

Principal Component Analysis(PCA)

Problem statement 1

Perform Principal component analysis and perform clustering using first

3 principal component scores (both heirarchial and k mean clustering(scree plot or elbow curve) and obtain

optimum number of clusters and check whether we have obtained same number of clusters with the original data

Answer:

Rcode:

```
wine <- read.csv(file.choose())
```

```
View(wine)
```

```
str(wine)
```

```
names(wine)
```

```
class(wine)
```

```
summary(wine)
```

```
attach(wine)
```

```
sd(Type)
```

```
sd(Alcohol)
```

```
sd(Malic)
```

```
sd(Alcalinity)
```

```
sd(Magnesium)
```

```
sd(Phenols)
```

```
sd(Flavanoids)
```

```
sd(Nonflavanoids)
```

```
sd(Proanthocyanins)
```

```
sd(Color)
```

sd(Hue)

sd(Dilution)

sd(Proline)

var(Type)

var(Alcohol)

var(Malic)

var(Alcalinity)

var(Magnesium)

var(Phenols)

var(Flavanoids)

var(Nonflavanoids)

var(Proanthocyanins)

var(Color)

var(Hue)

var(Dilution)

var(Proline)

library(moments)

skewness(Type)

skewness(Alcohol)

skewness(Malic)

skewness(Alcalinity)

skewness(Magnesium)

skewness(Phenols)

skewness(Flavanoids)

skewness(Nonflavanoids)

skewness(Proanthocyanins)
skewness(Color)
skewness(Hue)
skewness(Dilution)
skewness(Proline)

kurtosis(Type)
kurtosis(Alcohol)
kurtosis(Malic)
kurtosis(Alcalinity)
kurtosis(Magnesium)
kurtosis(Phenols)
kurtosis(Flavanoids)
kurtosis(Nonflavanoids)
kurtosis(Proanthocyanins)
kurtosis(Color)
kurtosis(Hue)
kurtosis(Dilution)
kurtosis(Proline)

names(wine)
plot(Alcohol)
plot(Malic)
plot(Ash)
plot(Alcalinity)
plot(Magnesium)
plot(Phenols)
plot(Flavanoids)

```
plot(Nonflavanoids)
plot(Proanthocyanins)
plot(Color)
plot(Hue)
plot(Dilution)
plot(Proline)
```

```
hist(Alcohol)
hist(Malic)
hist(Ash)
hist(Alcalinity)
hist(Magnesium)
hist(Phenols)
hist(Flavanoids)
hist(Nonflavanoids)
hist(Proanthocyanins)
hist(Color)
hist(Hue)
hist(Dilution)
hist(Proline)
```

```
boxplot(Alcohol, horizontal = T, xlab="Alcohol")
boxplot(Malic, horizontal = T, xlab="Malic")
boxplot(Ash, horizontal = T, xlab="Ash")
boxplot(Alcalinity, horizontal = T, xlab="Alcalinity")
boxplot(Magnesium, horizontal = T, xlab="Magnesium")
boxplot(Phenols, horizontal = T, xlab="Phenols")
```

```
boxplot(Flavanoids,horizontal = T,xlab="Flavanoids")
boxplot(Nonflavanoids,horizontal = T,xlab="Nonflavanoids")
boxplot(Proanthocyanins,horizontal = T,xlab="Proanthocyanins")
boxplot(Color,horizontal = T,xlab="Color")
boxplot(Hue,horizontal = T,xlab="Hue")
boxplot(Dilution,horizontal = T,xlab="Dilution")
boxplot(Proline,horizontal = T,xlab="Proline")
```

```
mydata <- wine[,-1]
cor(mydata)
```

```
pcaobj <- princomp(mydata,cor = T,scores = T,covmat = NULL)
str(pcaobj)
loadings(pcaobj)
plot(pcaobj)
biplot(pcaobj)
plot(cumsum(pcaobj$sdev*pcaobj$sdev)*100/(sum(pcaobj$sdev*pcaobj$sdev)),ty
pe="b")
pcaobj$scores[,1:3]
mydata1<-cbind(wine,pcaobj$scores[,1:3])
View(mydata1)
```

```
# Hierarchial Clustering
```

```
clus_data<-wine[,8:10]
```

```
# Normalizing the data
```

```

norm_clus<-scale(clus_data)
distance<-dist(norm_clus,method = "euclidean")

fit<-hclust(distance,method="complete")

plot(fit)

rect.hclust(fit, k=7, border="red")
groups <- cutree(fit,5)
clust_1 <- as.matrix(groups)
View(clust_1)
final <- cbind(clust_1,mydata1)
View(final)
aggregate(final[,-c(2,16:18)],by=list(clust_1),FUN = mean)
aggregate(final[,-c(2,16:18)],by=list(clust_1),FUN = max)
aggregate(final[,-c(2,16:18)],by=list(clust_1),FUN = min)
write.csv(final,file = "data_clust.txt",row.names = F,col.names = F)
write.csv(final,file = "data_clust.csv",row.names = F,col.names = F)
getwd()

# K-Means Clustering :
library(plyr)
mydata2 <- final
View(mydata2)

normalized_data<-scale(mydata2[,15:17])
kmeans_clust <- kmeans(normalized_data,7)

```

```

str(kmeans_clust)

final1<- cbind(kmeans_clust$cluster,mydata2)

View(final1)

wss = (nrow(normalized_data)-1)*sum(apply(normalized_data, 2, var))  #
Determine number of clusters by scree-plot

for (i in 1:7) wss[i] = sum(kmeans(normalized_data, centers=i)$withinss)

plot(1:7, wss, type="b", xlab="Number of Clusters", ylab="Within groups sum of
squares") # Look for an "elbow" in the scree plot #

title(sub = "K-Means Clustering Scree-Plot")

aggregate(mydata2[,2:12],by=list(kmeans_clust$cluster),FUN = mean)

aggregate(mydata2[,2:12],by=list(kmeans_clust$cluster),FUN = min)

aggregate(mydata2[,2:12],by=list(kmeans_clust$cluster),FUN = max)

kmeans_clust$centers

table(kmeans_clust$cluster)

write.csv(final1,file = "data_clust1.csv",row.names = F,col.names = F)

getwd()

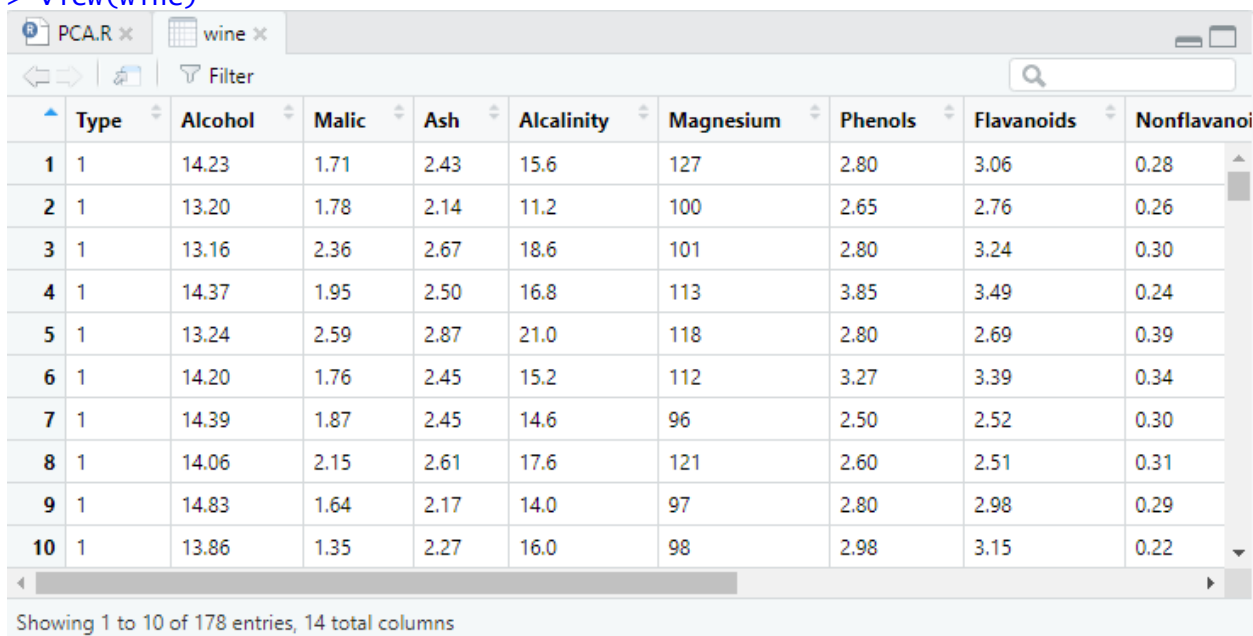
```

Console:

```

> wine <- read.csv(file.choose())
> View(wine)

```



	Type	Alcohol	Malic	Ash	Alkalinity	Magnesium	Phenols	Flavanoids	Nonflavonoids
1	1	14.23	1.71	2.43	15.6	127	2.80	3.06	0.28
2	1	13.20	1.78	2.14	11.2	100	2.65	2.76	0.26
3	1	13.16	2.36	2.67	18.6	101	2.80	3.24	0.30
4	1	14.37	1.95	2.50	16.8	113	3.85	3.49	0.24
5	1	13.24	2.59	2.87	21.0	118	2.80	2.69	0.39
6	1	14.20	1.76	2.45	15.2	112	3.27	3.39	0.34
7	1	14.39	1.87	2.45	14.6	96	2.50	2.52	0.30
8	1	14.06	2.15	2.61	17.6	121	2.60	2.51	0.31
9	1	14.83	1.64	2.17	14.0	97	2.80	2.98	0.29
10	1	13.86	1.35	2.27	16.0	98	2.98	3.15	0.22

Showing 1 to 10 of 178 entries, 14 total columns

```

> str(wine)
'data.frame': 178 obs. of 14 variables:
 $ Type      : int  1 1 1 1 1 1 1 1 1 1 ...
 $ Alcohol   : num  14.2 13.2 13.2 14.4 13.2 ...
 $ Malic     : num  1.71 1.78 2.36 1.95 2.59 1.76 1.87 2.15 1.64 1.35 ..
 .
 $ Ash       : num  2.43 2.14 2.67 2.5 2.87 2.45 2.45 2.61 2.17 2.27 ...
 $ Alcalinity : num  15.6 11.2 18.6 16.8 21 15.2 14.6 17.6 14 16 ...
 $ Magnesium  : int  127 100 101 113 118 112 96 121 97 98 ...
 $ Phenols    : num  2.8 2.65 2.8 3.85 2.8 3.27 2.5 2.6 2.8 2.98 ...
 $ Flavanoids : num  3.06 2.76 3.24 3.49 2.69 3.39 2.52 2.51 2.98 3.15 ..
 .
 $ Nonflavanoids : num  0.28 0.26 0.3 0.24 0.39 0.34 0.3 0.31 0.29 0.22 ...
 $ Proanthocyanins: num  2.29 1.28 2.81 2.18 1.82 1.97 1.98 1.25 1.98 1.85 ..
 .
 $ Color      : num  5.64 4.38 5.68 7.8 4.32 6.75 5.25 5.05 5.2 7.22 ...
 $ Hue        : num  1.04 1.05 1.03 0.86 1.04 1.05 1.02 1.06 1.08 1.01 ..
 .
 $ Dilution   : num  3.92 3.4 3.17 3.45 2.93 2.85 3.58 3.58 2.85 3.55 ...
 $ Proline     : int  1065 1050 1185 1480 735 1450 1290 1295 1045 1045 ...

> names(wine)
[1] "Type"           "Alcohol"        "Malic"          "Ash"
[5] "Alcalinity"     "Magnesium"      "Phenols"        "Flavanoids"
[9] "Nonflavanoids"  "Proanthocyanins" "Color"          "Hue"
[13] "Dilution"      "Proline"

> class(wine)
[1] "data.frame"

> summary(wine)
      Type      Alcohol      Malic      Ash
Min.   :1.000   Min.   :11.03   Min.   :0.740   Min.   :1.360
1st Qu.:1.000   1st Qu.:12.36   1st Qu.:1.603   1st Qu.:2.210
Median :2.000   Median :13.05   Median :1.865   Median :2.360
Mean   :1.938   Mean   :13.00   Mean   :2.336   Mean   :2.367
3rd Qu.:3.000   3rd Qu.:13.68   3rd Qu.:3.083   3rd Qu.:2.558
Max.   :3.000   Max.   :14.83   Max.   :5.800   Max.   :3.230
      Alcalinity  Magnesium  Phenols  Flavanoids
Min.   :10.60    Min.   : 70.00    Min.   :0.980    Min.   :0.340
1st Qu.:17.20    1st Qu.: 88.00    1st Qu.:1.742    1st Qu.:1.205
Median :19.50    Median : 98.00    Median :2.355    Median :2.135
Mean   :19.49    Mean   : 99.74    Mean   :2.295    Mean   :2.029
3rd Qu.:21.50    3rd Qu.:107.00   3rd Qu.:2.800    3rd Qu.:2.875
Max.   :30.00    Max.   :162.00   Max.   :3.880    Max.   :5.080
Nonflavanoids  Proanthocyanins  Color      Hue
Min.   :0.1300   Min.   :0.410    Min.   : 1.280    Min.   :0.4800
1st Qu.:0.2700   1st Qu.:1.250    1st Qu.: 3.220    1st Qu.:0.7825
Median :0.3400   Median :1.555    Median : 4.690    Median :0.9650
Mean   :0.3619   Mean   :1.591    Mean   : 5.058    Mean   :0.9574
3rd Qu.:0.4375   3rd Qu.:1.950    3rd Qu.: 6.200    3rd Qu.:1.1200
Max.   :0.6600   Max.   :3.580    Max.   :13.000    Max.   :1.7100
      Dilution  Proline
Min.   :1.270    Min.   : 278.0
1st Qu.:1.938    1st Qu.: 500.5
Median :2.780    Median : 673.5
Mean   :2.612    Mean   : 746.9
3rd Qu.:3.170    3rd Qu.: 985.0
Max.   :4.000    Max.   :1680.0

> attach(wine)
> sd(Type)
[1] 0.775035
> sd(Alcohol)
[1] 0.8118265
> sd(Malic)
[1] 1.117146
> sd(Alcalinity)
[1] 3.339564
> sd(Magnesium)
[1] 14.28248
> sd(Phenols)

```



```

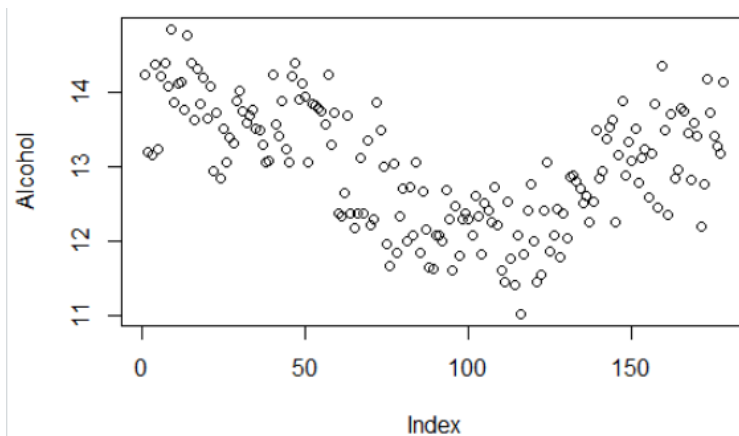
[1] 0.625851
> sd(Flavanoids)
[1] 0.9988587
> sd(Nonflavanoids)
[1] 0.1244533
> sd(Proanthocyanins)
[1] 0.5723589
> sd(Color)
[1] 2.318286
> sd(Hue)
[1] 0.2285716
> sd(Dilution)
[1] 0.7099904
> sd(Proline)
[1] 314.9075
>
> var(Type)
[1] 0.6006792
> var(Alcohol)
[1] 0.6590623
> var(Malic)
[1] 1.248015
> var(Alcalinity)
[1] 11.15269
> var(Magnesium)
[1] 203.9893
> var(Phenols)
[1] 0.3916895
> var(Flavanoids)
[1] 0.9977187
> var(Nonflavanoids)
[1] 0.01548863
> var(Proanthocyanins)
[1] 0.3275947
> var(Color)
[1] 5.374449
> var(Hue)
[1] 0.05224496
> var(Dilution)
[1] 0.5040864
> var(Proline)
[1] 99166.72
>
> library(moments)
> skewness(Type)
[1] 0.1065237
> skewness(Alcohol)
[1] -0.05104747
> skewness(Malic)
[1] 1.030869
> skewness(Alcalinity)
[1] 0.2112473
> skewness(Magnesium)
[1] 1.088915
> skewness(Phenols)
[1] 0.08590677
> skewness(Flavanoids)
[1] 0.02512948
> skewness(Nonflavanoids)
[1] 0.446349
> skewness(Proanthocyanins)
[1] 0.512769
> skewness(Color)
[1] 0.8612481
> skewness(Hue)
[1] 0.02091312
> skewness(Dilution)
[1] -0.3046899

```

```

> skewness(Proline)
[1] 0.7613362
> kurtosis(Type)
[1] 1.68056
> kurtosis(Alcohol)
[1] 2.13774
> kurtosis(Malic)
[1] 3.257348
> kurtosis(Alcalinity)
[1] 3.440823
> kurtosis(Magnesium)
[1] 5.012806
> kurtosis(Phenols)
[1] 2.154143
> kurtosis(Flavanoids)
[1] 2.110635
> kurtosis(Nonflavanoids)
[1] 2.347048
> kurtosis(Proanthocyanins)
[1] 3.505671
> kurtosis(Color)
[1] 3.33737
> kurtosis(Hue)
[1] 2.631975
> kurtosis(Dilution)
[1] 1.910325
> kurtosis(Proline)
[1] 2.725
> plot(Alcohol)

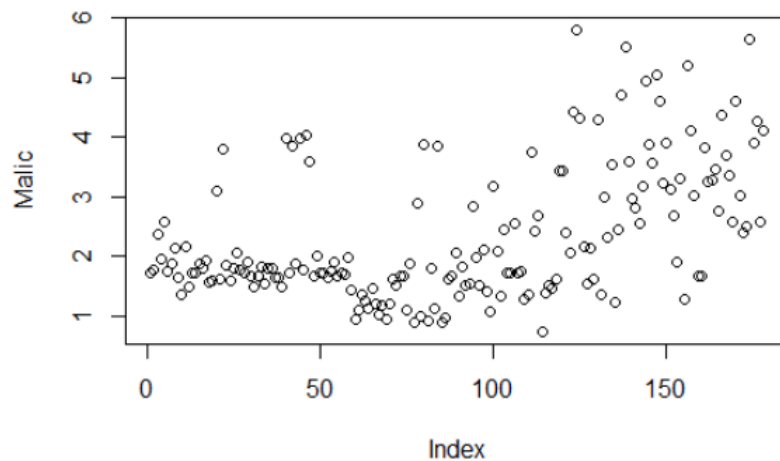
```



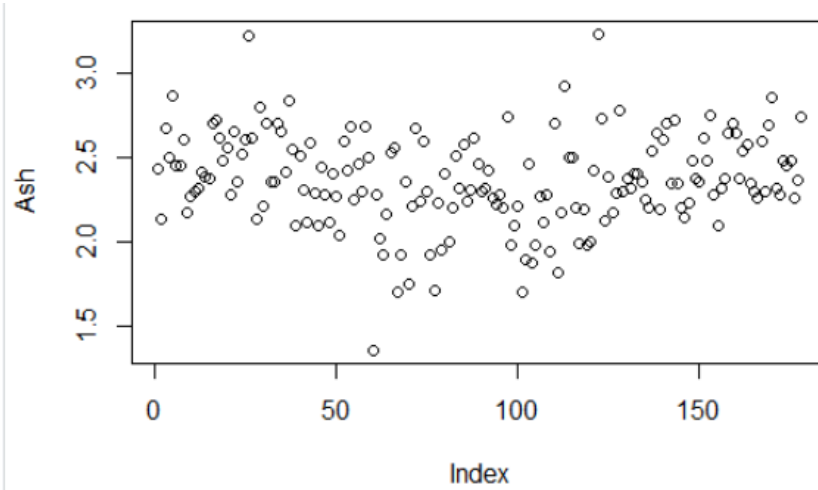
```

> plot(Malic)

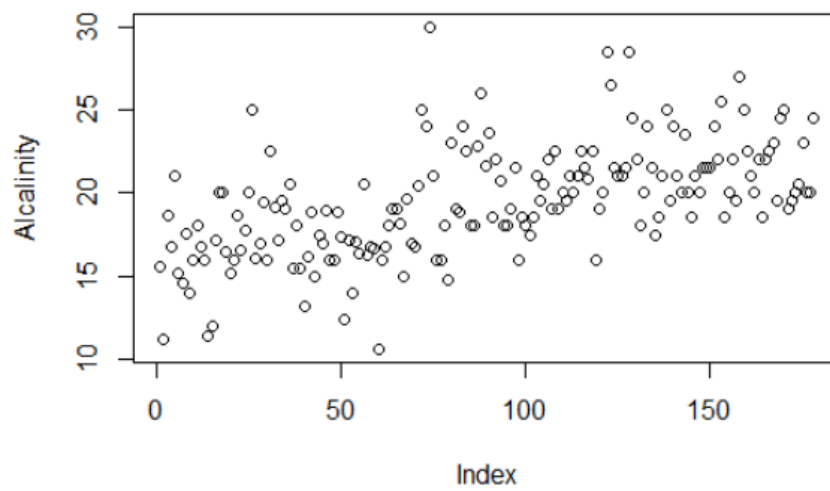
```



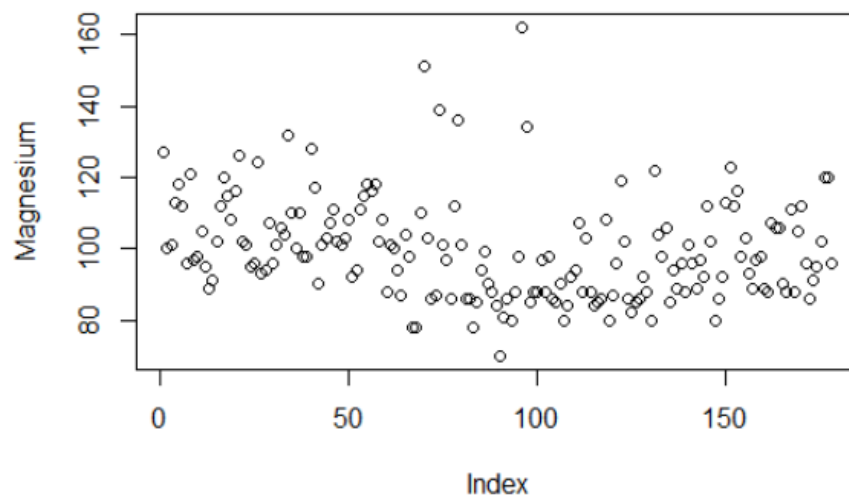
```
> plot(Ash)
```



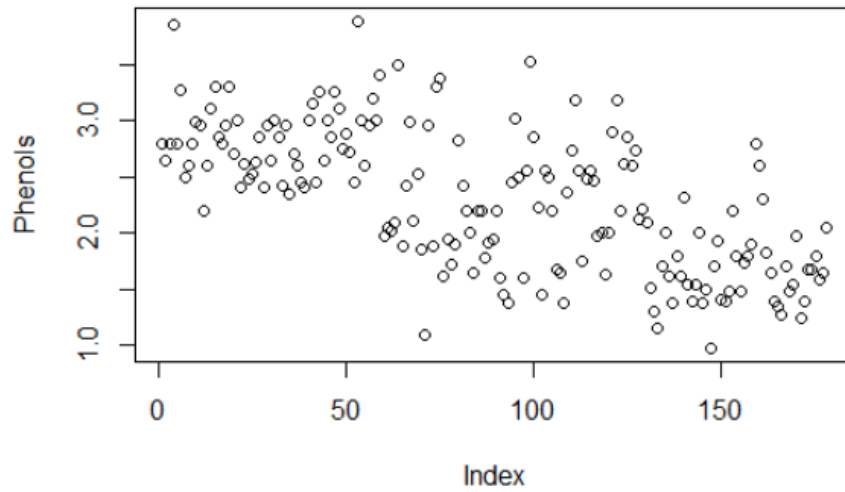
```
> plot(Alcalinity)
```



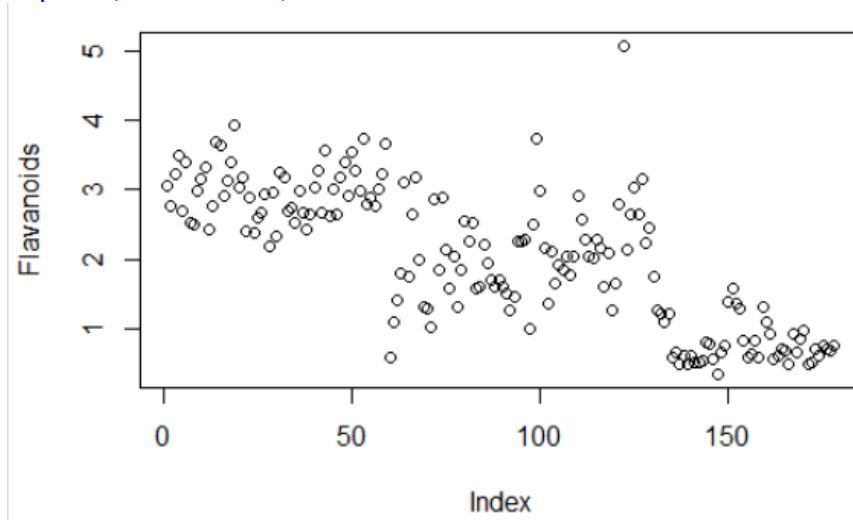
```
> plot(Magnesium)
```



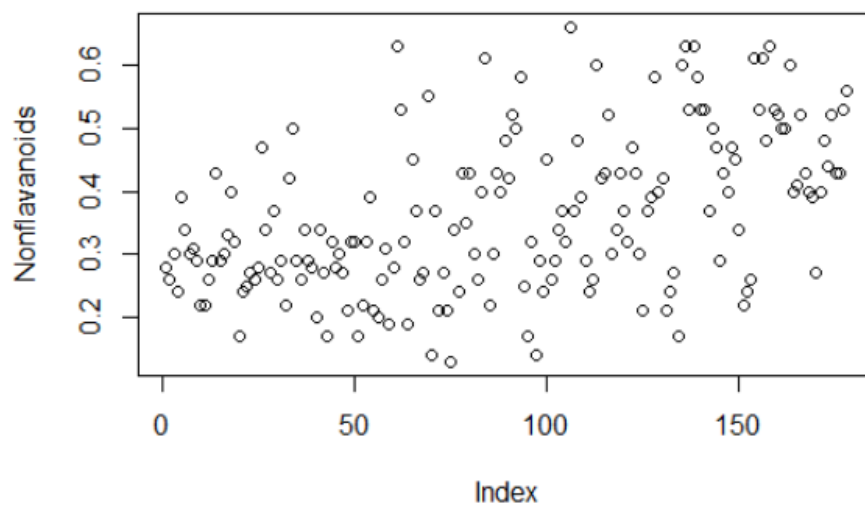
```
> plot(Phenols)
```



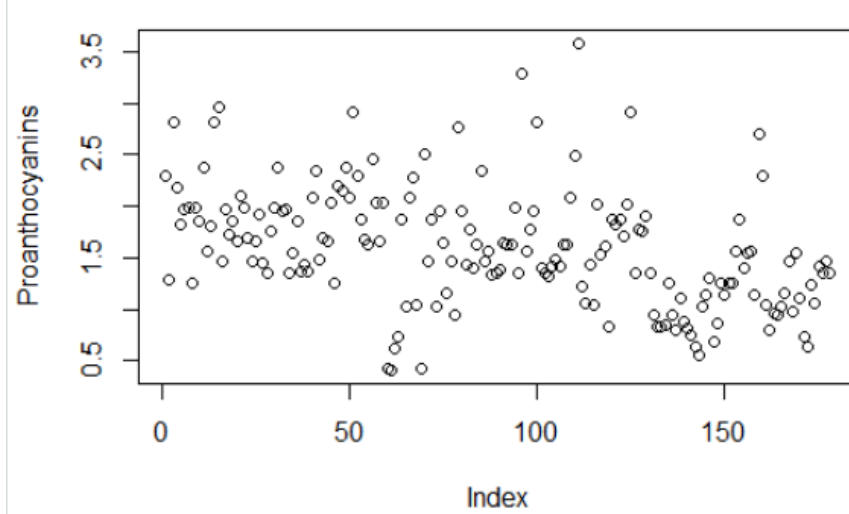
```
> plot(Flavanoids)
```



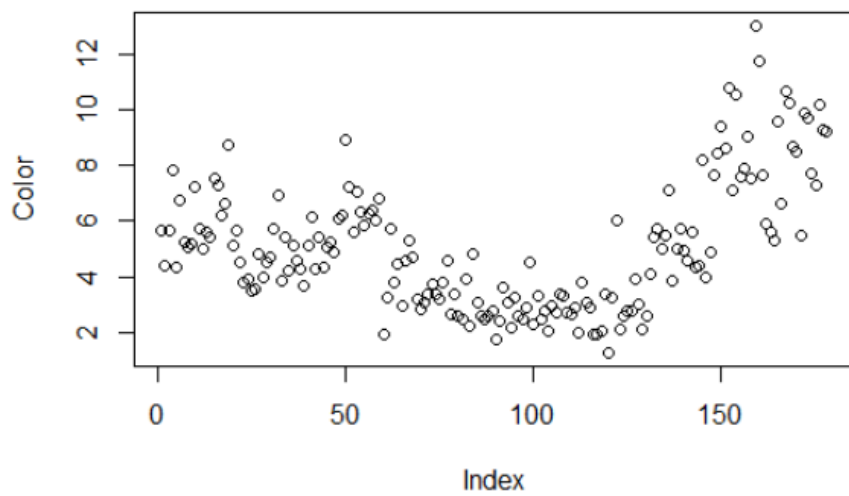
```
> plot(Nonflavanoids)
```



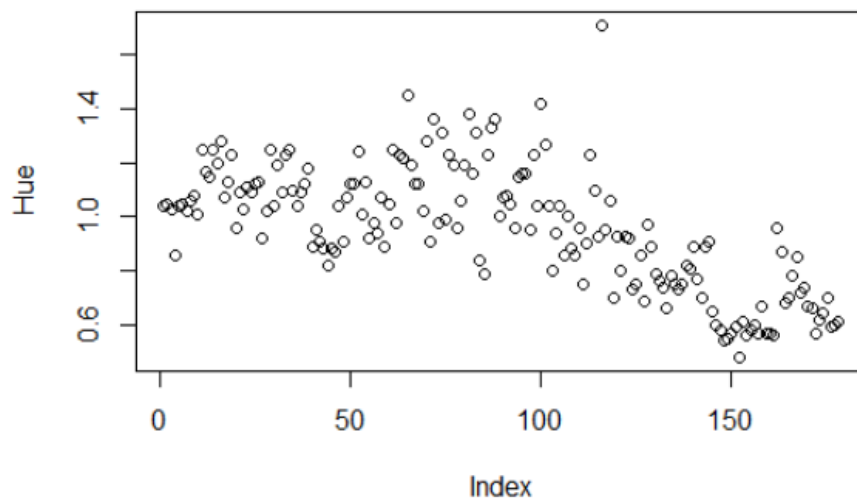
```
> plot(Proanthocyanins)
```



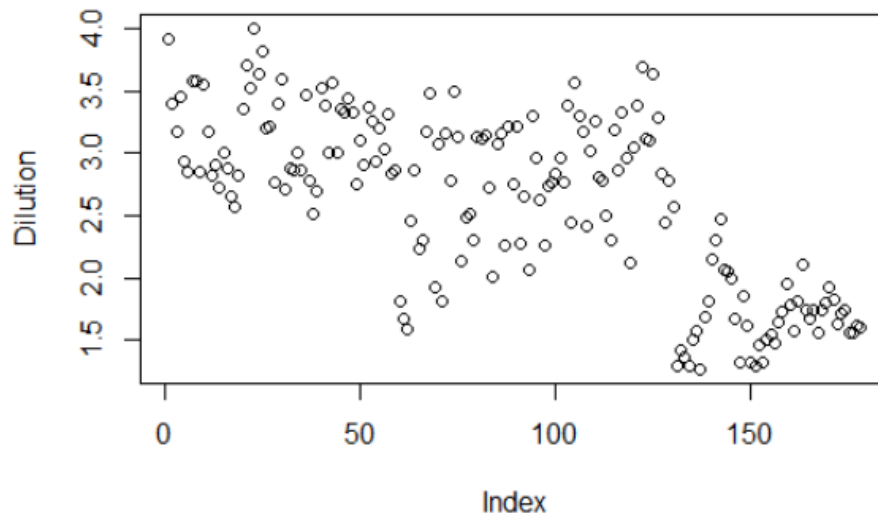
```
> plot(Color)
```



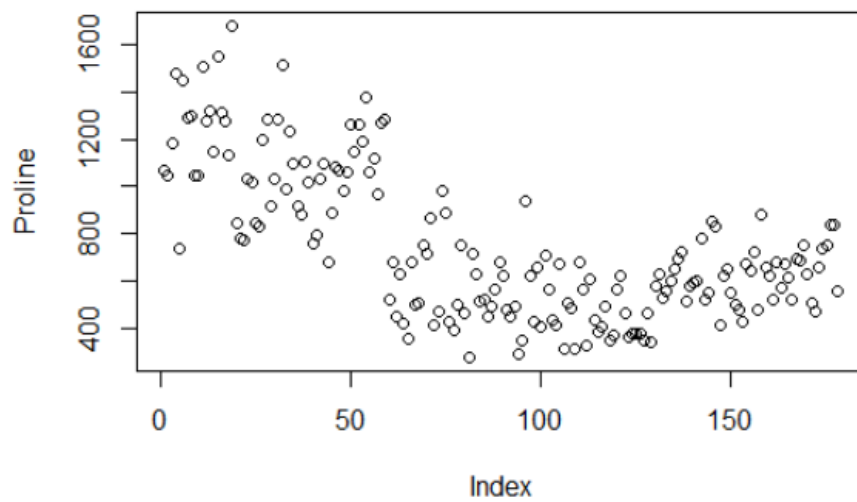
```
> plot(Hue)
```



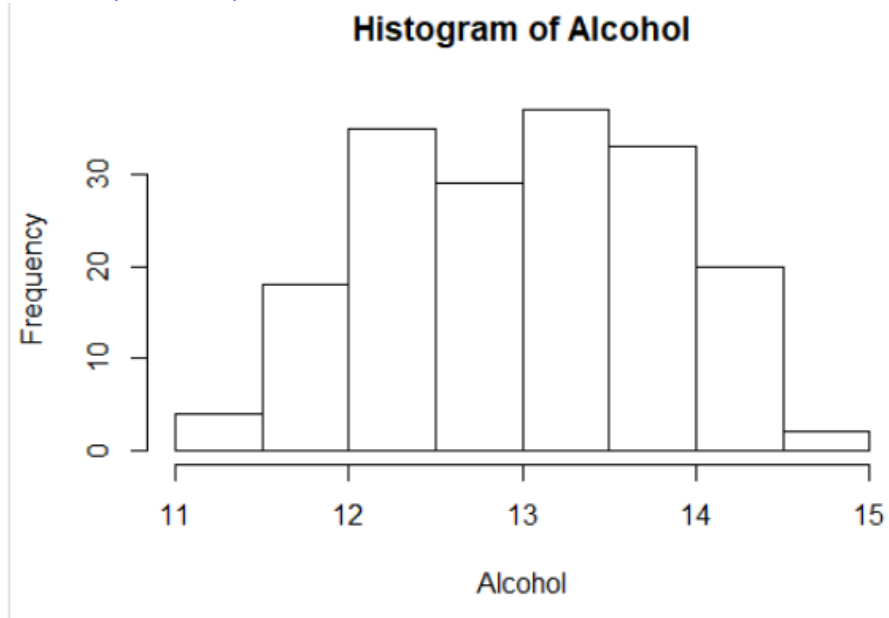
```
> plot(Dilution)
```



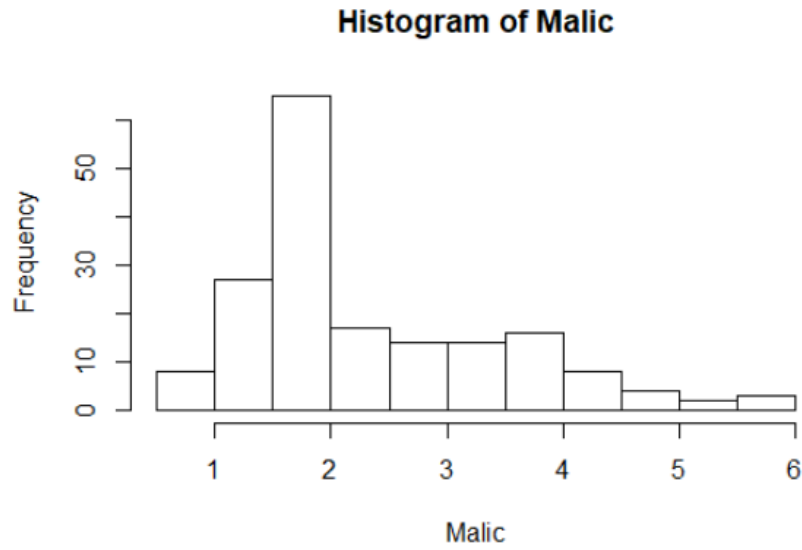
```
> plot(Proline)
```



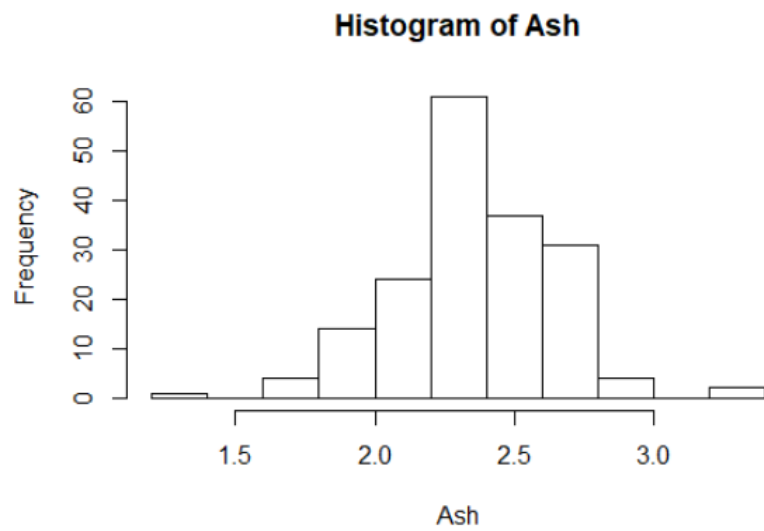
```
> hist(Alcohol)
```



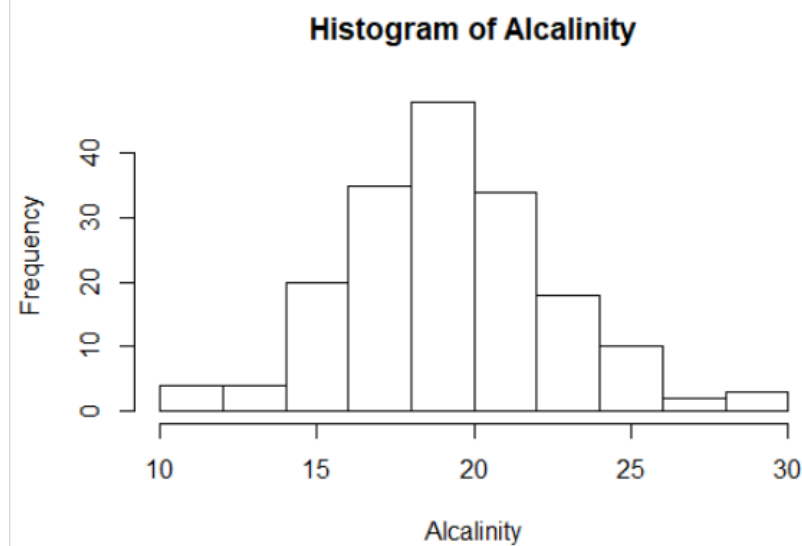
```
> hist(Malic)
```



```
> hist(Ash)
```

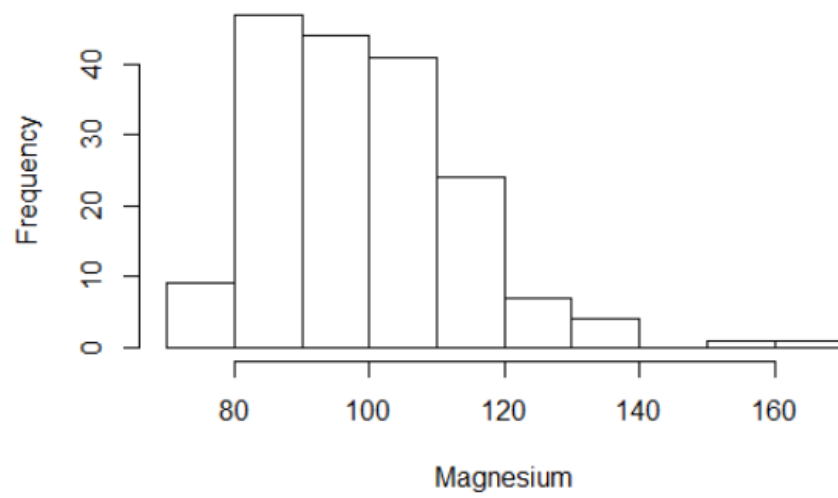


```
> hist(Alcalinity)
```



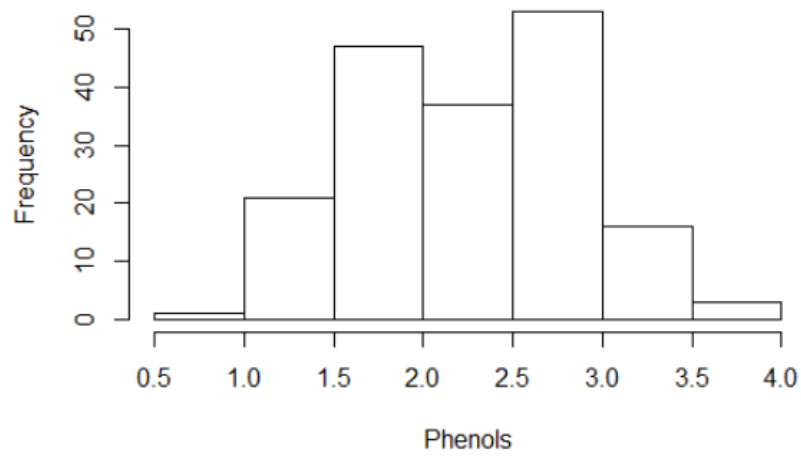
```
> hist(Magnesium)
```

Histogram of Magnesium



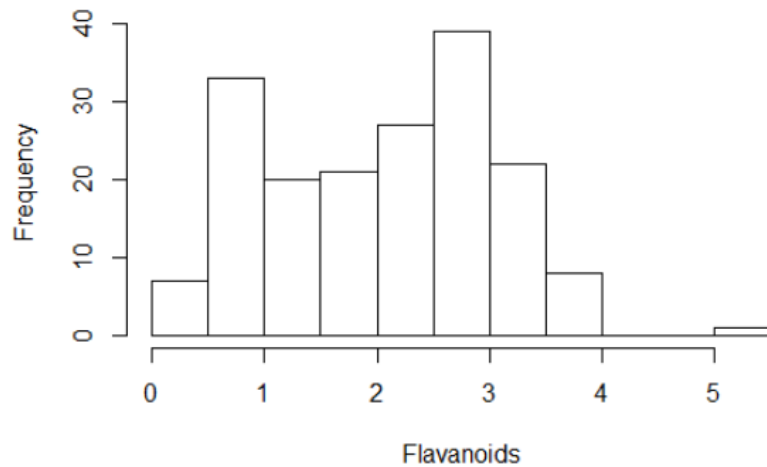
```
> hist(Phenols)
```

Histogram of Phenols

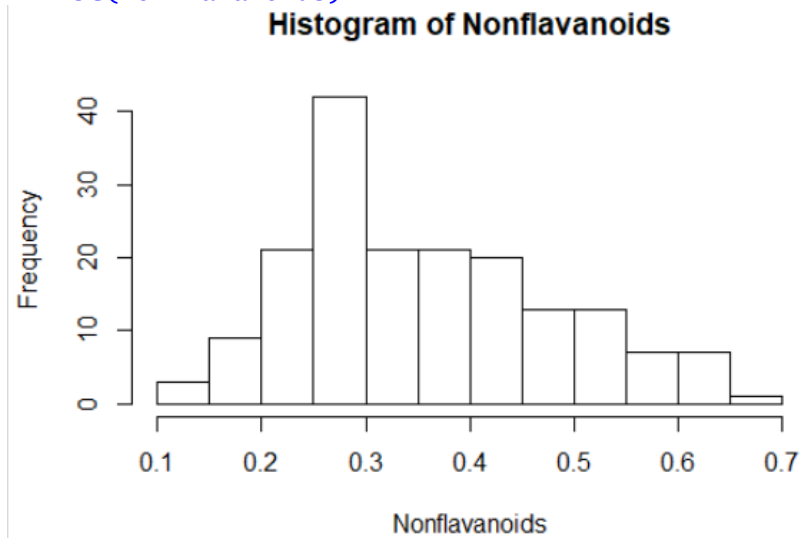


```
> hist(Flavanoids)
```

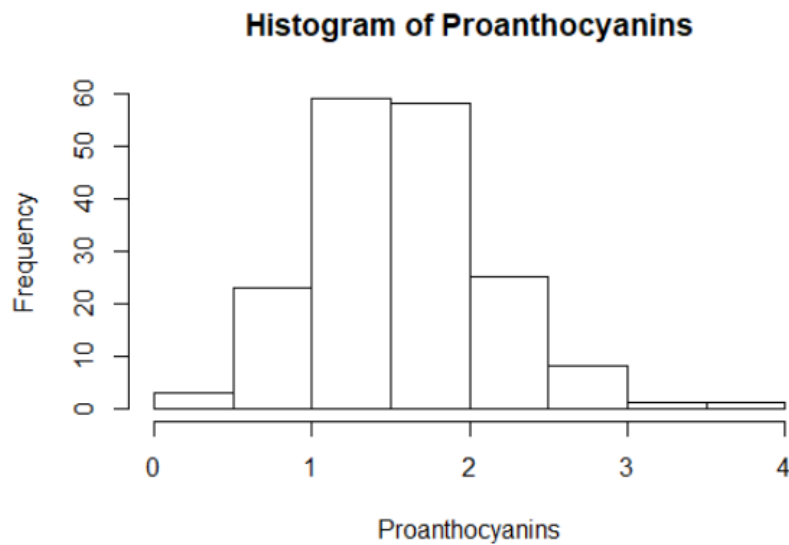
Histogram of Flavanoids



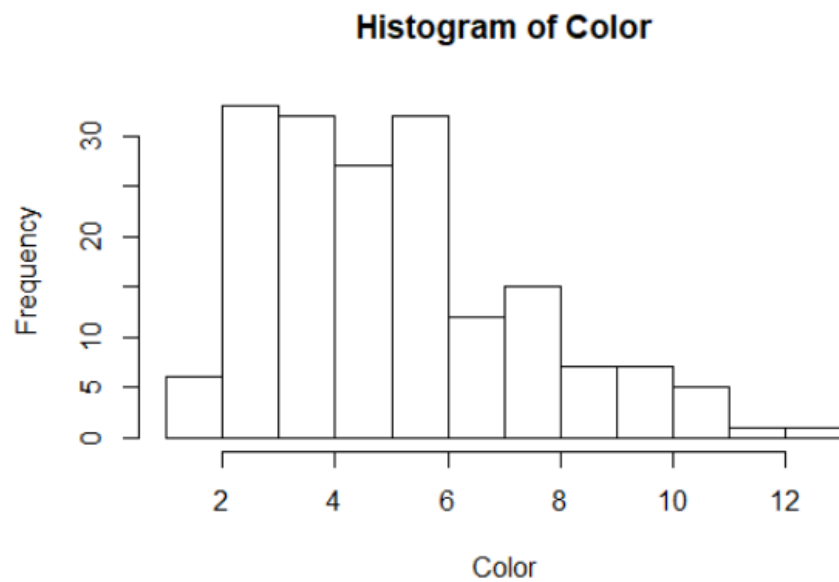

```
> hist(Nonflavanoids)
```



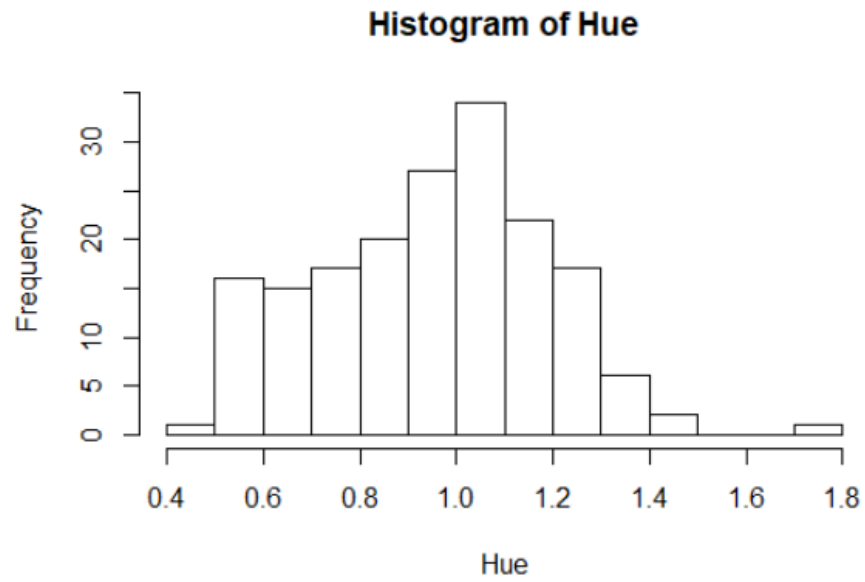
```
> hist(Proanthocyanins)
```



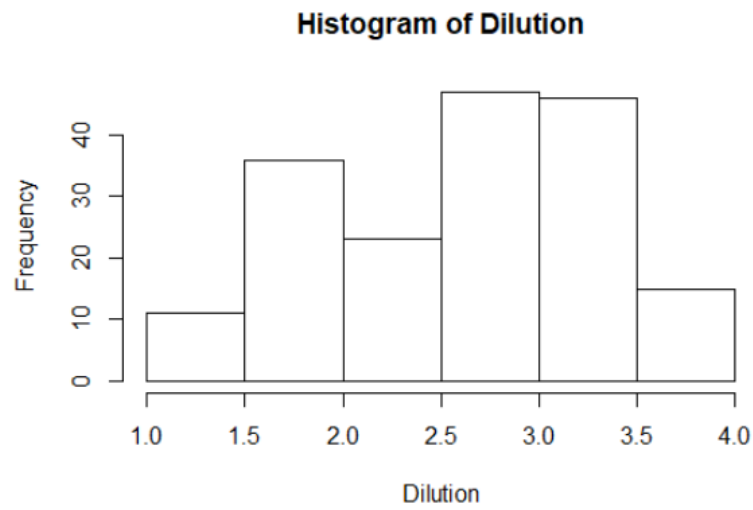
```
> hist(Color)
```



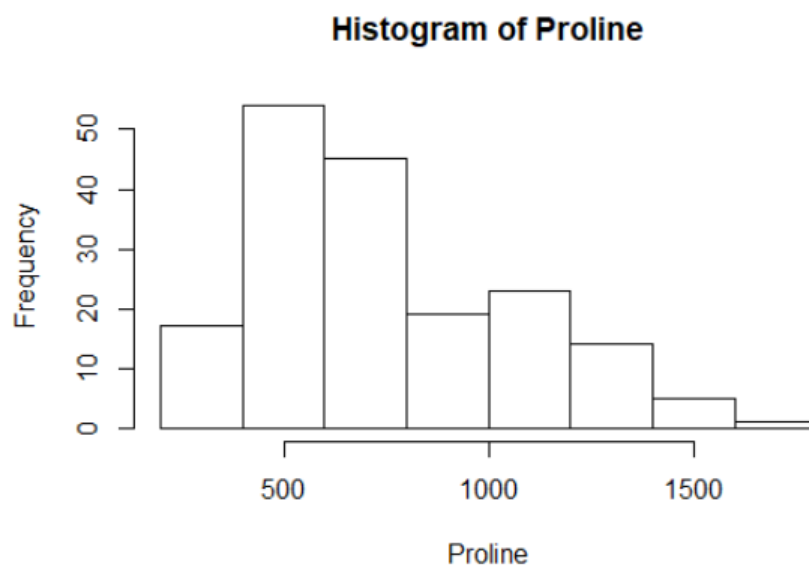
```
> hist(Hue)
```



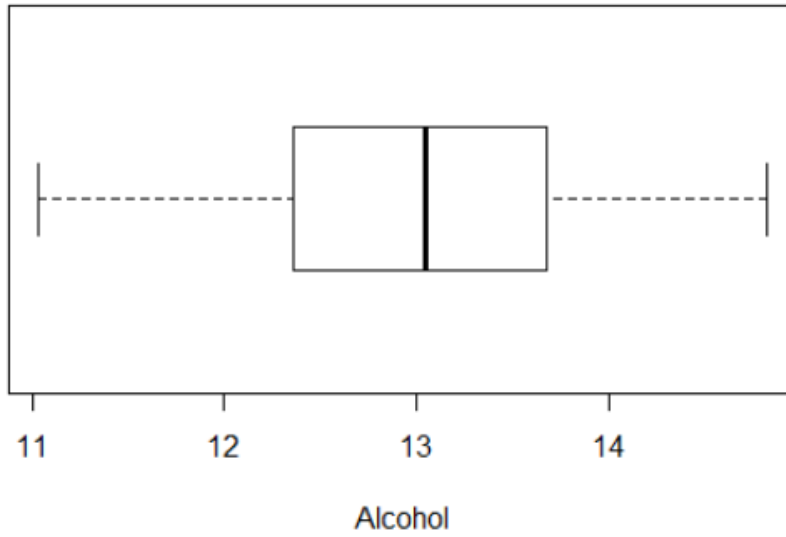
```
> hist(Dilution)
```



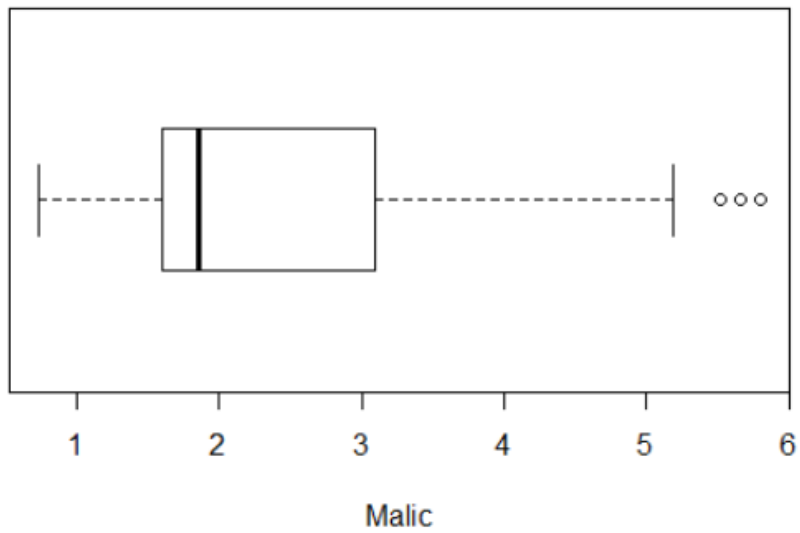
```
> hist(Proline)
```



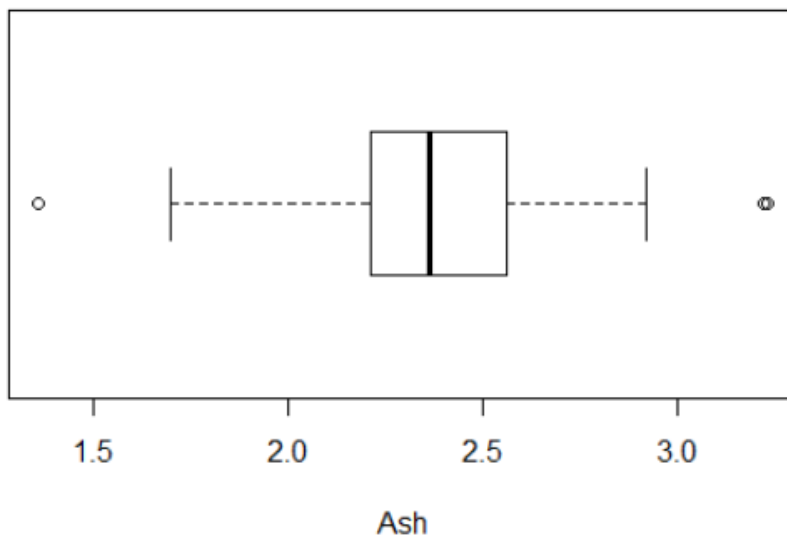
```
> boxplot(Alcohol, horizontal = T, xlab="Alcohol")
```



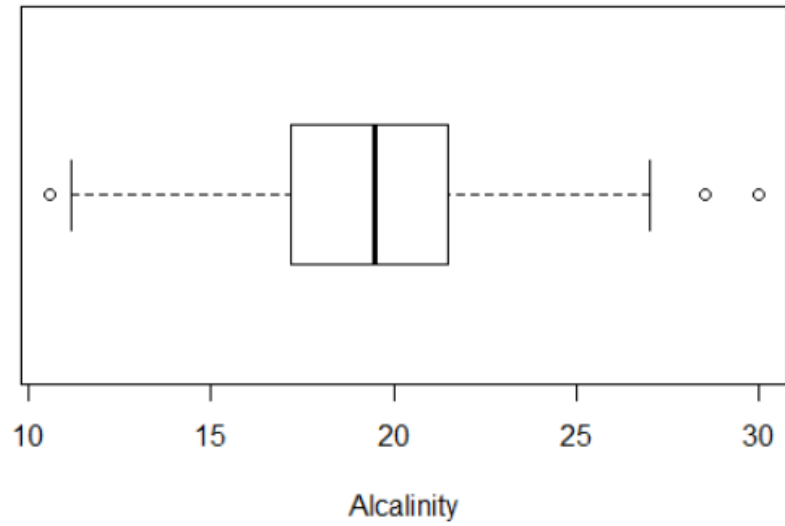
```
> boxplot(Malic, horizontal = T, xlab="Malic")
```



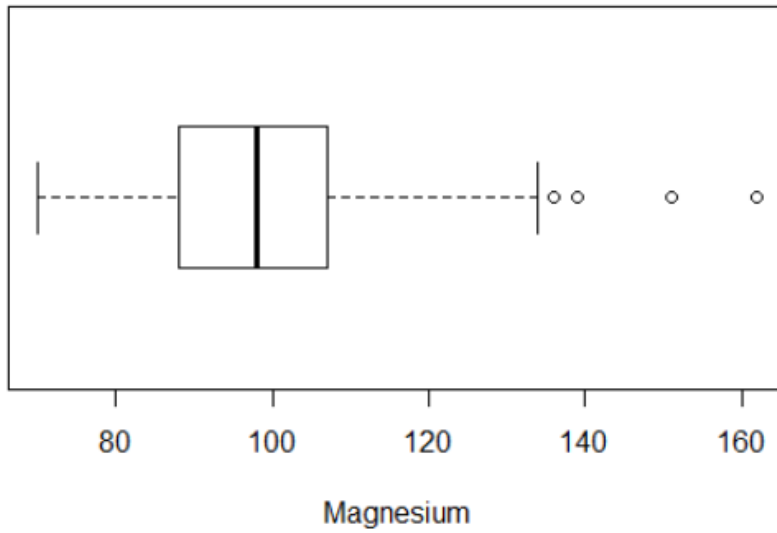
```
> boxplot(Ash, horizontal = T, xlab="Ash")
```



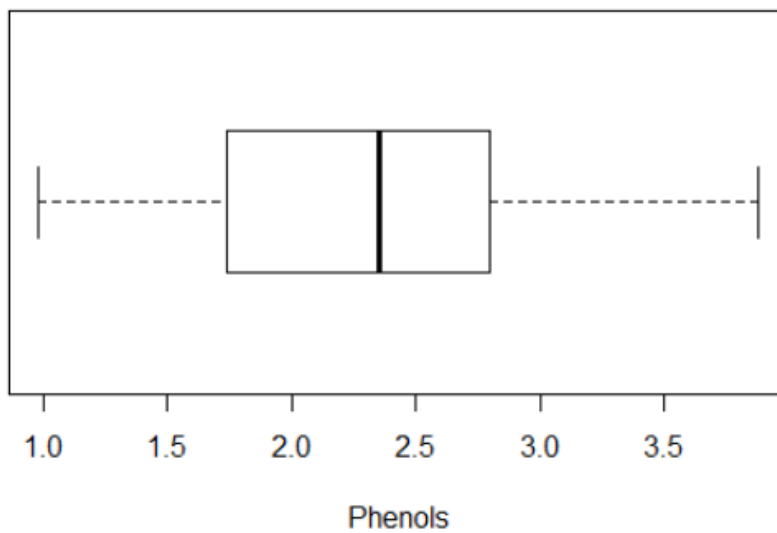
```
> boxplot(Alcalinity, horizontal = T, xlab="Alcalinity")
```



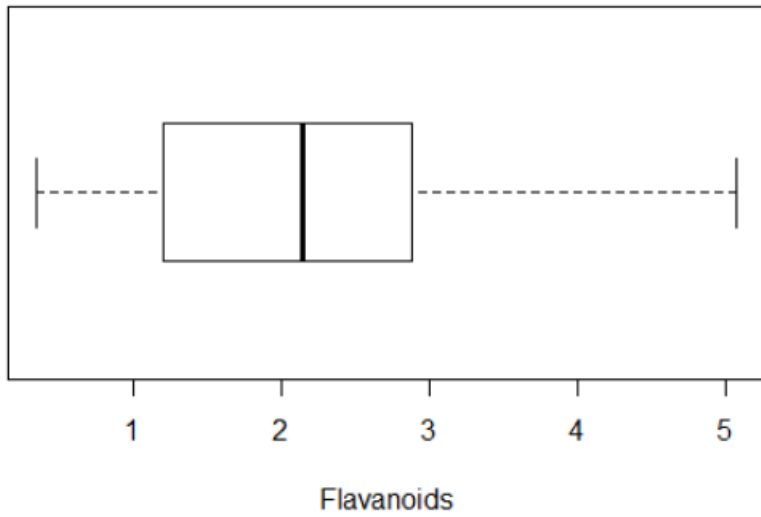
```
> boxplot(Magnesium, horizontal = T, xlab="Magnesium")
```



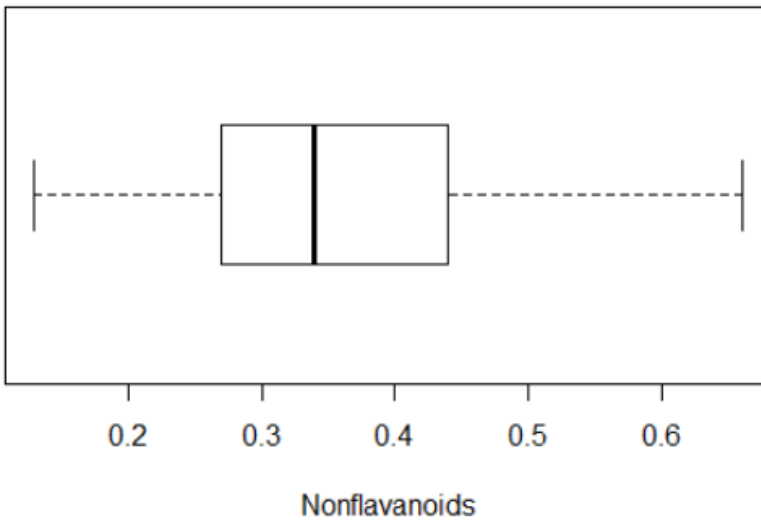
```
> boxplot(Phenols, horizontal = T, xlab="Phenols")
```



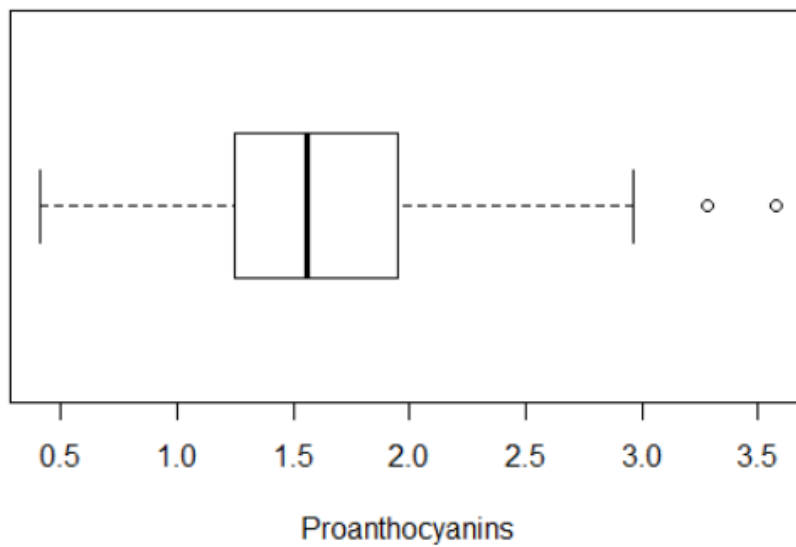
```
> boxplot(Flavanoids,horizontal = T,xlab="Flavanoids")
```



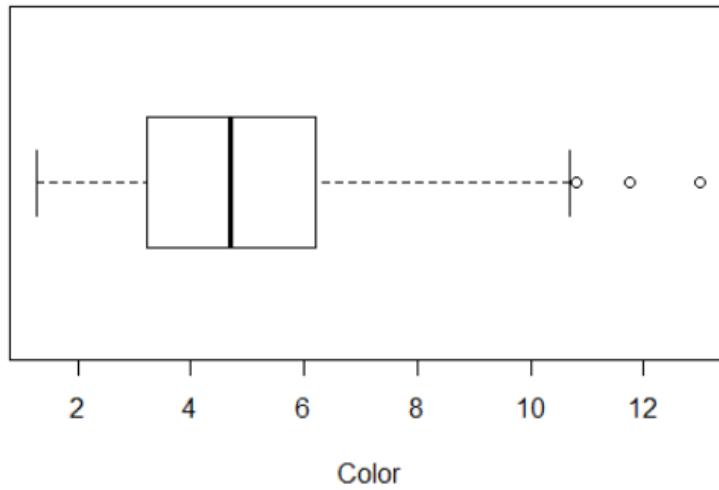
```
> boxplot(Nonflavanoids,horizontal = T,xlab="Nonflavanoids")
```



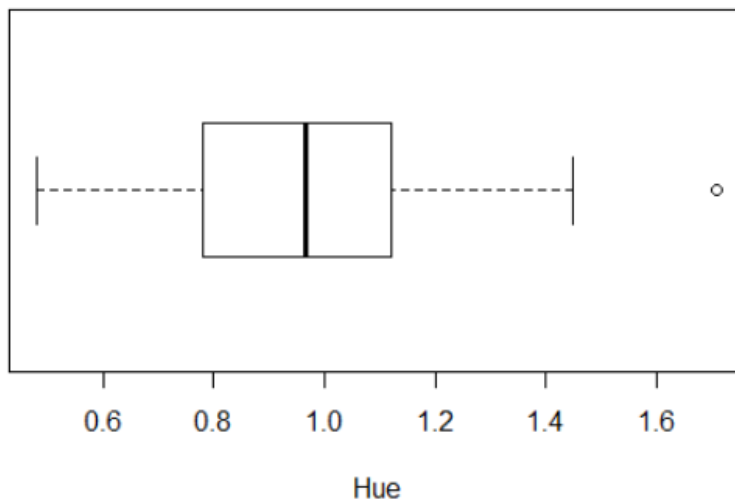
```
> boxplot(Proanthocyanins,horizontal = T,xlab="Proanthocyanins")
```



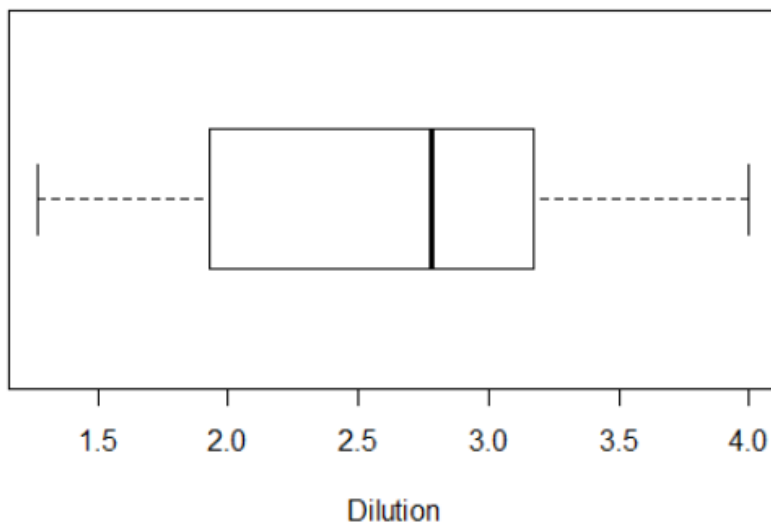
```
> boxplot(Color, horizontal = T, xlab="Color")
```



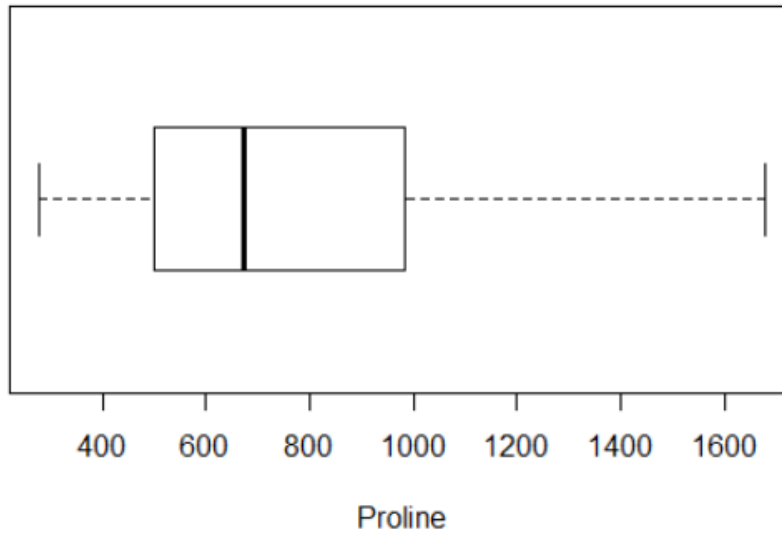
```
> boxplot(Hue, horizontal = T, xlab="Hue")
```



```
> boxplot(Dilution, horizontal = T, xlab="Dilution")
```



```
> boxplot(Proline,horizontal = T,xlab="Proline")
```



```
> mydata <- wine[,-1]
> cor(mydata)
```

	Alcohol	Malic	Ash	Alcalinity	Magnesium
Alcohol	1.00000000	0.09439694	0.211544596	-0.31023514	0.27079823
Malic	0.09439694	1.00000000	0.164045470	0.28850040	-0.05457510
Ash	0.21154460	0.16404547	1.000000000	0.44336719	0.28658669
Alcalinity	-0.31023514	0.28850040	0.443367187	1.00000000	-0.08333309
Magnesium	0.27079823	-0.05457510	0.286586691	-0.08333309	1.00000000
Phenols	0.28910112	-0.33516700	0.128979538	-0.32111332	0.21440123
Flavanoids	0.23681493	-0.41100659	0.115077279	-0.35136986	0.19578377
Nonflavanoids	-0.15592947	0.29297713	0.186230446	0.36192172	-0.25629405
Proanthocyanins	0.13669791	-0.22074619	0.009651935	-0.19732684	0.23644061
Color	0.54636420	0.24898534	0.258887259	0.01873198	0.19995001
Hue	-0.07174720	-0.56129569	-0.074666889	-0.27395522	0.05539820
Dilution	0.07234319	-0.36871043	0.003911231	-0.27676855	0.06600394
Proline	0.64372004	-0.19201056	0.223626264	-0.44059693	0.39335085

	Phenols	Flavanoids	Nonflavanoids	Proanthocyanins
Alcohol	0.28910112	0.2368149	-0.1559295	0.136697912
Malic	-0.33516700	-0.4110066	0.2929771	-0.220746187
Ash	0.12897954	0.1150773	0.1862304	0.009651935
Alcalinity	-0.32111332	-0.3513699	0.3619217	-0.197326836
Magnesium	0.21440123	0.1957838	-0.2562940	0.236440610
Phenols	1.00000000	0.8645635	-0.4499353	0.612413084
Flavanoids	0.86456350	1.0000000	-0.5378996	0.652691769
Nonflavanoids	-0.44993530	-0.5378996	1.0000000	-0.365845099
Proanthocyanins	0.61241308	0.6526918	-0.3658451	1.000000000
Color	-0.05513642	-0.1723794	0.1390570	-0.025249931
Hue	0.43368134	0.5434786	-0.2626396	0.295544253
Dilution	0.69994936	0.7871939	-0.5032696	0.519067096
Proline	0.49811488	0.4941931	-0.3113852	0.330416700

	Color	Hue	Dilution	Proline
Alcohol	0.54636420	-0.07174720	0.072343187	0.6437200
Malic	0.24898534	-0.56129569	-0.368710428	-0.1920106
Ash	0.25888726	-0.07466689	0.003911231	0.2236263
Alcalinity	0.01873198	-0.27395522	-0.276768549	-0.4405969
Magnesium	0.19995001	0.05539820	0.066003936	0.3933508
Phenols	-0.05513642	0.43368134	0.699949365	0.4981149
Flavanoids	-0.17237940	0.54347857	0.787193902	0.4941931
Nonflavanoids	0.13905701	-0.26263963	-0.503269596	-0.3113852
Proanthocyanins	-0.02524993	0.29554425	0.519067096	0.3304167
Color	1.00000000	-0.52181319	-0.428814942	0.3161001

```

Hue          -0.52181319  1.00000000  0.565468293  0.2361834
Dilution     -0.42881494  0.56546829  1.000000000  0.3127611
Proline       0.31610011  0.23618345  0.312761075  1.0000000
> pcaobj <- princomp(mydata,cor = T,scores = T,covmat = NULL)
> str(pcaobj)
List of 7
 $ sdev      : Named num [1:13] 2.169 1.58 1.203 0.959 0.924 ...
 ..- attr(*, "names")= chr [1:13] "Comp.1" "Comp.2" "Comp.3" "Comp.4" ...
 $ loadings: 'loadings' num [1:13, 1:13] 0.14433 -0.24519 -0.00205 -0.23932 0
 .14199 ...
 ..- attr(*, "dimnames")=List of 2
 .. ..$ : chr [1:13] "Alcohol" "Malic" "Ash" "Alcalinity" ...
 .. ..$ : chr [1:13] "Comp.1" "Comp.2" "Comp.3" "Comp.4" ...
 $ center   : Named num [1:13] 13 2.34 2.37 19.49 99.74 ...
 ..- attr(*, "names")= chr [1:13] "Alcohol" "Malic" "Ash" "Alcalinity" ...
 $ scale    : Named num [1:13] 0.81 1.114 0.274 3.33 14.242 ...
 ..- attr(*, "names")= chr [1:13] "Alcohol" "Malic" "Ash" "Alcalinity" ...
 $ n.obs    : int 178
 $ scores   : num [1:178, 1:13] 3.32 2.21 2.52 3.76 1.01 ...
 ..- attr(*, "dimnames")=List of 2
 .. ..$ : NULL
 .. ..$ : chr [1:13] "Comp.1" "Comp.2" "Comp.3" "Comp.4" ...
 $ call     : language princomp(x = mydata, cor = T, scores = T, covmat = NULL)
)
- attr(*, "class")= chr "princomp"
> loadings(pcaobj)

```

Loadings:

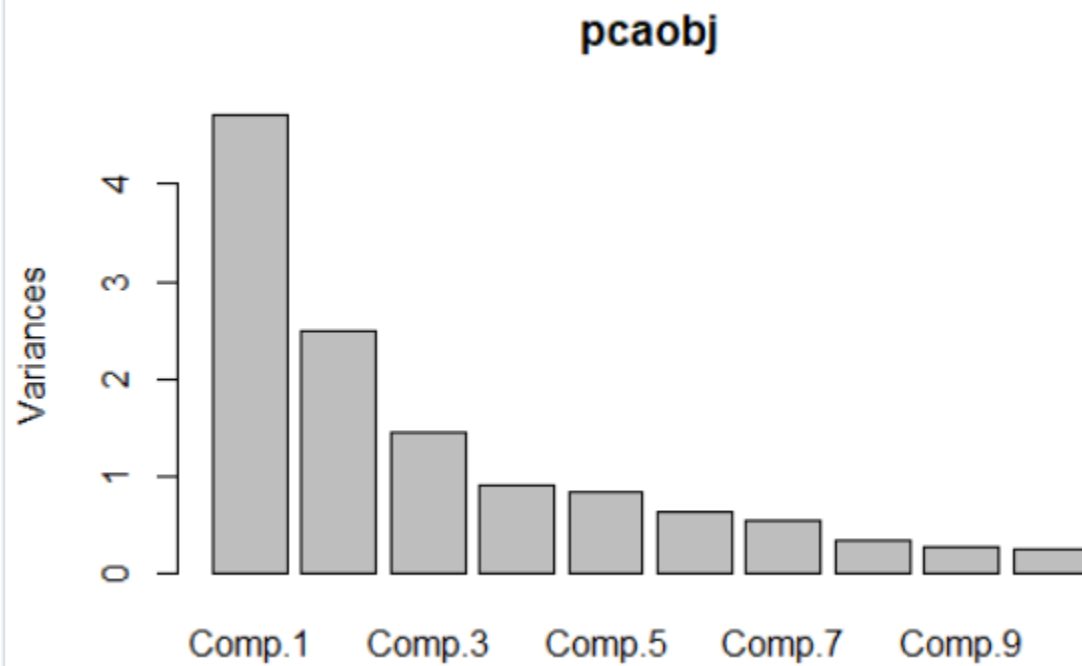
	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8
Alcohol	0.144	0.484	0.207		0.266	0.214		0.396
Malic	-0.245	0.225		-0.537		0.537	-0.421	
Ash		0.316	-0.626	0.214	0.143	0.154	0.149	-0.170
Alcalinity	-0.239		-0.612			-0.101	0.287	0.428
Magnesium	0.142	0.300	-0.131	0.352	-0.727		-0.323	-0.156
Phenols	0.395		-0.146	-0.198	0.149			-0.406
Flavanoids	0.423		-0.151	-0.152	0.109			-0.187
Nonflavanoids	-0.299		-0.170	0.203	0.501	-0.259	-0.595	-0.233
Proanthocyanins	0.313		-0.149	-0.399	-0.137	-0.534	-0.372	0.368
Color		0.530	0.137			-0.419	0.228	
Hue	0.297	-0.279		0.428	0.174	0.106	-0.232	0.437
Dilution	0.376	-0.164	-0.166	-0.184	0.101	0.266		
Proline	0.287	0.365	0.127	0.232	0.158	0.120		0.120

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8
Alcohol	0.509	0.212	0.226	0.266				
Malic		-0.309		-0.122				
Ash	-0.308		0.499		-0.141			
Alcalinity	0.200		-0.479					
Magnesium	0.271							
Phenols	0.286	-0.320	-0.304	0.304	-0.464			
Flavanoids		-0.163			0.832			
Nonflavanoids	0.196	0.216	-0.117		0.114			
Proanthocyanins	-0.209	0.134	0.237		-0.117			
Color		-0.291		-0.604				
Hue		-0.522		-0.259				
Dilution	0.137	0.524		-0.601	-0.157			
Proline	-0.576	0.162	-0.539					

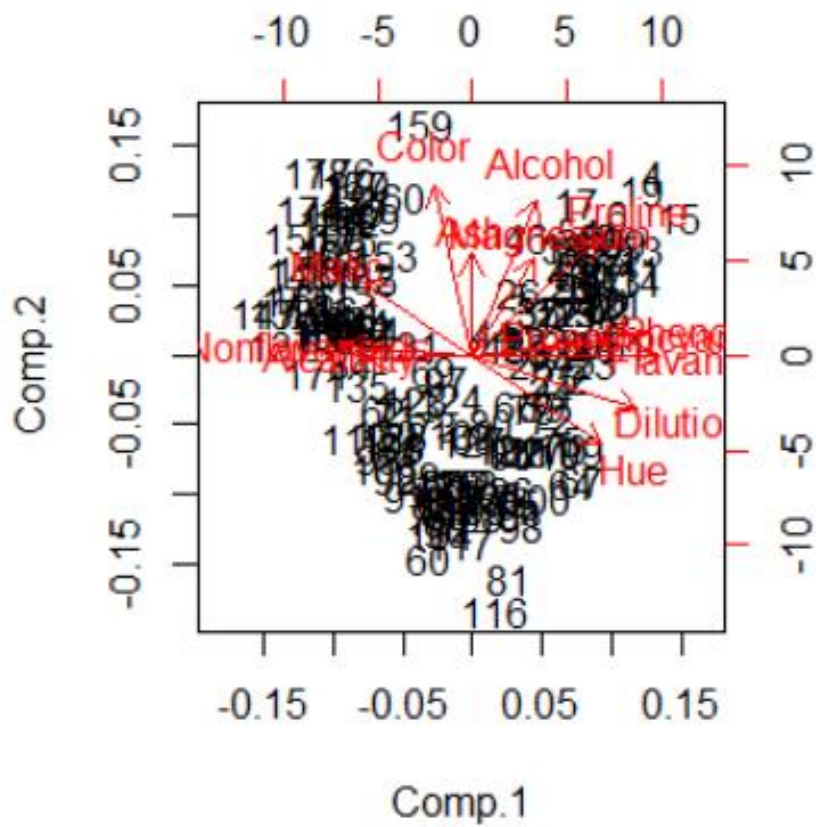
	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8
SS loadings	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Proportion Var	0.077	0.077	0.077	0.077	0.077	0.077	0.077	0.077
Cumulative Var	0.077	0.154	0.231	0.308	0.385	0.462	0.538	0.615

	Comp.9	Comp.10	Comp.11	Comp.12	Comp.13
SS loadings	1.000	1.000	1.000	1.000	1.000
Proportion Var	0.077	0.077	0.077	0.077	0.077
Cumulative Var	0.692	0.769	0.846	0.923	1.000

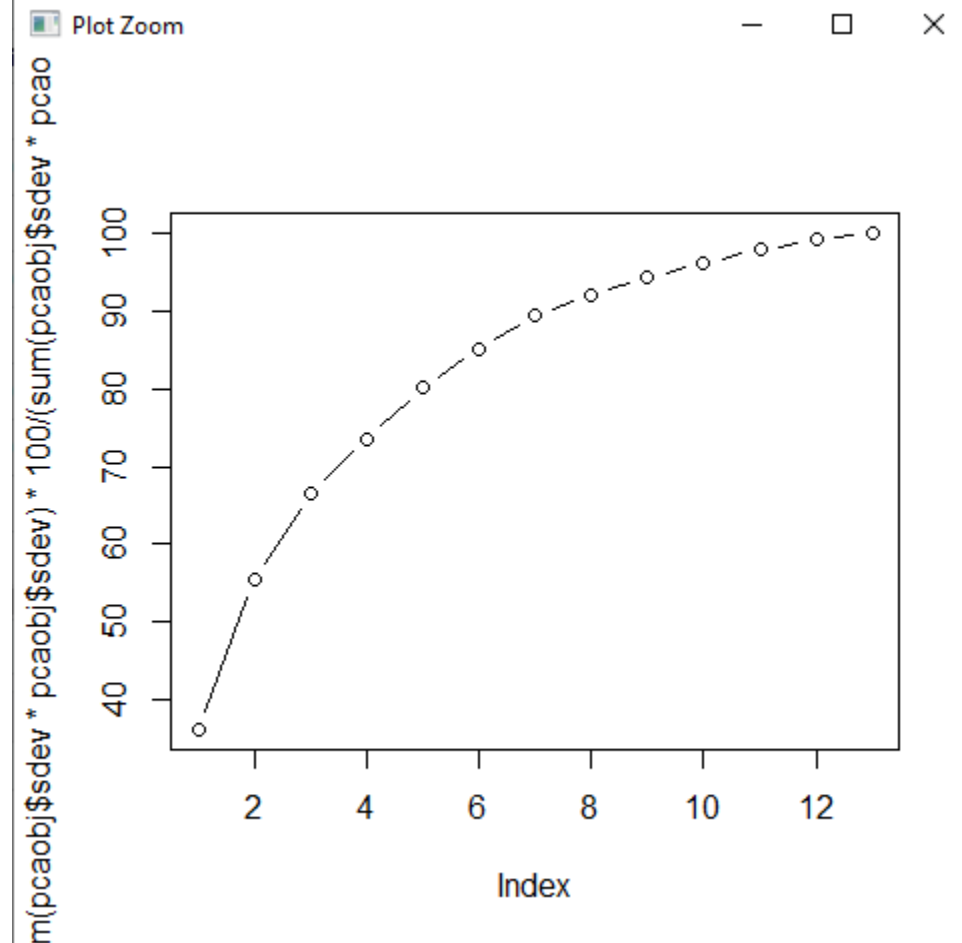

```
> plot(pcaobj)
```



```
> biplot(pcaobj)
```



```
> plot(cumsum(pcaobj$sdev*pcaobj$sdev)*100/(sum(pcaobj$sdev*pcaobj$sdev)),type="b")
```



```
> pcaobj$scores[,1:3]
```

	Comp.1	Comp.2	Comp.3
[1,]	3.31675081	1.44346263	0.165739045
[2,]	2.20946492	-0.33339289	2.026457374
[3,]	2.51674015	1.03115130	-0.982818670
[4,]	3.75706561	2.75637191	0.176191842
[5,]	1.00890849	0.86983082	-2.026688219
[6,]	3.05025392	2.12240111	0.629395827
[7,]	2.44908967	1.17485013	0.977094891
[8,]	2.05943687	1.60896307	-0.146281883
[9,]	2.51087430	0.91807096	1.770969027
[10,]	2.75362819	0.78943767	0.984247490
[11,]	3.47973668	1.30233324	0.422735217
[12,]	1.75475290	0.61197723	1.190878320
[13,]	2.11346234	0.67570634	0.865086426
[14,]	3.45815682	1.13062988	1.204276353
[15,]	4.31278391	2.09597558	1.263912752
[16,]	2.30518820	1.66255173	-0.217902616
[17,]	2.17195527	2.32730534	-0.831729866
[18,]	1.89897118	1.63136888	-0.794913792
[19,]	3.54198508	2.51834367	0.485458508
[20,]	2.08452220	1.06113799	0.164746678
[21,]	3.12440254	0.78689711	0.364887083
[22,]	1.08657007	0.24174355	-0.936961600
[23,]	2.53522408	-0.09184062	0.311932659
[24,]	1.64498834	-0.51627893	-0.143885095
[25,]	1.76157587	-0.31714893	-0.890285647
[26,]	0.99007910	0.94066734	-3.820908008
[27,]	1.77527763	0.68617513	0.086700406
[28,]	1.23542396	-0.08980704	1.386896545
[29,]	2.18840633	0.68956962	-1.394566881

[30,]	2.25610898	0.19146194	1.092657258
[31,]	2.50022003	1.24083383	-1.386017855
[32,]	2.67741105	1.47187365	0.332261728
[33,]	1.62857912	0.05270445	0.167128706
[34,]	1.90269086	1.63306043	-1.172082119
[35,]	1.41038853	0.69793432	-0.479743025
[36,]	1.90382623	0.17671095	-0.450835040
[37,]	1.38486223	0.65863985	-0.458438581
[38,]	1.12220741	0.11410976	0.039107277
[39,]	1.50219450	-0.76943201	1.426177346
[40,]	2.52980109	1.80300198	0.343152389
[41,]	2.58809543	0.77961630	0.118477466
[42,]	0.66848199	0.16996094	0.783362548
[43,]	3.07080699	1.15591896	0.312758084
[44,]	0.46220914	0.33074213	0.201476496
[45,]	2.10135193	-0.07100892	0.655849415
[46,]	1.13616618	1.77710739	-0.028705736
[47,]	2.72660096	1.19133469	0.539773261
[48,]	2.82133927	0.64625860	1.155552411
[49,]	2.00985085	1.24702946	0.057293988
[50,]	2.70749130	1.75196741	0.643113612
[51,]	3.21491747	0.16699199	1.973571680
[52,]	2.85895983	0.74527880	-0.004719502
[53,]	3.50560436	1.61273386	0.520774530
[54,]	2.22479138	1.87516800	-0.339549850
[55,]	2.14698782	1.01675154	0.957762762
[56,]	2.46932948	1.32900831	-0.513437453
[57,]	2.74151791	1.43654878	0.612473396
[58,]	2.17374092	1.21219984	-0.261779593
[59,]	3.13938015	1.73157912	0.285661413
[60,]	-0.92858197	-3.07348616	4.585064007
[61,]	-1.54248014	-1.38144351	0.874683112
[62,]	-1.83624976	-0.82998412	1.605702186
[63,]	0.03060683	-1.26278614	1.784408010
[64,]	2.05026161	-1.92503260	0.007368777
[65,]	-0.60968083	-1.90805881	-0.679357938
[66,]	0.90022784	-0.76391147	-0.573361302
[67,]	2.24850719	-1.88459248	2.031840193
[68,]	0.18338403	-2.42714611	1.069745560
[69,]	-0.81280503	-0.22051399	0.707005396
[70,]	1.97562050	-1.40328323	1.238276220
[71,]	-1.57221622	-0.88498314	0.628997950
[72,]	1.65768181	-0.95671220	-1.952584217
[73,]	-0.72537239	-1.06364540	-0.080332229
[74,]	2.56222717	0.26019855	-3.374393962
[75,]	1.83256757	-1.28787820	-0.458280027
[76,]	-0.86799290	-2.44410119	1.563333179
[77,]	0.37001440	-2.15390698	2.449386348
[78,]	-1.45737704	-1.38335177	0.227306902
[79,]	1.26293085	-0.77084953	1.184224517
[80,]	0.37615037	-1.02704340	-1.794466295
[81,]	0.76206390	-3.37505381	0.357470056
[82,]	1.03457797	-1.45070974	0.363011773
[83,]	-0.49487676	-2.38124353	-1.335743176
[84,]	-2.53897708	-0.08744336	-0.474251393
[85,]	0.83532015	-1.47367055	-0.610093576
[86,]	0.78790461	-2.02662652	0.254723404
[87,]	-0.80683216	-2.23383039	-0.772855797
[88,]	-0.55804262	-2.37298543	-2.307611404
[89,]	-1.11511104	-1.80224719	-0.959253308
[90,]	-0.55572283	-2.65754004	-0.849126898
[91,]	-1.34928528	-2.11800147	0.047652321
[92,]	-1.56448261	-1.85221452	-0.781067031
[93,]	-1.93255561	-1.55949546	0.089274676
[94,]	0.74666594	-2.31293171	-0.114679769
[95,]	0.95745536	-2.22352843	-0.142444774
[96,]	2.54386518	0.16927402	-0.788696991
[97,]	-0.54395259	-0.36892655	-1.308895932

[98,]	1.03104975	-2.56556935	1.086390174
[99,]	2.25190942	-1.43274138	0.230208244
[100,]	1.41021602	-2.16619177	-0.748896411
[101,]	0.79771979	-2.37694880	1.568112531
[102,]	-0.54953173	-2.29312864	1.498935323
[103,]	-0.16117374	-1.16448332	-1.003713103
[104,]	-0.65979494	-2.67996119	0.764920868
[105,]	0.39235441	-2.09873171	0.471850008
[106,]	-1.77249908	-1.71728847	-0.947033174
[107,]	-0.36626736	-2.16935330	0.481324235
[108,]	-1.62067257	-1.35558339	-0.287159001
[109,]	0.08253578	-2.30623459	0.463574989
[110,]	1.57827507	-1.46203429	-1.779645955
[111,]	1.42056925	-1.41820664	-0.139275829
[112,]	-0.27870275	-1.93056809	-0.078670553
[113,]	-1.30314497	-0.76317231	-1.999596510
[114,]	-0.45707187	-2.26941561	-1.061338968
[115,]	-0.49418585	-1.93904505	-1.323938072
[116,]	0.48207441	-3.87178385	-1.344271223
[117,]	-0.25288888	-2.82149237	0.302639785
[118,]	-0.10722764	-1.92892204	-0.690148243
[119,]	-2.43301260	-1.25714104	1.903027404
[120,]	-0.55108954	-2.22216155	0.356228830
[121,]	0.73962193	-1.40895667	-1.125345492
[122,]	1.33632173	0.25333693	-5.345388179
[123,]	-1.17708700	-0.66396684	-3.010221888
[124,]	-0.46233501	-0.61828818	-0.483442366
[125,]	0.97847408	-1.44557050	-1.481236975
[126,]	-0.09680973	-2.10999799	-0.434826116
[127,]	0.03848715	-1.26676211	-0.687577913
[128,]	-1.59715850	-1.20814357	-3.361175555
[129,]	-0.47956492	-1.93884066	-1.296507519
[130,]	-1.79283347	-1.15028810	-0.782800173
[131,]	-1.32710166	0.17038923	1.180013355
[132,]	-2.38450083	0.37458261	0.723822595
[133,]	-2.93694010	0.26386183	0.167639816
[134,]	-2.14681113	0.36825495	0.453301301
[135,]	-2.36986949	-0.45963481	1.101399789
[136,]	-3.06384157	0.35341284	1.099124104
[137,]	-3.91575378	0.15458252	-0.221827800
[138,]	-3.93646339	0.65968723	-1.712215419
[139,]	-3.09427612	0.34884276	1.026831413
[140,]	-2.37447163	0.29198035	-1.241914333
[141,]	-2.77881295	0.28680487	-0.609670124
[142,]	-2.28656128	0.37250784	0.971643032
[143,]	-2.98563349	0.48921791	-0.946952932
[144,]	-2.37519470	0.48233372	0.252883994
[145,]	-2.20986553	1.16005250	1.245125226
[146,]	-2.62562100	0.56316076	0.855961082
[147,]	-4.28063878	0.64967096	1.458196962
[148,]	-3.58264137	1.27270275	0.110784038
[149,]	-2.80706372	1.57053379	0.472527935
[150,]	-2.89965933	2.04105701	0.495959810
[151,]	-2.32073698	2.35636608	-0.437681744
[152,]	-2.54983095	2.04528309	0.312267999
[153,]	-1.81254128	1.52764595	-1.362589782
[154,]	-2.76014464	2.13893235	0.964628688
[155,]	-2.73715050	0.40988627	1.190404684
[156,]	-3.60486887	1.80238422	0.094036861
[157,]	-2.88982600	1.92521861	0.782322556
[158,]	-3.39215608	1.31187639	-1.602025969
[159,]	-1.04818190	3.51508969	-1.160038566
[160,]	-1.60991228	2.40663816	-0.548559697
[161,]	-3.14313097	0.73816104	0.090998724
[162,]	-2.24015690	1.17546529	0.101376932
[163,]	-2.84767378	0.55604397	-0.804215218
[164,]	-2.59749706	0.69796554	0.884939521
[165,]	-2.94929937	1.55530896	0.983400727

```

[166,] -3.53003227  0.88252680  0.466029128
[167,] -2.40611054  2.59235618 -0.428226211
[168,] -2.92908473  1.27444695  1.213358272
[169,] -2.18141278  2.07753731 -0.763782552
[170,] -2.38092779  2.58866743 -1.418044029
[171,] -3.21161722 -0.25124910  0.847129152
[172,] -3.67791872  0.84774784  1.339420231
[173,] -2.46555580  2.19379830  0.918780960
[174,] -3.37052415  2.21628914  0.342569512
[175,] -2.60195585  1.75722935 -0.207581355
[176,] -2.67783946  2.76089913  0.940941877
[177,] -2.38701709  2.29734668  0.550696197
[178,] -3.20875816  2.76891957 -1.013913664

```

```

> mydata1<-cbind(wine,pcaobj$scores[,1:3])
> View(mydata1)

```

anoids	Proanthocyanins	Color	Hue	Dilution	Proline	Comp.1	Comp.2	Comp.3
	2.29	5.64	1.040	3.92	1065	3.31675081	1.44346263	0.165739045
	1.28	4.38	1.050	3.40	1050	2.20946492	-0.33339289	2.026457374
	2.81	5.68	1.030	3.17	1185	2.51674015	1.03115130	-0.982818670
	2.18	7.80	0.860	3.45	1480	3.75706561	2.75637191	0.176191842
	1.82	4.32	1.040	2.93	735	1.00890849	0.86983082	-2.026688219
	1.97	6.75	1.050	2.85	1450	3.05025392	2.12240111	0.629395827
	1.98	5.25	1.020	3.58	1290	2.44908967	1.17485013	0.977094891
	1.25	5.05	1.060	3.58	1295	2.05943687	1.60896307	-0.146281883
	1.98	5.20	1.080	2.85	1045	2.51087430	0.91807096	1.770969027

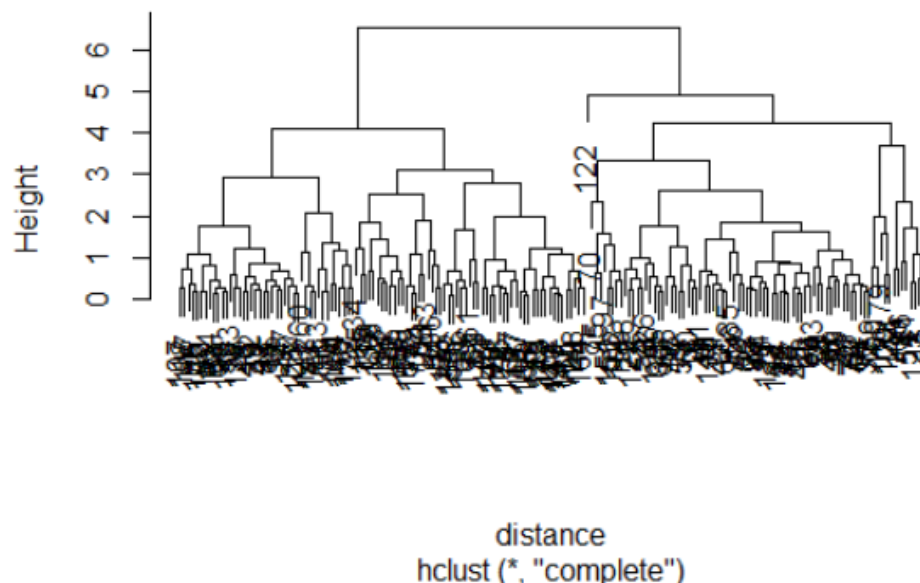
Showing 1 to 10 of 178 entries, 17 total columns

```

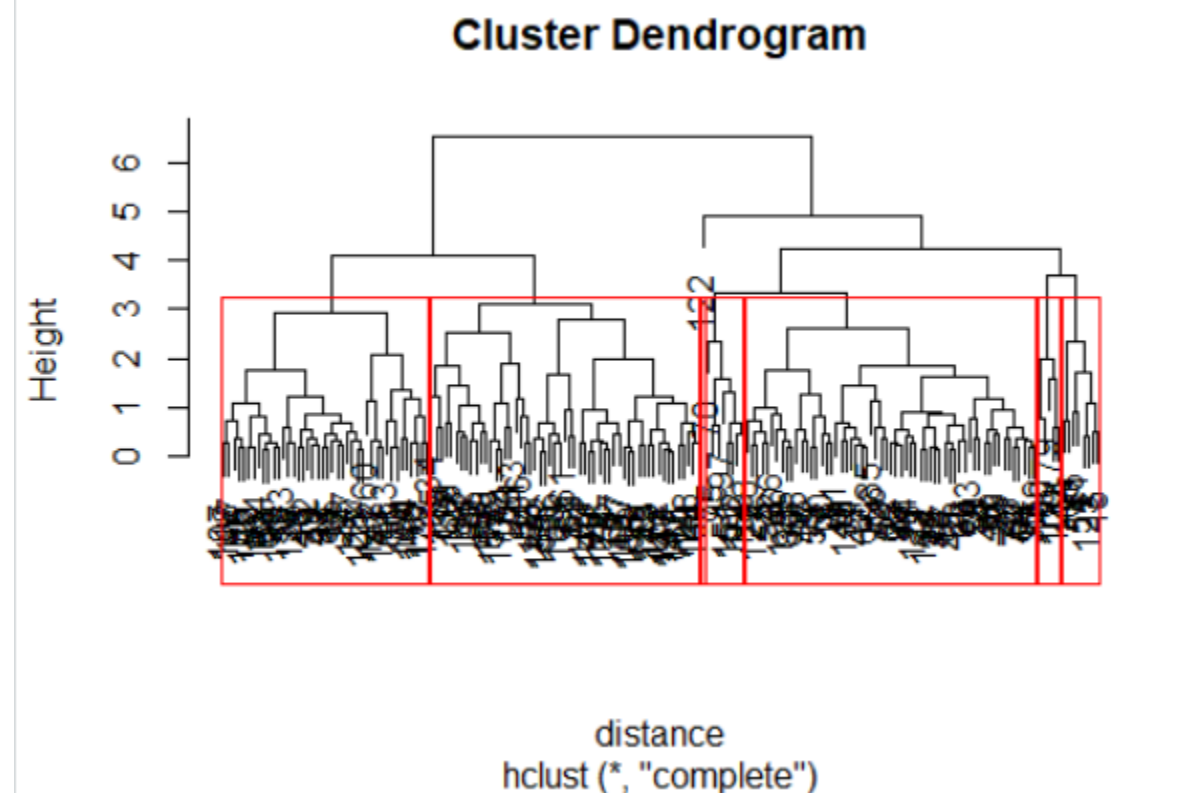
> clus_data<-wine[,8:10]
> # Normalizing the data
> norm_clus<-scale(clus_data)
> distance<-dist(norm_clus,method = "euclidean")
> fit<-hclust(distance,method="complete")
> plot(fit)

```

Cluster Dendrogram



```
> rect.hclust(fit, k=7, border="red")
```



```
> groups <- cutree(fit,5)
> clust_1 <- as.matrix(groups)
> View(clust_1)
```

PCA.R ×		clust_1 ×		final ×	
←	→	Filter			
	V1				
1	1				
2	2				
3	3				
4	1				
5	1				
6	1				
7	1				
8	2				
9	1				
10	1				
11	1				

Showing 1 to 11 of 178 entries, 1 total columns

```
> final <- cbind(clust_1,mydata1)
```

```
> view(final)
```

	clust_1	Type	Alcohol	Malic	Ash	Alcalinity	Magnesium	Phenols	Flavanoids
1	1	1	14.23	1.71	2.43	15.6	127	2.80	3.06
2	2	1	13.20	1.78	2.14	11.2	100	2.65	2.76
3	3	1	13.16	2.36	2.67	18.6	101	2.80	3.24
4	1	1	14.37	1.95	2.50	16.8	113	3.85	3.49
5	1	1	13.24	2.59	2.87	21.0	118	2.80	2.69
6	1	1	14.20	1.76	2.45	15.2	112	3.27	3.39
7	1	1	14.39	1.87	2.45	14.6	96	2.50	2.52
8	2	1	14.06	2.15	2.61	17.6	121	2.60	2.51
9	1	1	14.83	1.64	2.17	14.0	97	2.80	2.98

Showing 1 to 10 of 178 entries, 18 total columns

```
> aggregate(final[,-c(2,16:18)],by=list(clust_1),FUN = mean)
```

```

Group.1 clust_1 Alcohol Malic Ash Alcalinity Magnesium Phenols
1 1 1 13.33134 1.981343 2.419403 18.49851 105.02985 2.772687
2 2 2 12.77190 2.188571 2.238333 19.36190 95.19048 2.053095
3 3 3 12.90769 2.268462 2.256154 17.86154 102.07692 2.690000
4 4 4 12.82055 2.902909 2.410364 21.03273 95.87273 1.788727
5 5 5 11.56000 2.050000 3.230000 28.50000 119.00000 3.180000
Flavanoids Nonflavanoids Proanthocyanins Color Hue Dilution
1 2.808657 0.2762687 1.913284 5.102687 1.0310448 3.058955
2 1.811905 0.3090476 1.223095 4.071190 0.9807143 2.634286
3 2.511538 0.3515385 2.751538 5.270000 0.9746154 2.756923
4 1.076364 0.5069091 1.199636 5.690182 0.8464727 1.995636
5 5.080000 0.4700000 1.870000 6.000000 0.9300000 3.690000
Proline
1 901.6418
2 680.1190
3 786.3846
4 605.1636
5 465.0000

```

```
> aggregate(final[,-c(2,16:18)],by=list(clust_1),FUN = max)
```

```

Group.1 clust_1 Alcohol Malic Ash Alcalinity Magnesium Phenols Flavanoids
1 1 1 14.83 5.80 3.22 30.0 151 3.88 3.93
2 2 2 14.21 5.04 2.86 26.0 122 2.85 2.76
3 3 3 14.75 4.31 2.70 25.0 162 3.30 3.69
4 4 4 14.16 5.65 2.92 28.5 132 2.95 2.74
5 5 5 11.56 2.05 3.23 28.5 119 3.18 5.08
Nonflavanoids Proanthocyanins Color Hue Dilution Proline
1 0.47 2.50 10.80 1.36 4.00 1680
2 0.42 1.63 9.40 1.38 3.63 1295
3 0.53 3.58 13.00 1.42 3.64 1547
4 0.66 2.01 10.68 1.71 3.30 1235
5 0.47 1.87 6.00 0.93 3.69 465

```

```
> aggregate(final[,-c(2,16:18)],by=list(clust_1),FUN = min)
```

```

Group.1 clust_1 Alcohol Malic Ash Alcalinity Magnesium Phenols Flavanoids
1 1 1 11.45 0.89 1.70 13.2 78 1.40 0.99
2 2 2 11.65 0.90 1.36 10.6 70 0.98 0.34
3 3 3 11.46 0.99 1.82 11.4 82 1.90 1.10
4 4 4 11.03 0.74 1.98 16.0 80 1.10 0.47
5 5 5 11.56 2.05 3.23 28.5 119 3.18 5.08
Nonflavanoids Proanthocyanins Color Hue Dilution Proline
1 0.13 1.25 2.12 0.48 1.30 290
2 0.17 0.42 1.74 0.57 1.29 278
3 0.17 1.87 1.28 0.57 1.78 312

```

```

4          0.37          0.41  1.90 0.54          1.27          315
5          0.47          1.87  6.00 0.93          3.69          465
> write.csv(final,file = "data_clust.txt",row.names = F,col.names = F)
> write.csv(final,file = "data_clust.csv",row.names = F,col.names = F)
> getwd()
[1] "C:/Users/Arti Patel/Documents"
> # K-Means Clustering :
> library(plyr)
> mydata2 <- final
> normalized_data<-scale(mydata2[,15:17])
> kmeans_clust <- kmeans(normalized_data,7)
> str(kmeans_clust)
List of 9
 $ cluster      : int [1:178] 7 1 7 7 1 7 7 7 7 7 ...
 $ centers      : num [1:7, 1:3] 0.4252 -0.5863 0.0416 -0.3266 -0.8588 ...
 ..- attr(*, "dimnames")=List of 2
 .. ..$ : chr [1:7] "1" "2" "3" "4" ...
 .. ..$ : chr [1:3] "Proline" "Comp.1" "Comp.2"
 $ totss       : num 531
 $ withinss    : num [1:7] 15.17 3.29 3.84 6.18 19.14 ...
 $ tot.withinss: num 70.6
 $ betweenss   : num 460
 $ size        : int [1:7] 30 20 10 21 43 18 36
 $ iter        : int 4
 $ ifault      : int 0
 - attr(*, "class")= chr "kmeans"
> final1<- cbind(kmeans_clust$cluster,mydata2)
> View(final1)

```

PCAR.R × final1 × mydata × final × clust_1 × mydata1 × wine ×								
Filter								
	kmeans_clust\$cluster	clust_1	Type	Alcohol	Malic	Ash	Alcalinity	Magnesium
1	7	1	1	14.23	1.71	2.43	15.6	127
2	1	2	1	13.20	1.78	2.14	11.2	100
3	7	3	1	13.16	2.36	2.67	18.6	101
4	7	1	1	14.37	1.95	2.50	16.8	113
5	1	1	1	13.24	2.59	2.87	21.0	118
6	7	1	1	14.20	1.76	2.45	15.2	112
7	7	1	1	14.39	1.87	2.45	14.6	96
8	7	2	1	14.06	2.15	2.61	17.6	121
9	7	1	1	14.83	1.64	2.17	14.0	97

Showing 1 to 10 of 178 entries, 19 total columns

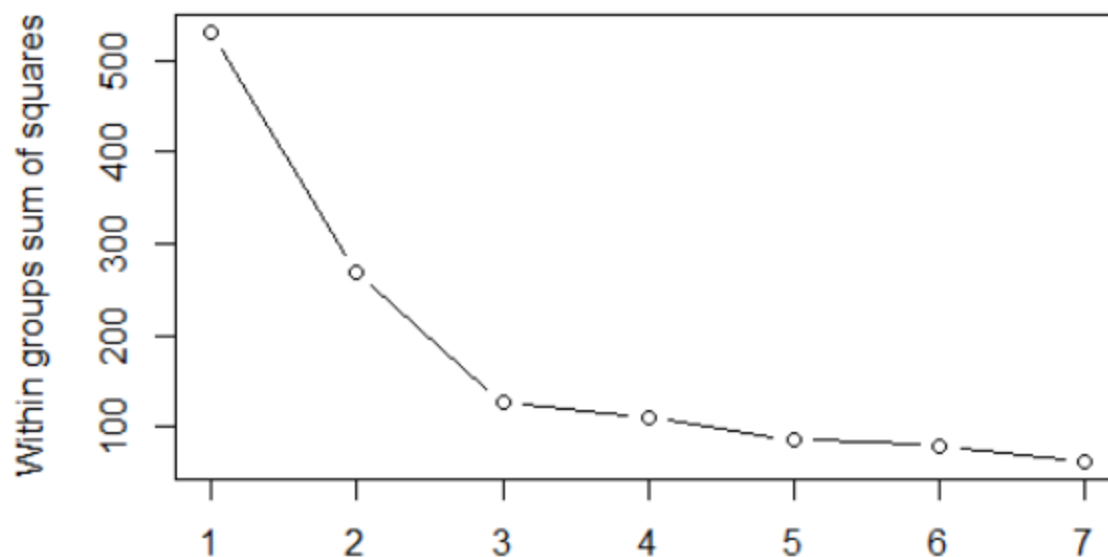
```

> wss = (nrow(normalized_data)-1)*sum(apply(normalized_data, 2, var)) # D
etermine number of clusters by scree-plot

> for (i in 1:7) wss[i] = sum(kmeans(normalized_data, centers=i)$withinss)
> plot(1:7, wss, type="b", xlab="Number of Clusters", ylab="Within groups sum
of squares") # Look for an "elbow" in the scree plot #

> title(sub = "K-Means Clustering Scree-Plot")

```

Number of Clusters
K-Means Clustering Scree-Plot

```
> aggregate(mydata2[,2:12],by=list(kmeans_clust$cluster),FUN = mean)
```

Group.1	Type	Alcohol	Malic	Ash	Alcalinity	Magnesium	Phenols
1	1	1.233333	13.18033	2.024000	2.439000	18.31333	111.76667
2	2	2.950000	12.92300	3.535000	2.403000	21.30000	93.95000
3	3	2.800000	12.88700	2.489000	2.337000	19.79000	103.30000
4	4	3.000000	13.46333	3.373333	2.497143	22.09524	102.52381
5	5	2.000000	12.21837	1.756047	2.162791	19.76047	89.04651
6	6	2.000000	12.32389	2.532222	2.383333	21.06111	93.33333
7	7	1.000000	13.92833	1.878611	2.452778	16.77778	106.30556

```
Flavanoids Nonflavanoids Proanthocyanins Color
```

Group.1	Flavanoids	Nonflavanoids	Proanthocyanins	Color
1	2.734667	0.2903333	1.889333	4.356333
2	0.7145000	0.4795000	0.921500	5.811000
3	0.7820000	0.4350000	1.003000	5.224000
4	0.9195238	0.4319048	1.448095	9.408571
5	2.2176744	0.3251163	1.680698	2.932093
6	1.6838889	0.4622222	1.326667	3.113333
7	3.1133333	0.2886111	1.985556	6.152500

```
> aggregate(mydata2[,2:12],by=list(kmeans_clust$cluster),FUN = min)
```

Group.1	Type	Alcohol	Malic	Ash	Alcalinity	Magnesium	Phenols	Flavanoids
1	1	11.56	0.99	1.75	11.2	90	1.85	1.28
2	2	12.20	1.29	2.10	18.5	80	0.98	0.34
3	3	12.25	0.94	2.15	17.0	85	1.10	0.50
4	4	12.79	1.67	2.26	18.5	88	1.35	0.61
5	5	11.03	0.74	1.36	10.6	70	1.45	0.57
6	6	11.64	1.10	1.98	16.0	80	1.38	0.99
7	7	13.05	1.35	2.04	11.4	89	2.20	2.19

```
Nonflavanoids Proanthocyanins Color
```

Group.1	Nonflavanoids	Proanthocyanins	Color
1	0.13	1.28	2.60
2	0.17	0.55	3.85
3	0.21	0.42	3.05
4	0.22	0.97	7.10
5	0.17	0.42	1.28
6	0.14	0.41	2.08
7	0.17	1.25	3.95

```
> aggregate(mydata2[,2:12],by=list(kmeans_clust$cluster),FUN = max)
```

Group.1	Type	Alcohol	Malic	Ash	Alcalinity	Magnesium	Phenols	Flavanoids
1	1	14.22	3.99	3.23	30.0	162	3.38	5.08
2	2	13.88	5.51	2.72	25.0	106	2.32	1.59
3	3	13.69	3.88	2.64	27.0	122	2.53	1.30
4	4	14.34	5.65	2.86	25.5	123	2.80	1.57

```

5      5      2    13.86  4.31 2.70      26.0      108      3.52      3.75
6      6      2    13.49  5.80 2.92      28.5      134      2.62      2.65
7      7      1    14.83  4.04 2.72      22.5      132      3.88      3.93
Nonflavanoids Proanthocyanins      Color
1      0.47      3.28  6.130000
2      0.63      1.62  9.899999
3      0.63      1.46  8.210000
4      0.61      2.70 13.000000
5      0.52      3.58  5.300000
6      0.66      2.01  5.750000
7      0.50      2.96  8.900000
> kmeans_clust$centers
      Proline      Comp.1      Comp.2
1  0.42522567  0.8186532  0.09610543
2 -0.58634130 -1.4144988  0.28686687
3  0.04162093 -0.9852757  0.24524737
4 -0.32659002 -1.1790025  1.36464749
5 -0.85875428  0.1510432 -1.31632360
6 -0.81171614 -0.6227871 -0.78032720
7  1.58193227  1.1960417  0.85881203
> table(kmeans_clust$cluster)

 1  2  3  4  5  6  7
30 20 10 21 43 18 36
> write.csv(final1,file = "data_clust1.csv",row.names = F,col.names = F)
> getwd()
[1] "C:/Users/Arti Patel/Documents"

```