STAT401\_HW1

김정현

library(MASS)  
# library(rgl)  
library(lattice)  
options(digits = 4)

## Q1.

Suppose mean vector and the variance-covariance matrix given below.

mu <- c(-1, 2, -5)  
sigma <- matrix(c(9, 4, 1,  
 4, 16, -1,  
 1, -1, 4), nrow = 3, byrow = TRUE)

### (a)

Note that population correlation matrix equals . Also, since and , We can calculate .

Therefore, the population correlation matrix equals

d <- matrix(c(1/3, 0, 0,  
 0, 1/4, 0,  
 0, 0, 1/2), nrow = 3, byrow = TRUE)  
  
ans <- d %\*% sigma %\*% d  
print(ans)

## [,1] [,2] [,3]  
## [1,] 1.0000 0.3333 0.1667  
## [2,] 0.3333 1.0000 -0.1250  
## [3,] 0.1667 -0.1250 1.0000

### (b)

Let and where .

a <- matrix(c(1, -1, 1), nrow = 3)  
e1 = t(a) %\*% mu  
var1 = t(a) %\*% sigma %\*% a  
  
cat("Mean of Z1:", e1[1], "\n")

## Mean of Z1: -8

cat("Variance of Z1:", var1[1], "\n")

## Variance of Z1: 25

The mean of Z1 is -8, and the variance is 25.

### (c)

Let and where .

b <- matrix(c(-1, 2, 3), nrow = 3)  
  
e2 = t(b) %\*% mu  
var2 = t(b) %\*% sigma %\*% b  
  
cat("Mean of Z2:", e2[1], "\n")

## Mean of Z2: -10

cat("Variance of Z2:", var2[1], "\n")

## Variance of Z2: 75

The mean of Z2 is -10, and the variance is 75.

### (d)

2) :

Variance for Z\_1 and Z\_2 is given as and .

cov\_z1z2 = t(a) %\*% sigma %\*% b  
cat("Covariance for Z1, Z2: ", cov\_z1z2)

## Covariance for Z1, Z2: -14

Therefore

### (e)

n <- 50  
sample\_data <- mvrnorm(n, mu, sigma)  
cat("Sample mean", colMeans(sample\_data), '\n')

## Sample mean -0.5363 1.331 -4.561

cat("Sample Variance", '\n')

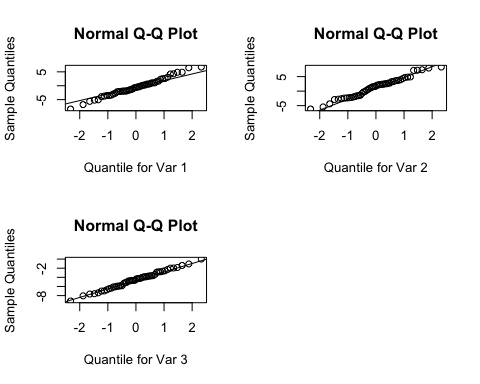
## Sample Variance

cov(sample\_data)

## [,1] [,2] [,3]  
## [1,] 10.510 1.420 1.124  
## [2,] 1.420 11.826 -2.193  
## [3,] 1.124 -2.193 3.870

### (f)

p <- ncol(sample\_data)  
par(mfrow=c(2,2))  
for(i in 1:p){qqnorm(sample\_data[,i],xlab=paste("Quantile for Var",i))  
 qqline(sample\_data[,i])}

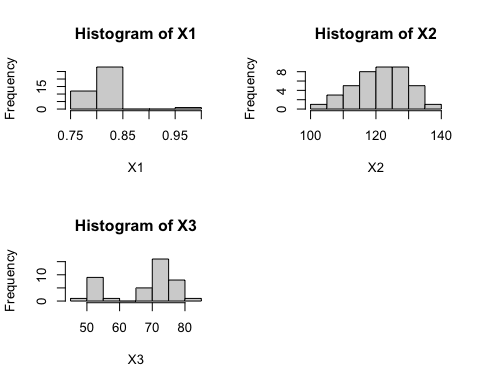


The plot shows that the distribution follows the multivariate normal distribution, which means that the plot follows the straight line.

## Q2.

### (a)

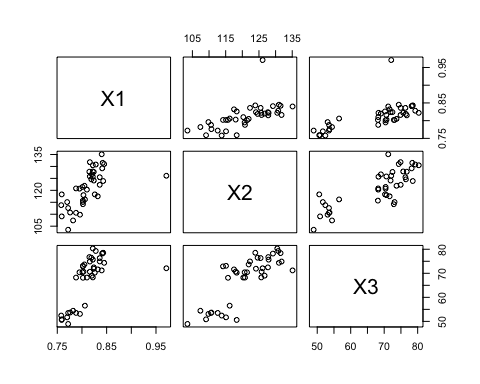
par(mfrow=c(2,2))  
data <- read.table('Paper.dat', header = T)  
attach(data)  
hist(X1)  
hist(X2)  
hist(X3)



One can observe that X1 (Density) has an outlier by boxplot. Also by histogram, We can find that X2 follows the normal distribution similarly and X3 does not.

### (b)

pairs(data)



X1 and X2 are strongly related (except 1 outlier). X1 and X3 are strongly related (except 1 outlier). X2 and X3 are strongly related.

### (c)

# plot3d(data)

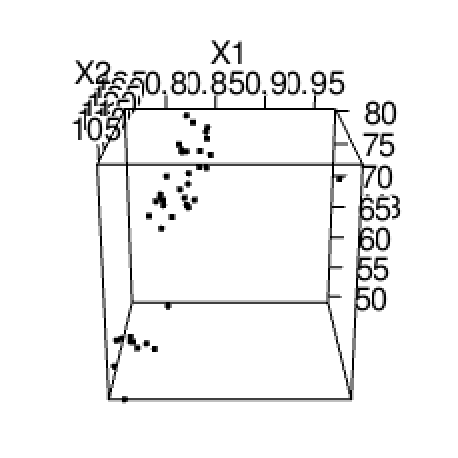


Image Title

We can identify one outlier by drawing the graph.

### (d)

cat("Sample mean:", colMeans(data), '\n')

## Sample mean: 0.8119 121 67.72

cat("Sample Variance", '\n')

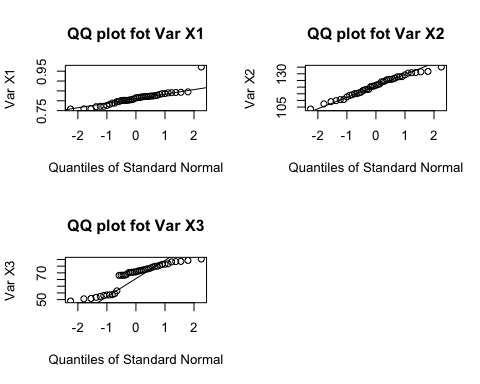
## Sample Variance

cov(data)

## X1 X2 X3  
## X1 0.001265 0.1684 0.2252  
## X2 0.168447 59.3211 60.9925  
## X3 0.225248 60.9925 95.8567

### (e)

p <- ncol(data)  
par(mfrow=c(2,2))  
for(i in 1:p){qqnorm(data[,i],xlab="Quantiles of Standard Normal",  
 ylab=paste("Var", colnames(data)[i]),  
 main=paste("QQ plot fot Var", colnames(data)[i]))  
 qqline(data[,i])}



The plot for X1 shows that the distribution follows the multivariate normal distribution (except 1 outlier), which means that the plot follows the straight line. This has same result with the box plot drew in (a).

The plot for X2 shows that the distribution follows the multivariate normal distribution, which means that the plot follows the straight line. This has same result with the histogram drew in (a) which follow the normal distribution similarly.

The plot for X3 shows that the distribution does not follow the multivariate normal distribution, which means that the plot does not follow the straight line. This has same result with the histogram drew in (a) which did not follow the normal distribution.

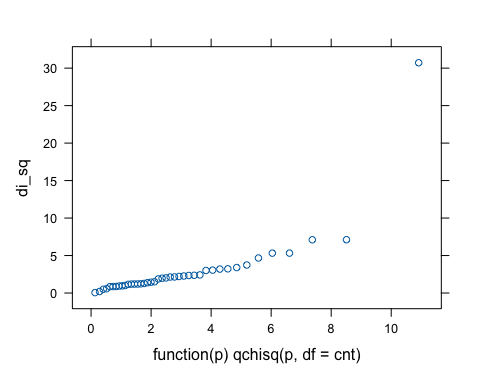
### (f)

di\_sq = mahalanobis(data, colMeans(data), cov(data))  
di\_sq

## [1] 0.50077 0.93008 1.27637 3.05282 7.10350 1.87632 1.19955 4.67091  
## [9] 1.97711 1.15470 1.20315 5.31942 0.05396 1.50274 2.34111 3.00743  
## [17] 3.22980 3.73736 7.11137 2.20634 2.01760 1.37747 2.26966 0.95797  
## [25] 30.71541 0.58239 0.85130 1.44447 2.41814 1.00082 2.13386 0.19841  
## [33] 0.86159 3.19152 0.87509 1.19042 1.22714 2.36489 5.32404 2.14504  
## [41] 3.39802

### (g)

cnt <- 3  
qqmath(di\_sq,distribution = function(p) qchisq(p,df=cnt))



This data set follows the multivariate normal distribution since the plot follows the chi-Squared distribution (except 1 outlier).