## **Application**

1. First, in order to use this set for modeling, we need to deal with its missing data. Since this is not a class on imputation, we deal with them in the easiest but most biased way: case deletion. When you have missing data in real life, *learn how to deal with them better than this!!!* Run the code below to clean the data and report the results

```
AQ = na.omit (airquality[,1:4])
dim(AQ)
```

```
#1
AQ = na.omit(airquality[, 1:4])
dim(AQ)
# 111 4 which means that there are 111 rows and 4 columns
```

```
#2
set.seed(4099183)
n = nrow(AQ)
reorder = sample.int(n)
size_train = floor(n*0.75)
ind_train = reorder[1:size_train]
ind_valid = reorder[(size_train+1):n]
data_train = AQ[ind_train, ]
data_valid = AQ[ind_valid, ]
#print(data_valid)
# Ozone Solar.R Wind Temp
# 94
         9
                24 13.8
                           81
                167 6.9
255 4.0
# 124
        96
                           91
# 99
        122
                           89
# 140
        18
                224 13.8
                           67
# 24
         32
                92 12.0
                           61
# 17
         34
                307 12.0
                           66
                223 5.7
# 30
       115
                           79
                    9.7
# 20
                44
        11
                           62
        65
                    9.7
# 106
                157
                           80
# 63
                    9.2
        49
                248
                           85
# 31
         37
                279 7.4
                           76
# 82
        16
                  7
                    6.9
                           74
        36
                139 10.3
# 146
                           81
# 4
        18
                313 11.5
                           62
        45
# 116
                212 9.7
                           79
                274 10.9
# 14
        14
                           68
        23
# 110
                115 7.4
                           76
# 29
        45
                252 14.9
                           81
         23
# 44
                148 8.0
                           82
# 120
        76
                203 9.7
                           97
# 8
        19
                99 13.8
                           59
# 66
        64
                175 4.6
                           83
# 109
        59
                51 6.3
                           79
# 113
        21
                259 15.5
                           77
# 22
        11
                320 16.6
                           73
                127 9.7
# 38
         29
                           82
# 92
         59
                254 9.2
                           81
# 149
        30
               193 6.9
```

```
#3
fit.solar = lm(Ozone ~ Solar.R, data = data_train)
fit.wind = lm(Ozone ~ Wind, data = data_train)
fit.temp = lm(ozone ~ Temp, data = data_train)
fit.all = lm(Ozone ~ Temp + Wind + Solar.R, data = data_train)
Temp*Wind + Temp*Solar.R + Wind*Solar.R, data = data_train)
pred.solar = predict(fit.solar, data_valid)
pred.wind = predict(fit.wind, data_valid)
pred.temp = predict(fit.temp, data_valid)
pred.all = predict(fit.all, data_valid)
pred.int = predict(fit.int, data_valid)
get.MSPE = function(Y, Y.hat){
  residuals = Y-Y.hat
  resid.sq = residuals^2
 SSPE = sum(resid.sq)
 MSPE = SSPE / length(Y)
 return (MSPE)
Y.valid = data_valid$0zone
MSPE.solar = get.MSPE(Y.valid, pred.solar)
MSPE.wind = get.MSPE(Y.valid, pred.wind)
MSPE.temp = get.MSPE(Y.valid, pred.temp)
MSPE.all = get.MSPE(Y.valid, pred.all)
MSPE.int = get.MSPE(Y.valid, pred.int)
print(MSPE.solar)
print (MSPE.wind)
print(MSPE.temp)
print (MSPE. all)
print(MSPE.int)
# > print(MSPE.solar)
# [1] 934.5455
# > print(MSPE.wind)
# [1] 574.8944
# > print(MSPE.temp)
# [1] 586.7952
# > print(MSPE.all)
# [1] 387.3352
# > print(MSPE.int)
# [1] 327.3287
# The model that allows curvature and interactions wins this competition
```