

Written Assignment 06: Joint probability distributions

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Assignment Goal

The overarching goal of this assignment is to deepen your understanding of bivariate and multivariate joint probability distributions.

Instructions

Please type your response to a question right below the question text. Compile this document to generate PDF output. Upload your PDF document to MS Teams. In the preamble above, change V.N. Gudivada to your name.

Questions

Marginal Probability Density Function (PDF)

The random variables X and Y are uniformly distributed over the interior of a circle of radius 1 centered at the origin. Find the marginal probability density function $f_X(x)$.

Solution:

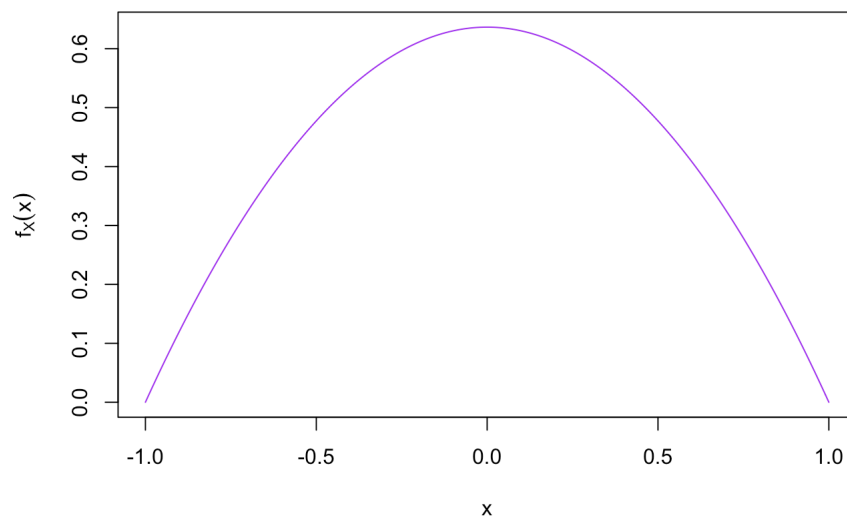
This question is about joint probability distribution. We have x and y axis. Create a circle and the origin on the $(0,0)$. First we should find the joint probability density function. If you think about the area of the circle, it is going to be πr^2 and because r is 1 so area of the circle would be π . Therefore the density is $\frac{1}{\text{area}=\pi}$.

$f(x, y) = \frac{1}{\pi}$ for all area $x^2 + y^2 < 1$. This is a uniform density and every point in that area of the circle should follow this $\text{area} \times \text{density} = 1$.

We need to find the marginal density, and for this we have to integrate out to get marginal density. $f_X(x) = \int \frac{dy}{\pi}$ and the upper and lower limit would be $-\sqrt{1-x^2}$ and $+\sqrt{1-x^2}$. The answer would be $\frac{2\sqrt{1-x^2}}{\pi}$ and for range $-1 < x < 1$. If you apply push in directions up and down. The marginal density of x would be -1 to 0 and 0 to 1 . The point of this question is if you try to plot this curve to special region.

```
library(latex2exp)
x <- seq(-1,1, by=0.01)
y <- (2*(1 - x^2)) / pi
plot(x,y, type="l", col="purple", main="Marginal Density of X", xlab=TeX(r'($x$)'), ylab=TeX(r'($f_X(x)$')))
```

Marginal Density of X



Joint Cumulative Distribution Function (CDF)

Consider the random variables X and Y with joint PDF

$$f(x, y) = 2, \quad 0 < x < y < 1$$

Find $F(x, y)$ for any point (x, y) in the support set \mathcal{A} . Use a calculus and a non-calculus (geometrical) approach to solve this problem.

Solution:

Here we need to find the cumulative distribution function. If we plot this special range, $0 < x < y < 1$ and the area where $x = y$, we will have $x < y$ and all other regions $x > y$. $f(x) = 2, 0 < x < y < 1$, to find CDF, we need to calculate $F(x, y) = P(X \leq x, Y \leq y)$. We are gonna use double integral. If we want to do algebra first: $P(X \leq x, Y \leq y) = x(y - x) + \frac{1}{2}x \times x = 2[\frac{1}{2}x^2 + x(y - x)]$,

$F(0, 0) = 1/2(0) + 0(0 - 0) = 0$ So entire area should be equal to 1. $F(1, 1) = 1/2(1) + 1(1 - 1) = 1/2$.

$\int \int 2dxdy = \int_0^x 2(y - x)dy = 2xy - x^2, F(0, 0) = 0, F(1, 1) = 2 - 1 = 1$ If you go back and look at the geometric distribution, we have it.

```
x <- seq(0,1, by=0.01)
length(x)
```

```
## [1] 101
```

```
library(plotly)
```

```
## Loading required package: ggplot2
```

```
##
## Attaching package: 'plotly'
```

```
## The following object is masked from 'package:ggplot2':
##
## last_plot
```

```
## The following object is masked from 'package:latex2exp':
##
## TeX
```

```
## The following object is masked from 'package:stats':
##
## filter
```

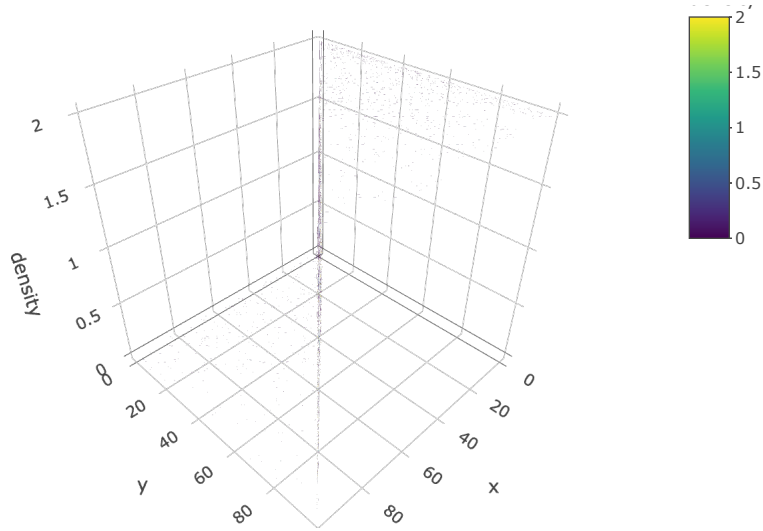
```
## The following object is masked from 'package:graphics':
##
## layout
```

```
x <- seq(0, 1, by=0.01)
# length(x)
y <- seq(0, 1, by=0.01)
# length(y)
density <- matrix(rep(0,10201), nrow=101, ncol=101)
# density
for(i in 1:(length(x)-1)){
  # print(i)
  for(j in (i+1):length(y)){
    density[j,i] <- 2
  }
}
print(density[2,3])
```

```
## [1] 0
```

```
biVarDensity <- plot_ly(z = ~density)
biVarDensity <- biVarDensity %>% add_surface()
biVarDensity
```

density



Joint Probability Density Function

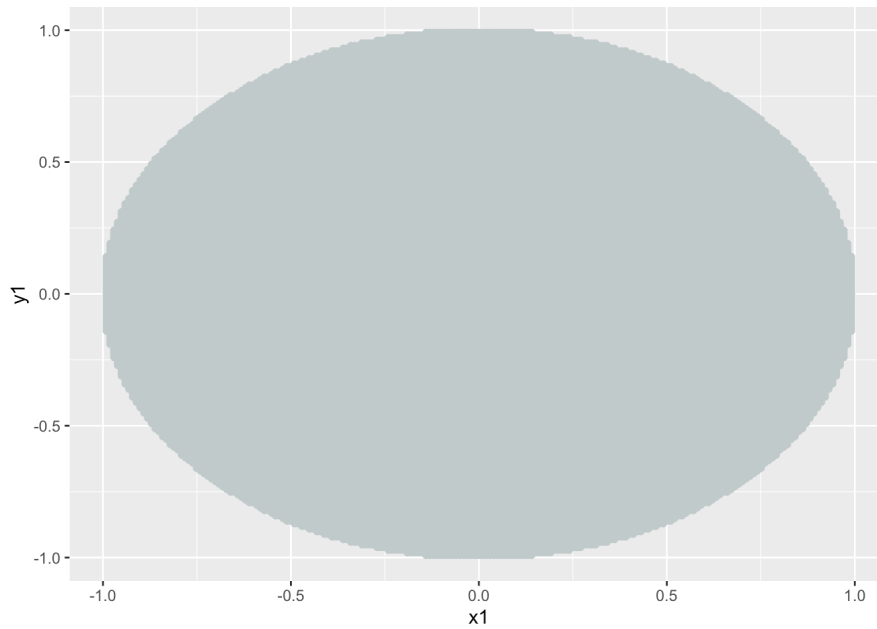
Let the random variables X and Y be uniformly distributed over the interior of the unit circle centered at $(0, 0)$.

- Write the joint probability density function $f(x, y)$.

Solution:

We need to answer the exact value of x and y . The question wants us to find the joint probability density function. The support is less than 1. We have a unit circle here and $X^2 + Y^2 < 1$.

```
library(ggplot2)
x <- seq(-1, 1, by=0.01)
y <- seq(-1, 1, by=0.01)
x1 <- numeric()
y1 <- numeric()
for(i in 1:length(x)){
  for(j in 1:length(y)){
    if ( (x[i]^2 + y[j]^2) < 1){
      x1 <- c(x1, x[i])
      y1 <- c(y1, y[j])
    }
  }
}
# plot(x1,y1, col="azure3")
df <- data.frame(x1,y1)
ggplot(df, aes(x=x1, y=y1)) + geom_point(color="azure3")
```



b. Are X and Y independent?

The question here is we look at the product space and it should be a rectangle. X and Y are not independent.

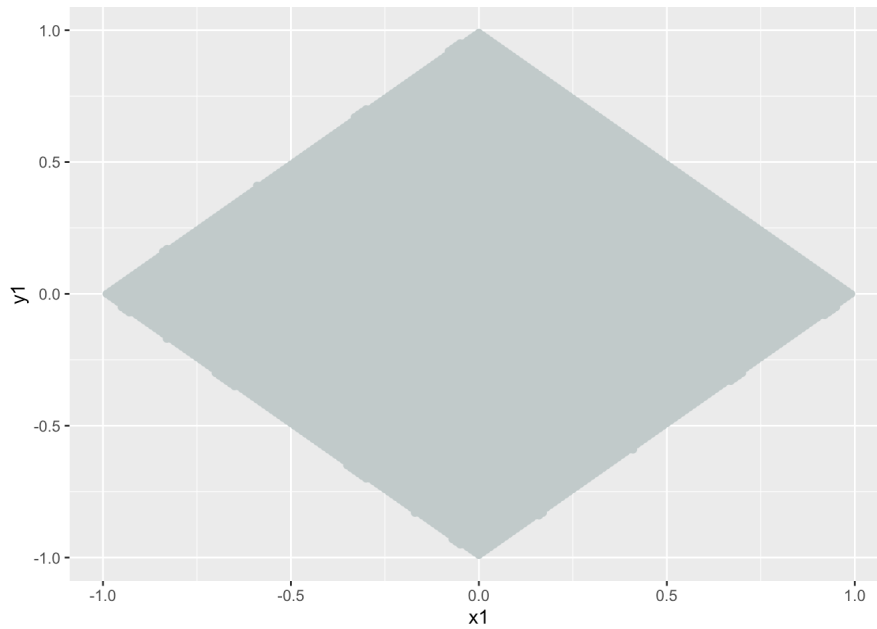
Solution:

c. Use no calculus to find the exact value of $\mathbb{P}(|X| + |Y| < 1)$.

Solution:

We need to know within this circle. Which area corresponds to absolute value of X and absolute value of Y is less than 1. You can actually see this is going to be a special area that we are looking at. This area is equal to 1. All the sites are kind of squared. Therefore, this probability area is $2 \times \text{density} = 1/\pi$

```
library(ggplot2)
x <- seq(-1, 1, by=0.01)
y <- seq(-1, 1, by=0.01)
x1 <- numeric()
y1 <- numeric()
for(i in 1:length(x)){
  for(j in 1:length(y)){
    if ( (abs(x[i]) + abs(y[j])) < 1){
      x1 <- c(x1, x[i])
      y1 <- c(y1, y[j])
    }
  }
}
# plot(x1,y1, col="azure3")
df <- data.frame(x1,y1)
ggplot(df, aes(x=x1, y=y1)) + geom_point(color="azure3")
```

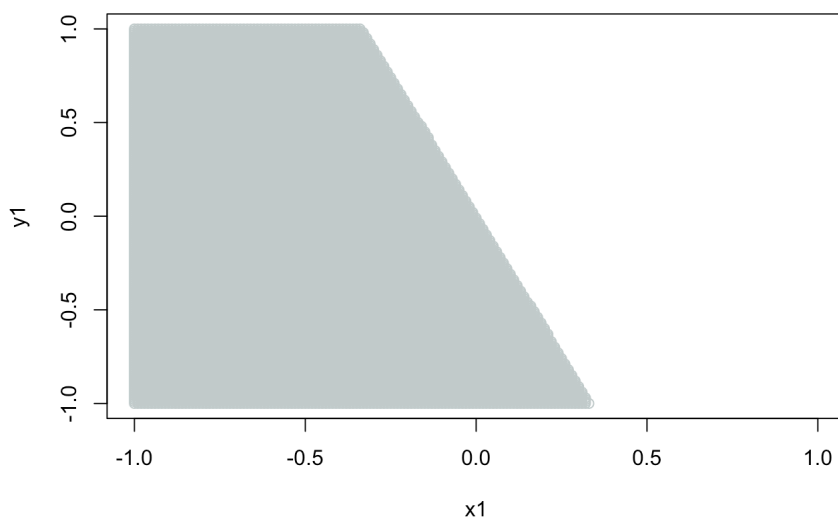


d. Use no calculus to find the exact value of $\mathbb{P}(3X + Y < 0)$.

Solution:

This probability would be an area, $\frac{1}{2}$. Now it is time to use some R codes to see the results. $\mathbb{P}(3X + Y < 0) = \pi/2 \times 1/\pi = 1/2$

```
library(ggplot2)
x <- seq(-1, 1, by=0.01)
y <- seq(-1, 1, by=0.01)
x1 <- numeric()
y1 <- numeric()
for(i in 1:length(x)){
  for(j in 1:length(y)){
    if ( (3*x[i] + y[j]) < 0){
      x1 <- c(x1, x[i])
      y1 <- c(y1, y[j])
    }
  }
}
plot(x1,y1, col="azure3", xlim=c(-1,1), ylim=c(-1,1))
```



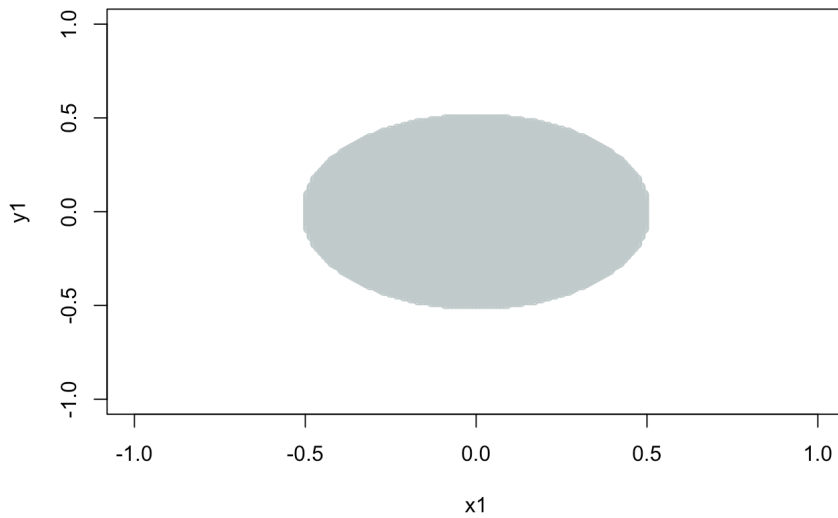
```
# df <- data.frame(x1,y1)
# ggplot(df, aes(x=x1, y=y1)) + geom_point(color="azure3")
```

e. Use no calculus to find the exact value of $\mathbb{P}(X^2 + Y^2 < 1/4)$.

Solution:

We have a unit circle which $x^2 + y^2 = 1$. $\mathbb{P}(X^2 + Y^2 < 1/4) = \pi(1/2)^2 = \pi/4$

```
library(ggplot2)
x <- seq(-1, 1, by=0.01)
y <- seq(-1, 1, by=0.01)
x1 <- numeric()
y1 <- numeric()
for(i in 1:length(x)){
  for(j in 1:length(y)){
    if ( (x[i]^2 + y[j]^2) < 1/4){
      x1 <- c(x1, x[i])
      y1 <- c(y1, y[j])
    }
  }
}
plot(x1,y1, col="azure3", xlim=c(-1,1), ylim=c(-1,1))
```



```
# df <- data.frame(x1,y1)
# ggplot(df, aes(x=x1, y=y1)) + geom_point(color="azure3")
```

■

Independent random variables

Let the independent random variables X and Y have marginal PMFs

$$f_X(x) = \frac{1}{2}, \quad x = 1, 2$$

and

$$f_Y(y) = \frac{1}{3}, \quad y = 1, 2, 3$$

Find $\mathbb{P}(X = Y)$

Solution:

We need to find the probability of x equal to 1. We need to compute the joint distribution of X and Y then we can compute the probability of $X=Y$. So the first thing we need to do here is dependent random variables.

$f_X(x) = 1/2$ where $X = 1, 2$ and the marginal distributions of y is also $f_Y(y) = 1/3$ where $Y = 1, 2, 3$. Therefore we need to find the probability joint distribution, $f(x, y) = 1/2 \times 1/3$ where $X = 1, 2$ and $Y = 1, 2, 3$. $P(X = Y) = P(X = 1, Y = 1) + P(X = 2, Y = 2)$, because they are independent, the joint distribution can be written as a product of marginal distributions. This can be written as $P(X = 1) \cdot P(Y = 1) + P(X = 2) \cdot P(Y = 2)$. We have the marginal distributions and the probability X takes the value 1 is $1/2$. and the probability Y takes the value 1 is $1/3$. So the result would be: $1/2 \times 1/3 + 1/2 \times 1/3 = 1/6 + 1/6 = 1/3$

■

Independence determination

Determine, by inspection, whether X and Y with distribution described by $f(x, y)$ given below, are independent or dependent random variables.

a. $f(x, y) = \frac{1}{27}(x^3 + y)$, $x = 0, 1, 2; y = 1, 2$

Solution:

We have the joint distribution $f(x, y) = \frac{1}{27}(x^3 + y)$, $x = 0, 1, 2; y = 1, 2$. Now we need to say whether X and Y are dependent or independent. There are two steps to test this. The first is looking at product of the space. If the product space is not a rectangle, we stop and say they are not independent. If the product space is rectangle, then we need to second check which is writing the joint distribution as a product of the two functions where the first one is the function of X only and where the second one is the function of Y only. In this case, is the support product of space? X takes the value of 0, 1.

$$f(x, y) = g(x) \cdot b(y), \frac{1}{27}(x^3 + y) = g(x) \cdot b(y), X \text{ and } Y \text{ are not independent.}$$

b. $f(x, y) = \frac{1}{40}x^2y$, $(x, y) = (1, 1), (3, 1), (1, 3), (3, 3)$

Solution:

X takes the value 1 and 3 also Y takes the value 1 and 3. We can actually say $f(x, y) = \frac{1}{40}x^2y = g(x)h(y)$ We have a function g which is the function of x only and y is the function of y only. So X and Y are independent.

c. $f(x, y) = e^{-x}$, $x > 0; 0 < y < 1$

Solution:

All positive value is x and if you look at y , y is between 0 and 1. So therefore we can say the support is the product space. Second test would be to write the joint probability function of $g(x)$ and $h(y)$. $f(x, y) = e^{-x} = g(x)h(y)$. So the first one is $g(x) = e^{-x}$ and the second one is $h(y) = 1$. So X and Y here are independent.

d. $f(x, y) = 1/2$, $|x| + |y| < 1$

Solution:

If we plot the function it would be a diamond. Support is not the product space so you do not need to go to the next step. X and Y are not independent.

e. $f(x, y) = 1/4$, $-1 < x < 1; -1 < y < 1$

Solution:

The support is the product space so we just need to check the second condition or step. The first function is $1/4$ and the second function would be 1 no matter what the value is for y . So X and Y are independent.

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Joint moment generating functions

Let the random variables X and Y have joint moment generating function:

$$M(t_1, t_2) = \frac{1}{2}e^{t_1+t_2} + \frac{1}{4}e^{2t_1+t_2} + \frac{1}{12}e^{t_2} + \frac{1}{6}e^{4t_1+3t_2}$$

for all real values of t_1 and t_2 .

a. Find $V[X]$

Solution:

$$V[X] = E[X^2] - (E[X])^2.$$

$$M(t) = \frac{1}{2}e^t + \frac{1}{4}e^{2t} + \frac{1}{12} + \frac{1}{6}e^{4t}$$

Now we need to find the variance of X we can differentiate the moment generating function one time and set t equal to 0. First derivative would be

$$M'_X(t) = \frac{1}{2}e^t + \frac{2}{4}e^{2t} + \frac{4}{6}e^{4t}$$

Now we need to evaluate this in 0, this would be: $1/2 + 1/2 + 4/6 = 5/3$.

$$M''_X(t) = \frac{1}{2}e^t + \frac{4}{4}e^{2t} + \frac{16}{6}e^{4t}$$

Now we need to evaluate this based on $t = 0$, it would be $E[X] = 1/2 + 1 + 8/3 = 25/6$

$$V[X] = E[X^2] - (E[X])^2 = 25/6 - (5/3)^2 = 25/18 = 1.38$$

b. Find $\mathbb{P}(X < Y)$

Solution: Here we need to find $P(X < Y)$, what we can actually do is look at the joint moment generating function:

$$M(t_1, t_2) = \frac{1}{2}e^{t_1+t_2} + \frac{1}{4}e^{2t_1+t_2} + \frac{1}{12}e^{t_2} + \frac{1}{6}e^{4t_1+3t_2}$$

If we look at the function $f(x, y)$, it would be $1/2$ where $x=1, y=1$ and $1/4$ where $x=2, y=1$ and then $1/12$ where $x=0, y=1$ and lastly $1/6$ where $x=4$ and $y=3$. $\mathbb{P}(X < Y)$ just when happens that $X=0$ and $Y=1/12$.

■

Variance

Let the continuous random variables X and Y be uniformly distributed over the triangular-shaped support region with vertices $(0, 2)$, $(1, 0)$, and $(0, -1)$

a. Find $E[X]$

We need to find the area of the triangle, base is 3 and height is 1. Therefore it is going to be $f(x, y) = 2/3$ where $x > 0, y > x-1$ and $y < -2x+2$.
 $(y - y_1) = m(x - x_1)$ so $m = \frac{y_2 - y_1}{x_2 - x_1} = \frac{0+1}{1-0} = 1$ This has the positive slope. $(y + 1) = 1(x - x_1)$, $x = y + 1$ we can rewrite this as $y = x - 1$ and $y > x - 1$. Next one is finding the equation of the second line which is $y - 2 = -2(x - 0)$ and $y > -2x + 2$. We can define our support which is going to be all pairs x and y . $A = (x, y) | x > 0, y > x - 1, y < -2x + 2$ By integrating out we can say the joint density would be $f_X(x) = \int \frac{2}{3} dy$ in the range of $x-1$ and $-2x+2$. We calculated the marginal density for X now we can find the marginal density for Y as well. So, the first segment would be $f_Y(y)$ for range of $-1 < y < 0$ it would be $\int 2/3 dx$ and for range $0 < y < 2$ it would be $\int 2/3 dx$. Also we have $f_Y(y) = \frac{2(y+1)}{3}$ where $-1 < y < 0$ and $\frac{2-y}{3}$ where $0 < y < 2$.

Solution:

$$\int_0^1 2(1-x)dx = x^2 - \frac{2x^3}{3} = \frac{1}{3}$$

b. Find $V[X]$

Solution:

$$V[X] = E[X^2] - (E[X])^2$$

$$E[X^2] = \int x^2 \cdot 2(1-x)dx = [\frac{2x^3}{3} - \frac{1x^4}{2}] = \frac{1}{6}$$

$$V[X] = 1/6 - (1/3)^2 = 1/6 - 1/9 = 1/18$$

c. Find ρ

Solution:

We have a lot of work to do. If we remember ρ is going to be $\rho = \frac{E[XY] - E[X]E[Y]}{\sqrt{V[X]}\sqrt{V[Y]}}$.

$$E[Y] = \int_{-1}^0 \frac{y \times 2(y+1)}{3} dy + \int_0^2 y \times \frac{2-y}{3} dy = 2/9 - 1/3 + 4/3 - 8/9 = 1/3,$$

$$E[Y^2] = \int_{-1}^0 \frac{y^2 \times 2(y+1)}{3} dy + \int_0^2 y^2 \times \frac{2-y}{3} dy = 1/2$$

$$V[X] = E[X^2] - (E[X])^2 = 1/2 - 1/9 = 7/18,$$

$$E[XY] = \int \int xy \frac{2}{3} dy dx = \int \frac{xy^2}{3} dx = \int_0^1 (x^3 - 2x^2 + x) dx = [\frac{x^4}{4} - \frac{2x^3}{3} + \frac{x^2}{2}] = 1/12$$

$$\rho = \frac{1/12 - 1/9}{\sqrt{7/18}} = -0.1890$$

d. Find $V[X - Y]$

Solution:

$$V[X - Y] = E[(X - Y)^2] - (E[(X - Y)])^2 = E[X^2 - 2XY + Y^2] - (E[X] - E[Y])^2 = E[X^2] - 2E[XY] + E[Y^2] - (E[X] - E[Y])^2 = E[X^2] - 2E[XY] = 1$$

e. Write a Monte Carlo simulation that supports your solution to part (d)

```
nrep <- 1000000
xx <- runif(nrep)
yy <- runif(nrep, -1,2)
index <- which(yy > xx - 1 & yy < 2 - 2*xx)
x <- xx[index]
y <- yy[index]
cat(sprintf("Mean of x: %f\n", mean(x)))
```

```
## Mean of x: 0.333710
```

```
cat(sprintf("Variance of x: %f\n", var(x)))
```

```
## Variance of x: 0.055545
```

```
cat(sprintf("cor(x, y): %f\n", cor(x, y)))
```

```
## cor(x, y): -0.187407
```

```
cat(sprintf("var(x - y): %f\n", var(x-y)))
```

```
## var(x - y): 0.498516
```

Solution:



Functions of random variables

A bag contains 5 balls numbered 1,2,3,4,5. Arno selects three balls at random and with replacement from the bag. The numbers on the balls selected are X_1 , X_2 , and X_3 . Find the population mean and population variance of:

$$Y = X_1 + X_2 + X_3$$

Support your results by executing a Monte Carlo simulation experiment.

```
nrep <- 100000
y <- rep(0, nrep)
for(i in 1: nrep){
  x <- sample(5, 3, replace = TRUE)
  y[i] <- sum(x)
}
cat(sprintf("Mean of y: %f\n", mean(y)))
```

```
## Mean of y: 8.983590
```

```
cat(sprintf("Variance of y: %f\n", var(y)))
```

```
## Variance of y: 6.006741
```

Solution:

These are independent and identically distributed. (X_1 , X_2 , and X_3 all are iid) Selecting the all 5 balls are equally likely.

$$E[X_1] = \frac{a+b}{2} = \frac{1+5}{2} = 3 = E[X_2] = E[X_3], V[X_1] = \frac{(b-a+1)^2-1}{12} = 2, V[X_1] = V[X_2] = V[X_3] = 2,$$

$$E[Y] = E[X_1 + X_2 + X_3] = E[X_1] + E[X_2] + E[X_3] = 3 + 3 + 3 = 9,$$

$$V[Y] = V[X_1 + X_2 + X_3] = V[X_1] + V[X_2] + V[X_3] = 2 + 2 + 2 = 6$$



Variance-covariance matrix

Let the random variables X , Y and Z have a tri-variate probability distribution with population mean vector $\mu = (0, 1, 2)'$ and variance-covariance matrix:

$$\Sigma = \begin{bmatrix} 7 & 3 & 4 \\ 3 & 8 & 5 \\ 4 & 5 & 9 \end{bmatrix}$$

Let $W = 2X - Y + 3Z$.

a. Find $E[W]$

Solution:

$$E[W] = E[2X - Y + 3Z] = E[2X] - E[Y] + E[3Z] = 2E[X] - E[Y] + 3E[Z] = 2(0) - 1 + 3 \times 2 = 5$$

b. Find $V[W]$

Solution:

$$V[W] = V[2X - Y + 3Z] = V[2X] + V[Y] + V[3Z] + 2Cov(2X - Y) + 2Cov(2X, 3Z) + 2Cov(-Y, 3Z) = 4V[X] + V[Y] + 9V[Z] - 4 \times Cov(X, Y) + 12 \times Cov(X, Z) - 6 \times Cov(Y, Z)$$

■

Multivariate normal distribution

How many parameters are required to specify the n -dimensional multivariate normal distribution?

Solution:

We have $n \times n$ matrix also we have n elements on diagonal. This matrix is symmetric. First we need to get $\mu = \mu_1, \mu_2, \mu_3, \dots$. Also we have variance co variance matrix. The number of parameters we need is, we need n population means and it is gonna be

$$\frac{n+n(n+1)}{2} = \frac{2n+n^2+n}{2} = \frac{n(n+3)}{2},$$

There is a second approach here: n pop means and n pop variances. Also you have n number of random variables. The number of pairs is n choose from 2. Therefore you can sum up all of three. It is gonna be $\$ \$$. For bivariate the solution would be $\$ = 5 \$$ These 5 parameters are $\mu_1, \mu_2, \sigma_1, \sigma_2$, and ρ .

■