \frac{2}{a}\right) &= 1 - e^{-ax} \Big|_0^{2/a} \\ &= (1 - e^{-a \frac{2}{a}}) - (1 - e^0) \\ &= 1 - e^{-2}\end{aligned}$$

So the probability that X falls within one standard deviation of the mean is a constant, unrelated to the value of a .

Exercise 5

Generate $N = 50$ independent random numbers, each uniformly distributed between 0 and 1. Plot a histogram of the random numbers using 10 bins. What is the sample mean and standard deviation of the numbers that you generated? What would you expect to see for the mean and standard deviation? Repeat for $N = 500$ and $N = 5000$ random numbers. What changes do you see in the histogram as N increases?

Our uniformly distributed probability density function is defined as

$$f(x) = \begin{cases} c & x \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

Because $\int_{-\infty}^{\infty} f(x) = 1$, $c = 1$. The expected value of this pdf is

$$\begin{aligned} E[X] &= \int_{-\infty}^{\infty} x f(x) dx \\ &= \int_0^1 x dx \\ &= \frac{1}{2} x^2 \Big|_0^1 \\ &= \frac{1}{2} \end{aligned}$$

and the expected value of X^2 is

$$\begin{aligned} E[X^2] &= \int_{-\infty}^{\infty} x^2 f(x) dx \\ &= \int_0^1 x^2 dx \\ &= \frac{1}{3} x^3 \Big|_0^1 \\ &= \frac{1}{3} \end{aligned}$$

So the variance can be calculated as

$$\sigma^2 = E[X^2] - (E[X])^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}$$

which means the standard deviation is $\sqrt{\sigma^2} = 0.2887$.

With a sample size $N = 50$ we get $\mu = 0.4899$ and $\sigma = 0.2786$, which is fairly close to the theoretical values. See figure ??, below, for a histogram of the experimental values. From the histogram we can see that the actual distribution of values is not terribly uniform.

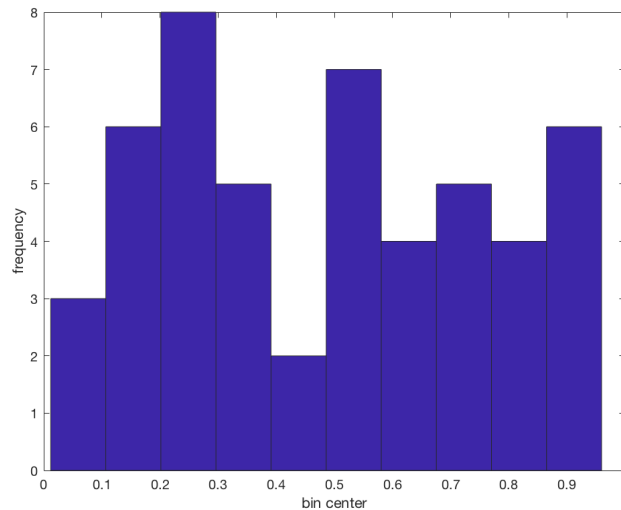


Figure 1: histogram for $N = 50$

With a sample size $N = 500$ we get $\mu = 0.4862$ and $\sigma = 0.2898$, which is fairly close to the theoretical values, though the μ value is actually further from the theoretical value than when $N = 50$. See figure ??, below, for a histogram of the experimental values. From the histogram we can see that the actual distribution of values is more uniform than in the previous case, though it still has distinct peaks that one would not expect with a truly uniform distribution.

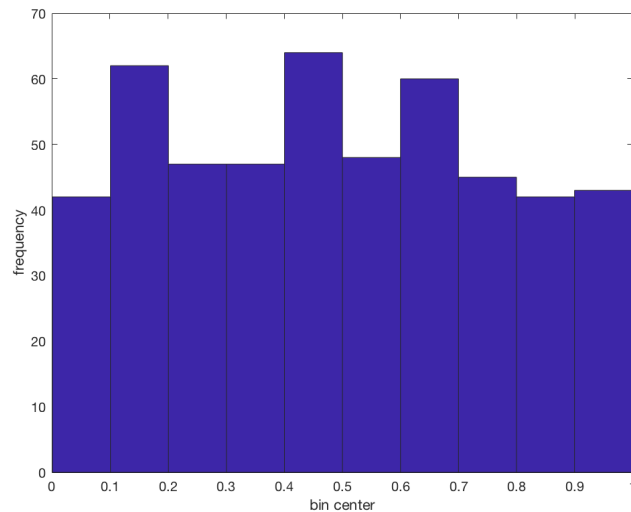


Figure 2: histogram for $N = 500$

With a sample size $N = 5000$ we get $\mu = 0.4999$ and $\sigma = 0.2892$, which is nearly the same as the theoretical values. See figure ??, below, for a histogram of the experimental values. From the histogram we can see that the actual distribution of values is very nearly uniform, with each of the 10 bins containing about 500 values.

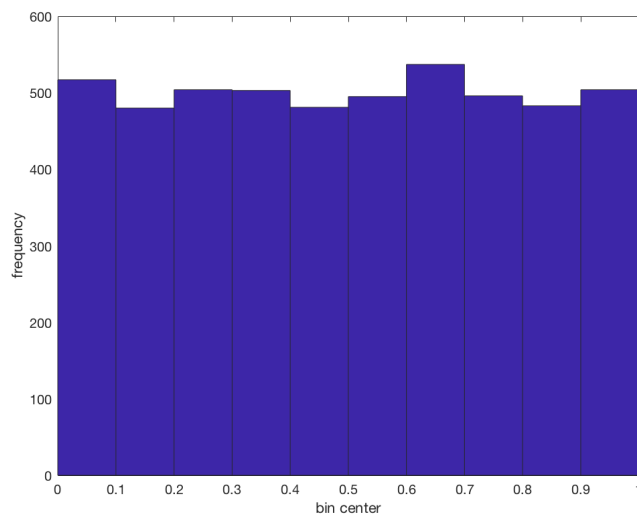


Figure 3: histogram for $N = 5000$

Exercise 6

AQ 1

AQ 2