

Assignment 4

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```
wine = read.csv("C:/Users/dgmur/Downloads/wineFl2022.csv")
str(wine)
```

```
## 'data.frame':  350 obs. of  13 variables:
## $ fixed_acidity      : num  5.9 7 6.9 9.9 6.7 7.6 8 6.5 6.8 12.2 ...
## $ volatile_acidity   : num  0.32 0.23 0.24 0.49 0.16 0.78 0.18 0.22 0.28 0.45 ...
## $ citric_acid        : num  0.2 0.42 0.49 0.23 0.28 0 0.37 0.27 0.37 0.49 ...
## $ residual_sugar     : num  14.4 1.1 1.3 2.4 2.5 1.7 0.9 1.6 7 1.4 ...
## $ chlorides          : num  0.05 0.062 0.032 0.087 0.046 0.076 0.049 0.039 0.057 0.075 ...
## $ free_sulfur_dioxide : num  29 35 35 19 40 33 36 36 35 3 ...
## $ total_sulfur_dioxide: num  144 100 148 115 153 45 109 116 208 6 ...
## $ density            : num  0.997 0.993 0.993 0.995 0.992 ...
## $ pH                 : num  3.24 3.04 3.45 2.77 3.38 3.31 2.89 3.38 3.57 3.13 ...
## $ sulphates          : num  0.41 0.4 0.57 0.44 0.51 0.62 0.44 0.57 0.55 0.63 ...
## $ alcohol            : num  10.3 9.2 10.7 9.4 11.4 10.7 12.7 11 10.2 10.4 ...
## $ quality             : int   6 5 7 6 7 6 6 7 5 5 ...
## $ type               : int   0 0 0 0 0 1 1 0 0 1 ...
```

```
wine$type=as.factor(wine$type)
wine$quality=as.numeric(wine$quality)
winec = wine[1:11]
View(winec)
#1

#a
cor(winec)
```

```
##          fixed_acidity volatile_acidity citric_acid residual_sugar
## fixed_acidity      1.00000000      0.225020854  0.38344924  -0.09864763
## volatile_acidity    0.22502085      1.000000000 -0.32286866  -0.17139713
## citric_acid         0.38344924    -0.322868655  1.00000000  0.16225541
## residual_sugar     -0.09864763    -0.171397128  0.16225541  1.00000000
## chlorides          0.32756202      0.341374902  0.08681857  -0.12546264
## free_sulfur_dioxide -0.28237859    -0.287109795  0.04245716  0.43035646
## total_sulfur_dioxide -0.36919701    -0.358818093  0.10598165  0.47333598
## density            0.53262405      0.244588159  0.19580930  0.54020970
## pH                 -0.23369013      0.228314657 -0.27738639  -0.22678197
## sulphates          0.30713463      0.186344437  0.10589346  -0.19561141
## alcohol            -0.20003736      0.004018921 -0.09567631  -0.34337904
##          chlorides free_sulfur_dioxide total_sulfur_dioxide
## fixed_acidity      0.32756202      -0.282378587      -0.3691970
## volatile_acidity    0.34137490      -0.287109795      -0.3588181
## citric_acid         0.08681857      0.042457162      0.1059817
## residual_sugar     -0.12546264      0.430356464      0.4733360
## chlorides          1.00000000      -0.248831028      -0.3270541
## free_sulfur_dioxide -0.24883103      1.000000000      0.7626948
## total_sulfur_dioxide -0.32705409      0.762694847      1.0000000
## density            0.37847916      0.005382083      -0.0229437
## pH                 0.01866647      -0.128416387      -0.1705793
## sulphates          0.39260836      -0.147085345      -0.2853768
## alcohol            -0.27335277      -0.101607695      -0.1933318
##          density      pH      sulphates      alcohol
## fixed_acidity      0.532624054 -0.233690126  0.307134630 -0.200037357
## volatile_acidity    0.244588159  0.228314657  0.186344437  0.004018921
## citric_acid         0.195809304 -0.277386386  0.105893456 -0.095676312
## residual_sugar     0.540209698 -0.226781965 -0.195611413 -0.343379037
## chlorides          0.378479157  0.018666467  0.392608361 -0.273352774
## free_sulfur_dioxide 0.005382083 -0.128416387 -0.147085345 -0.101607695
## total_sulfur_dioxide -0.022943702 -0.170579290 -0.285376799 -0.193331790
## density            1.000000000 -0.005440951  0.228176692 -0.717527825
## pH                 -0.005440951  1.000000000  0.169653803  0.129038264
## sulphates          0.228176692  0.169653803  1.000000000  0.008272161
## alcohol            -0.717527825  0.129038264  0.008272161  1.000000000
```

```
cov(winec)
```

```

##          fixed_acidity volatile_acidity   citric_acid
## fixed_acidity      1.890202538      0.0509847073  7.466255e-02
## volatile_acidity    0.050984707      0.0271597634 -7.535804e-03
## citric_acid         0.074662546     -0.0075358043  2.005772e-02
## residual_sugar     -0.642850594     -0.1338861850  1.089204e-01
## chlorides           0.014041393      0.0017541110  3.833680e-04
## free_sulfur_dioxide -7.137701187     -0.8699282440  1.105514e-01
## total_sulfur_dioxide -29.064296357     -3.3859831355  8.594466e-01
## density             0.002214716      0.0001219106  8.387203e-05
## pH                  -0.047560426      0.0055699075 -5.815366e-03
## sulphates           0.061132820      0.0044460147  2.171208e-03
## alcohol             -0.326777023      0.0007869709 -1.610020e-02
##          residual_sugar      chlorides free_sulfur_dioxide
## fixed_acidity      -0.642850594  1.404139e-02      -7.137701e+00
## volatile_acidity    -0.133886185  1.754111e-03      -8.699282e-01
## citric_acid         0.108920385  3.833680e-04      1.105514e-01
## residual_sugar      22.466657798 -1.854156e-02      3.750334e+01
## chlorides           -0.018541555  9.721314e-04      -1.426394e-01
## free_sulfur_dioxide  37.503344249 -1.426394e-01      3.380219e+02
## total_sulfur_dioxide 128.465413426 -5.838880e-01      8.029171e+02
## density             0.007744167  3.569008e-05      2.992719e-04
## pH                  -0.159121531  8.615407e-05      -3.494975e-01
## sulphates           -0.134231641  1.772203e-03      -3.915014e-01
## alcohol             -1.933878769 -1.012680e-02      -2.219654e+00
##          total_sulfur_dioxide      density      pH
## fixed_acidity      -2.906430e+01  2.214716e-03 -4.756043e-02
## volatile_acidity    -3.385983e+00  1.219106e-04  5.569907e-03
## citric_acid         8.594466e-01  8.387203e-05 -5.815366e-03
## residual_sugar      1.284654e+02  7.744167e-03 -1.591215e-01
## chlorides           -5.838880e-01  3.569008e-05  8.615407e-05
## free_sulfur_dioxide  8.029171e+02  2.992719e-04 -3.494975e-01
## total_sulfur_dioxide 3.278649e+03 -3.973325e-03 -1.445856e+00
## density             -3.973325e-03  9.147159e-06 -2.435954e-06
## pH                  -1.445856e+00 -2.435954e-06  2.191303e-02
## sulphates           -2.365689e+00  9.990929e-05  3.635853e-03
## alcohol             -1.315337e+01 -2.578506e-03  2.269635e-02
##          sulphates      alcohol
## fixed_acidity      6.113282e-02 -3.267770e-01
## volatile_acidity    4.446015e-03  7.869709e-04
## citric_acid         2.171208e-03 -1.610020e-02
## residual_sugar     -1.342316e-01 -1.933879e+00
## chlorides           1.772203e-03 -1.012680e-02
## free_sulfur_dioxide -3.915014e-01 -2.219654e+00
## total_sulfur_dioxide -2.365689e+00 -1.315337e+01
## density             9.990929e-05 -2.578506e-03
## pH                  3.635853e-03  2.269635e-02
## sulphates           2.095960e-02  1.422973e-03
## alcohol             1.422973e-03  1.411798e+00

```

#b

#I chose correlation because the variables are on a different scale

winec.pca = prcomp(winec, scale. = T)

winec.pca

Standard deviations (1, .., p=11):

[1] 1.7232494 1.6076239 1.2068558 0.9946672 0.8652731 0.8279042 0.7221999

[8] 0.7184746 0.5619289 0.4299401 0.1660712

##

Rotation (n x k) = (11 x 11):

	PC1	PC2	PC3	PC4
fixed_acidity	0.348489546	-0.29235631	0.32295067	-0.09387512
volatile_acidity	0.343846756	0.03276812	-0.40392763	-0.17954875
citric_acid	-0.009760994	-0.28502639	0.57385827	0.24915131
residual_sugar	-0.267480445	-0.40830325	-0.18326601	-0.09808649
chlorides	0.369862176	-0.19645234	-0.08349223	0.17270436
free_sulfur_dioxide	-0.418729255	-0.17289180	-0.14492089	0.36002435
total_sulfur_dioxide	-0.472907822	-0.18332739	-0.11478151	0.21541693
density	0.185040901	-0.53684289	-0.23345742	-0.05441968
pH	0.127515499	0.22004395	-0.46478704	0.39420304
sulphates	0.318732272	-0.07236050	0.02559981	0.70297463
alcohol	-0.028613894	0.47069391	0.23940149	0.16991977

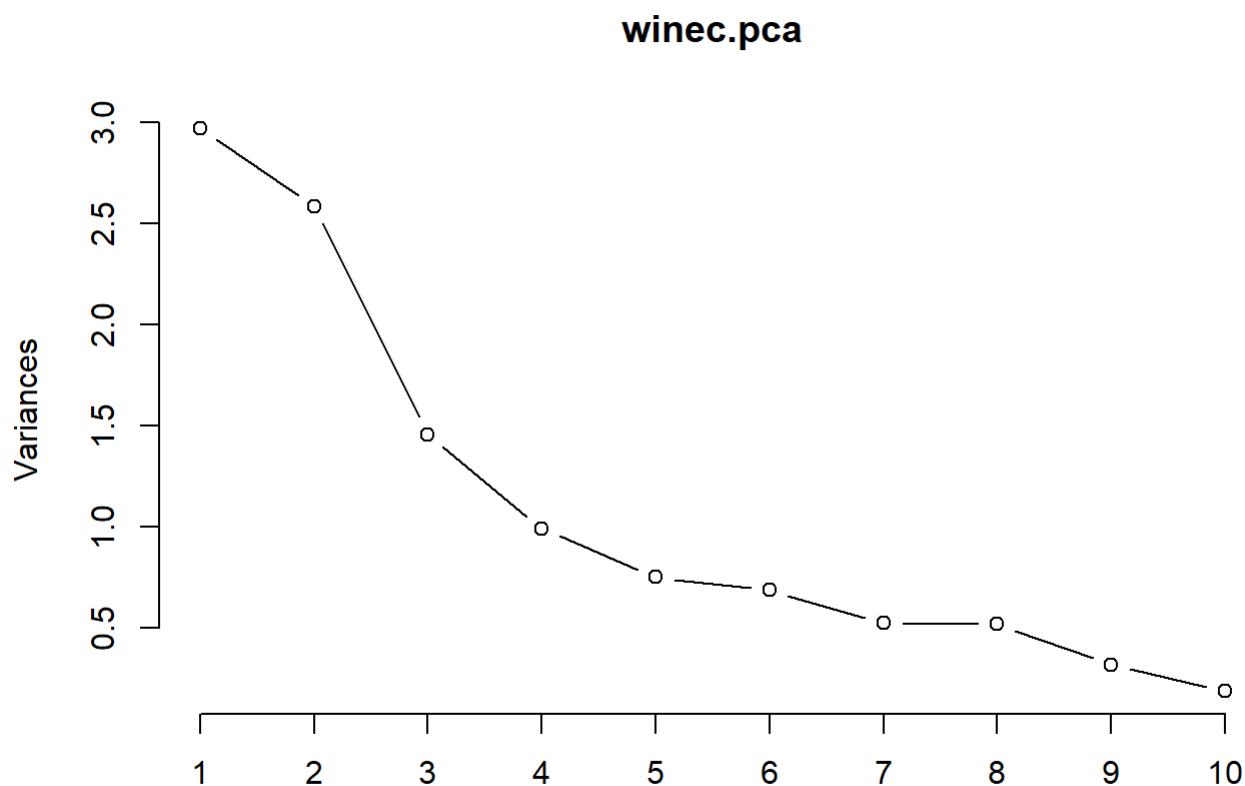
	PC5	PC6	PC7	PC8
fixed_acidity	0.04334508	-0.35893889	0.08376151	-0.42807410
volatile_acidity	0.39351946	-0.44374763	0.34170306	0.03221217
citric_acid	-0.26312144	-0.08990380	0.48010365	0.20787468
residual_sugar	-0.12195280	-0.30847793	-0.25710155	0.59045465
chlorides	0.35390879	0.51829635	0.30728852	0.42378092
free_sulfur_dioxide	0.35796425	-0.13909626	0.15758820	-0.22149241
total_sulfur_dioxide	0.19238345	0.01155103	0.23717318	-0.17675022
density	-0.20539242	-0.15702906	-0.08509948	-0.01480819
pH	-0.61576992	-0.08600367	0.32724636	-0.05467107
sulphates	0.14303805	-0.07636788	-0.53667765	-0.04357396
alcohol	0.16412954	-0.49498534	0.04190356	0.39858458

	PC9	PC10	PC11
fixed_acidity	0.40906345	0.27322565	0.342065272
volatile_acidity	-0.43991846	-0.12679298	0.078278568
citric_acid	-0.35278986	-0.22642878	-0.022744994
residual_sugar	0.07573828	0.05299393	0.430373805
chlorides	0.31160684	0.14293522	0.048809367
free_sulfur_dioxide	0.36690764	-0.53679357	-0.002643888
total_sulfur_dioxide	-0.24778764	0.70221161	-0.063510466
density	0.11286958	0.02068696	-0.729810143
pH	0.17856128	0.07100579	0.174135418
sulphates	-0.27646554	0.02552777	0.071039626
alcohol	0.30784128	0.21769828	-0.341627469

```
#c
summary(winec.pca)
```

```
## Importance of components:
##          PC1    PC2    PC3    PC4    PC5    PC6    PC7
## Standard deviation  1.723 1.6076 1.2069 0.99467 0.86527 0.82790 0.72220
## Proportion of Variance 0.270 0.2349 0.1324 0.08994 0.06806 0.06231 0.04742
## Cumulative Proportion 0.270 0.5049 0.6373 0.72726 0.79533 0.85764 0.90505
##          PC8    PC9    PC10    PC11
## Standard deviation  0.71847 0.56193 0.4299 0.16607
## Proportion of Variance 0.04693 0.02871 0.0168 0.00251
## Cumulative Proportion 0.95198 0.98069 0.9975 1.00000
```

```
screepplot(winec.pca, type = "lines")
```



```
#10 PCs
```

```
#d
```

```
c=cor(winec, winec.pca$x)
round(c,2)
```

##	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8	PC9
## fixed_acidity	0.60	-0.47	0.39	-0.09	0.04	-0.30	0.06	-0.31	0.23
## volatile_acidity	0.59	0.05	-0.49	-0.18	0.34	-0.37	0.25	0.02	-0.25
## citric_acid	-0.02	-0.46	0.69	0.25	-0.23	-0.07	0.35	0.15	-0.20
## residual_sugar	-0.46	-0.66	-0.22	-0.10	-0.11	-0.26	-0.19	0.42	0.04
## chlorides	0.64	-0.32	-0.10	0.17	0.31	0.43	0.22	0.30	0.18
## free_sulfur_dioxide	-0.72	-0.28	-0.17	0.36	0.31	-0.12	0.11	-0.16	0.21
## total_sulfur_dioxide	-0.81	-0.29	-0.14	0.21	0.17	0.01	0.17	-0.13	-0.14
## density	0.32	-0.86	-0.28	-0.05	-0.18	-0.13	-0.06	-0.01	0.06
## pH	0.22	0.35	-0.56	0.39	-0.53	-0.07	0.24	-0.04	0.10
## sulphates	0.55	-0.12	0.03	0.70	0.12	-0.06	-0.39	-0.03	-0.16
## alcohol	-0.05	0.76	0.29	0.17	0.14	-0.41	0.03	0.29	0.17
##	PC10	PC11							
## fixed_acidity	0.12	0.06							
## volatile_acidity	-0.05	0.01							
## citric_acid	-0.10	0.00							
## residual_sugar	0.02	0.07							
## chlorides	0.06	0.01							
## free_sulfur_dioxide	-0.23	0.00							
## total_sulfur_dioxide	0.30	-0.01							
## density	0.01	-0.12							
## pH	0.03	0.03							
## sulphates	0.01	0.01							
## alcohol	0.09	-0.06							

#PC1: Sulfur Dioxide

#PC2: Density of Alcohol by amount of sugar

#PC3: PH Level of acidity

#PC4: PH Level of sulfur containing bases

#PC5: Ph Level of volatile acidity and chlorides

#PC6: Chlorides residual sugar and acidity based on alcohol percentage

#PC7: PH Level of volatile, citric acid and sulphate

#PC8: fixed acidity and residual sugar based on percentage of alcohol

#PCA9: acidity, chlorides and sulfur dioxide

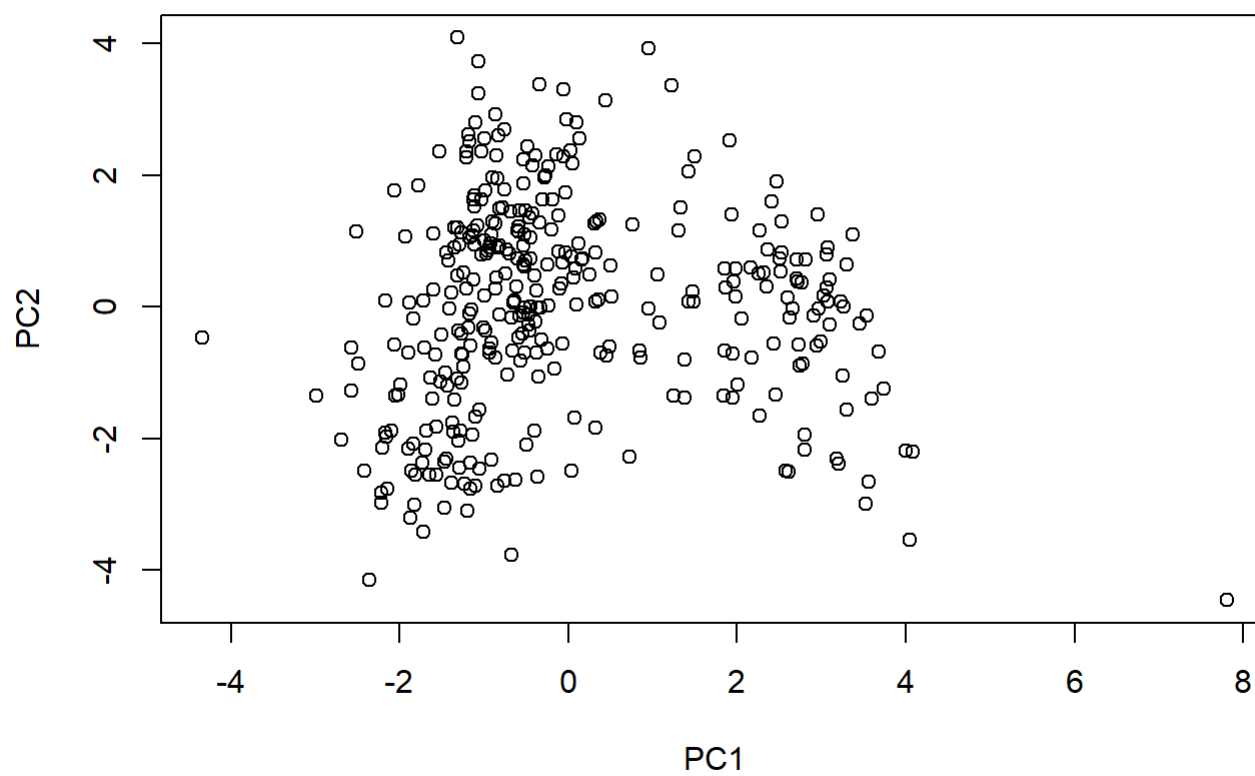
#PCA10: Sulfur dioxide

#e

#No, the proportion of variance is too low, and too close to PC2. PC1 has a 27% proportion of variance, while PC has 23.4%

#F

`plot(winec.pca$x[,1:2])`

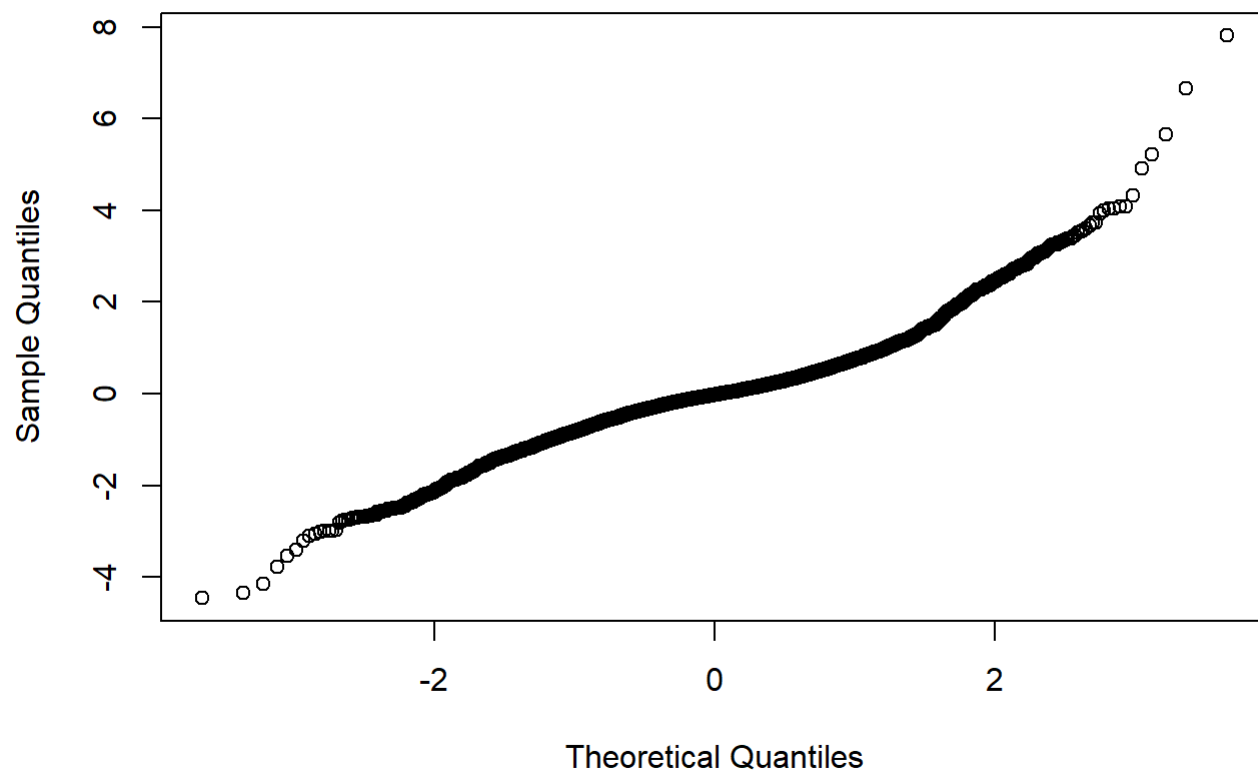


#There are no distinguishing groups, however, there is an outlier that is skewed to the higher end of PC1

#bonus

```
qqnorm(winec.pca$x)
```


Normal Q-Q Plot



#The plot shows that the distribution is not normal, this shows evidence of high variation.

#2

#a

```
wined = wine[-13]
```

```
cor(wined)
```

```

##          fixed_acidity volatile_acidity citric_acid residual_sugar
## fixed_acidity      1.00000000      0.22502085      0.38344924      -0.09864763
## volatile_acidity    0.22502085      1.00000000     -0.32286866     -0.17139713
## citric_acid         0.38344924     -0.32286865      1.00000000      0.16225541
## residual_sugar     -0.09864763     -0.17139712      0.16225541      1.00000000
## chlorides          0.32756202      0.34137490      0.08681857     -0.12546264
## free_sulfur_dioxide -0.28237859     -0.28710979      0.04245716      0.43035646
## total_sulfur_dioxide -0.36919701     -0.35881809      0.10598165      0.47333598
## density            0.53262405      0.24458815      0.19580930      0.54020970
## pH                 -0.23369013      0.22831465     -0.27738639     -0.22678197
## sulphates          0.30713463      0.18634443      0.10589346     -0.19561141
## alcohol            -0.20003736      0.00401892     -0.09567631     -0.34337904
## quality            -0.11163094     -0.23423052      0.02093266     -0.01517474
##          chlorides free_sulfur_dioxide total_sulfur_dioxide
## fixed_acidity      0.32756202     -0.282378587      -0.36919701
## volatile_acidity    0.34137490     -0.287109795      -0.35881809
## citric_acid         0.08681857      0.042457162      0.10598165
## residual_sugar     -0.12546264      0.430356464      0.47333598
## chlorides          1.00000000     -0.248831028      -0.32705409
## free_sulfur_dioxide -0.24883103      1.000000000      0.76269485
## total_sulfur_dioxide -0.32705409      0.762694847      1.00000000
## density            0.37847916      0.005382083      -0.02294370
## pH                 0.01866647     -0.128416387      -0.17057929
## sulphates          0.39260836     -0.147085345      -0.28537680
## alcohol            -0.27335277     -0.101607695      -0.19333179
## quality            -0.21776159      0.020542135      -0.01026426
##          density      pH      sulphates      alcohol
## fixed_acidity      0.532624054 -0.233690126  0.307134630 -0.200037357
## volatile_acidity    0.244588159  0.228314657  0.186344437  0.004018921
## citric_acid         0.195809304 -0.277386386  0.105893456 -0.095676312
## residual_sugar      0.540209698 -0.226781965 -0.195611413 -0.343379037
## chlorides           0.378479157  0.018666467  0.392608361 -0.273352774
## free_sulfur_dioxide  0.005382083 -0.128416387 -0.147085345 -0.101607695
## total_sulfur_dioxide -0.022943702 -0.170579290 -0.285376799 -0.193331790
## density             1.000000000 -0.005440951  0.228176692 -0.717527825
## pH                 -0.005440951  1.000000000  0.169653803  0.129038264
## sulphates           0.228176692  0.169653803  1.000000000  0.008272161
## alcohol             -0.717527825  0.129038264  0.008272161  1.000000000
## quality             -0.278927117  0.019394617  0.124581349  0.396693357
##          quality
## fixed_acidity      -0.11163094
## volatile_acidity    -0.23423053
## citric_acid         0.02093266
## residual_sugar     -0.01517474
## chlorides          -0.21776159
## free_sulfur_dioxide  0.02054213
## total_sulfur_dioxide -0.01026426
## density            -0.27892712
## pH                 0.01939462
## sulphates          0.12458135

```

```
## alcohol      0.39669336
## quality     1.00000000
```

```
cov(wined)
```

```

##          fixed_acidity volatile_acidity  citric_acid
## fixed_acidity      1.890202538      0.0509847073  7.466255e-02
## volatile_acidity    0.050984707      0.0271597634 -7.535804e-03
## citric_acid         0.074662546     -0.0075358043  2.005772e-02
## residual_sugar     -0.642850594     -0.1338861850  1.089204e-01
## chlorides          0.014041393      0.0017541110  3.833680e-04
## free_sulfur_dioxide -7.137701187     -0.8699282440  1.105514e-01
## total_sulfur_dioxide -29.064296357    -3.3859831355  8.594466e-01
## density            0.002214716      0.0001219106  8.387203e-05
## pH                 -0.047560426      0.0055699075 -5.815366e-03
## sulphates          0.061132820      0.0044460147  2.171208e-03
## alcohol            -0.326777023      0.0007869709 -1.610020e-02
## quality            -0.132358576     -0.0332904625  2.556693e-03
##          residual_sugar      chlorides free_sulfur_dioxide
## fixed_acidity      -0.642850594  1.404139e-02      -7.137701e+00
## volatile_acidity    -0.133886185  1.754111e-03      -8.699282e-01
## citric_acid         0.108920385  3.833680e-04      1.105514e-01
## residual_sugar      22.466657798 -1.854156e-02      3.750334e+01
## chlorides           -0.018541555  9.721314e-04      -1.426394e-01
## free_sulfur_dioxide  37.503344249 -1.426394e-01      3.380219e+02
## total_sulfur_dioxide 128.465413426 -5.838880e-01      8.029171e+02
## density             0.007744167  3.569008e-05      2.992719e-04
## pH                 -0.159121531  8.615407e-05      -3.494975e-01
## sulphates           -0.134231641  1.772203e-03      -3.915014e-01
## alcohol             -1.933878769 -1.012680e-02      -2.219654e+00
## quality             -0.062030291 -5.855407e-03      3.257102e-01
##          total_sulfur_dioxide      density      pH
## fixed_acidity      -2.906430e+01  2.214716e-03 -4.756043e-02
## volatile_acidity    -3.385983e+00  1.219106e-04  5.569907e-03
## citric_acid         8.594466e-01  8.387203e-05 -5.815366e-03
## residual_sugar      1.284654e+02  7.744167e-03 -1.591215e-01
## chlorides           -5.838880e-01  3.569008e-05  8.615407e-05
## free_sulfur_dioxide  8.029171e+02  2.992719e-04 -3.494975e-01
## total_sulfur_dioxide 3.278649e+03 -3.973325e-03 -1.445856e+00
## density             -3.973325e-03  9.147159e-06 -2.435954e-06
## pH                 -1.445856e+00 -2.435954e-06  2.191303e-02
## sulphates           -2.365689e+00  9.990929e-05  3.635853e-03
## alcohol             -1.315337e+01 -2.578506e-03  2.269635e-02
## quality             -5.068604e-01 -7.275238e-04  2.475972e-03
##          sulphates      alcohol      quality
## fixed_acidity      6.113282e-02 -3.267770e-01 -0.1323585755
## volatile_acidity    4.446015e-03  7.869709e-04 -0.0332904625
## citric_acid         2.171208e-03 -1.610020e-02  0.0025566926
## residual_sugar     -1.342316e-01 -1.933879e+00 -0.0620302906
## chlorides          1.772203e-03 -1.012680e-02 -0.0058554073
## free_sulfur_dioxide -3.915014e-01 -2.219654e+00  0.3257101924
## total_sulfur_dioxide -2.365689e+00 -1.315337e+01 -0.5068604175
## density            9.990929e-05 -2.578506e-03 -0.0007275238
## pH                 3.635853e-03  2.269635e-02  0.0024759722
## sulphates          2.095960e-02  1.422973e-03  0.0155545641

```

```
## alcohol      1.422973e-03  1.411798e+00  0.4064943375
## quality      1.555456e-02  4.064943e-01  0.7437494883
```

#b I chose correlation maxtrix because the variables are not one the same scale.

```
wine.pca = prcomp(wined,scale. = T)
```

```
wine.pca
```

```
## Standard deviations (1, .., p=12):
## [1] 1.7324119 1.6357004 1.2467205 1.0580945 0.9067586 0.8621958 0.7981909
## [8] 0.7194678 0.6417402 0.5537722 0.4277998 0.1658754
##
## Rotation (n x k) = (12 x 12):
##
```

	PC1	PC2	PC3	PC4
fixed_acidity	0.37384753	-0.20920790	0.34944193	0.101481689
volatile_acidity	0.34605826	0.06241715	-0.38344695	-0.006200858
citric_acid	0.01550334	-0.24883481	0.57469924	0.007293069
residual_sugar	-0.21073912	-0.43076891	-0.08045098	-0.163775376
chlorides	0.39301292	-0.13510339	-0.03755022	-0.114788780
free_sulfur_dioxide	-0.38691015	-0.23592001	-0.09789778	-0.287638032
total_sulfur_dioxide	-0.43615175	-0.25936172	-0.09887064	-0.145936568
density	0.25687987	-0.49174628	-0.11498773	-0.167868006
pH	0.10042460	0.22184714	-0.39264147	-0.448034762
sulphates	0.30735851	0.01054705	0.17042324	-0.613979534
alcohol	-0.10540151	0.47361058	0.21246419	-0.082411454
quality	-0.14068511	0.21960665	0.36299199	-0.482597698

	PC5	PC6	PC7	PC8
fixed_acidity	0.13882760	-0.12231102	0.37554679	-0.32337939
volatile_acidity	0.17854081	-0.49137727	0.39081629	0.20269632
citric_acid	-0.21716706	0.30102439	0.34520552	0.42157398
residual_sugar	0.45985782	-0.04057706	-0.03621810	0.36955879
chlorides	-0.38370342	-0.19562045	-0.44540196	0.52910531
free_sulfur_dioxide	-0.29102109	-0.32440834	0.22851724	-0.10881479
total_sulfur_dioxide	-0.25210986	-0.13491619	0.13311385	-0.02804903
density	0.29496975	0.11660006	0.04826955	-0.05610048
pH	-0.02773031	0.61105097	0.31137342	0.12066866
sulphates	-0.27804516	-0.13216386	-0.04388815	-0.32078443
alcohol	0.05413015	-0.26854548	0.36266713	0.35593643
quality	0.47615444	-0.11424921	-0.29101433	0.01265819

	PC9	PC10	PC11	PC12
fixed_acidity	0.29643825	0.3518858	0.27428184	-0.342475359
volatile_acidity	0.11151836	-0.4705296	-0.15021984	-0.075668433
citric_acid	0.02304073	-0.3421192	-0.23390324	0.023508793
residual_sugar	-0.41382212	0.1504179	0.07075169	-0.432262885
chlorides	0.26011021	0.2458745	0.13589718	-0.047862076
free_sulfur_dioxide	0.20186465	0.3555843	-0.53021442	0.003308181
total_sulfur_dioxide	0.20364509	-0.3097345	0.68535808	0.063560810
density	0.01043673	0.1040176	0.02101411	0.730305663
pH	0.20849284	0.1460274	0.07056923	-0.174230911
sulphates	-0.51367297	-0.1692788	0.04347695	-0.073442865
alcohol	-0.24893466	0.3759888	0.24899992	0.337711305
quality	0.45675359	-0.1735244	-0.06215342	0.009685102

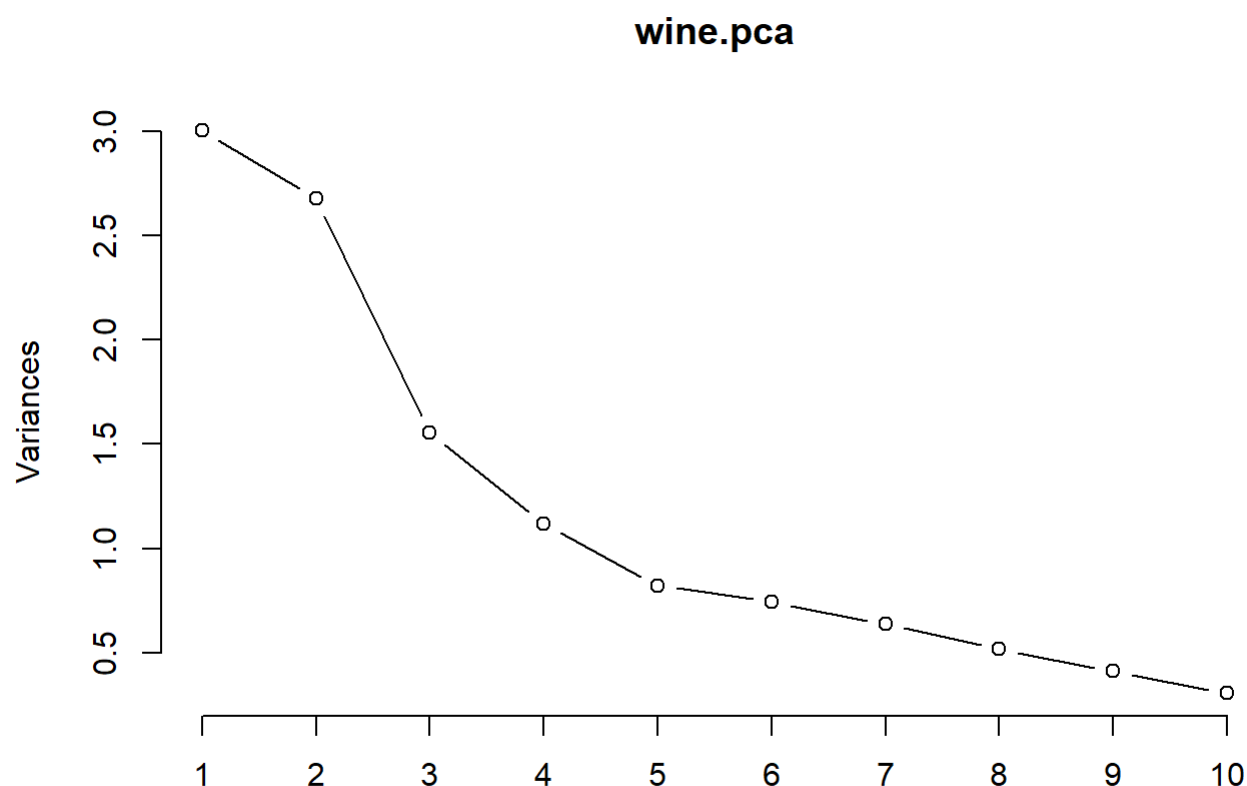
```
#c
summary(wine.pca)
```

```
## Importance of components:
```

	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	1.7324	1.6357	1.2467	1.0581	0.90676	0.86220	0.79819
Proportion of Variance	0.2501	0.2230	0.1295	0.0933	0.06852	0.06195	0.05309
Cumulative Proportion	0.2501	0.4731	0.6026	0.6959	0.76440	0.82635	0.87945

	PC8	PC9	PC10	PC11	PC12
Standard deviation	0.71947	0.64174	0.55377	0.42780	0.16588
Proportion of Variance	0.04314	0.03432	0.02556	0.01525	0.00229
Cumulative Proportion	0.92258	0.95690	0.98246	0.99771	1.00000

```
screepplot(wine.pca, type="lines")
```



```
#d
```

#10 PCs, however, I could get away with 9 PCs in comparison to question 1 due to PC9 in PC2 has a lower variance explained

```
#e
```

```
d=cor(wined, wine.pca$x)
```

```
round(d,2)
```

##	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8	PC9
## fixed_acidity	0.65	-0.34	0.44	0.11	0.13	-0.11	0.30	-0.23	0.19
## volatile_acidity	0.60	0.10	-0.48	-0.01	0.16	-0.42	0.31	0.15	0.07
## citric_acid	0.03	-0.41	0.72	0.01	-0.20	0.26	0.28	0.30	0.01
## residual_sugar	-0.37	-0.70	-0.10	-0.17	0.42	-0.03	-0.03	0.27	-0.27
## chlorides	0.68	-0.22	-0.05	-0.12	-0.35	-0.17	-0.36	0.38	0.17
## free_sulfur_dioxide	-0.67	-0.39	-0.12	-0.30	-0.26	-0.28	0.18	-0.08	0.13
## total_sulfur_dioxide	-0.76	-0.42	-0.12	-0.15	-0.23	-0.12	0.11	-0.02	0.13
## density	0.45	-0.80	-0.14	-0.18	0.27	0.10	0.04	-0.04	0.01
## pH	0.17	0.36	-0.49	-0.47	-0.03	0.53	0.25	0.09	0.13
## sulphates	0.53	0.02	0.21	-0.65	-0.25	-0.11	-0.04	-0.23	-0.33
## alcohol	-0.18	0.77	0.26	-0.09	0.05	-0.23	0.29	0.26	-0.16
## quality	-0.24	0.36	0.45	-0.51	0.43	-0.10	-0.23	0.01	0.29
##	PC10	PC11	PC12						
## fixed_acidity	0.19	0.12	-0.06						
## volatile_acidity	-0.26	-0.06	-0.01						
## citric_acid	-0.19	-0.10	0.00						
## residual_sugar	0.08	0.03	-0.07						
## chlorides	0.14	0.06	-0.01						
## free_sulfur_dioxide	0.20	-0.23	0.00						
## total_sulfur_dioxide	-0.17	0.29	0.01						
## density	0.06	0.01	0.12						
## pH	0.08	0.03	-0.03						
## sulphates	-0.09	0.02	-0.01						
## alcohol	0.21	0.11	0.06						
## quality	-0.10	-0.03	0.00						


```

#PC1: Wine type based of acidity, chlorides and sulfur dioxide

#PC2: Amount of density based on percentage of alcohol and residual sugar

#PC3: Acidity based on PH level and quality

#PC4: Amount of sulphates based on quality and PH level

#PC5: Quality based on Residual sugar and chlorides

#PC6: Average Ph level based on percentage of volatile acidity and free sulfur dioxide

#PC7: PH level of acidity

#PC8: Total Amount of residual sugars, chlorides and citric acid based on percentage of alcohol

#PC9: Quality based Sulphates and residual sugar

#PC10: Amount of Volatile Acidity based on quality of alcohol

#Yes, my interpretations do differ, categorical variable dominated the PCs

#F

summary(wine.pca)

```

```

## Importance of components:
##              PC1    PC2    PC3    PC4    PC5    PC6    PC7
## Standard deviation  1.7324 1.6357 1.2467 1.0581 0.90676 0.86220 0.79819
## Proportion of Variance 0.2501 0.2230 0.1295 0.0933 0.06852 0.06195 0.05309
## Cumulative Proportion 0.2501 0.4731 0.6026 0.6959 0.76440 0.82635 0.87945
##              PC8    PC9    PC10    PC11    PC12
## Standard deviation  0.71947 0.64174 0.55377 0.42780 0.16588
## Proportion of Variance 0.04314 0.03432 0.02556 0.01525 0.00229
## Cumulative Proportion 0.92258 0.95690 0.98246 0.99771 1.00000

```

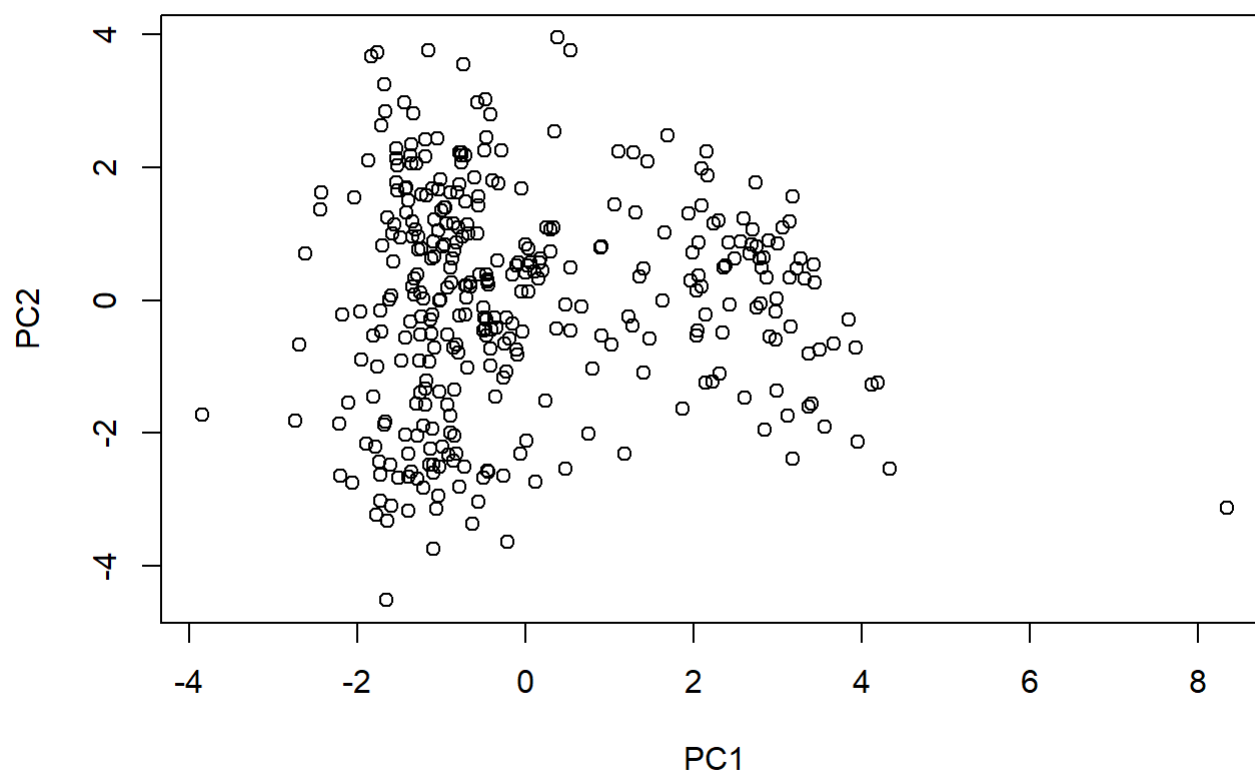
```

#No, the proportion of variance is too close in proportion of variance to PC2

#G

plot(wine.pca$x[,1:2])

```



#There are distinguishable groups, and there is one outlier on the PC1 axis. There is a difference compared to question one because the groups are distinguishable

#Bonus

#I would prefer with categorical variables because it's more easier to interpret relationships in each PC. An alternative method would be creating subsets for the data based on factors of the categorical variables, and then use PCA for each subset.

#3

```
library(rpart)
library(rpart.plot)
```

#a

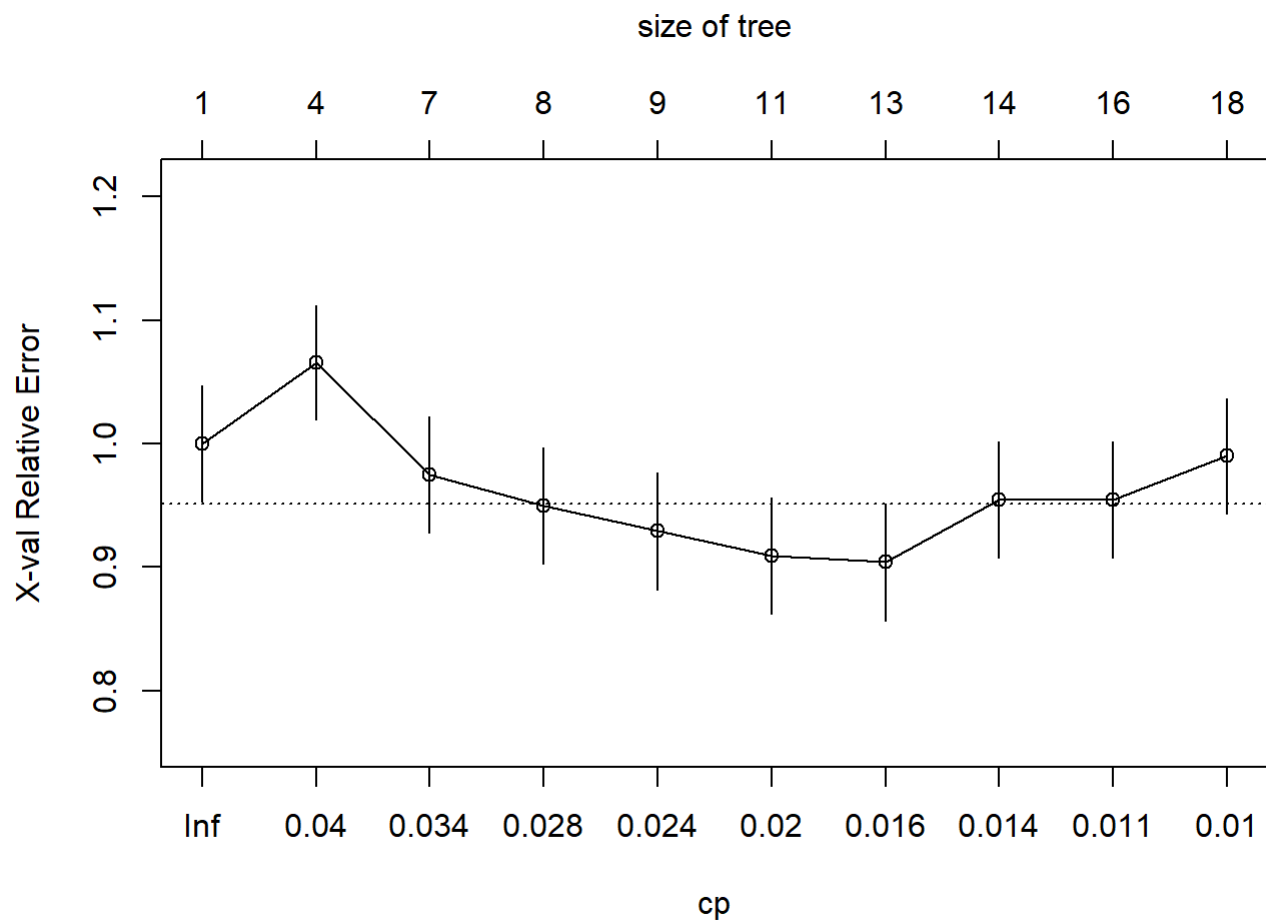
```
wine$quality=as.factor(wine$quality)

fit= rpart(quality~., method = "class", data=wine)

printcp(fit)
```

```
##
## Classification tree:
## rpart(formula = quality ~ ., data = wine, method = "class")
##
## Variables actually used in tree construction:
## [1] alcohol      chlorides      density
## [4] fixed_acidity pH          sulphates
## [7] total_sulfur_dioxide volatile_acidity
##
## Root node error: 198/350 = 0.56571
##
## n= 350
##
##      CP nsplit rel error  xerror    xstd
## 1  0.043771     0   1.00000 1.00000 0.046833
## 2  0.037037     3   0.86869 1.06566 0.046233
## 3  0.030303     6   0.75758 0.97475 0.046993
## 4  0.025253     7   0.72727 0.94949 0.047113
## 5  0.022727     8   0.70202 0.92929 0.047181
## 6  0.017677    10   0.65657 0.90909 0.047224
## 7  0.015152    12   0.62121 0.90404 0.047231
## 8  0.012626    13   0.60606 0.95455 0.047092
## 9  0.010101    15   0.58081 0.95455 0.047092
## 10 0.010000    17   0.56061 0.98990 0.046902
```

```
plotcp(fit)
```



```
summary(fit)
```

```
## Call:
## rpart(formula = quality ~ ., data = wine, method = "class")
##   n= 350
##
##           CP nsplit rel error   xerror   xstd
## 1  0.04377104      0 1.0000000 1.0000000 0.04683334
## 2  0.03703704      3 0.8686869 1.0656566 0.04623272
## 3  0.03030303      6 0.7575758 0.9747475 0.04699258
## 4  0.02525253      7 0.7272727 0.9494949 0.04711262
## 5  0.02272727      8 0.7020202 0.9292929 0.04718063
## 6  0.01767677     10 0.6565657 0.9090909 0.04722386
## 7  0.01515152     12 0.6212121 0.9040404 0.04723081
## 8  0.01262626     13 0.6060606 0.9545455 0.04709173
## 9  0.01010101     15 0.5808081 0.9545455 0.04709173
## 10 0.01000000     17 0.5606061 0.9898990 0.04690176
##
## Variable importance
##           alcohol          density total_sulfur_dioxide
##              18              13              13
##          sulphates          chlorides          fixed_acidity
##              12              11              8
## volatile_acidity free_sulfur_dioxide residual_sugar
##              6              4              4
##              type              pH          citric_acid
##              4              4              2
##
## Node number 1: 350 observations,    complexity param=0.04377104
## predicted class=6 expected loss=0.5657143 P(node) =1
## class counts:   14   108   152   67    9
## probabilities: 0.040 0.309 0.434 0.191 0.026
## left son=2 (218 obs) right son=3 (132 obs)
## Primary splits:
## alcohol < 10.75 to the left, improve=20.929650, (0 missing)
## density < 0.99222 to the right, improve=10.996840, (0 missing)
## chlorides < 0.0445 to the right, improve= 7.709710, (0 missing)
## free_sulfur_dioxide < 17.5 to the left, improve= 3.759377, (0 missing)
## volatile_acidity < 0.455 to the right, improve= 3.660000, (0 missing)
## Surrogate splits:
## density < 0.992935 to the right, agree=0.820, adj=0.523, (0 split)
## chlorides < 0.0365 to the right, agree=0.731, adj=0.288, (0 split)
## fixed_acidity < 6.05 to the right, agree=0.649, adj=0.068, (0 split)
## residual_sugar < 1.25 to the right, agree=0.649, adj=0.068, (0 split)
## sulphates < 0.335 to the right, agree=0.646, adj=0.061, (0 split)
##
## Node number 2: 218 observations,    complexity param=0.04377104
## predicted class=6 expected loss=0.559633 P(node) =0.6228571
## class counts:   11    95    96    11    5
## probabilities: 0.050 0.436 0.440 0.050 0.023
## left son=4 (209 obs) right son=5 (9 obs)
## Primary splits:
## total_sulfur_dioxide < 220 to the left, improve=4.775559, (0 missing)
## volatile_acidity < 0.2275 to the right, improve=4.748625, (0 missing)
```

```

##      density          < 0.99369  to the left,  improve=2.768169, (0 missing)
##      pH              < 3.055    to the right, improve=2.679951, (0 missing)
##      sulphates       < 0.375    to the left,  improve=2.469719, (0 missing)
##
## Node number 3: 132 observations,    complexity param=0.03703704
## predicted class=6 expected loss=0.5757576 P(node) =0.3771429
## class counts:      3    13    56    56    4
## probabilities: 0.023 0.098 0.424 0.424 0.030
## left son=6 (125 obs) right son=7 (7 obs)
## Primary splits:
##      chlorides          < 0.023    to the right, improve=3.863152, (0 missing)
##      total_sulfur_dioxide < 80.5    to the left,  improve=3.237374, (0 missing)
##      volatile_acidity    < 0.485    to the right, improve=3.213442, (0 missing)
##      sulphates          < 0.375    to the left,  improve=3.043466, (0 missing)
##      free_sulfur_dioxide < 15.5    to the left,  improve=2.932734, (0 missing)
##
## Node number 4: 209 observations,    complexity param=0.04377104
## predicted class=5 expected loss=0.5454545 P(node) =0.5971429
## class counts:      11    95    87    11    5
## probabilities: 0.053 0.455 0.416 0.053 0.024
## left son=8 (154 obs) right son=9 (55 obs)
## Primary splits:
##      volatile_acidity    < 0.2275   to the right, improve=5.501982, (0 missing)
##      sulphates          < 0.555    to the left,  improve=3.023857, (0 missing)
##      pH                < 3.295    to the left,  improve=2.873411, (0 missing)
##      total_sulfur_dioxide < 22.5    to the left,  improve=2.214736, (0 missing)
##      density           < 0.99369   to the left,  improve=2.144072, (0 missing)
## Surrogate splits:
##      residual_sugar      < 14        to the left,  agree=0.770, adj=0.127, (0 split)
##      alcohol             < 8.95     to the right, agree=0.756, adj=0.073, (0 split)
##      free_sulfur_dioxide < 75.5    to the left,  agree=0.742, adj=0.018, (0 split)
##      density            < 0.999405 to the left,  agree=0.742, adj=0.018, (0 split)
##
## Node number 5: 9 observations
## predicted class=6 expected loss=0 P(node) =0.02571429
## class counts:      0    0    9    0    0
## probabilities: 0.000 0.000 1.000 0.000 0.000
##
## Node number 6: 125 observations,    complexity param=0.03703704
## predicted class=6 expected loss=0.552 P(node) =0.3571429
## class counts:      3    13    56    49    4
## probabilities: 0.024 0.104 0.448 0.392 0.032
## left son=12 (32 obs) right son=13 (93 obs)
## Primary splits:
##      total_sulfur_dioxide < 80.5    to the left,  improve=2.975253, (0 missing)
##      chlorides          < 0.0485    to the right, improve=2.779764, (0 missing)
##      residual_sugar      < 1.55     to the left,  improve=2.767268, (0 missing)
##      volatile_acidity    < 0.485    to the right, improve=2.697455, (0 missing)
##      free_sulfur_dioxide < 14        to the left,  improve=2.552000, (0 missing)
## Surrogate splits:
##      type                splits as RL,          agree=0.928, adj=0.719, (0 split)
##      chlorides          < 0.0585    to the right, agree=0.896, adj=0.594, (0 split)

```

```

##      free_sulfur_dioxide < 18.5      to the left,  agree=0.896, adj=0.594, (0 split)
##      density              < 0.99372 to the right, agree=0.864, adj=0.469, (0 split)
##      fixed_acidity        < 8.05     to the right, agree=0.840, adj=0.375, (0 split)
##
## Node number 7: 7 observations
##   predicted class=7   expected loss=0   P(node) =0.02
##   class counts:      0      0      0      7      0
##   probabilities: 0.000 0.000 0.000 1.000 0.000
##
## Node number 8: 154 observations,      complexity param=0.03030303
##   predicted class=5   expected loss=0.4675325   P(node) =0.44
##   class counts:      9      82      56      5      2
##   probabilities: 0.058 0.532 0.364 0.032 0.013
##   left son=16 (88 obs) right son=17 (66 obs)
##   Primary splits:
##     sulphates          < 0.555      to the left,  improve=3.987013, (0 missing)
##     alcohol            < 9.85       to the left,  improve=3.692264, (0 missing)
##     total_sulfur_dioxide < 174      to the right, improve=2.783800, (0 missing)
##     fixed_acidity      < 6.25       to the right, improve=2.257234, (0 missing)
##     free_sulfur_dioxide < 36.5      to the right, improve=1.979978, (0 missing)
##   Surrogate splits:
##     type                splits as LR,      agree=0.721, adj=0.348, (0 split)
##     total_sulfur_dioxide < 92.5      to the right, agree=0.714, adj=0.333, (0 split)
##     chlorides           < 0.069      to the left,  agree=0.675, adj=0.242, (0 split)
##     fixed_acidity       < 8.05       to the left,  agree=0.662, adj=0.212, (0 split)
##     free_sulfur_dioxide < 18.5      to the right, agree=0.630, adj=0.136, (0 split)
##
## Node number 9: 55 observations,      complexity param=0.01262626
##   predicted class=6   expected loss=0.4363636   P(node) =0.1571429
##   class counts:      2      13      31      6      3
##   probabilities: 0.036 0.236 0.564 0.109 0.055
##   left son=18 (37 obs) right son=19 (18 obs)
##   Primary splits:
##     pH                  < 3.275      to the left,  improve=3.896970, (0 missing)
##     fixed_acidity       < 7.05       to the left,  improve=3.797332, (0 missing)
##     sulphates          < 0.525      to the right, improve=3.431768, (0 missing)
##     total_sulfur_dioxide < 137.5    to the left,  improve=3.131276, (0 missing)
##     chlorides          < 0.039      to the right, improve=2.855303, (0 missing)
##   Surrogate splits:
##     sulphates          < 0.565      to the left,  agree=0.782, adj=0.333, (0 split)
##     volatile_acidity   < 0.1175     to the right, agree=0.709, adj=0.111, (0 split)
##     residual_sugar     < 1.7         to the right, agree=0.709, adj=0.111, (0 split)
##     total_sulfur_dioxide < 167.5    to the left,  agree=0.709, adj=0.111, (0 split)
##     density            < 0.99908    to the left,  agree=0.709, adj=0.111, (0 split)
##
## Node number 12: 32 observations,      complexity param=0.01010101
##   predicted class=6   expected loss=0.4375   P(node) =0.09142857
##   class counts:      1      7      18      6      0
##   probabilities: 0.031 0.219 0.562 0.188 0.000
##   left son=24 (14 obs) right son=25 (18 obs)
##   Primary splits:
##     sulphates          < 0.605      to the left,  improve=3.393849, (0 missing)

```

```

##      free_sulfur_dioxide < 14      to the left,  improve=2.988366, (0 missing)
##      density             < 0.9938  to the left,  improve=2.598128, (0 missing)
##      chlorides           < 0.0695  to the left,  improve=2.473214, (0 missing)
##      volatile_acidity    < 0.325   to the right, improve=1.687500, (0 missing)
##  Surrogate splits:
##      density             < 0.9938  to the left,  agree=0.844, adj=0.643, (0 split)
##      chlorides           < 0.0565  to the left,  agree=0.781, adj=0.500, (0 split)
##      type                 splits as LR,          agree=0.781, adj=0.500, (0 split)
##      fixed_acidity        < 6.7     to the left,  agree=0.750, adj=0.429, (0 split)
##      total_sulfur_dioxide < 37.5    to the right, agree=0.719, adj=0.357, (0 split)
##
## Node number 13: 93 observations,      complexity param=0.03703704
##  predicted class=7  expected loss=0.5376344  P(node) =0.2657143
##  class counts:      2      6      38      43      4
##  probabilities: 0.022 0.065 0.409 0.462 0.043
##  left son=26 (70 obs) right son=27 (23 obs)
##  Primary splits:
##      sulphates           < 0.565    to the left,  improve=5.525893, (0 missing)
##      residual_sugar      < 1.55     to the left,  improve=3.730862, (0 missing)
##      density             < 0.990985 to the right, improve=2.280156, (0 missing)
##      total_sulfur_dioxide < 153.5   to the right, improve=2.126679, (0 missing)
##      pH                  < 3.055    to the left,  improve=1.542593, (0 missing)
##  Surrogate splits:
##      citric_acid         < 0.165    to the right, agree=0.785, adj=0.130, (0 split)
##      alcohol             < 13.7     to the left,  agree=0.785, adj=0.130, (0 split)
##      volatile_acidity    < 0.115    to the right, agree=0.774, adj=0.087, (0 split)
##      free_sulfur_dioxide < 60.5     to the left,  agree=0.763, adj=0.043, (0 split)
##
## Node number 16: 88 observations,      complexity param=0.01515152
##  predicted class=5  expected loss=0.3636364  P(node) =0.2514286
##  class counts:      6      56      24      2      0
##  probabilities: 0.068 0.636 0.273 0.023 0.000
##  left son=32 (73 obs) right son=33 (15 obs)
##  Primary splits:
##      fixed_acidity       < 6.25     to the right, improve=3.040349, (0 missing)
##      total_sulfur_dioxide < 156.5   to the right, improve=2.786830, (0 missing)
##      alcohol             < 10.55    to the left,  improve=1.503987, (0 missing)
##      volatile_acidity    < 0.69     to the right, improve=1.448292, (0 missing)
##      residual_sugar      < 5.65     to the left,  improve=1.415388, (0 missing)
##  Surrogate splits:
##      density             < 0.99234  to the right, agree=0.875, adj=0.267, (0 split)
##      pH                  < 3.395    to the left,  agree=0.852, adj=0.133, (0 split)
##      residual_sugar      < 16.15    to the left,  agree=0.841, adj=0.067, (0 split)
##
## Node number 17: 66 observations,      complexity param=0.02525253
##  predicted class=6  expected loss=0.5151515  P(node) =0.1885714
##  class counts:      3      26      32      3      2
##  probabilities: 0.045 0.394 0.485 0.045 0.030
##  left son=34 (8 obs) right son=35 (58 obs)
##  Primary splits:
##      total_sulfur_dioxide < 22.5     to the left,  improve=2.452194, (0 missing)
##      citric_acid         < 0.425    to the right, improve=2.201154, (0 missing)

```



```

##      alcohol          < 9.85      to the left,  improve=1.836398, (0 missing)
##      sulphates        < 0.625     to the right, improve=1.601399, (0 missing)
##      chlorides        < 0.093     to the right, improve=1.566234, (0 missing)
## Surrogate splits:
##      free_sulfur_dioxide < 4.5      to the left,  agree=0.909, adj=0.25, (0 split)
##
## Node number 18: 37 observations,      complexity param=0.01262626
## predicted class=6 expected loss=0.5945946 P(node) =0.1057143
## class counts:      2      12      15      5      3
## probabilities: 0.054 0.324 0.405 0.135 0.081
## left son=36 (15 obs) right son=37 (22 obs)
## Primary splits:
##      fixed_acidity      < 6.85      to the left,  improve=2.872727, (0 missing)
##      sulphates          < 0.415     to the left,  improve=2.132867, (0 missing)
##      density            < 0.99495   to the right, improve=1.807692, (0 missing)
##      pH                 < 3.05      to the right, improve=1.767241, (0 missing)
##      total_sulfur_dioxide < 137.5    to the left,  improve=1.417647, (0 missing)
## Surrogate splits:
##      sulphates          < 0.4        to the left,  agree=0.730, adj=0.333, (0 split)
##      density            < 0.992415   to the left,  agree=0.703, adj=0.267, (0 split)
##      citric_acid        < 0.245      to the left,  agree=0.676, adj=0.200, (0 split)
##      pH                 < 3.16      to the right, agree=0.676, adj=0.200, (0 split)
##      alcohol            < 10.45      to the right, agree=0.676, adj=0.200, (0 split)
##
## Node number 19: 18 observations
## predicted class=6 expected loss=0.1111111 P(node) =0.05142857
## class counts:      0      1      16      1      0
## probabilities: 0.000 0.056 0.889 0.056 0.000
##
## Node number 24: 14 observations
## predicted class=5 expected loss=0.5 P(node) =0.04
## class counts:      1      7      5      1      0
## probabilities: 0.071 0.500 0.357 0.071 0.000
##
## Node number 25: 18 observations
## predicted class=6 expected loss=0.2777778 P(node) =0.05142857
## class counts:      0      0      13      5      0
## probabilities: 0.000 0.000 0.722 0.278 0.000
##
## Node number 26: 70 observations,      complexity param=0.02272727
## predicted class=6 expected loss=0.5 P(node) =0.2
## class counts:      2      5      35      25      3
## probabilities: 0.029 0.071 0.500 0.357 0.043
## left son=52 (43 obs) right son=53 (27 obs)
## Primary splits:
##      density            < 0.990985   to the right, improve=2.926935, (0 missing)
##      alcohol            < 11.05      to the right, improve=2.385714, (0 missing)
##      chlorides          < 0.0345     to the left,  improve=2.035047, (0 missing)
##      residual_sugar     < 1.65      to the left,  improve=1.768627, (0 missing)
##      free_sulfur_dioxide < 22.5      to the right, improve=1.330939, (0 missing)
## Surrogate splits:
##      alcohol            < 12.1      to the left,  agree=0.871, adj=0.667, (0 split)

```

```

##      residual_sugar      < 3.6      to the right, agree=0.771, adj=0.407, (0 split)
##      chlorides          < 0.0355   to the right, agree=0.729, adj=0.296, (0 split)
##      total_sulfur_dioxide < 112.5   to the right, agree=0.729, adj=0.296, (0 split)
##      pH                 < 3.03     to the right, agree=0.657, adj=0.111, (0 split)
##
## Node number 27: 23 observations
##   predicted class=7   expected loss=0.2173913   P(node) =0.06571429
##   class counts:      0      1      3      18      1
##   probabilities: 0.000 0.043 0.130 0.783 0.043
##
## Node number 32: 73 observations
##   predicted class=5   expected loss=0.3150685   P(node) =0.2085714
##   class counts:       6     50     15      2      0
##   probabilities: 0.082 0.685 0.205 0.027 0.000
##
## Node number 33: 15 observations
##   predicted class=6   expected loss=0.4   P(node) =0.04285714
##   class counts:       0      6      9      0      0
##   probabilities: 0.000 0.400 0.600 0.000 0.000
##
## Node number 34: 8 observations
##   predicted class=5   expected loss=0.25   P(node) =0.02285714
##   class counts:       0      6      1      0      1
##   probabilities: 0.000 0.750 0.125 0.000 0.125
##
## Node number 35: 58 observations,      complexity param=0.01767677
##   predicted class=6   expected loss=0.4655172   P(node) =0.1657143
##   class counts:       3     20     31      3      1
##   probabilities: 0.052 0.345 0.534 0.052 0.017
##   left son=70 (39 obs) right son=71 (19 obs)
##   Primary splits:
##     total_sulfur_dioxide < 47      to the right, improve=3.011215, (0 missing)
##     alcohol              < 9.85    to the left,  improve=2.359824, (0 missing)
##     free_sulfur_dioxide  < 8.5     to the right, improve=1.943591, (0 missing)
##     citric_acid          < 0.48    to the right, improve=1.685157, (0 missing)
##     sulphates            < 0.72    to the right, improve=1.605187, (0 missing)
##   Surrogate splits:
##     free_sulfur_dioxide < 11      to the right, agree=0.828, adj=0.474, (0 split)
##     chlorides           < 0.0715   to the left,  agree=0.793, adj=0.368, (0 split)
##     residual_sugar      < 2.3      to the right, agree=0.776, adj=0.316, (0 split)
##     volatile_acidity    < 0.525    to the left,  agree=0.759, adj=0.263, (0 split)
##     type                splits as LR,      agree=0.759, adj=0.263, (0 split)
##
## Node number 36: 15 observations
##   predicted class=5   expected loss=0.4   P(node) =0.04285714
##   class counts:       1      9      4      0      1
##   probabilities: 0.067 0.600 0.267 0.000 0.067
##
## Node number 37: 22 observations
##   predicted class=6   expected loss=0.5   P(node) =0.06285714
##   class counts:       1      3     11      5      2
##   probabilities: 0.045 0.136 0.500 0.227 0.091

```

```

##
## Node number 52: 43 observations,    complexity param=0.02272727
## predicted class=7 expected loss=0.5581395 P(node) =0.1228571
## class counts:      2      5      16      19      1
## probabilities: 0.047 0.116 0.372 0.442 0.023
## left son=104 (16 obs) right son=105 (27 obs)
## Primary splits:
## total_sulfur_dioxide < 150.5 to the right, improve=2.620155, (0 missing)
## fixed_acidity < 5.95 to the left, improve=1.924917, (0 missing)
## density < 0.99217 to the right, improve=1.820155, (0 missing)
## alcohol < 11.05 to the right, improve=1.820155, (0 missing)
## pH < 3.135 to the right, improve=1.765610, (0 missing)
## Surrogate splits:
## fixed_acidity < 7.05 to the right, agree=0.721, adj=0.25, (0 split)
## volatile_acidity < 0.42 to the right, agree=0.721, adj=0.25, (0 split)
## residual_sugar < 10.75 to the right, agree=0.721, adj=0.25, (0 split)
## chlorides < 0.049 to the right, agree=0.721, adj=0.25, (0 split)
## free_sulfur_dioxide < 36.5 to the right, agree=0.721, adj=0.25, (0 split)
##
## Node number 53: 27 observations,    complexity param=0.01010101
## predicted class=6 expected loss=0.2962963 P(node) =0.07714286
## class counts:      0      0      19      6      2
## probabilities: 0.000 0.000 0.704 0.222 0.074
## left son=106 (18 obs) right son=107 (9 obs)
## Primary splits:
## alcohol < 12.84667 to the left, improve=3.370370, (0 missing)
## free_sulfur_dioxide < 37.5 to the left, improve=1.919577, (0 missing)
## density < 0.98916 to the right, improve=1.411306, (0 missing)
## total_sulfur_dioxide < 108 to the right, improve=1.348148, (0 missing)
## residual_sugar < 1.45 to the left, improve=1.038258, (0 missing)
## Surrogate splits:
## volatile_acidity < 0.36 to the left, agree=0.815, adj=0.444, (0 split)
## density < 0.98916 to the right, agree=0.815, adj=0.444, (0 split)
## fixed_acidity < 5.55 to the right, agree=0.778, adj=0.333, (0 split)
## residual_sugar < 2.7 to the left, agree=0.778, adj=0.333, (0 split)
## free_sulfur_dioxide < 37.5 to the left, agree=0.778, adj=0.333, (0 split)
##
## Node number 70: 39 observations,    complexity param=0.01767677
## predicted class=5 expected loss=0.5641026 P(node) =0.1114286
## class counts:      3      17      16      3      0
## probabilities: 0.077 0.436 0.410 0.077 0.000
## left son=140 (20 obs) right son=141 (19 obs)
## Primary splits:
## sulphates < 0.635 to the right, improve=2.653576, (0 missing)
## free_sulfur_dioxide < 24.5 to the left, improve=2.289817, (0 missing)
## chlorides < 0.0705 to the right, improve=2.289817, (0 missing)
## total_sulfur_dioxide < 140.5 to the left, improve=1.880769, (0 missing)
## residual_sugar < 5.55 to the left, improve=1.788701, (0 missing)
## Surrogate splits:
## citric_acid < 0.33 to the right, agree=0.718, adj=0.421, (0 split)
## chlorides < 0.043 to the right, agree=0.692, adj=0.368, (0 split)
## pH < 3.355 to the left, agree=0.641, adj=0.263, (0 split)

```

```

##      fixed_acidity      < 7.05      to the right, agree=0.615, adj=0.211, (0 split)
##      free_sulfur_dioxide < 21      to the left, agree=0.615, adj=0.211, (0 split)
##
## Node number 71: 19 observations
##   predicted class=6   expected loss=0.2105263   P(node) =0.05428571
##   class counts:      0      3      15      0      1
##   probabilities: 0.000 0.158 0.789 0.000 0.053
##
## Node number 104: 16 observations
##   predicted class=6   expected loss=0.4375   P(node) =0.04571429
##   class counts:      1      2      9      3      1
##   probabilities: 0.062 0.125 0.562 0.188 0.062
##
## Node number 105: 27 observations
##   predicted class=7   expected loss=0.4074074   P(node) =0.07714286
##   class counts:      1      3      7      16      0
##   probabilities: 0.037 0.111 0.259 0.593 0.000
##
## Node number 106: 18 observations
##   predicted class=6   expected loss=0.1111111   P(node) =0.05142857
##   class counts:      0      0      16      1      1
##   probabilities: 0.000 0.000 0.889 0.056 0.056
##
## Node number 107: 9 observations
##   predicted class=7   expected loss=0.4444444   P(node) =0.02571429
##   class counts:      0      0      3      5      1
##   probabilities: 0.000 0.000 0.333 0.556 0.111
##
## Node number 140: 20 observations
##   predicted class=5   expected loss=0.35   P(node) =0.05714286
##   class counts:      1     13      6      0      0
##   probabilities: 0.050 0.650 0.300 0.000 0.000
##
## Node number 141: 19 observations
##   predicted class=6   expected loss=0.4736842   P(node) =0.05428571
##   class counts:      2      4     10      3      0
##   probabilities: 0.105 0.211 0.526 0.158 0.000

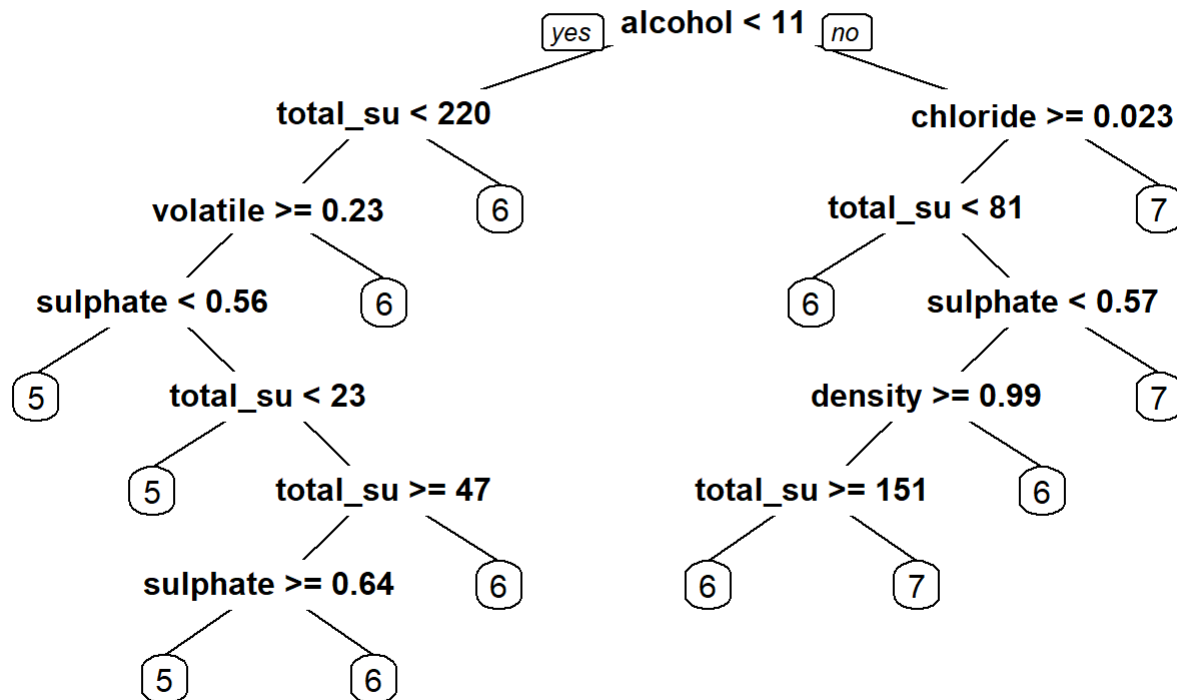
```

```
pfit<- prune(fit, cp=fit$cpstable[which.min(fit$cpstable[, "xerror"]), "CP"])
```

```
#b
```

```
prp(pfit, main = "Pruned Classification Tree for Wine Qualities")
```

Pruned Classification Tree for Wine Qualities



#The predicted the wine quality would be a 6

```
#c
library(randomForest)
```

```
## randomForest 4.7-1.1
```

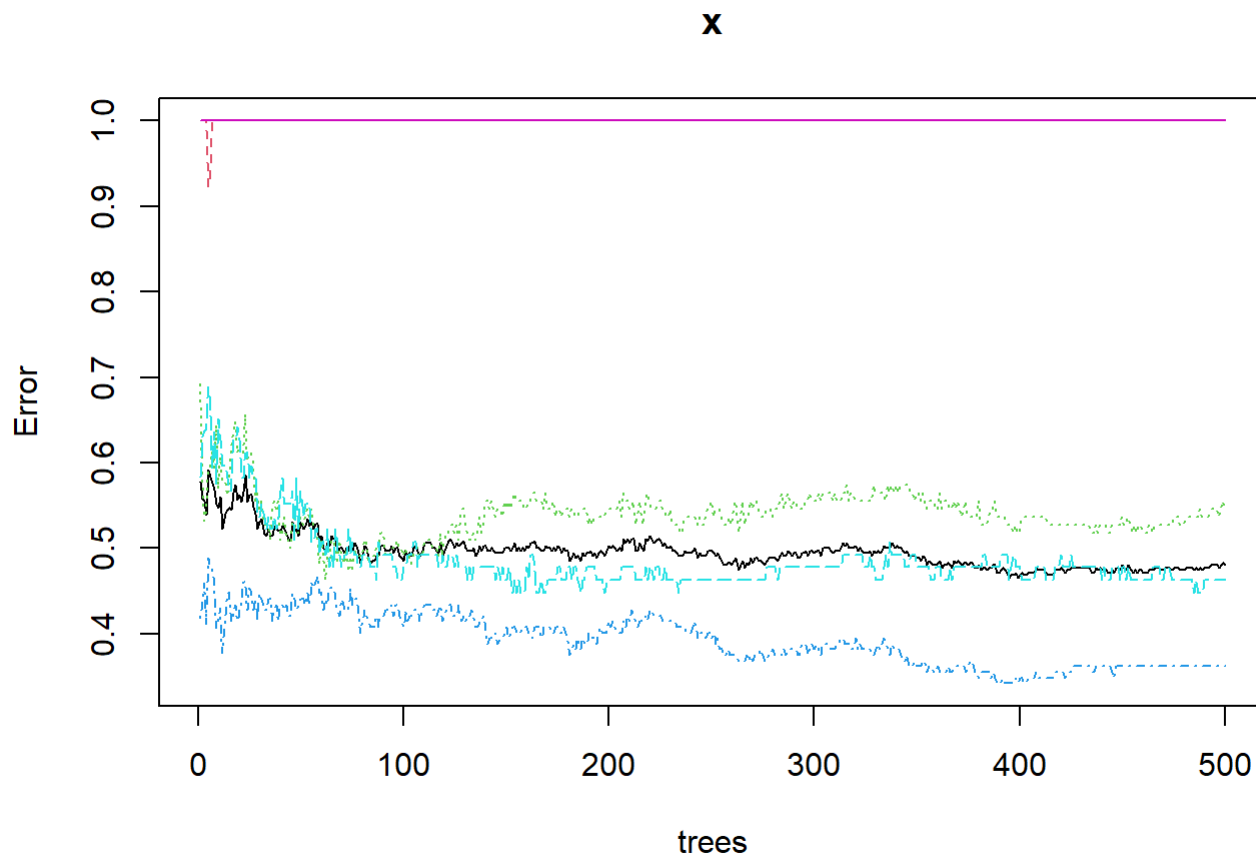
```
## Type rfNews() to see new features/changes/bug fixes.
```

```
set.seed(478)

wine$quality=as.factor(wine$quality)

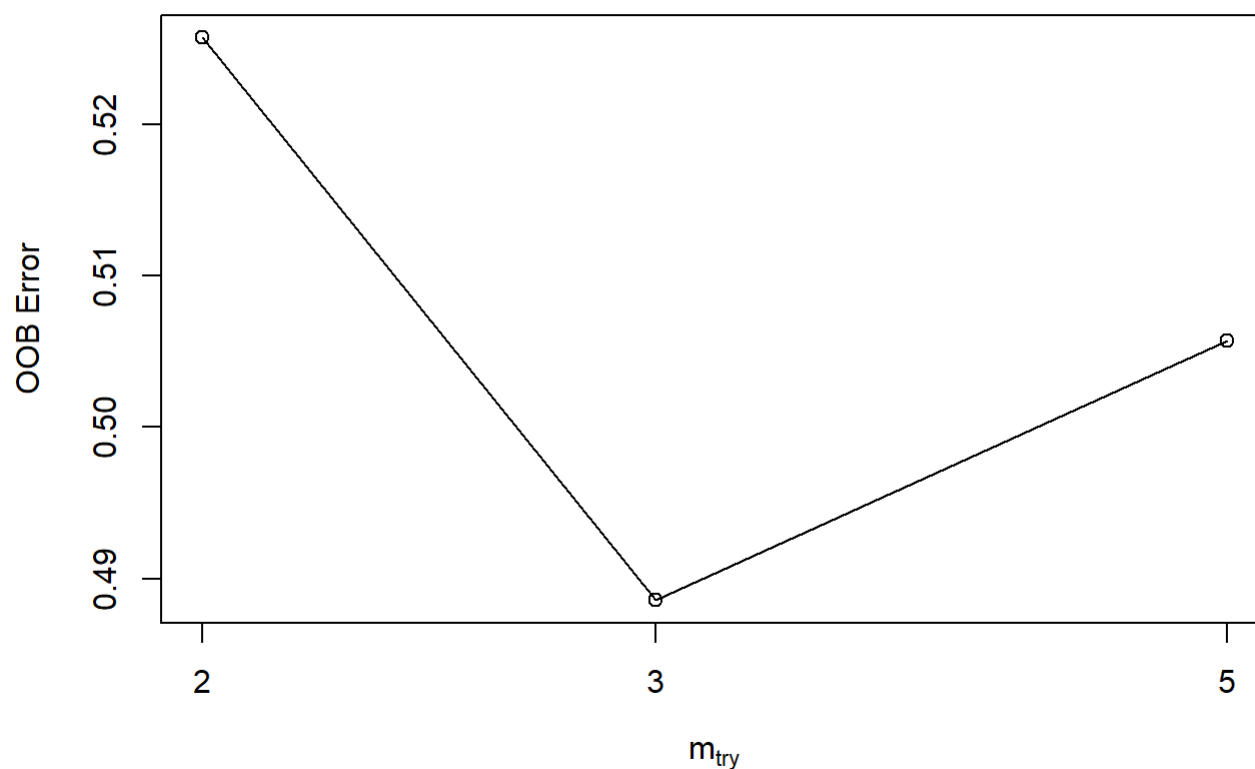
x = randomForest(quality~., data=wine)

plot(x)
```



```
t = tuneRF(wine[,-12],wine[,12],
           stepFactor = 0.7, plot = TRUE,
           ntreeTry = 300,
           trace = TRUE,
           improve=1)
```

```
## mtry = 3  OOB error = 48.86%
## Searching left ...
## mtry = 5    OOB error = 50.57%
## -0.03508772 1
## Searching right ...
## mtry = 2    OOB error = 52.57%
## -0.07602339 1
```



```
x1 = randomForest(quality~., ntree=300,mtry=3,nodesize=2, data=wine)
print(x1)
```

```
##
## Call:
## randomForest(formula = quality ~ ., data = wine, ntree = 300,      mtry = 3, nodesize = 2)
##           Type of random forest: classification
##           Number of trees: 300
## No. of variables tried at each split: 3
##
##           OOB estimate of  error rate: 50%
## Confusion matrix:
##   4  5  6  7  8 class.error
## 4 0  7  7  0  0  1.0000000
## 5 0 51 51  6  0  0.5277778
## 6 0 42 91 19  0  0.4013158
## 7 0  4 30 33  0  0.5074627
## 8 0  4  4  1  0  1.0000000
```

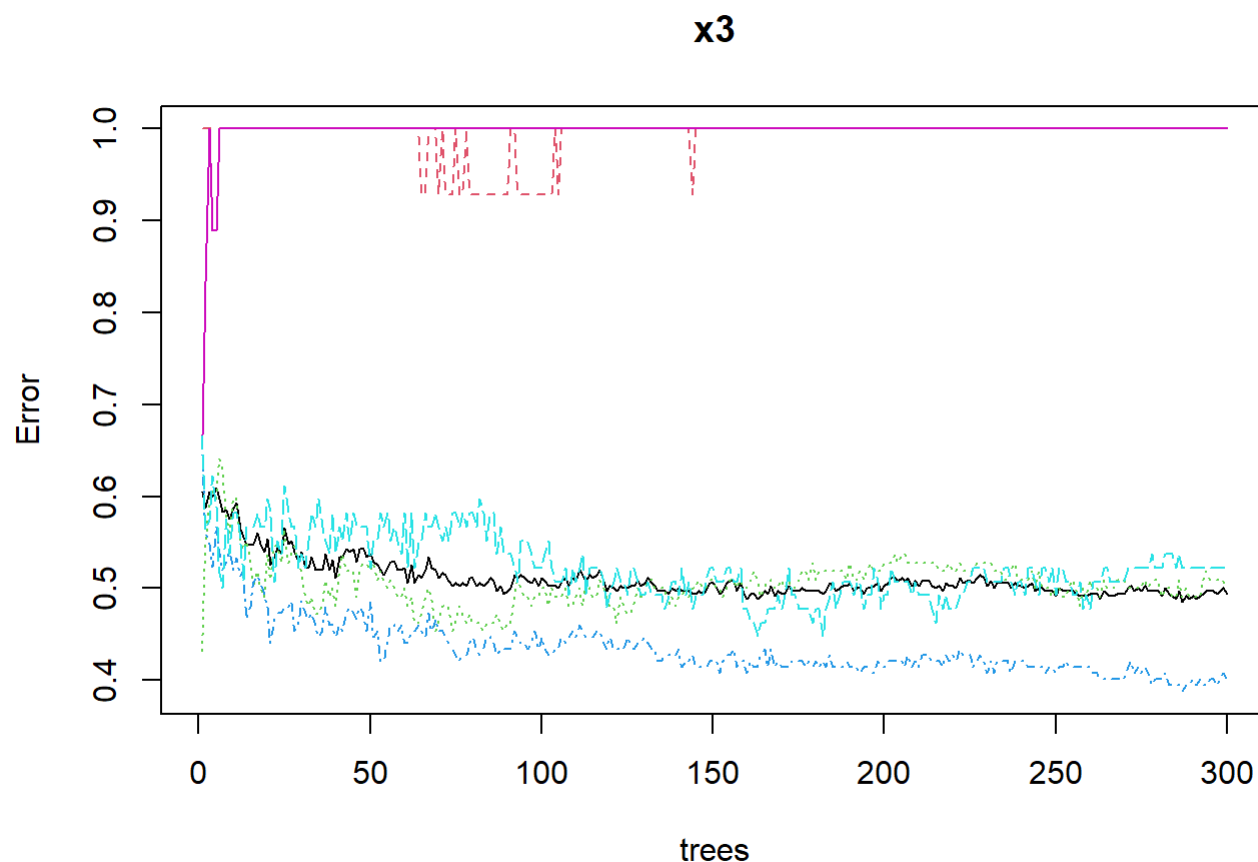
```
x2 = randomForest(quality~., ntree=300,mtry=2,nodesize=2,data=wine)
print(x2)
```

```
##
## Call:
##  randomForest(formula = quality ~ ., data = wine, ntree = 300,      mtry = 2, nodesize = 2)
##              Type of random forest: classification
##              Number of trees: 300
## No. of variables tried at each split: 2
##
##      OOB estimate of  error rate: 50.29%
## Confusion matrix:
##   4  5  6  7  8 class.error
## 4 0  7  6  1 0    1.0000000
## 5 0 49 55  4 0    0.5462963
## 6 0 38 93 21 0    0.3881579
## 7 0  4 31 32 0    0.5223881
## 8 0  4  3  2 0    1.0000000
```

```
x3 = randomForest(quality~., ntree=300,mtry=5,data=wine)
print(x3)
```

```
##
## Call:
##  randomForest(formula = quality ~ ., data = wine, ntree = 300,      mtry = 5)
##              Type of random forest: classification
##              Number of trees: 300
## No. of variables tried at each split: 5
##
##      OOB estimate of  error rate: 49.43%
## Confusion matrix:
##   4  5  6  7  8 class.error
## 4 0  7  7  0 0    1.0000000
## 5 0 54 48  6 0    0.5000000
## 6 0 36 91 25 0    0.4013158
## 7 0  3 32 32 0    0.5223881
## 8 0  4  4  1 0    1.0000000
```

```
plot(x3)
```

#x3 had the lowest oob error rate

#I chose x3

#d

```
a = data.frame(fixed_acidity = 8, volatile_acidity = 0.6, citric_acid = 0.6, residual_sugar = 1
5, chlorides = 0.3,
               free_sulfur_dioxide = 50, total_sulfur_dioxide = 200, density = 1, pH = 3.5, sul
phates = 0.75,
               alcohol = 10, type=factor("1",c(0:1)))
```

```
b=predict(x3, a, interval="predict", level=0.95)
```

b

1

5

Levels: 4 5 6 7 8

#E

printcp(pfit)

```
##
## Classification tree:
## rpart(formula = quality ~ ., data = wine, method = "class")
##
## Variables actually used in tree construction:
## [1] alcohol          chlorides          density
## [4] sulphates        total_sulfur_dioxide volatile_acidity
##
## Root node error: 198/350 = 0.56571
##
## n= 350
##
##          CP nsplit rel error  xerror    xstd
## 1 0.043771     0   1.00000 1.00000 0.046833
## 2 0.037037     3   0.86869 1.06566 0.046233
## 3 0.030303     6   0.75758 0.97475 0.046993
## 4 0.025253     7   0.72727 0.94949 0.047113
## 5 0.022727     8   0.70202 0.92929 0.047181
## 6 0.017677    10   0.65657 0.90909 0.047224
## 7 0.015152    12   0.62121 0.90404 0.047231
```

print(x3)

```
##
## Call:
## randomForest(formula = quality ~ ., data = wine, ntree = 300,      mtry = 5)
##          Type of random forest: classification
##          Number of trees: 300
## No. of variables tried at each split: 5
##
##          OOB estimate of  error rate: 49.43%
## Confusion matrix:
##   4  5  6  7  8 class.error
## 4 0  7  7  0  0   1.0000000
## 5 0 54 48  6  0   0.5000000
## 6 0 36 91 25  0   0.4013158
## 7 0  3 32 32  0   0.5223881
## 8 0  4  4  1  0   1.0000000
```

*#The classification tree predicted 6 while the random forest predicted 5. The error rate for
the CT was about 56.5% while, RF was about 48%*

```
bikes = read.csv("C:/Users/dgmur/Downloads/SeoulBikes_Fl2022.csv")

dim(bikes)
```

```
## [1] 730 16
```

```
bikes = bikes[4:16]

x.new <- data.frame(Holiday=factor("Holiday", c("Holiday", "No Holiday")),
                    Seasons=factor("Spring", c("Autumn", "Spring", "Summer", "Winter")),
                    Functioning.Day=factor("Yes",c("No","Yes")),
                    Temperature=12.1, Humidity=29, WindSpeed=2.3,
                    Visibility=1734, DewPointTemp=-5.4, SolarRadiation=2.26,
                    Rainfall=0, Snowfall=0,
                    DayF=factor("21",c(1:31)), MonthF=factor("3",c(1:12)),
                    Time=factor("Morning",c("Morning","Evening"))))

str(bikes)
```

```
## 'data.frame': 730 obs. of 13 variables:
## $ RentedBikeCount: int 930 219 85 937 812 802 754 780 222 35 ...
## $ Temperature : num -7.6 -4.2 3.2 -0.8 -8.1 -4.2 -0.2 -7 -5.4 -0.2 ...
## $ Humidity : int 37 79 92 69 36 78 96 59 56 90 ...
## $ WindSpeed : num 1.1 2.1 1.8 3.5 1.6 0 0.8 0.8 0.7 1.2 ...
## $ Visibility : int 2000 1436 244 954 2000 1518 173 1559 1940 378 ...
## $ DewPointTemp : num -19.8 -7.3 2 -5.7 -20.5 -7.4 -0.7 -13.6 -12.8 -1.6 ...
## $ SolarRadiation : num 0.01 0.01 0 0.01 0.01 0.01 0.01 0.01 0.01 0 ...
## $ Rainfall : num 0 0 0 0 0 0 0 0 0 ...
## $ Snowfall : num 0 0 0 0 0 0.2 0.8 0 0 4.1 ...
## $ Seasons : chr "Winter" "Winter" "Winter" "Winter" ...
## $ Holiday : chr "No Holiday" "No Holiday" "No Holiday" "No Holiday" ...
## $ Functioning.Day: chr "Yes" "Yes" "Yes" "Yes" ...
## $ Time : chr "Morning" "Morning" "Morning" "Morning" ...
```

```

bikes$Seasons= as.factor(bikes$Seasons)

bikes$Holiday= as.factor(bikes$Holiday)

bikes$Functioning.Day=as.factor(bikes$Functioning.Day)

bikes$Time=as.factor(bikes$Time)

#a

fit2= rpart(RentedBikeCount~., method = "anova", data=bikes)

printcp(fit2)

```

```

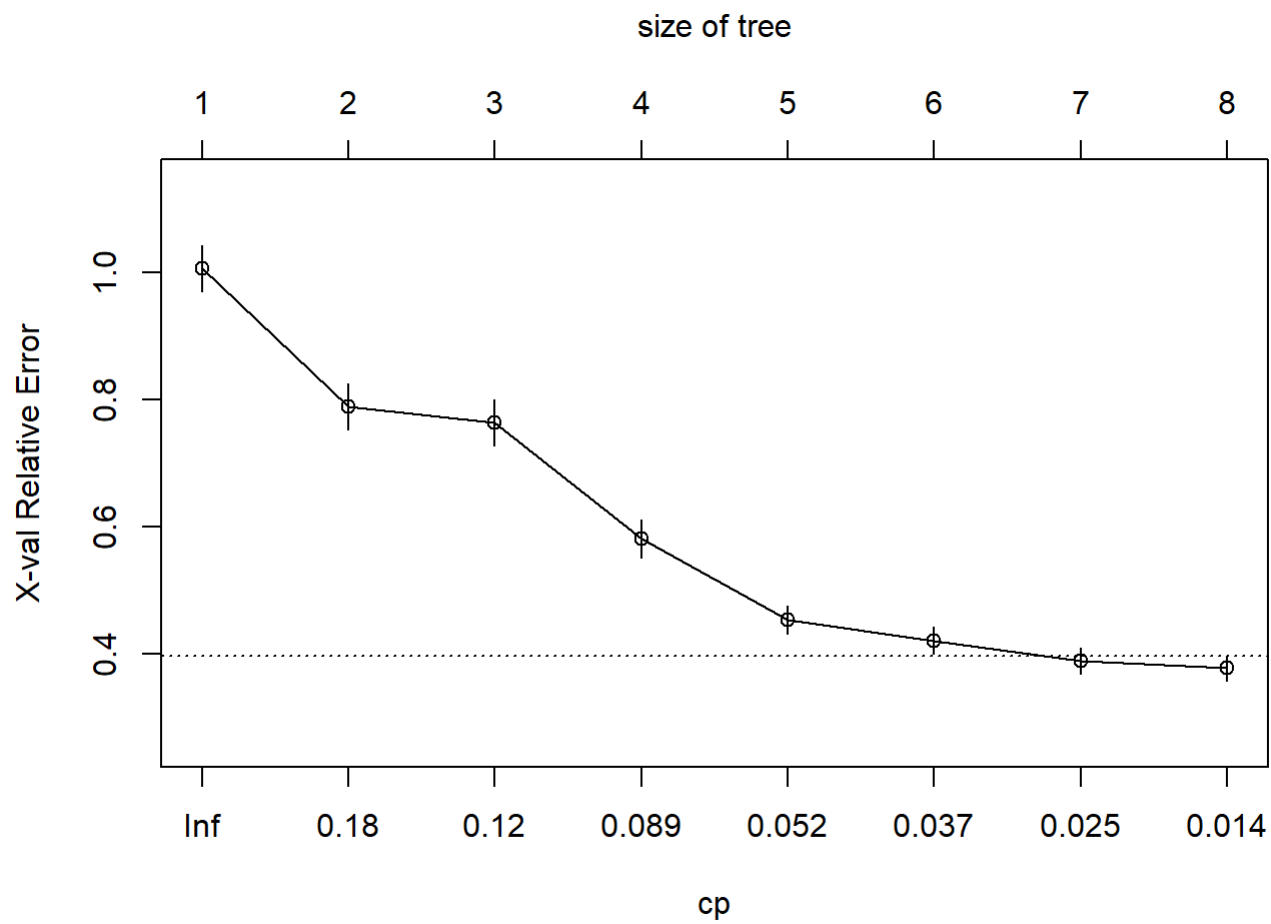
##
## Regression tree:
## rpart(formula = RentedBikeCount ~ ., data = bikes, method = "anova")
##
## Variables actually used in tree construction:
## [1] Functioning.Day Humidity      Seasons      SolarRadiation
## [5] Temperature      Time
##
## Root node error: 640097003/730 = 876845
##
## n= 730
##
##      CP nsplit rel error  xerror    xstd
## 1 0.269598     0   1.00000 1.00621 0.036343
## 2 0.116823     1   0.73040 0.78870 0.036018
## 3 0.114820     2   0.61358 0.76427 0.036189
## 4 0.068987     3   0.49876 0.58097 0.030018
## 5 0.039809     4   0.42977 0.45350 0.022044
## 6 0.034160     5   0.38996 0.42065 0.020941
## 7 0.018238     6   0.35580 0.38881 0.020175
## 8 0.010000     7   0.33756 0.37699 0.019396

```

```

plotcp(fit2)

```



```
summary(fit2)
```

```
## Call:
## rpart(formula = RentedBikeCount ~ ., data = bikes, method = "anova")
##   n= 730
##
##           CP nsplit rel error   xerror   xstd
## 1 0.26959787    0 1.0000000 1.0062097 0.03634349
## 2 0.11682322    1 0.7304021 0.7887003 0.03601774
## 3 0.11481981    2 0.6135789 0.7642709 0.03618950
## 4 0.06898737    3 0.4987591 0.5809696 0.03001769
## 5 0.03980925    4 0.4297717 0.4534989 0.02204431
## 6 0.03415988    5 0.3899625 0.4206475 0.02094054
## 7 0.01823837    6 0.3558026 0.3888055 0.02017526
## 8 0.01000000    7 0.3375642 0.3769880 0.01939600
##
## Variable importance
##      Temperature      Seasons  DewPointTemp  SolarRadiation      Humidity
##              22             16             15             12             11
## Functioning.Day      Time      Rainfall      Snowfall      WindSpeed
##              8             5             4             3             3
##      Visibility
##              2
##
## Node number 1: 730 observations,    complexity param=0.2695979
##   mean=1259.314, MSE=876845.2
##   left son=2 (266 obs) right son=3 (464 obs)
##   Primary splits:
##     Temperature < 7.85  to the left, improve=0.26959790, (0 missing)
##     Seasons      splits as RRRL,      improve=0.25654450, (0 missing)
##     SolarRadiation < 0.205 to the left, improve=0.25609000, (0 missing)
##     DewPointTemp < -2.25 to the left, improve=0.17980060, (0 missing)
##     Humidity     < 84.5  to the right, improve=0.09142905, (0 missing)
##   Surrogate splits:
##     DewPointTemp < 2.75  to the left, agree=0.892, adj=0.703, (0 split)
##     Seasons      splits as RRRL,      agree=0.882, adj=0.677, (0 split)
##     SolarRadiation < 0.075 to the left, agree=0.818, adj=0.500, (0 split)
##     Snowfall     < 0.1   to the right, agree=0.688, adj=0.143, (0 split)
##     Humidity     < 34.5  to the left, agree=0.653, adj=0.049, (0 split)
##
## Node number 2: 266 observations,    complexity param=0.03980925
##   mean=617.1617, MSE=239589
##   left son=4 (221 obs) right son=5 (45 obs)
##   Primary splits:
##     Seasons      splits as RL-L,      improve=0.39983550, (0 missing)
##     Temperature < 3.25  to the left, improve=0.14399360, (0 missing)
##     DewPointTemp < -10.25 to the left, improve=0.07283931, (0 missing)
##     SolarRadiation < 0.115 to the left, improve=0.07059487, (0 missing)
##     Holiday      splits as LR,      improve=0.05290116, (0 missing)
##   Surrogate splits:
##     Temperature < 6.75  to the left, agree=0.842, adj=0.067, (0 split)
##     Functioning.Day splits as RL,      agree=0.835, adj=0.022, (0 split)
##
## Node number 3: 464 observations,    complexity param=0.1168232
```

```

## mean=1627.444, MSE=870253.3
## left son=6 (57 obs) right son=7 (407 obs)
## Primary splits:
## Humidity < 84.5 to the right, improve=0.1851873, (0 missing)
## Rainfall < 0.15 to the right, improve=0.1700205, (0 missing)
## Functioning.Day splits as LR, improve=0.1587288, (0 missing)
## Time splits as RL, improve=0.1203948, (0 missing)
## SolarRadiation < 0.155 to the left, improve=0.1168390, (0 missing)
## Surrogate splits:
## Rainfall < 0.4 to the right, agree=0.933, adj=0.456, (0 split)
## Visibility < 283.5 to the left, agree=0.897, adj=0.158, (0 split)
## DewPointTemp < 25.45 to the right, agree=0.884, adj=0.053, (0 split)
## WindSpeed < 0.1 to the left, agree=0.881, adj=0.035, (0 split)
##
## Node number 4: 221 observations
## mean=477.4977, MSE=91569.86
##
## Node number 5: 45 observations
## mean=1303.067, MSE=400265.6
##
## Node number 6: 57 observations, complexity param=0.01823837
## mean=554.7193, MSE=395396.7
## left son=12 (43 obs) right son=13 (14 obs)
## Primary splits:
## SolarRadiation < 0.185 to the left, improve=0.51799300, (0 missing)
## Rainfall < 0.75 to the right, improve=0.30090910, (0 missing)
## Humidity < 90.5 to the right, improve=0.14186150, (0 missing)
## Temperature < 19.3 to the left, improve=0.07291409, (0 missing)
## DewPointTemp < 18.55 to the left, improve=0.07291409, (0 missing)
## Surrogate splits:
## Temperature < 24.5 to the left, agree=0.825, adj=0.286, (0 split)
## DewPointTemp < 23.8 to the left, agree=0.807, adj=0.214, (0 split)
##
## Node number 7: 407 observations, complexity param=0.1148198
## mean=1777.678, MSE=753026.4
## left son=14 (22 obs) right son=15 (385 obs)
## Primary splits:
## Functioning.Day splits as LR, improve=0.23980490, (0 missing)
## Time splits as RL, improve=0.14092870, (0 missing)
## Temperature < 20.45 to the left, improve=0.11122760, (0 missing)
## WindSpeed < 2.25 to the left, improve=0.07788779, (0 missing)
## Humidity < 57.5 to the right, improve=0.06370545, (0 missing)
##
## Node number 12: 43 observations
## mean=296.4884, MSE=99407.6
##
## Node number 13: 14 observations
## mean=1347.857, MSE=470625.8
##
## Node number 14: 22 observations
## mean=0, MSE=0
##

```

```

## Node number 15: 385 observations,    complexity param=0.06898737
##   mean=1879.26, MSE=605158.2
##   left son=30 (164 obs) right son=31 (221 obs)
##   Primary splits:
##       Time           splits as  RL,           improve=0.18953340, (0 missing)
##       WindSpeed    < 1.35   to the left,  improve=0.10008420, (0 missing)
##       Temperature  < 17.05  to the left,  improve=0.08926161, (0 missing)
##       Humidity     < 57.5   to the right, improve=0.06494347, (0 missing)
##       Rainfall     < 0.05   to the right, improve=0.04771395, (0 missing)
##   Surrogate splits:
##       WindSpeed    < 1.25   to the left,  agree=0.769, adj=0.457, (0 split)
##       Humidity     < 57.5   to the right, agree=0.735, adj=0.378, (0 split)
##       Visibility   < 1064   to the left,  agree=0.631, adj=0.134, (0 split)
##       Temperature  < 13.85  to the left,  agree=0.590, adj=0.037, (0 split)
##       Rainfall     < 0.05   to the right, agree=0.577, adj=0.006, (0 split)
##
## Node number 30: 164 observations
##   mean=1486.116, MSE=463720.8
##
## Node number 31: 221 observations,    complexity param=0.03415988
##   mean=2171.005, MSE=510303.6
##   left son=62 (55 obs) right son=63 (166 obs)
##   Primary splits:
##       Temperature  < 16.55  to the left,  improve=0.19388370, (0 missing)
##       DewPointTemp < 0.4     to the left,  improve=0.09247077, (0 missing)
##       SolarRadiation < 0.685 to the left,  improve=0.08959975, (0 missing)
##       Humidity     < 72.5    to the right, improve=0.08502754, (0 missing)
##       Visibility   < 642.5   to the left,  improve=0.04144956, (0 missing)
##   Surrogate splits:
##       SolarRadiation < 0.075 to the left,  agree=0.860, adj=0.436, (0 split)
##       DewPointTemp  < 2.85   to the left,  agree=0.855, adj=0.418, (0 split)
##       Visibility    < 505.5   to the left,  agree=0.769, adj=0.073, (0 split)
##
## Node number 62: 55 observations
##   mean=1624.545, MSE=444930.4
##
## Node number 63: 166 observations
##   mean=2352.06, MSE=400242.7

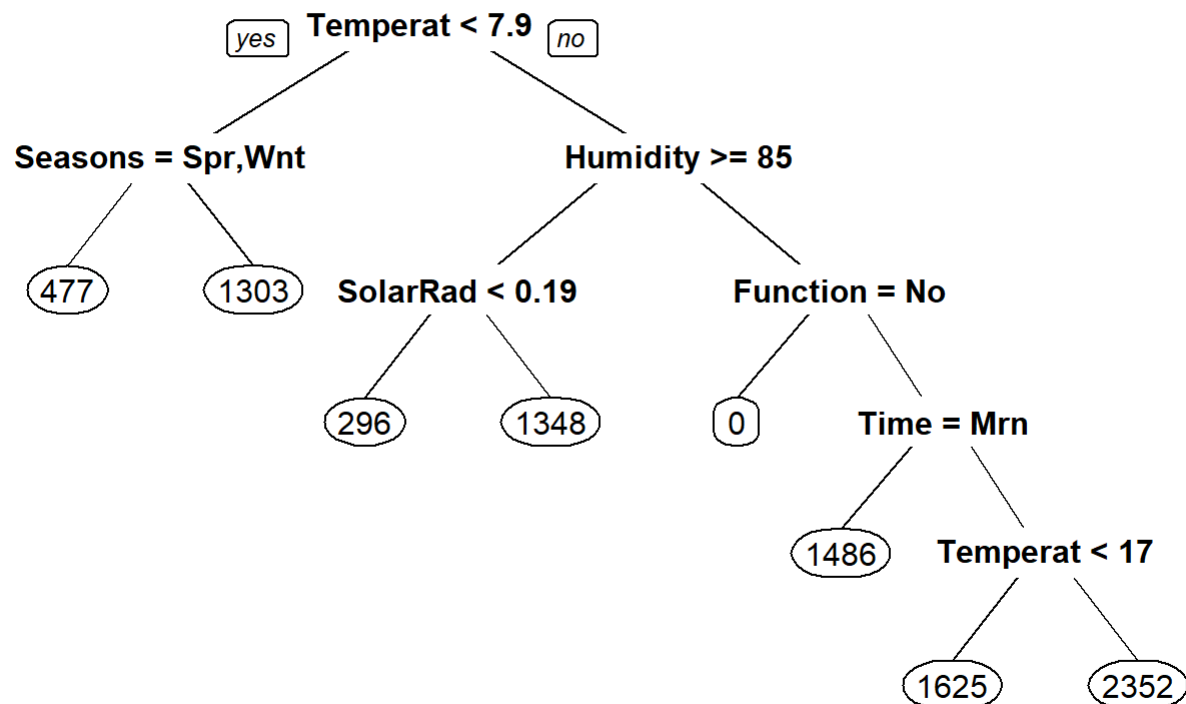
```

```
pfit2<- prune(fit2, cp=fit$cptable[which.min(fit$cptable[, "xerror"]), "CP"])
```

```
#b
```

```
prp(pfit2, main = "Pruned Classification Tree for Rented Bike Count")
```

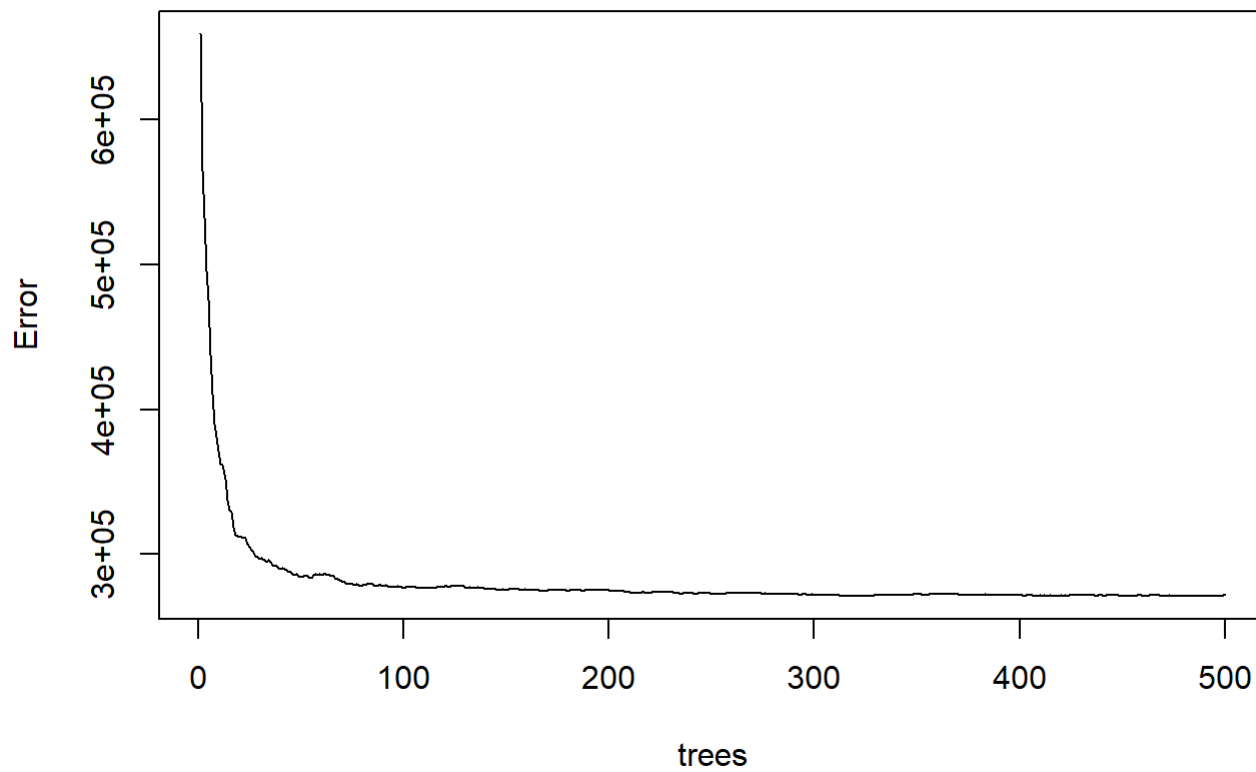

Pruned Classification Tree for Rented Bike Count



```
#tree predicted 1486
```

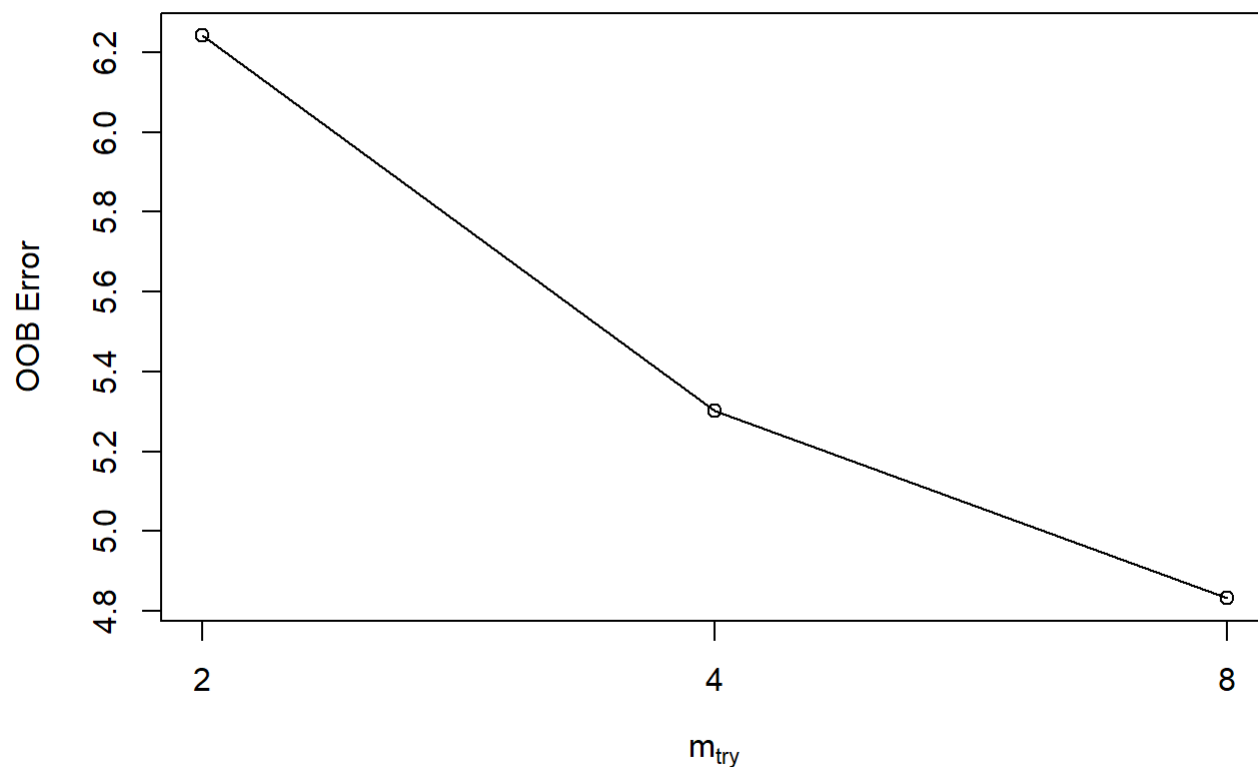
```
#c
set.seed(55)
```

```
z = randomForest(RentedBikeCount~., data=bikes)
plot(z)
```

z

```
t2 = tuneRF(wine[,-4],wine[,4],
            stepFactor = .5, plot = TRUE,
            ntreeTry = 400,
            trace = TRUE,
            improve=1)
```

```
## mtry = 4  OOB error = 5.302175
## Searching left ...
## mtry = 8    OOB error = 4.832046
## 0.0886673 1
## Searching right ...
## mtry = 2    OOB error = 6.241474
## -0.1771534 1
```



```
z1 = randomForest(RentedBikeCount~., ntree=400,mtry=8,nodesize=2, data=bikes)
print(z1)
```

```
##
## Call:
## randomForest(formula = RentedBikeCount ~ ., data = bikes, ntree = 400,      mtry = 8, nodesi
ze = 2)
##              Type of random forest: regression
##              Number of trees: 400
## No. of variables tried at each split: 8
##
##              Mean of squared residuals: 278243
##              % Var explained: 68.27
```

```
z2 = randomForest(RentedBikeCount~., ntree=400,mtry=8,nodesize=1, data=bikes)
print(z2)
```

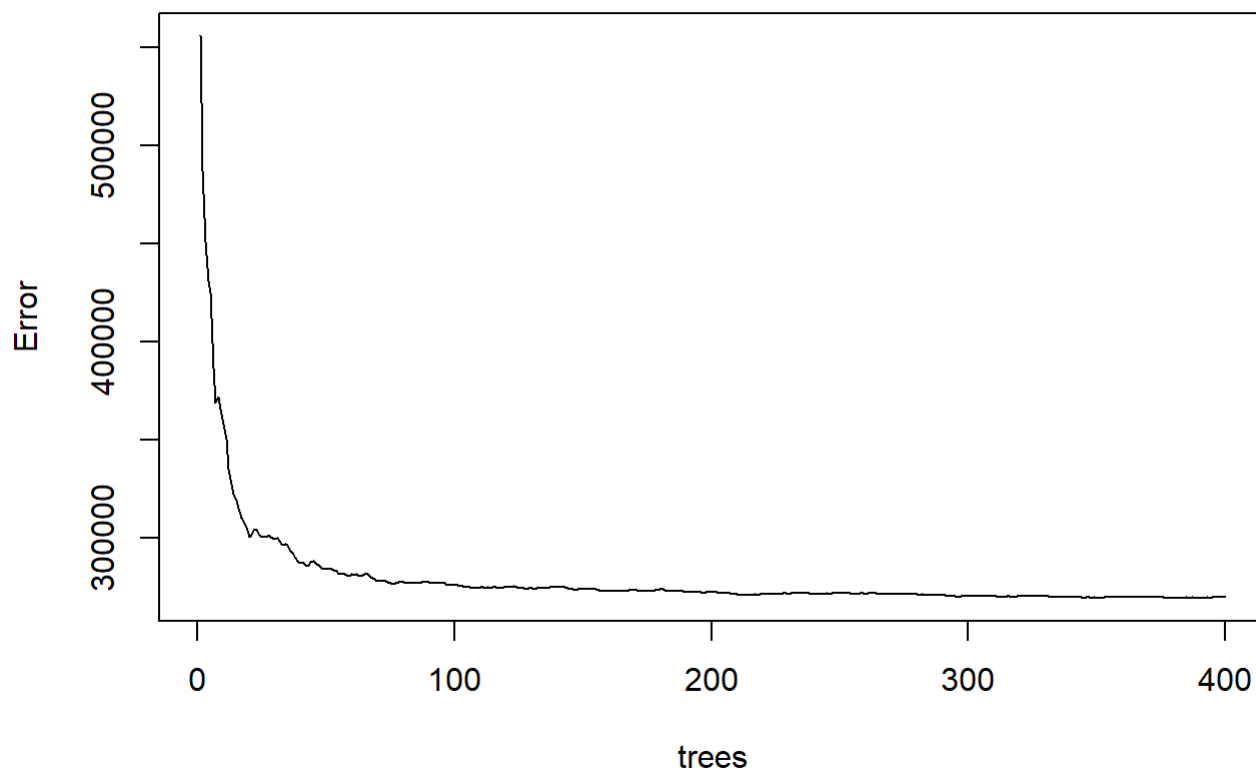
```
##
## Call:
##  randomForest(formula = RentedBikeCount ~ ., data = bikes, ntree = 400,      mtry = 8, nodesi
ze = 1)
##              Type of random forest: regression
##              Number of trees: 400
## No. of variables tried at each split: 8
##
##              Mean of squared residuals: 270018.4
##              % Var explained: 69.21
```

```
z3 = randomForest(RentedBikeCount~., ntree=400,mtry=8,nodesize=5, data=bikes)
print(z3)
```

```
##
## Call:
##  randomForest(formula = RentedBikeCount ~ ., data = bikes, ntree = 400,      mtry = 8, nodesi
ze = 5)
##              Type of random forest: regression
##              Number of trees: 400
## No. of variables tried at each split: 8
##
##              Mean of squared residuals: 271002.1
##              % Var explained: 69.09
```

```
#I selected z2
```

```
plot(z2)
```

z2

```
#d
zp=predict(z1, x.new, interval="predict", level=0.95)
zp
```

```
##          1
## 1192.925
```

```
#RF predicted 1147.564
```

```
#E
printcp(pfit2)
```

```
##
## Regression tree:
## rpart(formula = RentedBikeCount ~ ., data = bikes, method = "anova")
##
## Variables actually used in tree construction:
## [1] Functioning.Day Humidity Seasons SolarRadiation
## [5] Temperature Time
##
## Root node error: 640097003/730 = 876845
##
## n= 730
##
##      CP nsplit rel error  xerror  xstd
## 1 0.269598    0  1.00000 1.00621 0.036343
## 2 0.116823    1  0.73040 0.78870 0.036018
## 3 0.114820    2  0.61358 0.76427 0.036189
## 4 0.068987    3  0.49876 0.58097 0.030018
## 5 0.039809    4  0.42977 0.45350 0.022044
## 6 0.034160    5  0.38996 0.42065 0.020941
## 7 0.018238    6  0.35580 0.38881 0.020175
## 8 0.010000    7  0.33756 0.37699 0.019396
```

```
print(z2)
```

```
##
## Call:
## randomForest(formula = RentedBikeCount ~ ., data = bikes, ntree = 400, mtry = 8, nodesize = 1)
##
##      Type of random forest: regression
##      Number of trees: 400
## No. of variables tried at each split: 8
##
##      Mean of squared residuals: 270018.4
##      % Var explained: 69.21
```

#The regression tree had a rootnode error rate of 100% and rf had a 69% variance explained
zp

```
##      1
## 1192.925
```

regression tree predicted 1486 while rf was 1147.564