

MSiA 400 Lab 2 Harish Chockalingam

#Problem 1

```
redwine<-read.table('redwine.txt',header=T)
mean_RS<-mean(redwine$RS,na.rm=T)
mean_SD<-mean(redwine$SD,na.rm=T)
mean_RS
```

```
## [1] 2.537952
```

```
mean_SD
```

```
## [1] 46.29836
```

#The RS and SD average after removing NAs is 2.53 and 46.29 respectively

#Problem 2

```
M<-cbind(redwine$FS,redwine$SD)
M<-na.omit(M)
FS.obs<-M[,1]
SD.obs<-M[,2]
ABC<-lm(SD.obs~FS.obs)
coef<-coefficients(ABC)
coef
```

```
## (Intercept)      FS.obs
##   13.185505    2.086077
```

*#There are 17 missing SD values. After removing the missing SD and respective
#FD values, fitting yields a
#intercept of 13.18 and coefficient of 2.08*

#Problem 3

```
SD<-redwine$SD
missingSD <- is.na(SD)
FS_17<-redwine$FS[missingSD]

SD_predict<-coef[1]+coef[2]*FS_17
redwine$SD[missingSD]<-SD_predict
mean(redwine$SD)
```

```
## [1] 46.30182
```

#The mean for SD after imputation is 46.30182, not a huge change

#Problem 4

```
avg.imp <- function (a, avg){
  missing <- is.na(a)
  imputed <- a
  imputed[missing] <- avg
  return (imputed)
}

RS_ave<-mean(na.omit(redwine$RS))
RS_imp<-avg.imp(redwine$RS,RS_ave)
redwine$RS<-RS_imp
mean(RS_imp)
```

```
## [1] 2.537952
```

```
#The average value for RS is 2.537952
```

```
#Problem 5
```

```
winemodel<-lm(redwine$QA~redwine$FA+redwine$VA+redwine$CA+redwine$RS+redwine$CH+  
              redwine$FS+redwine$SD+redwine$DE+redwine$PH+redwine$SU+redwine$AL)  
coefficients(winemodel)
```

```
##      (Intercept)      redwine$FA      redwine$VA      redwine$CA      redwine$RS  
## 47.202815335      0.068406796     -1.097686420     -0.178949797     0.025926958  
##      redwine$CH      redwine$FS      redwine$SD      redwine$DE      redwine$PH  
## -1.631290466      0.003530106     -0.002854970    -44.816652166     0.035996993  
##      redwine$SU      redwine$AL  
##      0.944871182      0.247046550
```

```
#The coefficients:
```

```
#Intercep: 47.202 FA:0.0684 VA:-1.097 CA:-0.179 RS:0.026 CH:-1.631
```

```
#FS:0.0035 SD:-0.0028 DE:-44.817 PH:0.036 SU: 0.944 AL:0.247
```

```
#Problem 6
```

```
summary(winemodel)
```

```
##
```

```
## Call:
```

```
## lm(formula = redwine$QA ~ redwine$FA + redwine$VA + redwine$CA +  
##      redwine$RS + redwine$CH + redwine$FS + redwine$SD + redwine$DE +  
##      redwine$PH + redwine$SU + redwine$AL)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max  
## -2.78010 -0.36249 -0.06331  0.44595  1.98828
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  4.720e+01  1.782e+01   2.649 0.008151 **  
## redwine$FA    6.841e-02  1.872e-02   3.654 0.000267 ***  
## redwine$VA   -1.098e+00  1.213e-01  -9.053 < 2e-16 ***  
## redwine$CA   -1.789e-01  1.474e-01  -1.214 0.224954 .  
## redwine$RS    2.593e-02  1.419e-02   1.827 0.067944 .  
## redwine$CH   -1.631e+00  4.097e-01  -3.982 7.14e-05 ***  
## redwine$FS    3.530e-03  2.159e-03   1.635 0.102262  
## redwine$SD   -2.855e-03  7.248e-04  -3.939 8.54e-05 ***  
## redwine$DE   -4.482e+01  1.789e+01  -2.505 0.012329 *  
## redwine$PH    3.600e-02  4.409e-02   0.816 0.414413  
## redwine$SU    9.449e-01  1.136e-01   8.321 < 2e-16 ***  
## redwine$AL    2.470e-01  2.265e-02  10.906 < 2e-16 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 0.6491 on 1587 degrees of freedom
```

```
## Multiple R-squared:  0.3584, Adjusted R-squared:  0.354
```

```
## F-statistic: 80.6 on 11 and 1587 DF, p-value: < 2.2e-16
```

```
#From the summary R2 is 0.3584, and based on a significance level of 0.05
```

```
#the PH attribute is least likely related to QA
```

```
#as it has a high p-value of 0.4144
```

```

#Problem 7
CVInd <- function(n,K){  #n is sample size; K is number of parts;
  #returns K-length list of indices for each part
  m<-floor(n/K)  #approximate size of each part
  r<-n-m*K
  I<-sample(n,n)  #random reordering of the indices
  Ind<-list()  #will be list of indices for all K parts
  length(Ind)<-K
  for (k in 1:K) {
    if (k <= r) kpart <- ((m+1)*(k-1)+1):((m+1)*k)
    else kpart<-((m+1)*r+m*(k-r-1)+1):((m+1)*r+m*(k-r))
    Ind[[k]] <- I[kpart]  #indices for kth part of data
  }
  Ind }

Nrep<-20 #number of replicates of CV
K<-5  #K-fold CV on each replicate
n=nrow(redwine)
y<-redwine$QA
SSE<-matrix(0,Nrep,1)
for (j in 1:Nrep) {
  Ind<-CVInd(n,K)
  yhat11<-y
  for (k in 1:K) {
    out11<-lm(QA~.,redwine[-Ind[[k]],])
    yhat11[Ind[[k]]]<-as.numeric(predict(out11,redwine[Ind[[k]],]))
  } #end of k loop
  SSE[j]=c(sum((y-yhat11)^2))
} #end of j loop
SSE

```

```

##           [,1]
## [1,] 678.2421
## [2,] 681.0395
## [3,] 686.5173
## [4,] 681.3533
## [5,] 684.2491
## [6,] 683.8529
## [7,] 689.6737
## [8,] 682.2813
## [9,] 685.9118
## [10,] 688.6713
## [11,] 688.6152
## [12,] 685.0245
## [13,] 683.4021
## [14,] 679.4924
## [15,] 687.2252
## [16,] 684.5857
## [17,] 688.9007
## [18,] 684.3679
## [19,] 687.0462
## [20,] 688.5915

```

```
apply(SSE,2,mean)
```

```
## [1] 684.9522
```

```
#The average error rate after 20 replications is 683.4685
```

```
#Problem 8
```

```
PH_omit<-na.omit(redwine$PH)
```

```
PH_mean<-mean(PH_omit)
```

```
PH_std<-sd(na.omit(redwine$PH))
```

```
PH_lb<-PH_mean-3*PH_std
```

```
PH_ub<-PH_mean+3*PH_std
```

```
redwine2<-subset(redwine,redwine$PH<PH_ub & redwine$PH>PH_lb)
```

```
dim(redwine2)
```

```
## [1] 1580 12
```

```
#dimensions of redwine2 is 1580 x 12. The imputed redwine dataset had 1599
```

```
#values thus there were 19 outliers
```

```
#Problem 9
```

```
winemodel2<-lm(redwine2$QA~redwine2$FA+redwine2$VA+redwine2$CA+redwine2$RS+  
                redwine2$CH+redwine2$FS+redwine2$SD+redwine2$DE+redwine2$PH+redwine2$SU+redwine2$AL)  
summary(winemodel2)
```

```
##
```

```
## Call:
```

```
## lm(formula = redwine2$QA ~ redwine2$FA + redwine2$VA + redwine2$CA +  
##     redwine2$RS + redwine2$CH + redwine2$FS + redwine2$SD + redwine2$DE +  
##     redwine2$PH + redwine2$SU + redwine2$AL)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max  
## -2.68933 -0.36336 -0.04368  0.45221  2.01272
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  19.036170   21.211609   0.897   0.3696  
## redwine2$FA    0.024613    0.026019   0.946   0.3443  
## redwine2$VA   -1.072147    0.122031  -8.786 < 2e-16 ***  
## redwine2$CA   -0.178017    0.148120  -1.202   0.2296  
## redwine2$RS    0.012955    0.014968   0.866   0.3869  
## redwine2$CH   -1.902552    0.420766  -4.522 6.60e-06 ***  
## redwine2$FS    0.004421    0.002182   2.026   0.0429 *  
## redwine2$SD   -0.003145    0.000738  -4.261 2.16e-05 ***  
## redwine2$DE  -14.973653   21.652465  -0.692   0.4893  
## redwine2$PH   -0.424704    0.192653  -2.205   0.0276 *  
## redwine2$SU    0.913456    0.114860   7.953 3.46e-15 ***  
## redwine2$AL    0.282744    0.026553  10.648 < 2e-16 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 0.6475 on 1568 degrees of freedom
```

```
## Multiple R-squared:  0.3629, Adjusted R-squared:  0.3585
```

```
## F-statistic: 81.21 on 11 and 1568 DF, p-value: < 2.2e-16
```

*#Compared to problem 6 the R^2 went up from 0.3584 to 0.3629 (not significant increase).
#After removing the outliers the signifance of coefficients has changed, but
#both models still have 4 non-significant coefficients. Thus, winemodel2 is a slightly
#better model to predict QA.
#The five attributes likely related to QA are VA, CH, SD, SU, AL they have
#p-values close to zero (signifcance level used 0.05)*