
tle: "Devoir_2"
thor: "EL_Hadrami"
te: "10/11/2020"
tput: pdf_document

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0 --

## v ggplot2 3.3.2      v purrr 0.3.4
## v tibble 3.0.4       v dplyr 1.0.2
## v tidyr 1.1.2        v stringr 1.4.0
## v readr 1.4.0        v forcats 0.5.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

Exercice.13

```
# Creation d'un data-frame "acteur"
Mort.à <- c(93,53,72,68,68,53)
Années.de.carrière <- c(66,25,48,37,31,32)
Nombre.de.films <- c(211,58,98,140,74,81)
Prénom <- c("Michel","André","Jean","Louis","Lino","Jacques")
Nom <- c("Galabru","Raimbourg","Gabin","De Funès","Ventura","Villeret")
Date.du.décès <- c("04-01-2016","23-09-1970","15-10-1976","27-01-1983","22-10-1987","28-01-2005")
data.acteur <- data.frame(Mort.à,Années.de.carrière,Nombre.de.films,Prénom,Nom,Date.du.décès)
#utilisation d'un dplyer pour renommer la premiere variable
data.acteur.r <- rename(data.acteur,"Age.du.décès"=Mort.à)
# extraction de la colonne Prénom
prenom.extract <- data.acteur$Prénom
data.acteur.arrange <- arrange(data.acteur.r,Age.du.décès)
```

Exercice.14

Question 1.

```
w <- read.delim(file="data/fromages1-TP-M1.txt")
```

Question 2.

```
w <- rename(w,"mean.score"=Y,"c.a.a"=X1,"c.h.s"=X2,"c.a.l"=X3)
w$X1
```

```
## NULL
```

Question 3.:Les caracteristiques de w:

```
print(w)
```

```
##      mean.score c.a.a c.h.s c.a.l
## 1          12.3 4.543 3.135 0.86
## 2          20.9 5.159 5.043 1.53
## 3          39.0 5.366 5.438 1.57
## 4          47.9 5.759 7.496 1.81
## 5           5.6 4.663 3.807 0.99
## 6          25.9 5.697 7.601 1.09
```

```
## 7      37.3 5.892  8.726  1.29
## 8      21.9 6.078  7.966  1.78
## 9      18.1 4.898  3.850  1.29
## 10     21.0 5.242  4.174  1.58
## 11     34.9 5.740  6.142  1.68
## 12     57.2 6.446  7.908  1.90
## 13       0.7 4.477  2.996  1.06
## 14     25.9 5.236  4.942  1.30
## 15     54.9 6.151  6.752  1.52
## 16     40.9 6.365  9.588  1.74
## 17     15.9 4.787  3.912  1.16
## 18       6.4 5.412  4.700  1.49
## 19     18.0 5.247  6.174  1.63
## 20     38.9 5.438  9.064  1.99
## 21     14.0 4.564  4.949  1.15
## 22     15.2 5.298  5.220  1.33
## 23     32.0 5.455  9.242  1.44
## 24     56.7 5.855 10.199  2.01
## 25     16.8 5.366  3.664  1.31
## 26     11.6 6.043  3.219  1.46
## 27     26.5 6.458  6.962  1.72
## 28       0.7 5.328  3.912  1.25
## 29     13.4 5.802  6.685  1.08
## 30       5.5 6.176  4.787  1.25
```

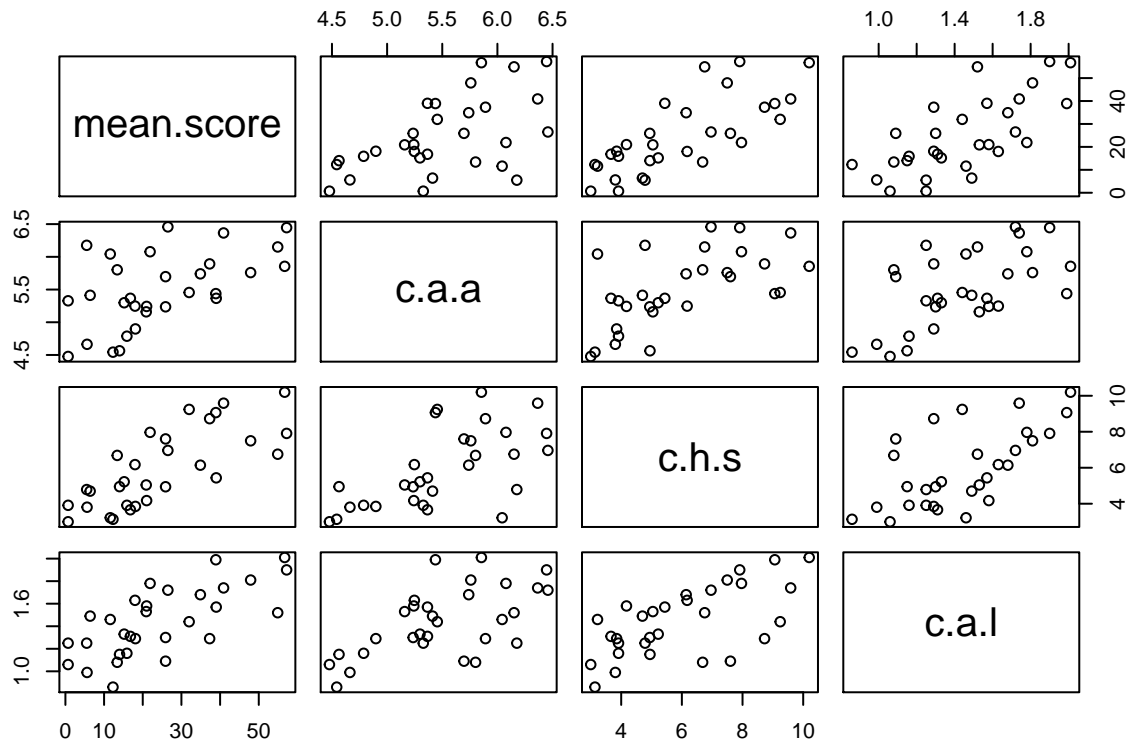
Question 4: Les paramètres statistiques des variables

```
summary(w)
```

```
##      mean.score      c.a.a      c.h.s      c.a.l
## Min.   : 0.70   Min.   :4.477   Min.   : 2.996   Min.   :0.860
## 1st Qu.:13.55   1st Qu.:5.237   1st Qu.: 3.978   1st Qu.:1.250
## Median :20.95   Median :5.425   Median : 5.329   Median :1.450
## Mean   :24.53   Mean   :5.498   Mean   : 5.942   Mean   :1.442
## 3rd Qu.:36.70   3rd Qu.:5.883   3rd Qu.: 7.575   3rd Qu.:1.667
## Max.   :57.20   Max.   :6.458   Max.   :10.199   Max.   :2.010
```

Question 5.:

```
pairs(w)
```



la commande pairs permet de tracer un nuage de point pour chaque variable afin de voir les différentes corrélations qui peuvent exister

Question 6.: Construction d'une nouvelle data frame

```
nv.c.a.a <- c(w$c.a.a[w$c.a.a > 5.1], rep(NA, 6))
nv.c.a.l <- c(w$c.a.l[w$c.a.l < 1.77], rep(NA, 5))
ww <- data.frame(w$mean.score, nv.c.a.a, w$c.h.s, nv.c.a.l)
ww <- rename(ww, "mean.score"=w.mean.score, "c.a.a"=nv.c.a.a, "c.h.s"=w.c.h.s, "c.a.l"=nv.c.a.l)
```

Question 7.: Les caractéristiques de ww

```
print(ww)
```

```
##   mean.score c.a.a  c.h.s c.a.l
## 1      12.3 5.159  3.135 0.86
## 2      20.9 5.366  5.043 1.53
## 3      39.0 5.759  5.438 1.57
## 4      47.9 5.697  7.496 0.99
## 5       5.6 5.892  3.807 1.09
## 6      25.9 6.078  7.601 1.29
## 7      37.3 5.242  8.726 1.29
## 8      21.9 5.740  7.966 1.58
## 9      18.1 6.446  3.850 1.68
## 10     21.0 5.236  4.174 1.06
## 11     34.9 6.151  6.142 1.30
## 12     57.2 6.365  7.908 1.52
## 13       0.7 5.412  2.996 1.74
## 14     25.9 5.247  4.942 1.16
## 15     54.9 5.438  6.752 1.49
## 16     40.9 5.298  9.588 1.63
## 17     15.9 5.455  3.912 1.15
```

```
## 18      6.4 5.855  4.700  1.33
## 19     18.0 5.366  6.174  1.44
## 20     38.9 6.043  9.064  1.31
## 21     14.0 6.458  4.949  1.46
## 22     15.2 5.328  5.220  1.72
## 23     32.0 5.802  9.242  1.25
## 24     56.7 6.176 10.199  1.08
## 25     16.8   NA  3.664  1.25
## 26     11.6   NA  3.219   NA
## 27     26.5   NA  6.962   NA
## 28       0.7   NA  3.912   NA
## 29     13.4   NA  6.685   NA
## 30       5.5   NA  4.787   NA
```

Question 8.: Les parametres statistiques de la variable ww

```
summary(ww)
```

```
##      mean.score      c.a.a      c.h.s      c.a.l
## Min.   : 0.70   Min.   :5.159   Min.   : 2.996   Min.   :0.860
## 1st Qu.:13.55   1st Qu.:5.356   1st Qu.: 3.978   1st Qu.:1.160
## Median :20.95   Median :5.718   Median : 5.329   Median :1.310
## Mean   :24.53   Mean   :5.709   Mean   : 5.942   Mean   :1.351
## 3rd Qu.:36.70   3rd Qu.:6.052   3rd Qu.: 7.575   3rd Qu.:1.530
## Max.   :57.20   Max.   :6.458   Max.   :10.199   Max.   :1.740
##                      NA's   :6                      NA's   :5
```

Exercice 15.

1.

```
data(airquality) # charger les données airquality
df.airquality <- data.frame(airquality)
```

2.Affichage des noms des variables

```
names(df.airquality)
```

```
## [1] "Ozone" "Solar.R" "Wind" "Temp" "Month" "Day"
```

3.nombre de ligne et de colonne

```
nrow(df.airquality)
```

```
## [1] 153
```

```
ncol(df.airquality)
```

```
## [1] 6
```

4.Les parametres statistiques

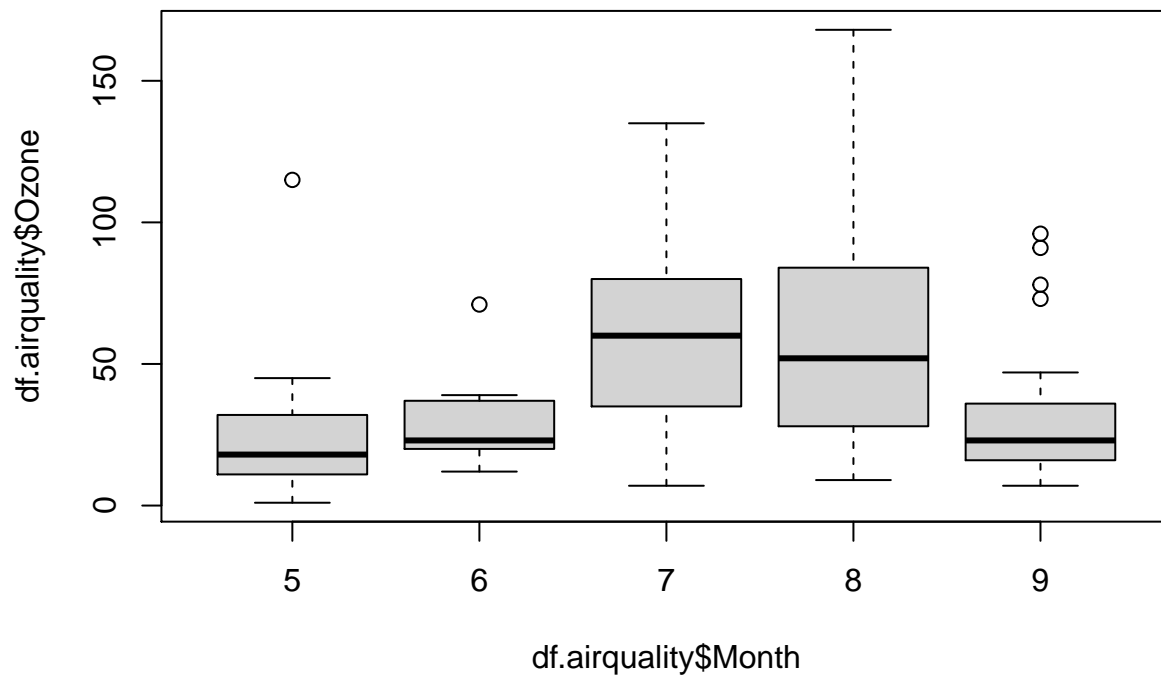
```
summary(df.airquality)
```

```
##      Ozone      Solar.R      Wind      Temp
## Min.   : 1.00   Min.   : 7.0   Min.   : 1.700   Min.   :56.00
## 1st Qu.:18.00   1st Qu.:115.8   1st Qu.: 7.400   1st Qu.:72.00
## Median :31.50   Median :205.0   Median : 9.700   Median :79.00
## Mean   :42.13   Mean   :185.9   Mean   : 9.958   Mean   :77.88
## 3rd Qu.:63.25   3rd Qu.:258.8   3rd Qu.:11.500   3rd Qu.:85.00
```

```
## Max. :168.00 Max. :334.0 Max. :20.700 Max. :97.00
## NA's :37 NA's :7
## Month Day
## Min. :5.000 Min. : 1.0
## 1st Qu.:6.000 1st Qu.: 8.0
## Median :7.000 Median :16.0
## Mean :6.993 Mean :15.8
## 3rd Qu.:8.000 3rd Qu.:23.0
## Max. :9.000 Max. :31.0
##
```

5.representation de la boite a moustache

```
boxplot(df.airquality$Ozone~df.airquality$Month)
```



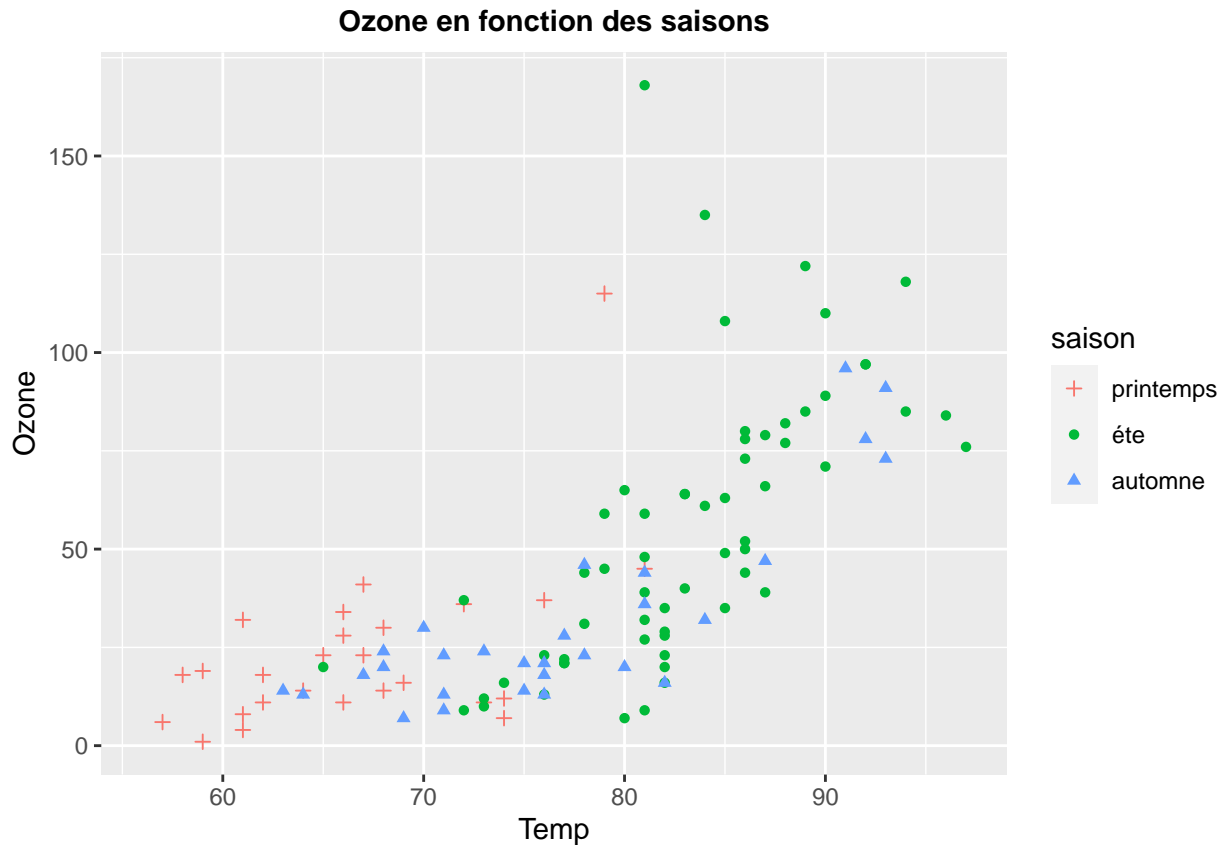
6.Creation d'une variable qualitative "saison"

```
saison <-factor(df.airquality$Month,levels=c(5:9))
levels(saison)[levels(saison)==5] <- "printemps"
levels(saison)[levels(saison)==6] <- "été"
levels(saison)[levels(saison)==7] <- "été"
levels(saison)[levels(saison)==8] <- "été"
levels(saison)[levels(saison)==9] <- "automne"
df.airquality$season = saison
```

7.

```
g <- ggplot(data = df.airquality) +
  geom_point(mapping = aes(x =Temp, y = Ozone, shape = saison,color=saison)) + scale_shape_manual(values=
  ggtitle("Ozone en fonction des saisons")
g + theme (plot.title = element_text(size=11,face="bold",hjust = 0.5))
```

```
## Warning: Removed 37 rows containing missing values (geom_point).
```



Exercice 16

1. Simulation de 100 valeurs suivant une loi normale

```
n <- 100
e <- rnorm(n,0,25)
```

2. Pour tout $i \in 1, \dots, 100$, on pose $y_i = 1.7 + 2.1i + e_i$

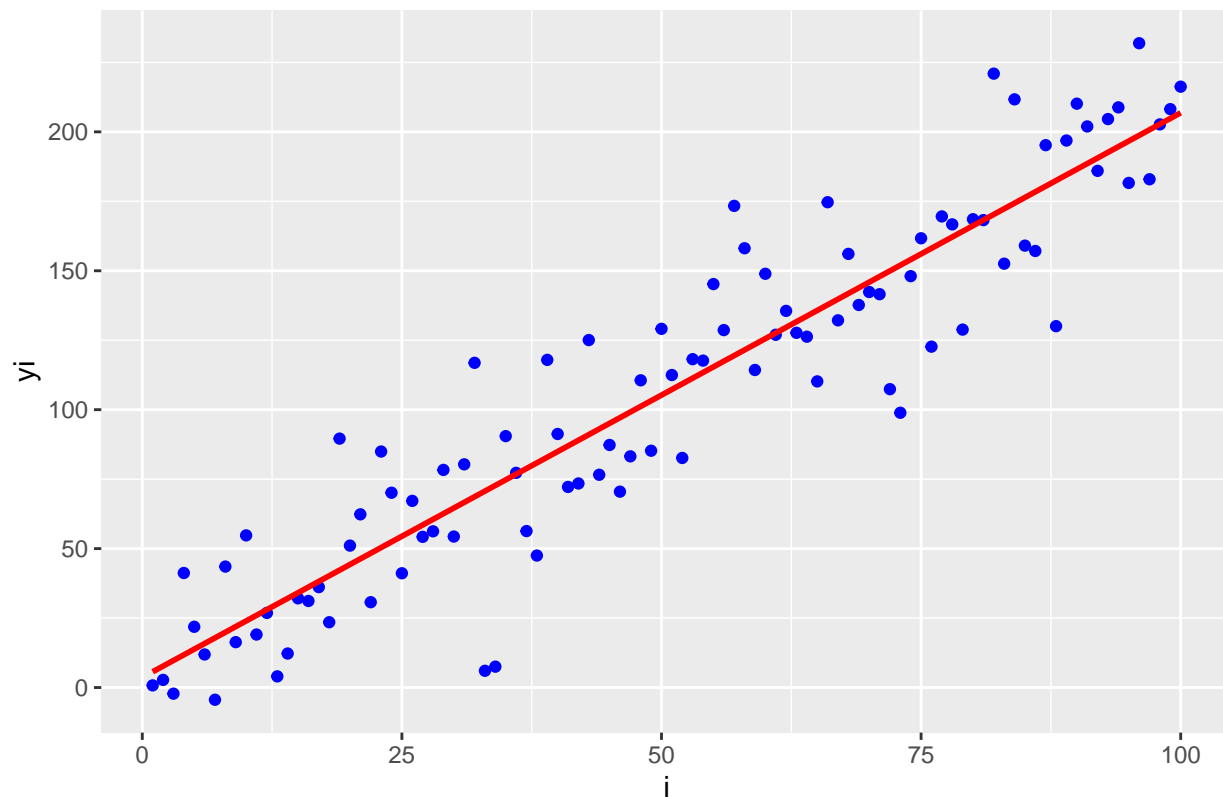
```
i <- c(1:100)
yi <- 1.7 + 2.1 * i + e[i]
```

2.a et 2.b: representation d'un nuage de point et une droite (i, y_i)

```
data.f <- data.frame(i,yi)
g <- ggplot(data = data.f) +
  geom_point(mapping = aes(x = i, y = yi),colour="blue") +
  geom_smooth(mapping = aes(x = i, y = yi),se = FALSE,colour="red",fill="red",method = lm) +
  ggtitle("nuage de points et sa droite de regression")
g + theme (plot.title = element_text(size=11,face="bold",hjust = 0.5))
```

```
## `geom_smooth()` using formula 'y ~ x'
```

nuage de points et sa droite de regression



Exercice 17

Creation d'une matrice

```
ligne1 <- c(68,119,26,7)
```

```
ligne2 <- c(15,54,14,10)
```

```
ligne3 <- c(5,29,14,16)
```

```
ligne4 <- c(20,84,17,94)
```

```
mat <- matrix(c(ligne1,ligne2,ligne3,ligne4),nrow = 4,ncol=4,byrow = T,dimnames = list(c("marron","noisette","vert","bleu"),c("brun","chatin","roux","blond")))
```

2. Calculer la matrice des fréquences (arrondi au 100ème près)

```
matfreq <- mat / sum(mat)
```

```
round(matfreq*100,2)
```

```
##          brun chatin roux blond
## marron  11.49  20.10  4.39  1.18
## noisette  2.53   9.12  2.36  1.69
## vert     0.84   4.90  2.36  2.70
## bleu     3.38  14.19  2.87 15.88
```

3. les lois marginales (nommer c pour le vecteur colonne et r pour le vecteur ligne)

```
l <- round(apply(matfreq,1,sum),2)
```

```
c <- round(apply(matfreq,2,sum),2)
```

4.Profils lignes

```
L <- round(sweep(mat,1,rowSums(mat),'/'),2)
```

5.Profils colonnes

```
C <- round(sweep(mat,2,colSums(mat),'/'),2)
```

6. La distance de chi-deux entre les profils lignes

```
d.chi <- 0
distancechideux <- function(L){
  for(i in 1:nrow(L)-1){

    d.chi <- d.chi + sum(((L[i,] - L[i+1,]) ^ 2) /c)
  }
  return (d.chi)
}
```

7. la matrice des taux de liaison

```
t <- round((matfreq - (1%*%t(c))) / (1%*%t(c)),2)
```

Exercice 18

1.

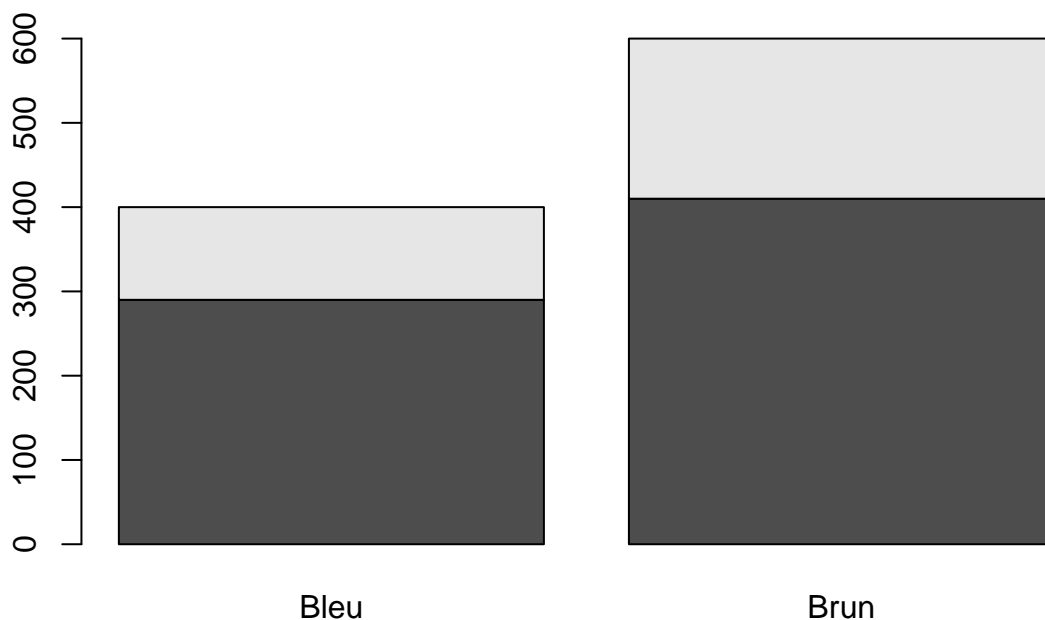
```
tableau <- matrix(c(290,410,110,190), ncol=2, byrow=TRUE)
colnames(tableau) <- c("Bleu","Brun")
rownames(tableau) <- c("Celib","Marie")
tableau <- as.table(tableau)
```

2.

```
print(tableau)
```

```
##      Bleu Brun
## Celib 290  410
## Marie 110  190
```

```
barplot(tableau)
```



3.


```
n <- margin.table(tableau)
m1 <- margin.table(tableau,1)
m2 <- margin.table(tableau,2)
prop.table(tableau)
```

```
##      Bleu Brun
## Celib 0.29 0.41
## Marie 0.11 0.19
```

4. Le test du chi-deux

```
chisq.test(tableau)
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data:  tableau
## X-squared = 1.7907, df = 1, p-value = 0.1808
```

4.a

```
tab0 <- as.array(m1) %*% t(as.array(m2))/n
tab0 <- as.table(tab0)
```

4.b

5.b

```
tab1 <- as.matrix(tableau)
tab1[2,1] <- tab1[1,1] + tab1[2,1]
tab1[1,1] <- 0
tab1[1,2] <- tab1[1,2] + tab1[2,2]
tab1[2,2] <- 0
tab1tab <- as.matrix(tab1)
chisq.test(tab1tab)
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data:  tab1tab
## X-squared = 995.84, df = 1, p-value < 2.2e-16
```

6. Test de khi-deux sur quelques échantillons de R

```
data(HairEyeColor)
dfH <- as.data.frame(HairEyeColor)
tH <- xtabs(Freq~Hair+Eye,dfH)
chisq.test(tH)
```

```
##
## Pearson's Chi-squared test
##
## data:  tH
## X-squared = 138.29, df = 9, p-value < 2.2e-16
```

```
data(Titanic)
dfT <- as.data.frame(Titanic)
dfTtab <- xtabs(Freq~Sex+Survived,dfT)
chisq.test(dfTtab)
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: dfTtab
## X-squared = 454.5, df = 1, p-value < 2.2e-16
```

```
data(UCBAdmissions)
dfU <- as.data.frame(UCBAdmissions)
dfUtab <- xtabs(Freq~Admit+Gender,dfU)
chisq.test(as.table(dfUtab))
```

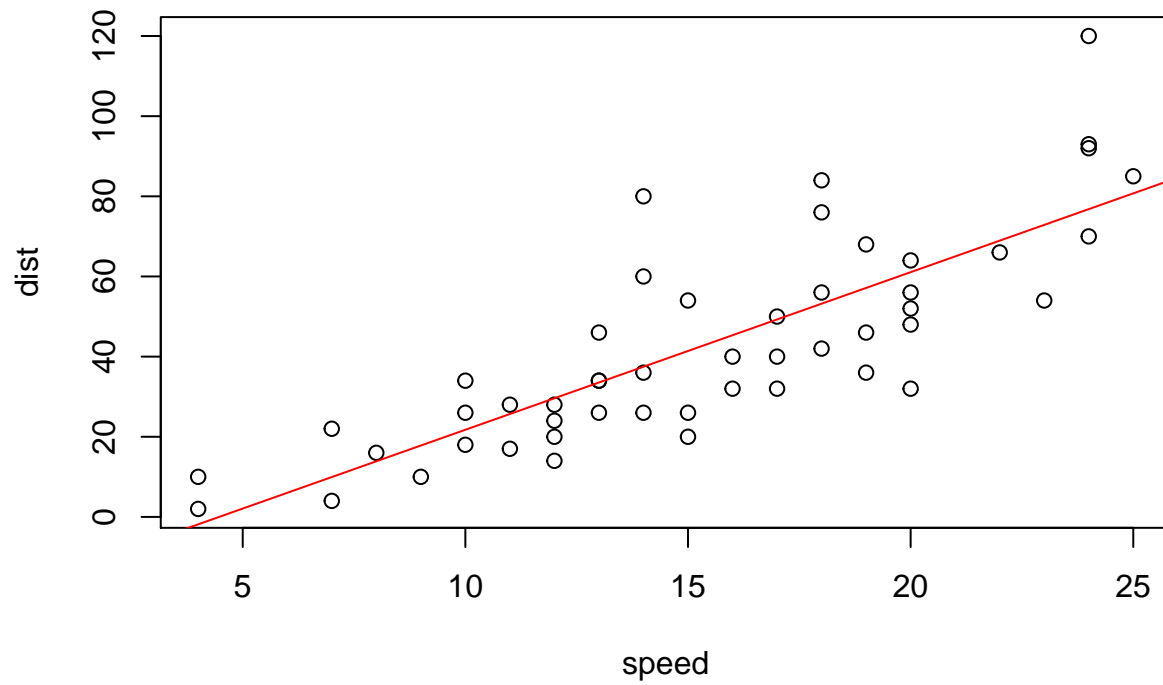
```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: as.table(dfUtab)
## X-squared = 91.61, df = 1, p-value < 2.2e-16
```

Exercice 19

```
data(cars)
# regression lineaire
reg <- lm(dist~speed,cars)
summary(reg)
```

```
##
## Call:
## lm(formula = dist ~ speed, data = cars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -29.069  -9.525  -2.272   9.215  43.201
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -17.5791     6.7584  -2.601   0.0123 *
## speed         3.9324     0.4155   9.464 1.49e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15.38 on 48 degrees of freedom
## Multiple R-squared:  0.6511, Adjusted R-squared:  0.6438
## F-statistic: 89.57 on 1 and 48 DF,  p-value: 1.49e-12
```

```
plot(cars)
abline(reg,col="red")
```



```
#abline(reg$coefficients,col="yellow")
```

19.a: La valeur prédite pour une vitesse de 20

```
predict(reg,newdata = data.frame(speed=20))
```

```
##          1  
## 61.06908
```