

Mathematics applied to economics and management

Foundations of Descriptive and Inferential Statistics

December 2015 - Final assessment - Session 1 - Semester 1

Time allowed : 1h30 - All documents allowed

Exercise 1 *Answer :* 7pts

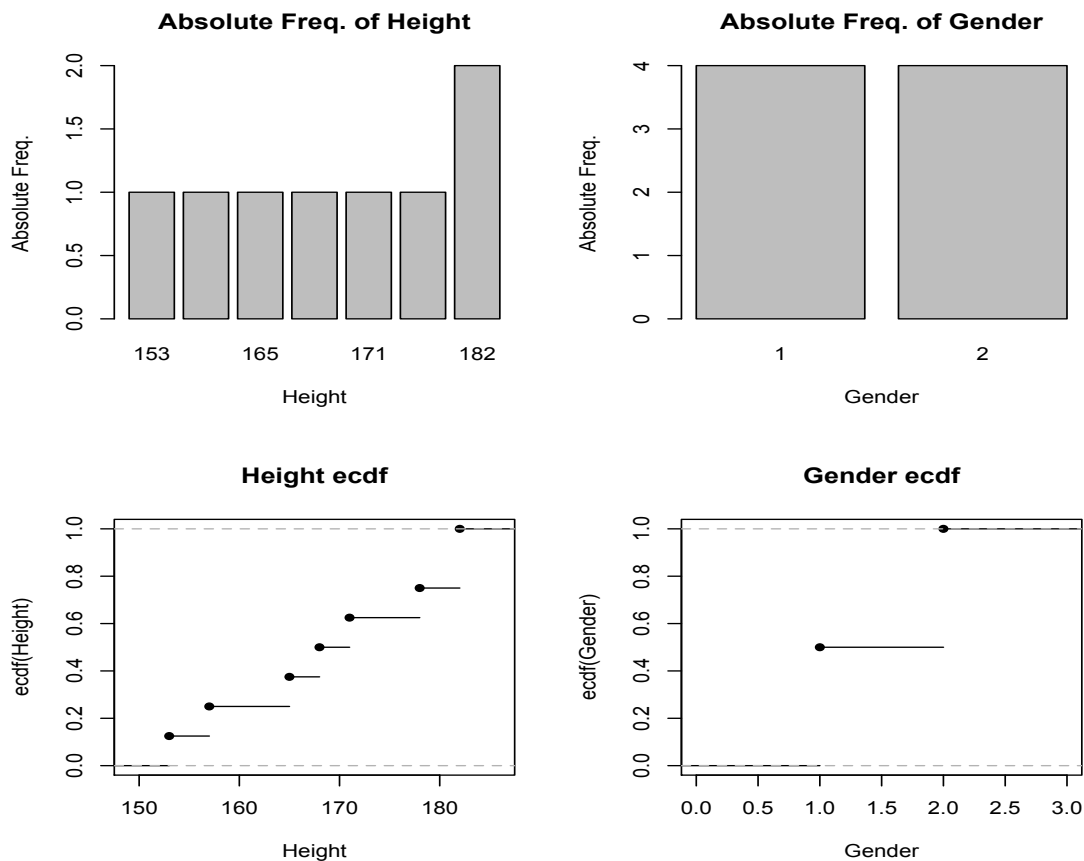
- 2pts Frequency tables

```
> Height<-c(165,157,168,178,171,182,182,153)
> Gender<-c("1","2","2","1","2","1","1","2")
> d1<-data.frame(table(Height))
> d11<-data.frame(d1,"Percent.Freq."=d1$Freq/length(Height)*100,
+ "Cumulative.Freq."=cumsum(d1$Freq))
> # At last, we may modify the label of the second column
> names(d11)[2]<-"Absolute.Freq."
> d11
  Height Absolute.Freq. Percent.Freq. Cumulative.Freq.
1   153             1         12.5             1
2   157             1         12.5             2
3   165             1         12.5             3
4   168             1         12.5             4
5   171             1         12.5             5
6   178             1         12.5             6
7   182             2         25.0             8

> d2<-data.frame(table(Gender))
> d22<-data.frame(d2,"Percent.Freq."=d2$Freq/length(Height)*100,
+ "Cumulative.Freq."=cumsum(d2$Freq))
> # At last, we may modify the label of the second column
> names(d22)[2]<-"Absolute.Freq."
> d22
  Gender Absolute.Freq. Percent.Freq. Cumulative.Freq.
1     1             4         50             4
2     2             4         50             8
```

- 2pts Graphs

```
> par(mfrow=c(2,2))
> barplot(table(Height),xlab="Height",ylab="Absolute Freq.",main="Absolute
+ Freq. of Height")
> barplot(table(Gender),xlab="Gender",ylab="Absolute Freq.",main="Absolute
+ Freq. of Gender")
> plot(ecdf(Height),xlab="Height",ylab="ecdf(Height)",main="Height ecdf")
> plot(ecdf(Gender),xlab="Gender",ylab="ecdf(Gender)",main="Gender ecdf")
```



- 1pt Contingency table

```
> table(Gender, Height)
      Height
Gender  153  157  165  168  171  178  182
    1     0    0    1    0    0    1    2
    2     1    1    0    1    1    0    0
```

- 2pts The two variables we want to compare do not belong to the same scale levels. The variable "Height" is metrically scaled while the variable "Gender" is nominally scaled. In order to describe the relation between Gender and Height, we may refer to <http://pbil.univ-lyon1.fr/R/pdf/tdr208.pdf> and use the following commands :

```
> vartot <- function(x) {res <- sum((x-mean(x))^2) ; return(res)}
> varinter <- function(x,gpe){
moyennes <- tapply(x,gpe,mean)
effectifs <- tapply(x,gpe,length)
res <- (sum(effectifs*(moyennes-mean(x))^2))
return(res)
}
> eta2 <- function(x,gpe){res <- varinter(x,gpe)/vartot(x) ; return(res)}
> eta2(Height, Gender)
[1] 0.50179
```

The result is not significant enough to state or not a link between the two variables. Nevertheless, we may provide some descriptive measures for each distribution with respect to the gender :

```
> split(Height, Gender)
```

```
$'1'
[1] 165 178 182 182
```

```
$'2'
[1] 157 168 171 153
```

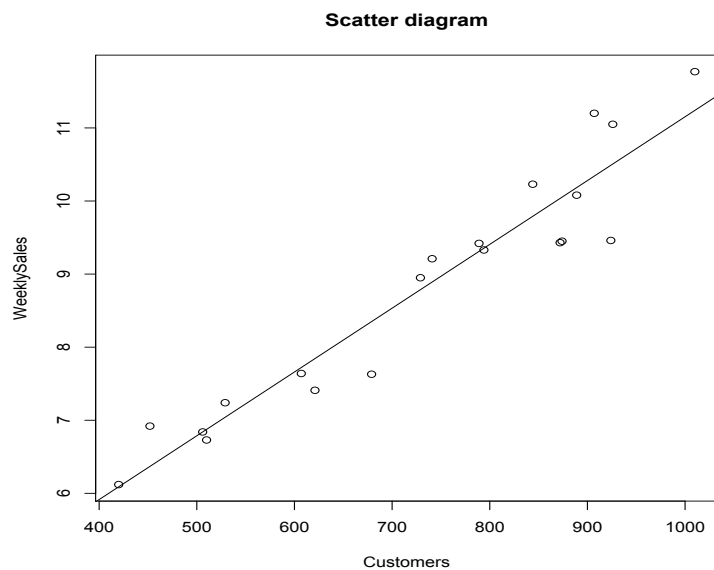
```
> summary(sp$'1')
  Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
165.0   174.8   180.0   176.8   182.0   182.0
> summary(sp$'2')
  Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
153.0   156.0   162.5   162.2   168.8   171.0
```

Since the measures are quite different, we may conclude that relatively to our distribution, the gender affects strongly the height.

Exercise 2 *Answer :* 4pts

1. 1pt

```
> Customers<-c(907,926,506,741,789,889,874,510,529,1010,
+ 420,679,872,924,607,452,729,794,844,621)
> WeeklySales<-c(11.20,11.05,6.84,9.21,9.42,10.08,9.45,6.73,7.24,11.77,
+ 6.12,7.63,9.43,9.46,7.64,6.92,8.95,9.33,10.23,7.41)
> plot(WeeklySales~Customers,main="Scatter diagram")
```



2. 1pt

```
> model<-lm(WeeklySales~Customers)
> model
```

Call :
lm(formula = WeeklySales~Customers)

Coefficients :
(Intercept) Customers
2.423044 0.008729

We deduce from the previous results that $b_0 = 2.423044$ and $b_1 = 0.008729$. We may plot the trend line whose equation is $Y = 0.008729X + 2.423044$ help to the command :

```
> abline(model)
```

3. 1pt The slope b_1 may be interpreted as the marginal value : when Customers increase to $t\%$, the Weekly Sales increase to $0.008729 \times t\%$.
4. 1pt The average weekly Sales (in thousands in dollars) for stores that have 600 customers may be estimated help to the following command :

```
> predict(model,list(Customers=c(600)))
```

```
1
7.660647
```

Exercise 3 *Answer :* 4pts

1. 1pt

```
> dim(iris)
```



```
[1]150 5
```



```
> str(iris)
```



```
> str(iris)
```



```
'data.frame' :      150 obs. of 5 variables :
$ Sepal.Length :   num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
$ Sepal.Width :    num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
$ Petal.Length :   num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
$ Petal.Width :    num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
$ Species :        Factor w/ 3 levels "setosa","versicolor",... : 1 1 1 1 1 1 1 1 1 1 ...
```
2. 1pt

```
> summary(iris)
```

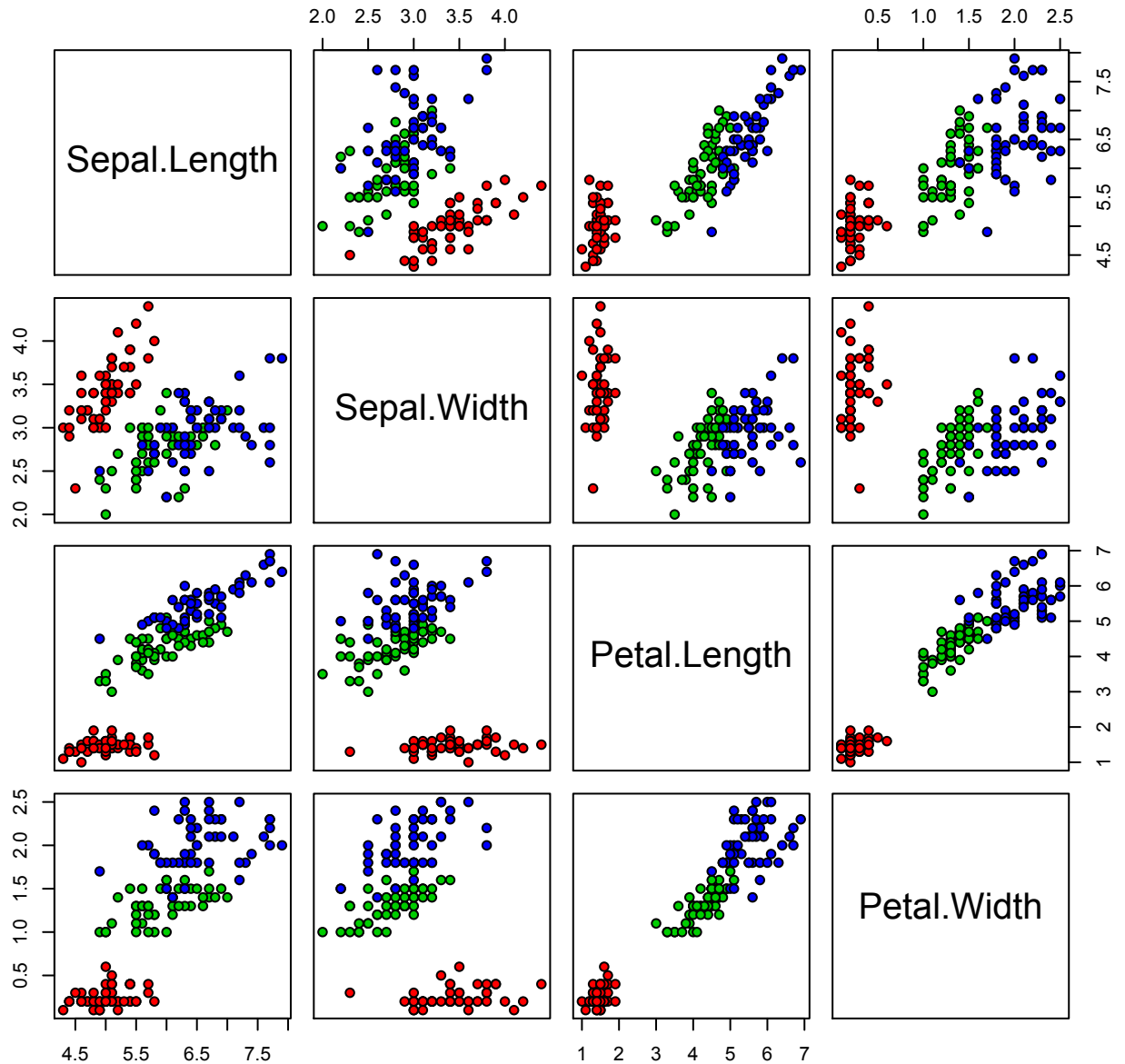

Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
Min. :4.300	Min. :2.000	Min. :1.000	Min. :0.100	setosa :50
1st Qu. :5.100	1st Qu. :2.800	1st Qu. :1.600	1st Qu. :0.300	versicolor :50
Median :5.800	Median :3.000	Median :4.350	Median :1.300	virginica :50
Mean :5.843	Mean :3.057	Mean :3.758	Mean :1.199	
3rd Qu. :6.400	3rd Qu. :3.300	3rd Qu. :5.100	3rd Qu. :1.800	
Max. :7.900	Max. :4.400	Max. :6.900	Max. :2.500	
3. 1pt The answer was available on internet, see

http://www2.warwick.ac.uk/fac/sci/moac/people/students/peter_cock/r/iris_plots/

for example. We may use the command :

```
> pairs(iris[1 :4], main = "Edgar Anderson's Iris Data", pch = 21, bg = c("red",
+ "green3","blue")[unclass(iris$Species)])
```

Edgar Anderson's Iris Data



4. `1pt` `> cor(iris[1:4])`

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width
Sepal.Length	1.0000000	-0.1175698	0.8717538	0.8179411
Sepal.Width	-0.1175698	1.0000000	-0.4284401	-0.3661259
Petal.Length	0.8717538	-0.4284401	1.0000000	0.9628654
Petal.Width	0.8179411	-0.3661259	0.9628654	1.0000000

Exercise 4 *Answer :* `5pts`

- `2,5pts` Frequency distributions

```
> Facebook<-c(0,11,17,16,22,17,25,30,27,27,31,35,30,45,60)
> Statistics<-c(25,22,28,30,22,27,26,21,27,28,23,29,30,24,27)
> class1X<-c(0,10,20,30,40,50,60)
> class2X<-c(0,15,30,60)
> class1Y<-c(18,22,26,30)
> class2Y<-c(18,21,24,27,30)
```

```

> class3Y<-c(18,20,22,24,26,28,30)
> table(cut(Facebook,class1X,right=F,include.lowest=T))

[0,10)  [10,20)  [20,30)  [30,40)  [40,50)  [50,60]
      1         4         4         4         1         1

> table(cut(Facebook,class2X,right=F,include.lowest=T))

[0,15)  [15,30)  [30,60]
      2         7         6

> table(cut(Statistics,class1Y,right=F,include.lowest=T))

[18,22)  [22,26)  [26,30]
      1         5         9

> table(cut(Statistics,class2Y,right=F,include.lowest=T))

[18,21)  [21,24)  [24,27)  [27,30]
      0         4         3         8

> table(cut(Statistics,class3Y,right=F,include.lowest=T))

[18,20)  [20,22)  [22,24)  [24,26)  [26,28)  [28,30]
      0         1         3         2         4         5

```

- 2,5pts Histograms

```

> split.screen( figs = c( 2, 1 ) )
[1] 1 2
> split.screen( figs = c( 1, 2 ), screen = 1 )
[1] 3 4
> split.screen( figs = c( 1, 3 ), screen = 2 )
[1] 5 6 7
> screen(3)
> hist(Facebook,breaks=class1X,right=F,main="Facebook frequencies Histogram\n
+ Grouping 1",xlab="Minutes",freq=T,include.lowest=T)
> screen(4)
> hist(Facebook,breaks=class2X,right=F,main="Facebook frequencies Histogram\n
+ Grouping 2",xlab="Minutes",freq=T,include.lowest=T)
> screen(5)
> hist(Statistics,breaks=class1Y,right=F,main="Statistics frequencies Histogram\n
+ Grouping 1",xlab="Grade",freq=T,include.lowest=T)
> screen(6)
> hist(Statistics,breaks=class2Y,right=F,main="Statistics frequencies Histogram\n
+ Grouping 2",xlab="Grade",freq=T,include.lowest=T)
> screen(7)
> hist(Statistics,breaks=class3Y,right=F,main="Statistics frequencies Histogram\n
+ Grouping 3",xlab="Grade",freq=T,include.lowest=T)

```

