## HW1: Regression Modeling

## Your Name

due on 10/11 (Tue)

Before working on HW1, you better complete the practice of ISLR ch2-3 R Labs. In particular, you may follow similar fitting steps demonstrated in ISLR ch3 R Lab for the following two data analysis.

## Problem1

The first data consist of 4 variables (daily measurements) measured at the same place:

- ozone: response variable
- radiation
- temperature
- wind (speed)

Goal: Predict ozone using regression models.

Your analysis should include the following:

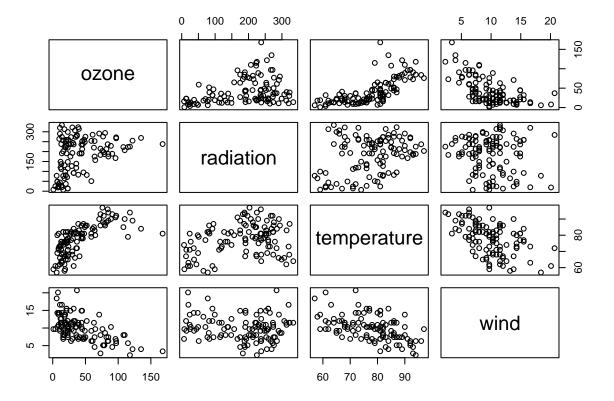
- (a) Exploratory data analysis (EDA) among 4 variables
- (b) Regression model fitting and model summaries.
- (c) Model selection and diagonostics
- (d) Comments on your prediction results and scientific findings. (state at least 3 viewpoints with data evidence)

```
dat1 <- read.csv("ozone.csv")
head(dat1)</pre>
```

##		ozone	radiation	temperature	wind
##	1	41	190	67	7.4
##	2	36	118	72	8.0
##	3	12	149	74	12.6
##	4	18	313	62	11.5
##	5	23	299	65	8.6
##	6	19	99	59	13.8

dim(dat1)

## [1] 111 4



View data and start your analysis here...

## Problem2

The second problem concerns about the Prostate data (more data descriptions can be found in ESL book chapter 3.2.1). There are 8 input variables and 1 response variable:

- Input variables (columns 1–8)
  - lcavol
  - lweight

  - age
  - lbph - svi
  - lcp
  - gleason
  - pgg45
- Response variable: lpsa (column 9)
- Indicator for training set (column 10): there are 97 observations in total, in which 70 obs'n are used as training data set and the rest of 27 observations are used as validation data. Note: our training index is different from those used in the ESL book chapter 3.2.1.

Goal: Predict lpsa using regression models.

Your analysis is based on the training data set only and should include the following:

- (a) EDA
- (b) Determine a good regression model for predicting lpsa
- (c) Describe the important main effects and interaction effects.
- (d) Predict lpsa for the validation data set based on the fitted model, with their prediction intervals. And compared the prediction results to the true observations. Comment on your model performance.

```
dat2 <- read.csv("Prostate.csv")
head(round(dat2,3))

## | lcavol | lweight age | lbph syi | lcp gleason pgg45 | lpsa train.idx</pre>
```

```
##
     lcavol lweight age
                           lbph svi
                                        1cp gleason pgg45
                                                              lpsa train.idx
## 1 -0.580
              2.769
                      50 -1.386
                                   0 -1.386
                                                   6
                                                         0 - 0.431
                                                                            1
                                                                            0
## 2 -0.994
              3.320
                      58 -1.386
                                   0 - 1.386
                                                   6
                                                         0 -0.163
                                                   7
                                                         20 -0.163
## 3 -0.511
              2.691
                      74 -1.386
                                   0 -1.386
                                                                            1
## 4 -1.204
                      58 -1.386
                                                   6
              3.283
                                   0 -1.386
                                                          0 - 0.163
                                                                            1
## 5 0.751
              3.432
                      62 -1.386
                                   0 -1.386
                                                   6
                                                          0
                                                             0.372
                                                                            0
## 6 -1.050
               3.229
                      50 -1.386
                                   0 - 1.386
                                                             0.765
                                                                            0
dim(dat2)
```

## [1] 97 10