— title: "INFX 573 Lab: 8b james.et.al" author: "Pierre Augustamar" date: "November 17th, 2016"

Collaborators:

Instructions:

Before beginning this assignment, please ensure you have access to R and/or RStudio.

- 1. Download the week8a_lab.Rmd file from Canvas. Open week8a_lab.Rmd in RStudio (or your favorite editor) and supply your solutions to the assignment by editing week8a_lab.Rmd.
- 2. Replace the "Insert Your Name Here" text in the author: field with your own full name.
- 3. Be sure to include code chucks, figures and written explanations as necessary. Any collaborators must be listed on the top of your assignment. Any figures should be clearly labeled and appropriately referenced within the text.
- 4. When you have completed the assignment and have **checked** that your code both runs in the Console and knits correctly when you click Knit, rename the R Markdown file to YourLastName_YourFirstName_lab6a.Rmd, and knit it into a PDF. Submit the compiled PDF on Canvas.

In this lab, you will need access to the following R packages:

Use the Stock Market data set to analyze the correlation of whether the market will be up or down

```
# Load some helpful libraries
library(ISLR)
```

Analyze the data

```
names (Smarket) #extract information from Smarket
## [1] "Year"
                                                                       "Lag5"
                    "Lag1"
                                 "Lag2"
                                              "Lag3"
                                                          "Lag4"
## [7] "Volume"
                    "Today"
                                 "Direction"
summary(Smarket) #qenerate results summaries
##
         Year
                         Lag1
                                              Lag2
##
           :2001
                           :-4.922000
                                                :-4.922000
    1st Qu.:2002
                    1st Qu.:-0.639500
                                         1st Qu.:-0.639500
##
   Median:2003
                   Median : 0.039000
                                         Median: 0.039000
    Mean
           :2003
                           : 0.003834
                                                 : 0.003919
##
                    Mean
                                         Mean
##
   3rd Qu.:2004
                    3rd Qu.: 0.596750
                                         3rd Qu.: 0.596750
           :2005
                           : 5.733000
##
    Max.
                    Max.
                                                : 5.733000
##
         Lag3
                              Lag4
                                                   Lag5
```

```
Min.
         :-4.922000
                     Min.
                           :-4.922000 Min.
                                             :-4.92200
   1st Qu.:-0.640000 1st Qu.:-0.640000 1st Qu.:-0.64000
## Median: 0.038500 Median: 0.038500 Median: 0.03850
## Mean
        : 0.001716 Mean
                           : 0.001636
                                      Mean
                                             : 0.00561
##
   3rd Qu.: 0.596750 3rd Qu.: 0.596750
                                       3rd Qu.: 0.59700
         : 5.733000 Max.
                           : 5.733000
                                             : 5.73300
##
  Max.
                                       Max.
                     Today
                                Direction
       Volume
## Min.
                                   Down:602
         :0.3561 Min. :-4.922000
##
   1st Qu.:1.2574 1st Qu.:-0.639500
                                   Up :648
  Median :1.4229
                  Median: 0.038500
## Mean
        :1.4783
                  Mean
                        : 0.003138
## 3rd Qu.:1.6417
                  3rd Qu.: 0.596750
## Max. :3.1525
                  Max. : 5.733000
```

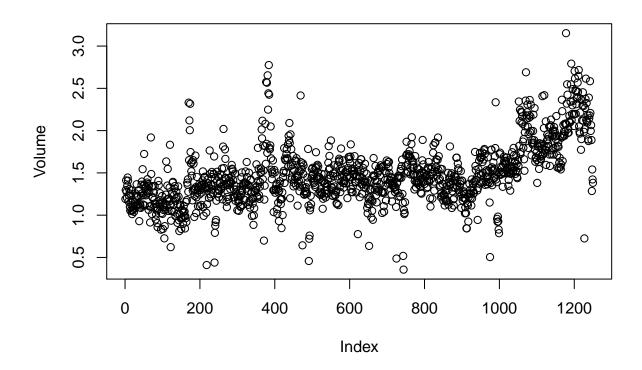
Generate a pairwise correlations

```
cor(Smarket[,-9]) #qenerate a matrix that contains all of the pairwise correlations
```

```
##
               Year
                            Lag1
                                         Lag2
                                                      Lag3
         1.00000000 0.029699649 0.030596422 0.033194581 0.035688718
## Year
         0.02969965 1.000000000 -0.026294328 -0.010803402 -0.002985911
## Lag1
         0.03059642 -0.026294328 1.000000000 -0.025896670 -0.010853533
## Lag2
## Lag3
        0.03319458 -0.010803402 -0.025896670 1.000000000 -0.024051036
         0.03568872 -0.002985911 -0.010853533 -0.024051036 1.000000000
## Lag4
## Lag5
        0.02978799 -0.005674606 -0.003557949 -0.018808338 -0.027083641
## Volume 0.53900647 0.040909908 -0.043383215 -0.041823686 -0.048414246
## Today 0.03009523 -0.026155045 -0.010250033 -0.002447647 -0.006899527
                 Lag5
                           Volume
                                         Today
## Year
          0.029787995 0.53900647 0.030095229
## Lag1 -0.005674606 0.04090991 -0.026155045
## Lag2 -0.003557949 -0.04338321 -0.010250033
## Lag3
         -0.018808338 -0.04182369 -0.002447647
## Lag4
         -0.027083641 -0.04841425 -0.006899527
## Lag5
          1.000000000 -0.02200231 -0.034860083
## Volume -0.022002315 1.00000000 0.014591823
## Today -0.034860083 0.01459182 1.000000000
```

Generate a graph of the volume as response

```
attach(Smarket)
plot(Volume) #ploting the correlation of volume to year
```



Logistic Regression

fits generalized linear model to predict direction using lag1 through lag5 and volume $glm.fit = glm(Direction \sim Lag1+Lag2+Lag3+Lag4+Lag5+Volume, data=Smarket, family=binomial)$ summary(glm.fit) #generate a summary of the generalized model

```
##
## Call:
## glm(formula = Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 +
       Volume, family = binomial, data = Smarket)
##
##
## Deviance Residuals:
##
      Min
               1Q Median
                                3Q
                                       Max
  -1.446 -1.203
                    1.065
                                     1.326
                             1.145
##
## Coefficients:
##
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.126000
                            0.240736
                                     -0.523
                                                0.601
## Lag1
               -0.073074
                            0.050167
                                      -1.457
                                                0.145
## Lag2
               -0.042301
                            0.050086
                                      -0.845
                                                0.398
## Lag3
                0.011085
                            0.049939
                                       0.222
                                                0.824
## Lag4
                0.009359
                            0.049974
                                       0.187
                                                0.851
                            0.049511
                                       0.208
                                                0.835
## Lag5
                0.010313
```

```
## Volume 0.135441 0.158360 0.855 0.392
##
## (Dispersion parameter for binomial family taken to be 1)
##
    Null deviance: 1731.2 on 1249 degrees of freedom
## Residual deviance: 1727.6 on 1243 degrees of freedom
## AIC: 1741.6
##
## Number of Fisher Scoring iterations: 3
```

The summary shows that the smallest p-value is associated with lag1. Thus, if we were to remove lag1 from the model then there might be a higher estimate overall.

Generate a coefficient

Generate probability

glm.pred[glm.probs >.5]="Up"

```
#predicting the problity of the market going up
glm.probs = predict(glm.fit , type = "response")
glm.probs[1:10] #printing the first 10 probabilityes

## 1 2 3 4 5 6 7
## 0.5070841 0.4814679 0.4811388 0.5152224 0.5107812 0.5069565 0.4926509
## 8 9 10
## 0.5092292 0.5176135 0.4888378

contrasts(Direction) #train R to generate a dummy variable for up and down... 1 = up and 0 = down

## Up
## Down 0
## Up
## Down 0
## Up 1
#creating a vector of class predictions based on whether the predictability of a market increase is gre
glm.pred = rep("Down",1250)
```

```
#determine total observations that were correctly or inconrectly classified.
table(glm.pred , Direction)

## Direction
## glm.pred Down Up
## Down 145 141
## Up 457 507

(507+145) /1250

## [1] 0.5216

## [1] 0.5216
```