Homework 3 Solutions - MATH 4322

Instructions

- 1. Due date: February 26, 2024
- 2. Scan or Type your answers and submit only one file. (If you submit several files only the recent one uploaded will be graded).
- 3. Preferably save your file as PDF before uploading. Submit in Canvas under Homework 3.
- 4. These questions are from An Introduction to Statistical Learning with Applications in R by James, et. al., chapter 4.

Problem 1

Suppose we collect data for a group of students in a statistics class with variables X_1 = hours studied, X_2 =undergrad GPA, and Y = receive an A. We fit a logistic regression and produce estimated coefficient, $\hat{\beta}_0 = -6$, $\hat{\beta}_1 = 0.05$, $\hat{\beta}_2 = 1$.

- (a) Estimate the probability that a student who studies for 40 h and has an undergrad GPA of 3.5 gets an A in the class.
- (b) How many hours would the student in part (a) need to study to have a 50% chance of getting an A in the class?

Answer

(a) The model is:

$$p(\hat{X}) = \frac{exp(-6 + 0.05 \times \text{hours} + \text{GPA})}{1 + exp(-6 + 0.05 \times \text{hours} + \text{GPA})}$$

Thus $p(\hat{X}) = 0.3775$.

(b) Use this as the model:

$$\log\left(\frac{p(X)}{1 - p(X)}\right) = -6 + 0.05h + 3.5$$

$$\log(1) = -2.5 + 0.05h$$
$$2.5 = 0.05h$$
$$h = 50$$

Problem 2

This question should be answered using the Weekly data set, which is part of the ISLR2 package. This data set consists of percentage returns for the S&P 500 stock index over 1,089 weekly returns for 21 years, from the beginning of 1990 to the end of 2010. For each week, we have recorded the percentage returns for each of the five previous trading weeks, Lag1 through Lag5. We have also recorded Volume (the average number of shares traded on the previous week, in billions), Today (the percentage return for the week in question) and Direction (whether the market was Up or Down on this week). Our goal is to predict Direction (a qualitative response) using the other features.

(a) Produce some numerical and graphical summaries of the Weekly data. Do there appear to be any patterns?

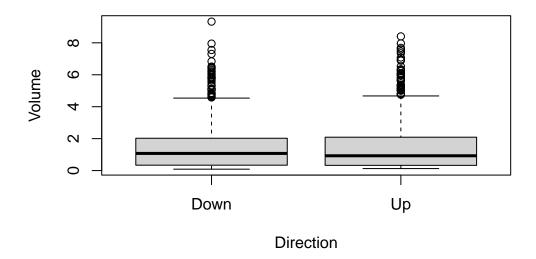
Answer

```
library(ISLR2)
summary(Weekly)
```

Year	Lag1	Lag2	Lag3
Min. :1990	Min. :-18.1950	Min. :-18.1950	Min. :-18.1950
1st Qu.:1995	1st Qu.: -1.1540	1st Qu.: -1.1540	1st Qu.: -1.1580
Median :2000	Median : 0.2410	Median : 0.2410	Median : 0.2410
Mean :2000	Mean : 0.1506	Mean : 0.1511	Mean : 0.1472
3rd Qu.:2005	3rd Qu.: 1.4050	3rd Qu.: 1.4090	3rd Qu.: 1.4090
Max. :2010	Max. : 12.0260	Max. : 12.0260	Max. : 12.0260
Lag4	Lag5	Volume	Today
Min. :-18.195	50 Min. :-18.19	950 Min. :0.0874	7 Min. :-18.1950
1st Qu.: -1.158	30 1st Qu.: -1.16	360 1st Qu.:0.3320	2 1st Qu.: -1.1540
Median : 0.238	30 Median: 0.23	340 Median :1.0026	8 Median: 0.2410
Mean : 0.145	58 Mean : 0.13	399 Mean :1.5746	2 Mean : 0.1499
3rd Qu.: 1.409	90 3rd Qu.: 1.40	050 3rd Qu.:2.0537	3 3rd Qu.: 1.4050
Max. : 12.026	60 Max. : 12.02	260 Max. :9.3282	1 Max. : 12.0260
Direction			
Down: 484			
Up :605			

```
cor(Weekly[,c(2,3,4,5,6,7)])
```

```
Lag1
                       Lag2
                                 Lag3
                                            Lag4
                                                       Lag5
       Lag1
      -0.074853051 \quad 1.00000000 \quad -0.07572091 \quad 0.05838153 \quad -0.072499482
Lag2
Lag3
       0.058635682 \ -0.07572091 \ 1.00000000 \ -0.07539587 \ 0.060657175
      Lag4
Lag5
      -0.008183096 \ -0.07249948 \quad 0.06065717 \ -0.07567503 \quad 1.000000000
Volume -0.064951313 -0.08551314 -0.06928771 -0.06107462 -0.058517414
          Volume
Lag1
      -0.06495131
Lag2
      -0.08551314
Lag3
      -0.06928771
Lag4
      -0.06107462
Lag5
      -0.05851741
Volume 1.00000000
  boxplot(Volume ~ Direction, data = Weekly)
```



There really is no correlation between the lag variables, and looking at the volume weather the direction is up or down is about the same.

(b) Use the full data set to perform a logistic regression with Direction as the response and the five lag variables plus Volume as predictors. Use the summary function to print the results. Do any of the predictors appear to be statistically significant? If so, which ones?

Call:

```
glm(formula = Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 +
Volume, family = "binomial", data = Weekly)
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)
(Intercept) 0.26686
                         0.08593
                                            0.0019 **
                                    3.106
            -0.04127
                         0.02641
                                  -1.563
                                            0.1181
Lag1
Lag2
             0.05844
                         0.02686
                                            0.0296 *
                                    2.175
Lag3
            -0.01606
                         0.02666
                                  -0.602
                                            0.5469
```

```
0.2937
Lag4
            -0.02779
                       0.02646 -1.050
            -0.01447
Lag5
                        0.02638 -0.549
                                          0.5833
Volume
            -0.02274
                        0.03690 -0.616
                                         0.5377
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 1496.2 on 1088 degrees of freedom
Residual deviance: 1486.4 on 1082 degrees of freedom
AIC: 1500.4
Number of Fisher Scoring iterations: 4
```

It appears that only Lag2 is significant in predicting if the stock will go "Up".

(c) Compute the confusion matrix and overall fraction of correct predictions. Explain what the confusion matrix is telling you about the types of mistakes made by logistic regression.

Answer

```
percent.w = predict.glm(week.fit,type = "response")
predict.w = ifelse(percent.w < 0.5, "Down", "Up")
table(predict.w, Weekly$Direction)

predict.w Down Up
    Down 54 48
    Up 430 557</pre>
```

Fraction of correct predictions: $\frac{611}{1089} = 0.5611$ This is not a good model to predict if a stock will go up or down.

(d) Now fit the logistic regression model using a training data period from 1990 to 2008, with Lag2 as the only predictor. Compute the confusion matrix and the overall fraction of correct predictions for the held out data (that is, the data from 2009 and 2010).

Answer

```
train.w = Weekly[Weekly$Year <2009,]
test.w = Weekly[Weekly$Year > 2008,]
week.fit2 = glm(Direction ~ Lag2, data = train.w, family = "binomial")
```

```
Call:
glm(formula = Direction ~ Lag2, family = "binomial", data = train.w)
Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept) 0.20326
                         0.06428
                                    3.162 0.00157 **
              0.05810
                                    2.024 0.04298 *
Lag2
                         0.02870
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 1354.7
                            on 984 degrees of freedom
Residual deviance: 1350.5 on 983 degrees of freedom
AIC: 1354.5
Number of Fisher Scoring iterations: 4
  percent.w = predict.glm(week.fit2,newdata = test.w, type = "response")
  predict.w = ifelse(percent.w < 0.5,"Down","Up")
table(predict.w,test.w$Direction)</pre>
predict.w Down Up
     Down
              9 5
            34 56
     Uр
```

Problem 3

This problem involves writing functions.

Fraction of correct predictions: $\frac{65}{104} = 0.625$

(a) Write a function, Power(), that prints out the result of raising 2 to the 3rd power. In other words, your function should compute 2^3 and print out the results. *Hint: Recall that* x^a raises x to the power a. Use the print() function to output the result.

Answer

```
Power = function(x) {
   print(x^3)
}
Power(2)
```

[1] 8

(b) Create a new function, Power2(), that allows you to pass any two numbers, x and a, and prints out the value of x^a . You can do this by beginning your function with the line

```
Power2 <- function(x, a) {
```

You should be able to call your function by entering, for instance,

```
Power2(3, 8)
```

on the command line. This should output the value of 3⁸, namely, 6,561.

Answer

```
Power2 = function(x,a) {
   print(x^a)
}
Power2(3,8)
```

[1] 6561

Answer

```
Power2(10,3)
```

[1] 1000

```
Power2(8,17)
```

[1] 2.2518e+15

```
Power2(131,3)
```

[1] 2248091

(d) Now create a new function, Power3(), that actually returns the result x^a as an R object, rather than simply printing it to the screen. That is, if you store the value x^a in an object called result within your function, then you can simply return() this result, using the following line:

return(result)

The line above should be the last line in your function, before the } symbol.

Answer

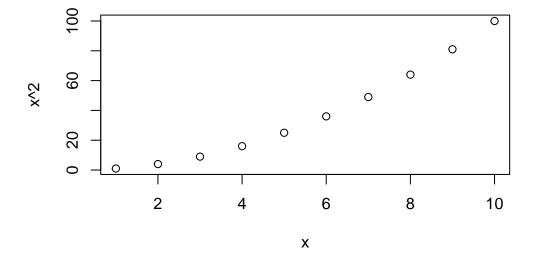
```
Power3 = function(x,a) {
  p3 = x^a
  return(p3)
}
```

(e) Now using the Power3() function, create a plot of $f(x) = x^2$. The x-axis should display a range of integers from 1 to 10, and the y-axis should display x^2 . Label the axes appropriately, and use an appropriate title for the figure.

Answer

```
plot(1:10, Power3(1:10,2), xlab = "x", ylab = "x^2", main = "Using the Power3 function")
```

Using the Power3 function



(f) Create a function, PlotPower(), that allows you to create a plot of x against x^a for a fixed a and for a range of values of x. For instance, if you call

PlotPower(1:10, 3)

then a plot should be created with an x-axis taking on values 1, 2, ..., 10, and a y-axis taking on values $1^3, 2^3, ..., 10^3$.

Answer

```
PlotPower = function(x,a) {
   plot(x,Power3(x,a),xlab = "x",ylab = "x^a",main = "")
}
PlotPower(1:10,3)
```

