

Homework 3-1, 3-2

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2018/4/10

1

density function for the multivariate normal distribution:

$$f(x; \mu, \Sigma) = \frac{1}{(2\pi)^{p/2} |\Sigma|^{1/2}} e^{-\frac{1}{2}(x-\mu)' \Sigma^{-1} (x-\mu)}$$

the likelihood function for the two independent sample:

$$\begin{aligned} L(\mu_1, \mu_2, \Sigma) &= \prod_{i=1}^{n_1} f(X_i; \mu_1, \Sigma) \prod_{j=1}^{n_2} f(X_j; \mu_2, \Sigma) \\ &= L(\mu_1, \Sigma) L(\mu_2, \Sigma) \end{aligned}$$

So the likelihood function can be defined as:

$$\begin{aligned} L(\mu_1, \mu_2, \Sigma) &= \frac{1}{(2\pi)^{Np/2} |\Sigma|^{N/2}} \exp\left(-\frac{1}{2} \text{tr}[\Sigma^{-1} (\sum_{i=1}^{n_1} \Phi_1(X_i) + \sum_{j=1}^{n_2} \Phi_2(X_j))]\right) \\ \Phi_i(x) &= (x - \mu_i)(x - \mu_i)' \end{aligned}$$

Using MLE, the maximum is:

$$\begin{aligned} \hat{\Sigma} &= \frac{1}{n_1 + n_2} \left(\sum_{i=1}^{n_1} \Phi_1(X_i) + \sum_{j=1}^{n_2} \Phi_2(X_j) \right) \\ &= \frac{1}{n_1 + n_2} [(n_1 - 1)S_1 + (n_2 - 1)S_2] = \frac{n_1 + n_2 - 1}{n_1 + n_2} S_{pooled} \end{aligned}$$

2

(a)

It can be calculated that:

$$\begin{aligned} \bar{x} &= [4.64, 45.4, 9.965]^T \\ S &= \begin{bmatrix} 2.879 & 10.01 & -1.81 \\ 10.01 & 199.788 & -5.64 \\ -1.81 & -5.64 & 3.682 \end{bmatrix} \end{aligned}$$

we can calculate S's eigen value and its respective eigen vector:

$$\begin{aligned} \text{value} &= [200.46209264, 1.31804876, 4.56885859] \\ e'_1 &= [-0.05084165, -0.82194341, -0.56729548] \\ e'_2 &= [-0.99828327, 0.02528569, 0.0528313] \\ e'_3 &= [0.02907988, -0.56900761, 0.82181792] \end{aligned}$$

The axes of the region are:

$$\sqrt{\lambda_i} \sqrt{\frac{p(n-1)}{n(n-p)} F_{p,n-p}(\alpha)}$$
$$\frac{p(n-1)}{n(n-p)} F_{p,n-p}(\alpha) = \frac{19 \times 3}{17 \times 20} \times F_{3,17}(0.1) = 0.167 \times 2.44 = 0.409$$

so the axes' lengths are:

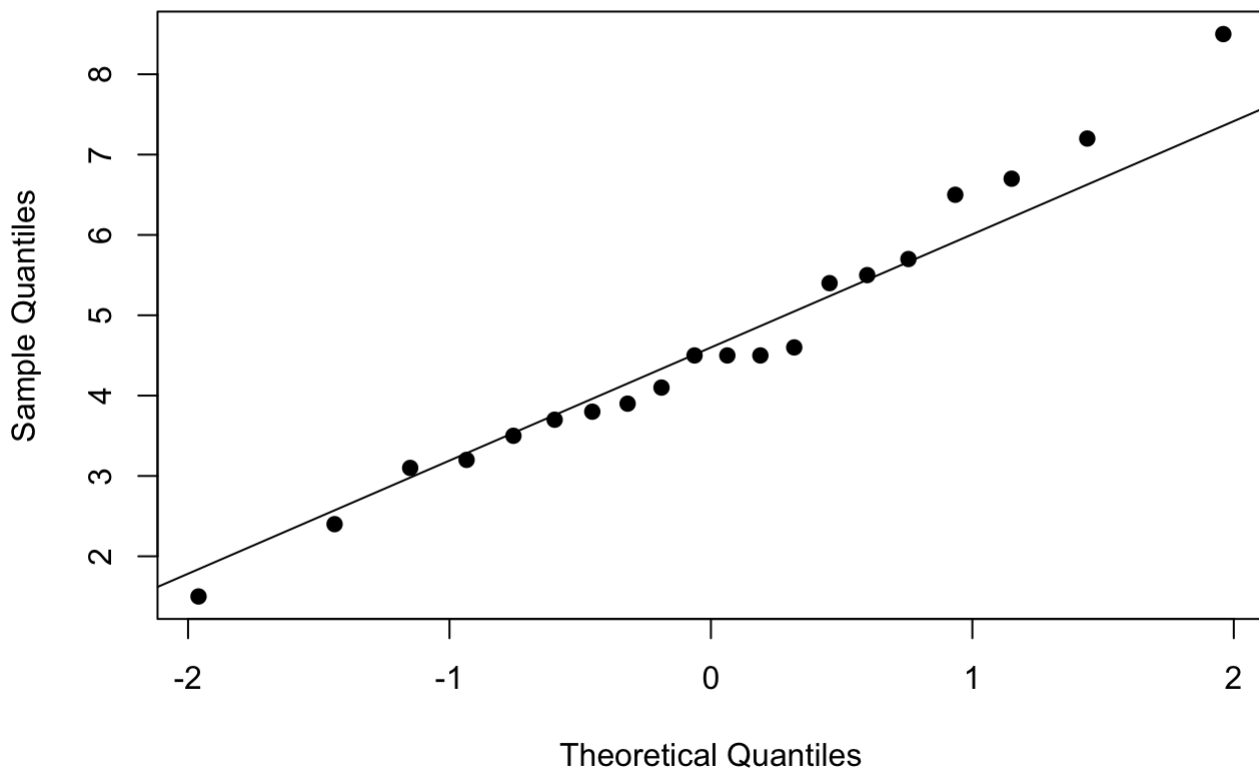
$$\sqrt{200.46209264} \times \sqrt{0.409} = 9.055$$
$$\sqrt{1.31804876} \times \sqrt{0.409} = 0.734$$
$$\sqrt{4.56885859} \times \sqrt{0.409} = 1.367$$

The directions of each axis is determined by its corresponding eigen vector shown above.

(b)

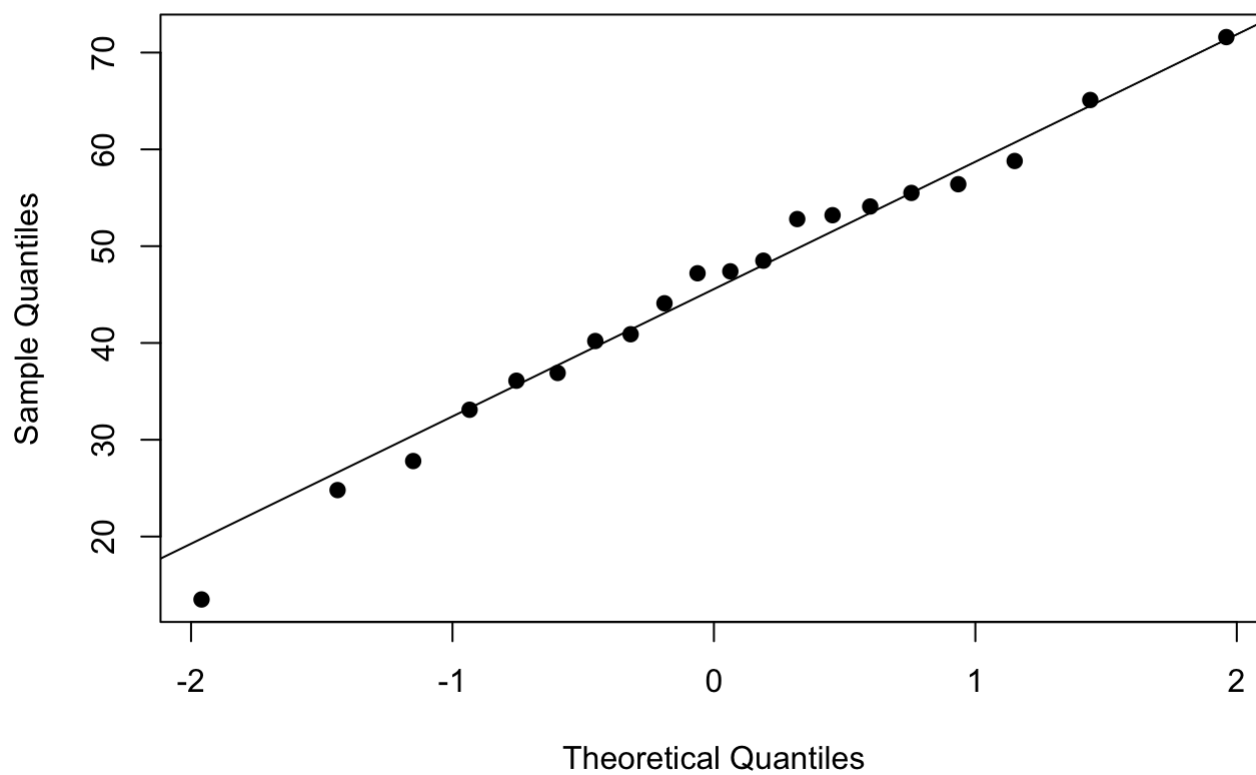
```
# read the data
setwd('~\\Desktop\\三春\\3多元统计分析\\作业\\作业3-1,3-2\\')
dat<-read.csv("data.csv")
x1<-dat$x1
x2<-dat$x2
x3<-dat$x3
qqnorm(x1, main="Normal Probability Plot", pch=19)
qqline(x1)
```

Normal Probability Plot



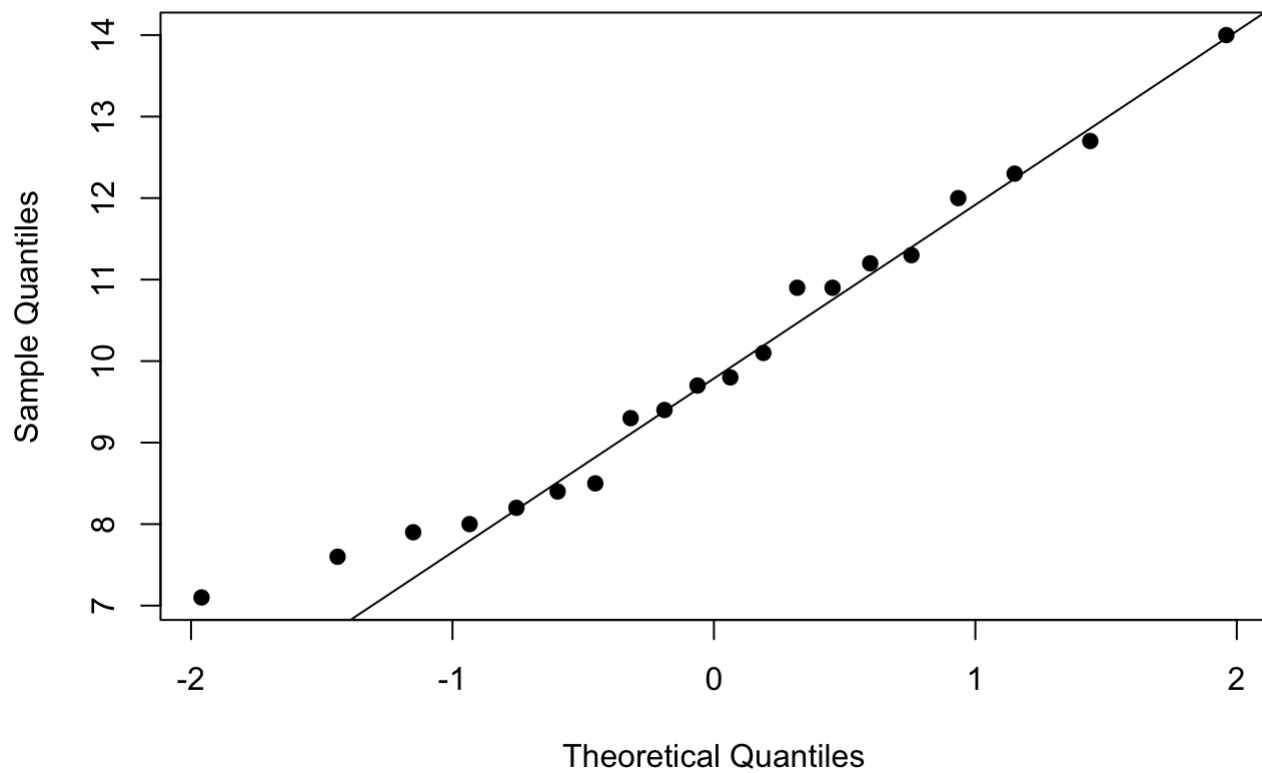
```
qqnorm(x2, main="Normal Probability Plot", pch=19)  
qqline(x2)
```

Normal Probability Plot

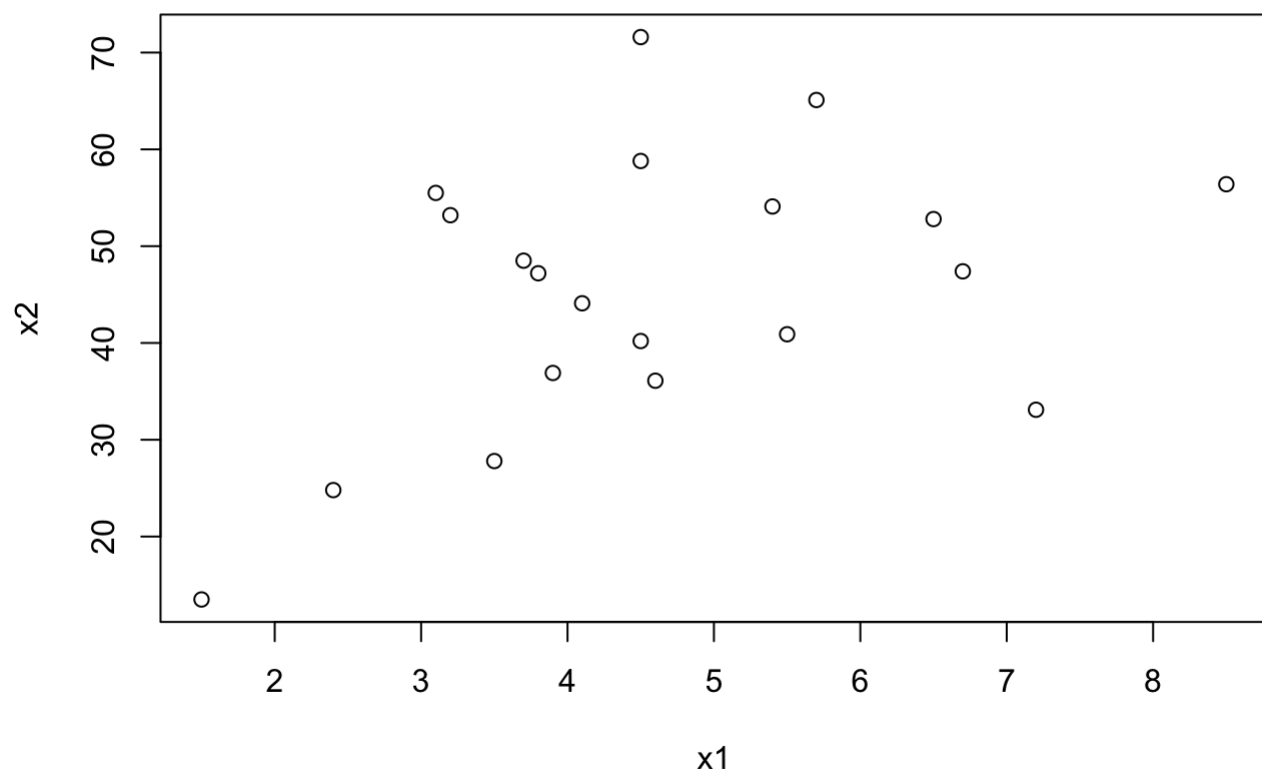


```
qqnorm(x3, main="Normal Probability Plot", pch=19)  
qqline(x3)
```

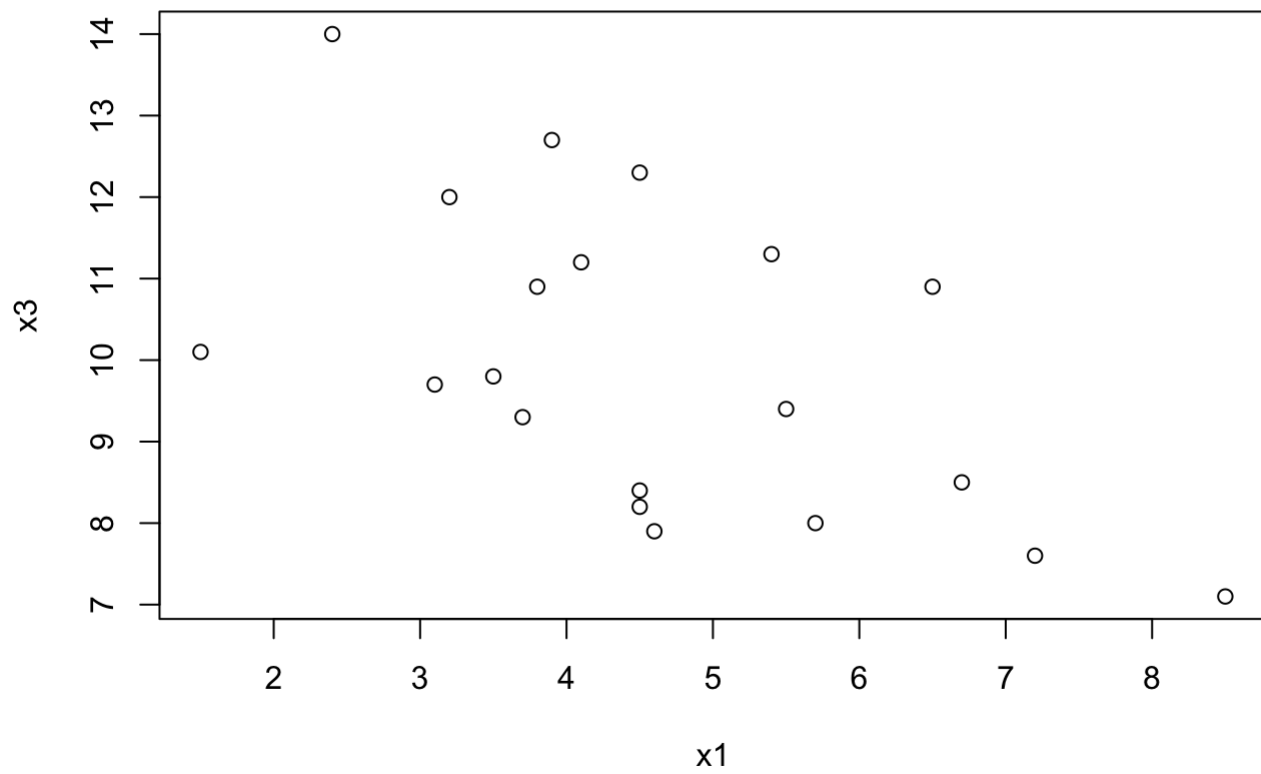
Normal Probability Plot



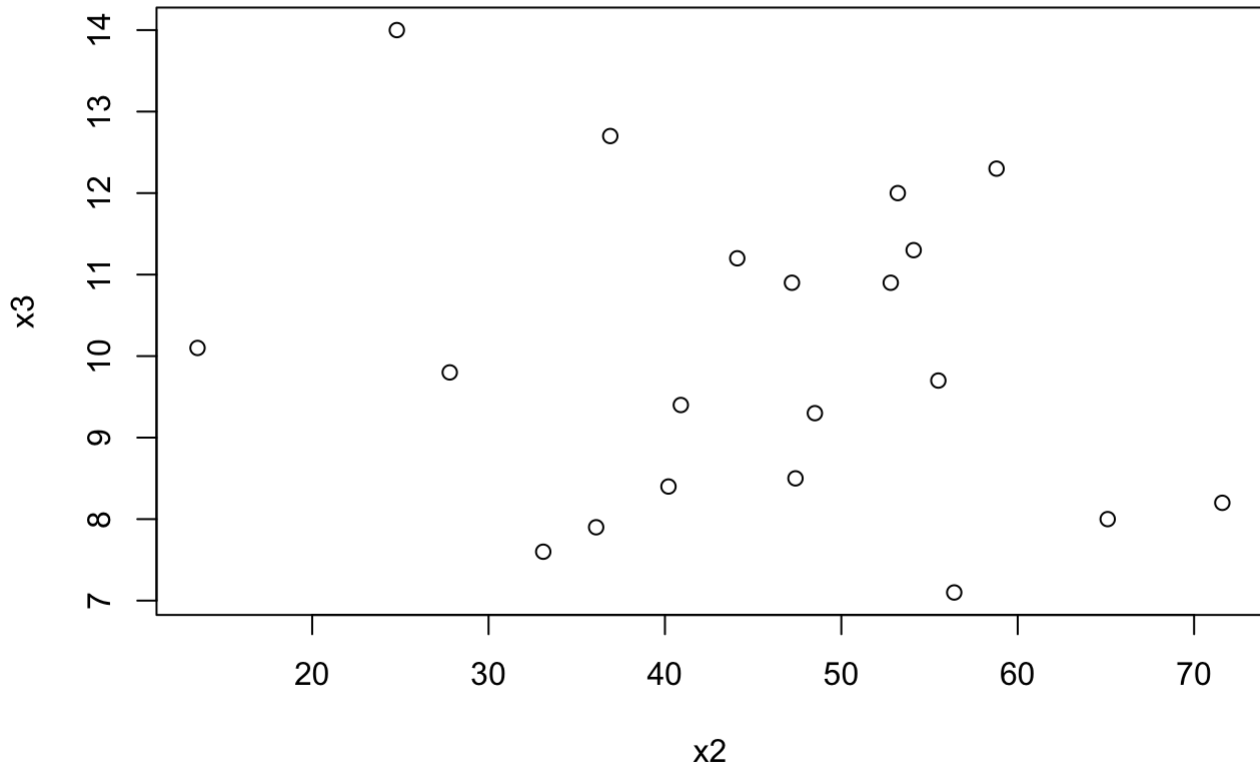
```
plot(x1,x2)
```



```
plot(x1,x3)
```



```
plot(x2,x3)
```



It

seems that each variable's normality is fine and they don't have a significant relationship with each other, so the multivariate normal assumption seems justified.

3

$$T^2 = \sqrt{n}(\bar{X} - \mu_0)' \left(\frac{\sum_{j=1}^n (X_j - \bar{X})(X_j - \bar{X})'}{n-1} \right)^{-1} \sqrt{n}(\bar{X} - \mu_0)$$

So it can be calculated that $T^2 = 9.74$

$$\frac{p(n-1)}{(n-p)} F_{p,n-p}(\alpha) = \frac{19 \times 3}{17} \times F_{3,17}(0.05) = \frac{19 \times 3}{17} \times 3.20 = 10.729$$

the confidence region is defined as :

$$(\bar{x}_i - \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{ii}}{n}}, \bar{x}_i + \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{ii}}{n}})$$

$$F_{p,n-p}(\alpha) = F_{3,17}(0.05) = 3.20, \bar{x}_1 = 4.64, \bar{x}_2 = 45.4, \bar{x}_3 = 9.965$$

$$s_{11} = 2.879, s_{22} = 199.788, s_{33} = 3.628$$

so the three regions are:

$$(\bar{x}_1 - \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{11}}{n}}, \bar{x}_1 + \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{11}}{n}}) \\ = [3.397, 5.882]$$

$$(\bar{x}_2 - \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{22}}{n}}, \bar{x}_2 + \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{22}}{n}}) \\ = [35.047, 55.752]$$

$$(\bar{x}_3 - \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{22}}{n}}, \bar{x}_3 + \sqrt{\frac{p(n-1)}{n(n-p)}} F_{p,n-p}(\alpha) \sqrt{\frac{s_{33}}{n}}) \\ = [8.569, 11.360]$$

(b) The Bonferroni region is defined as :

$$[\bar{x}_p - t_{n-1}(\frac{\alpha}{2p}) \sqrt{\frac{s_{pp}}{n}}, \bar{x}_p + t_{n-1}(\frac{\alpha}{2p}) \sqrt{\frac{s_{pp}}{n}}] \\ t_{19}(0.0083) = 2.625106$$

so the Bonferroni regions are:

$$[3.644, 5.635] \\ [37.103, 53.696] \\ [8.846, 11.083]$$

which are smaller than T^2 confidence region because it focus on single confidence interval.