

TD1

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2023-09-25

Contents

Exercice 3	2
Longueur des sépales	5
Exercice 4	7
Question 1	7
Question 2	8

Exercise 3

```
# Show the dataset
iris
```

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	5.1	3.5	1.4	0.2	setosa
## 2	4.9	3.0	1.4	0.2	setosa
## 3	4.7	3.2	1.3	0.2	setosa
## 4	4.6	3.1	1.5	0.2	setosa
## 5	5.0	3.6	1.4	0.2	setosa
## 6	5.4	3.9	1.7	0.4	setosa
## 7	4.6	3.4	1.4	0.3	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa
## 10	4.9	3.1	1.5	0.1	setosa
## 11	5.4	3.7	1.5	0.2	setosa
## 12	4.8	3.4	1.6	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 15	5.8	4.0	1.2	0.2	setosa
## 16	5.7	4.4	1.5	0.4	setosa
## 17	5.4	3.9	1.3	0.4	setosa
## 18	5.1	3.5	1.4	0.3	setosa
## 19	5.7	3.8	1.7	0.3	setosa
## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa
## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa

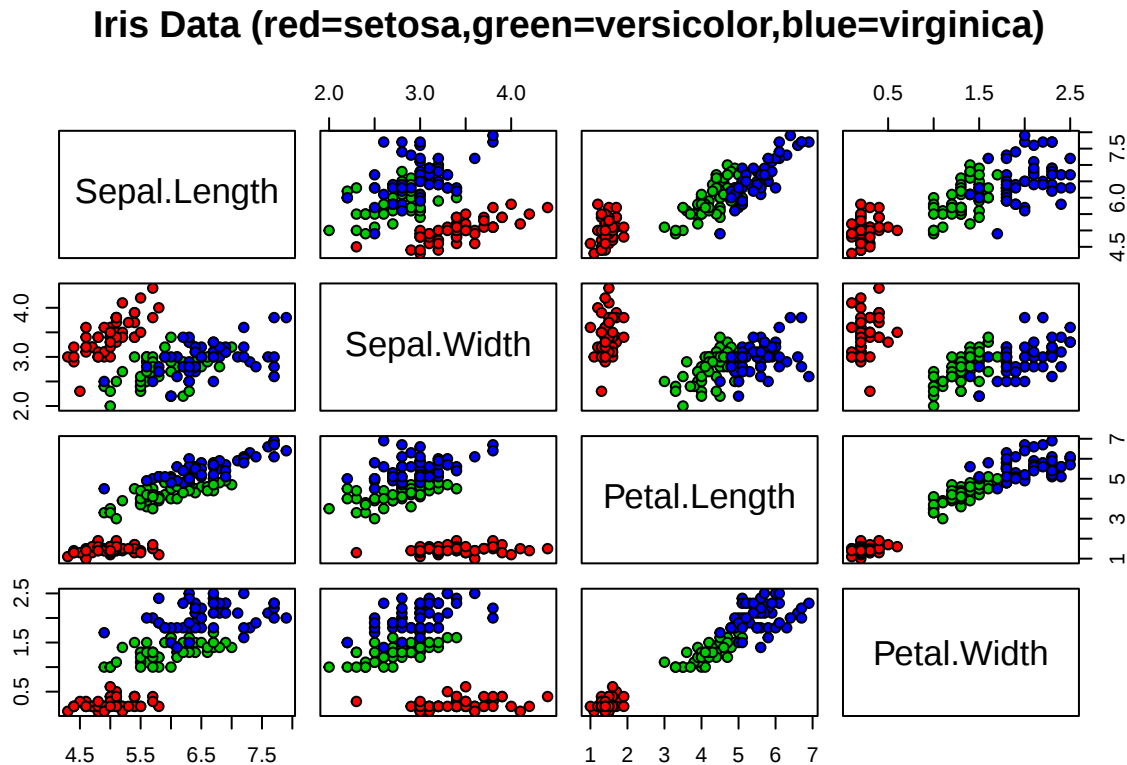
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 55	6.5	2.8	4.6	1.5	versicolor
## 56	5.7	2.8	4.5	1.3	versicolor
## 57	6.3	3.3	4.7	1.6	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 60	5.2	2.7	3.9	1.4	versicolor
## 61	5.0	2.0	3.5	1.0	versicolor
## 62	5.9	3.0	4.2	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 64	6.1	2.9	4.7	1.4	versicolor
## 65	5.6	2.9	3.6	1.3	versicolor
## 66	6.7	3.1	4.4	1.4	versicolor
## 67	5.6	3.0	4.5	1.5	versicolor
## 68	5.8	2.7	4.1	1.0	versicolor
## 69	6.2	2.2	4.5	1.5	versicolor
## 70	5.6	2.5	3.9	1.1	versicolor
## 71	5.9	3.2	4.8	1.8	versicolor
## 72	6.1	2.8	4.0	1.3	versicolor
## 73	6.3	2.5	4.9	1.5	versicolor
## 74	6.1	2.8	4.7	1.2	versicolor
## 75	6.4	2.9	4.3	1.3	versicolor
## 76	6.6	3.0	4.4	1.4	versicolor
## 77	6.8	2.8	4.8	1.4	versicolor
## 78	6.7	3.0	5.0	1.7	versicolor
## 79	6.0	2.9	4.5	1.5	versicolor
## 80	5.7	2.6	3.5	1.0	versicolor
## 81	5.5	2.4	3.8	1.1	versicolor
## 82	5.5	2.4	3.7	1.0	versicolor
## 83	5.8	2.7	3.9	1.2	versicolor
## 84	6.0	2.7	5.1	1.6	versicolor
## 85	5.4	3.0	4.5	1.5	versicolor
## 86	6.0	3.4	4.5	1.6	versicolor
## 87	6.7	3.1	4.7	1.5	versicolor
## 88	6.3	2.3	4.4	1.3	versicolor
## 89	5.6	3.0	4.1	1.3	versicolor
## 90	5.5	2.5	4.0	1.3	versicolor
## 91	5.5	2.6	4.4	1.2	versicolor
## 92	6.1	3.0	4.6	1.4	versicolor
## 93	5.8	2.6	4.0	1.2	versicolor
## 94	5.0	2.3	3.3	1.0	versicolor
## 95	5.6	2.7	4.2	1.3	versicolor
## 96	5.7	3.0	4.2	1.2	versicolor
## 97	5.7	2.9	4.2	1.3	versicolor
## 98	6.2	2.9	4.3	1.3	versicolor
## 99	5.1	2.5	3.0	1.1	versicolor
## 100	5.7	2.8	4.1	1.3	versicolor
## 101	6.3	3.3	6.0	2.5	virginica

## 102	5.8	2.7	5.1	1.9	virginica
## 103	7.1	3.0	5.9	2.1	virginica
## 104	6.3	2.9	5.6	1.8	virginica
## 105	6.5	3.0	5.8	2.2	virginica
## 106	7.6	3.0	6.6	2.1	virginica
## 107	4.9	2.5	4.5	1.7	virginica
## 108	7.3	2.9	6.3	1.8	virginica
## 109	6.7	2.5	5.8	1.8	virginica
## 110	7.2	3.6	6.1	2.5	virginica
## 111	6.5	3.2	5.1	2.0	virginica
## 112	6.4	2.7	5.3	1.9	virginica
## 113	6.8	3.0	5.5	2.1	virginica
## 114	5.7	2.5	5.0	2.0	virginica
## 115	5.8	2.8	5.1	2.4	virginica
## 116	6.4	3.2	5.3	2.3	virginica
## 117	6.5	3.0	5.5	1.8	virginica
## 118	7.7	3.8	6.7	2.2	virginica
## 119	7.7	2.6	6.9	2.3	virginica
## 120	6.0	2.2	5.0	1.5	virginica
## 121	6.9	3.2	5.7	2.3	virginica
## 122	5.6	2.8	4.9	2.0	virginica
## 123	7.7	2.8	6.7	2.0	virginica
## 124	6.3	2.7	4.9	1.8	virginica
## 125	6.7	3.3	5.7	2.1	virginica
## 126	7.2	3.2	6.0	1.8	virginica
## 127	6.2	2.8	4.8	1.8	virginica
## 128	6.1	3.0	4.9	1.8	virginica
## 129	6.4	2.8	5.6	2.1	virginica
## 130	7.2	3.0	5.8	1.6	virginica
## 131	7.4	2.8	6.1	1.9	virginica
## 132	7.9	3.8	6.4	2.0	virginica
## 133	6.4	2.8	5.6	2.2	virginica
## 134	6.3	2.8	5.1	1.5	virginica
## 135	6.1	2.6	5.6	1.4	virginica
## 136	7.7	3.0	6.1	2.3	virginica
## 137	6.3	3.4	5.6	2.4	virginica
## 138	6.4	3.1	5.5	1.8	virginica
## 139	6.0	3.0	4.8	1.8	virginica
## 140	6.9	3.1	5.4	2.1	virginica
## 141	6.7	3.1	5.6	2.4	virginica
## 142	6.9	3.1	5.1	2.3	virginica
## 143	5.8	2.7	5.1	1.9	virginica
## 144	6.8	3.2	5.9	2.3	virginica
## 145	6.7	3.3	5.7	2.5	virginica
## 146	6.7	3.0	5.2	2.3	virginica
## 147	6.3	2.5	5.0	1.9	virginica
## 148	6.5	3.0	5.2	2.0	virginica
## 149	6.2	3.4	5.4	2.3	virginica
## 150	5.9	3.0	5.1	1.8	virginica

```
# Show the help page, with information about the dataset
?iris
```

```
## démarrage du serveur d'aide httpd ... fini
```

```
# Create scatterplots of all pairwise combination of the 4 variables in the dataset
pairs(iris[1:4], main="Iris Data (red=setosa,green=versicolor,blue=virginica)",
      pch=21, bg=c("red","green3","blue")[unclass(iris$Species)])
```



Longueur des sépales

On prend $\alpha = 0.05$, le jeu de données est plutôt grand, on approxime par le TCL la répartition des valeurs par une loi normale $\mathcal{N}(\mu, \sigma^2)$.

Avec $\mu \approx 5.84$ et $\sigma^2 = 0.69$. On cherche les x_i tel que $x_i \notin I_c$ avec I_c l'intervalle de confiance de paramètre α .

```
tab_fleurs_bizarres <- vector("list", 4)

for (k in seq(1,4)) {
  mu <- mean(iris[,k])
  sigma2 <- var(iris[,k])

  v1 <- qnorm(0.025, mean = mu, sd = sqrt(sigma2))
  v2 <- qnorm(0.975, mean = mu, sd = sqrt(sigma2))

  indices <- c()

  for(i in seq(1,length(iris[,k]))) {
    if (iris[i,k]<v1 || iris[i,k]>v2) {
```

```

        indices <- c(indices, i)
    }
}
if (is.null(indices)) {
    indices <- c(-1)
}

tab_fleurs_bizarres[[k]] <- indices
}

tab_fleurs_bizarres

```

```

## [[1]]
## [1] 106 118 119 123 132 136
##
## [[2]]
## [1] 15 16 33 34 61 63 69 120
##
## [[3]]
## [1] -1
##
## [[4]]
## [1] -1

```

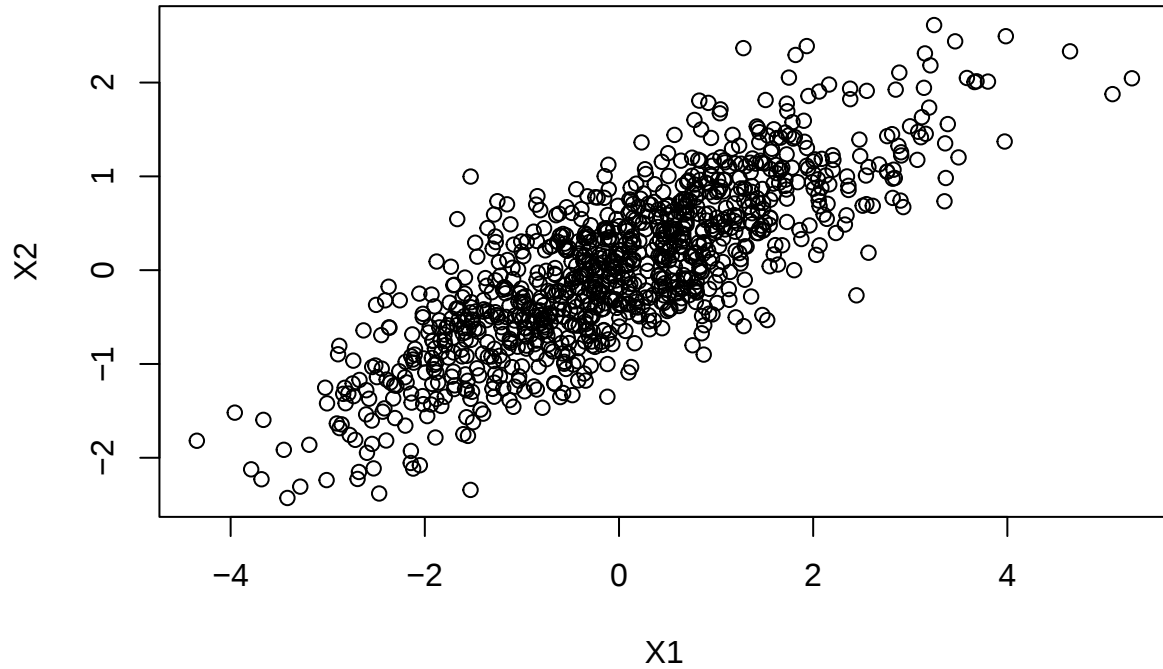
Ici, les -1 signifit que aucun spécimen de fleurs a des valeurs aberrantes concernant les pétales.

Exercice 4

Question 1

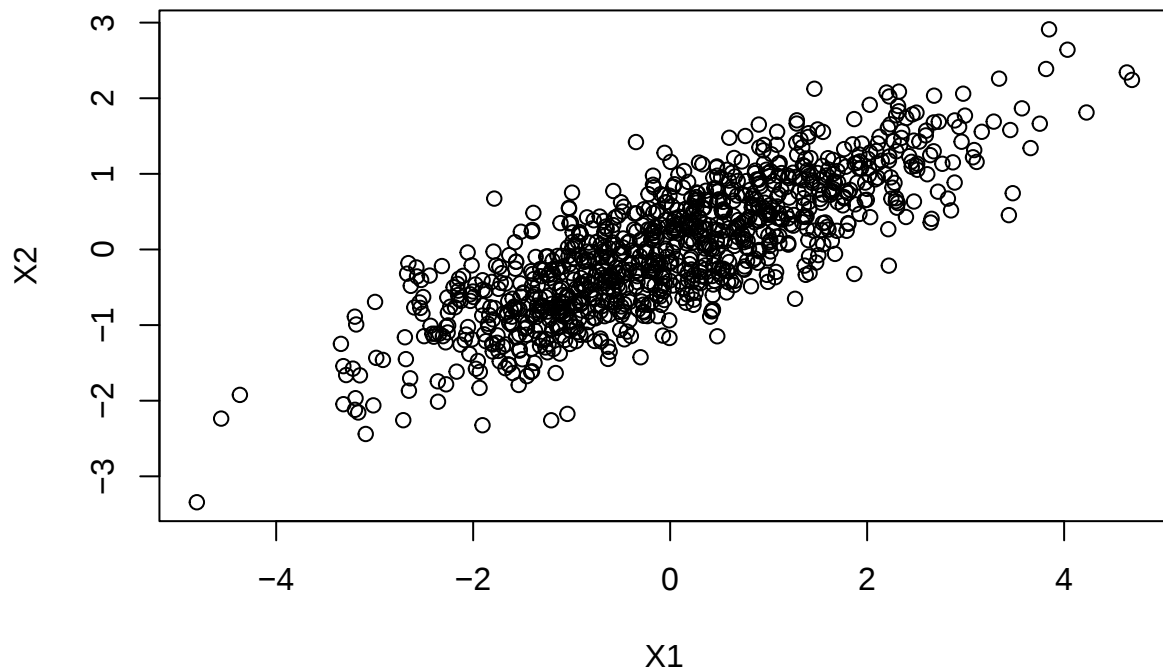
```
mu <- c(0, 0)
sigma <- matrix(c(2,1,1,0.75), ncol=2)

x <- matrix(rnorm(2000),1000)
sigm <- chol(sigma)
x <- x %*% sigm + mu
plot(x[, 1], x[, 2], xlab="X1", ylab="X2")
```



```
mu <- c(0, 0)
sigma <- matrix(c(2,1,1,0.75), ncol=2)

x <- mvrnorm(n=1000, mu=mu, Sigma=sigma)
plot(x[, 1], x[, 2], xlab="X1", ylab="X2")
```



Question 2

```
mu <- c(0, 0)
sigma <- matrix(c(2,1,1,0.75), ncol=2)

x <- matrix(rnorm(2000),1000)

racine_sigma <- chol(sigma)

y <- c()

for (i in seq(1,1000)) {
  yi <- racine_sigma %*% (x[i] - mu)
  y <- c(y, yi)
}

head(y)
```

```
## [1] -1.5493623 -0.3651882 -4.7940226 -1.1299620  0.3565664  0.0840435
```

```
data <- iris[,1:4]
n <- length(iris[,1])
C <- matrix(nrow = n, ncol=3)
```

```
for(i in seq(1,n)) {
  if (iris[i,]$Species == "setosa") {
    C[i,] <- c(1,0,0);
  } else if (iris[i,]$Species == "versicolor") {
    C[i,] <- c(0,1,0);
  } else {
    C[i,] <- c(0,0,1);
  }
}
```

Correction rapide du prof

```
library(nnet)
C<-class.ind(iris$Species)
```

Pour comprendre ce qui se passa dans le calcul du centre de gravité
`diag(t(C)%*%C)`

```
##      setosa versicolor  virginica
##      50          50          50
```

```
t(data)%*%C
```

```
##      setosa versicolor virginica
## Sepal.Length 250.3      296.8      329.4
## Sepal.Width  171.4      138.5      148.7
## Petal.Length  73.1      213.0      277.6
## Petal.Width   12.3       66.3      101.3
```

Et le centre de gravité de chaque classe :

```
t((t(data)%*%C))/diag(t(C)%*%C)
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## setosa           5.006       3.428       1.462       0.246
## versicolor       5.936       2.770       4.260       1.326
## virginica        6.588       2.974       5.552       2.026
```