STAT 602 - Homework 5

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Document External Libraries

- ISLR package for the homework data sets
- knitr package used for kable function used to format tables
- dplyr package for data formatting and cleaning
- MASS package for LDA and QDA models
- mclust package for Mclust models and graphs
- class package for KNN models

Reusable Functions

• The misclassification function created in homework 3 (*misclass.fun.JH*) will be reused for questions in this homework.

Exercises

Question 1 (ISLR 4.7.6 pg 170)

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, $\hat{\beta}_0 = -6$, $\hat{\beta}_1 = 0.05$, $\hat{\beta}_2 = 1$.

Question 1(a)

Estimate the probability that a student who studies for 40 hours and has an undergrad GPA of 3.5 gets an A in the class.

Answer

```
# Setting variables for the beta values for each each covariant 'x' and the intercept
b_0 = -6
b_1 = 0.05
b_2 = 1
# Setting variables for X values
x_1 = 40
x_2 = 3.5
# Calculating the "x" in phat_x
phat_x <- b_0+b_1*x_1+b_2*x_2
# Plug phat_x into our probability function to derive an answer</pre>
```

```
phat <- exp(phat_x)/(1+exp(phat_x))
cat('Probability that this student will get an A is',round(phat*100,2),'%')</pre>
```

Probability that this student will get an A is 37.75 %

In formulaic terms:

$$\hat{p}(X) = \frac{e^{\hat{\beta}_0 + \hat{\beta}_1 x_1 + \hat{\beta}_2 x_2}}{1 + e^{\hat{\beta}_0 + \hat{\beta}_1 x_1 + \hat{\beta}_2 x_2}}$$

Plugging in values to the variables:

$$\hat{p}(X) = \frac{e^{-6+0.05*40+1*3.5}}{1 + e^{-6+0.05*40+1*3.5}}$$

Simplified:

$$\hat{p}(X) = \frac{e^{-0.5}}{1 + e^{-0.5}}$$

Therefore:

$$\hat{p}(X) = 0.3775407$$

Question 1(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

Solve for x:

$$0.5 = \frac{e^{-6+0.05x+3.5}}{1 + e^{-6+0.05x+3.5}}$$

Simplify the exponent:

$$0.5 = \frac{e^{-2.5 + 0.05x}}{1 + e^{-2.5 + 0.05x}}$$

Multiply each side by the demoninator:

$$0.5(1 + e^{-2.5 + 0.05x}) = e^{-2.5 + 0.05x}$$

Simplify the left side of the equation:

$$0.5 + 0.5(e^{-2.5 + 0.05x}) = e^{-2.5 + 0.05x}$$

Subtract each side by $0.5(e^{-2.5+0.05x})$:

$$0.5 = e^{-2.5 + 0.05x} - 0.5e^{-2.5 + 0.05x}$$

$$0.5 = 0.5e^{-2.5 + 0.05x}$$

Multiplying each side by 2:

$$1 = e^{-2.5 + 0.05x}$$

$$0 = -2.5 + 0.05x$$

Take the natural log of each side:

$$ln(1) = ln(e^{-2.5 + 0.05x})$$

$$0 = -2.5 + 0.05x$$

$$x = 50$$

It would take this student 50 hours of studying to have a 50% chance of recieving a 4.0 GPA.

Source: An Introduction to Statistical Learning

Question 2

Continue from Homework #3 & 4 using the **Weekly** dataset from 4.7.10). Construct a model (using the predictors chosen for previous homework) and fit this model using the MclustDA function from the **mclust** library.

Question 2(i) Part I

Provide a summary of your model.

Answer

Using the all the Lag variables as was done in Homework 3 & 4 as the predictors, and Direction as the target with the MclustDA function to create the model. In addition, I set the # of Groups to 9. In order to calculate the training and test errors, I partitioned the data set into train/test with the same method as prior homeworks. The below summaires are just on the training set.

Table 1: Summary of Mclust Model (Train)

| | Down | Up |
|------------|------|------|
| n | 441 | 544 |
| Proportion | 0.45 | 0.55 |
| Model | VII | VII |
| G | 3 | 2 |

Table 2: Metric Summary of Mclust Model (Train)

| Metrics | Values |
|-------------|--------|
| Class Error | 0.424 |
| Brier Score | 0.242 |

| Metrics | Values |
|----------|------------|
| Log Like | -10473.330 |
| BIC | -21174.117 |

Table 3: Weekly MclustDA Confusion Matrix

| | Pred Down | Pred Up |
|----------|-----------|---------|
| Act Down | 185 | 256 |
| Act Up | 162 | 382 |

Question 2(i) Part II

What is the best model selected by BIC? Report the Model Name and the BIC. (See mclustModelNames)

Answer

Using the Mclust function, I examined the different BIC scores of groupings set between 1-9 for both the Up and Down class.

Table 4: Clustering Table Proportions for Up Class

| Group | Proportion |
|-------|------------|
| 1 | 0.817 |
| 2 | 0.183 |
| 3 | 0.817 |

Table 5: Model Metrics for Up Class

| Metrics | Values |
|----------|-----------|
| Model | VII |
| Log Like | -5795.5 |
| n | 544 |
| df | 5 |
| BIC | -11672.89 |
| ICL | -11743.48 |
| | |

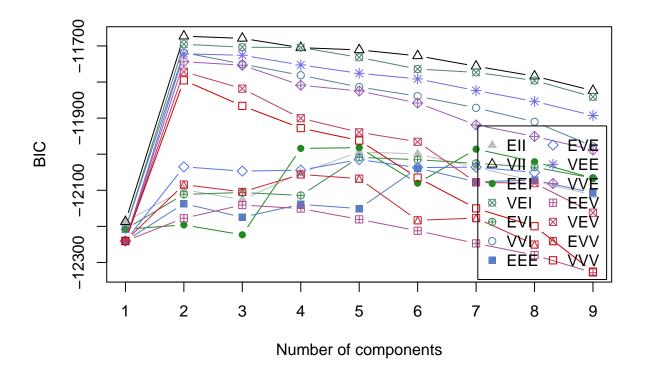
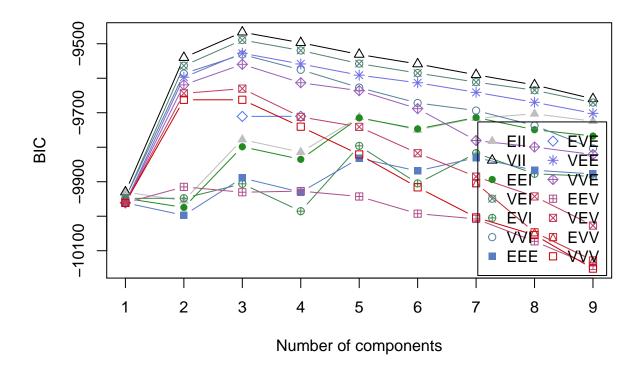


Table 6: Clustering Table Proportions for Down Class

| Group | Proportion |
|---------|------------|
| - Group | Troportion |
| 1 | 0.483 |
| 2 | 0.491 |
| 3 | 0.026 |

Table 7: Model Metrics for Down Class

| Metrics | Values |
|----------|----------|
| Model | VII |
| Log Like | -4672.34 |
| n | 441 |
| df | 5 |
| BIC | -9466.46 |
| ICL | -9646.07 |
| | |



According to the tables and graphs above, the Up class with the lowest BIC model has 2 groups with a spherical, varying volume model and a BIC of -11,672.89.

For the Down class, the lowest BIC model has 3 groups with a spherical, varying volume model and a BIC of -9,466.46.

This is displayed on the BIC plots with the VII model having the highest BIC for 2 and 3 components, respectfully.

Question 2(i) Part III

Report the true positive rate, true negative rate, training error, and test error. You can reuse the function written in Homework # 3.

Answer

Calculated the TP, TN, FP, and FN based of the confusion matrix in the mclust model. It was just built on the training data.

Table 8: Metrics for Training Model

| Metrics | Values |
|----------------|----------------|
| TPR TNR | 0.702 0.420 |
| Training Error | 0.424 |

Calculated the TP, TN, FP, and FN based of the confusion matrix in the mclust model. The summary of the model was updated with new data and class on the test data set.

Table 9: Metrics for Test Model

| Metrics | Values |
|---------|--------|
| TPR | 0.607 |
| TNR | 0.465 |

| Metrics | Values |
|------------|--------|
| Test Error | 0.452 |

Question 2(ii) Part I

Repeat the MclustDA analysis, but this time specify modelType = "EDDA". Provide a summary of this model.

Answer

Table 10: Summary of Mclust EDDA Model (Train)

| | Down | $_{\mathrm{Up}}$ |
|------------|------|------------------|
| n | 441 | 544 |
| Proportion | 0.45 | 0.55 |
| Model | EII | $_{ m EII}$ |
| G | 1 | 1 |

Table 11: Metric Summary of Mclust EDDA Model (Train)

| Metrics | Values |
|-------------|------------|
| Class Error | 0.445 |
| Brier Score | 0.245 |
| Log Like | -11026.144 |
| BIC | -22128.107 |

Table 12: Weekly EDDA MclustDA Confusion Matrix

| | Pred Down | Pred Up |
|----------|-----------|---------|
| Act Down | 65 | 376 |
| Act Up | 62 | 482 |

Question 2(ii) Part II

What is the best model using BIC as the model selection criteria?

Answer

According EDDA model above, is EII or spherical, equal volume with one group based on a BIC of -22,128.107.

Question 2(ii) Part III

Report the true positive rate, true negative rate, training error, and test error. You can reuse the function written in Homework # 3.

Answer

I calculated the TP, TN, FP, and FN based of the confusion matrix in the EDDA mclust model. It was built on the training data.

Table 13: Metrics for Training EDDA Model

| Metrics | Values |
|----------------|--------|
| TPR | 0.886 |
| TNR | 0.147 |
| Training Error | 0.445 |

Calculated the TP, TN, FP, and FN based of the confusion matrix in the EDDA mclust model. The summary of the model

was updated with new data and class on the test data set.

Table 14: Metrics for Test EDDA Model

| Metrics | Values |
|------------|--------|
| TPR | 0.607 |
| TNR | 0.465 |
| Test Error | 0.452 |
| | |

Question 2(iii)

Compare the results with Homework #3 & 4. Which method performed the best? Justify your answer. Present your results in a well formatted table; include the previous methods and their corresponding rates.

Answer

Using formulas and code from Homework 4 with all Lag variables as predictors for each of the models, I summarized the Test Error, TPR, and TNR rates for each in a data frame.

Table 15: Model Summaries for Weekly Data All Lag Variables

| Metrics | GLM | LDA | QDA | KNN | Mclust | Mclust_EDDA |
|------------------------------------|-------------------------|-----------------------|-----------------------|-----------------------|-----------------------|-------------------------|
| Test Error Test TPR Test TNR | 0.558 0.098 0.930 | 0.452 0.787 0.209 | 0.538 0.623 0.233 | 0.481 0.541 0.488 | 0.452 0.702 0.420 | 0.452 0.607 0.465 |

According to the summary above, LDA, Mclust, and Mclust_EDDA have the same and lowest test error rate, however the Mclust_EDDA have a more evenly distributed TPR and TNR than the other models. However, the Mclust model may be the best fit since the loss on the TNR rate comapred to the EDDA model isn't much, but the gain on the TPR rate is faily high.

Question 2(iv)

From the original model variables, construct a new set of variables, fit a model using MclustDA and repeat i-iii. *Hint: new variables may be interactions, polynomials, and/or splines*. Do these new variables give an improvement in error rates compared to previous models? Explain how the new variables were constructed.

Answer

Using the following variables to build the model based on Homework 4, which were chosen with all Lag variables, their interactions, squared and cubed. I used each of these variables as predictors in the new models and summazied the metrics in a data frame.

- Lag1
- Lag2
- Lag4
- · Lag1 Squared
- Lag3 Squared
- Lag5 Squared
- Lag1:Lag3 Interaction
- $\bullet \quad Lag1: Lag3: Lag4 \ Interaction$
- $\bullet \quad Lag2: Lag3: Lag4 \ Interaction$
- $\bullet \quad Lag2: Lag3: Lag5 \ Interaction$

I created a new data frame with the above variables and split the data into training and test data sets. These variables will be used for all the models in the previous question. In addition, I converted the *Direction* variable to to 1 for Up and 0 for Down for model simplicity.

Created the MclustDA model on the training data with the new variables using the same method as in the first model.

Table 16: Summary of Mclust Model 2 (Train)

| | 0 | 1 |
|------------|------|------|
| n | 441 | 544 |
| Proportion | 0.45 | 0.55 |
| Model | VVI | VVI |
| G | 14 | 12 |

Table 17: Metric Summary of Mclust Model 2 (Train)

| ues |
|---------------------------------|
| $\frac{1}{328}$ $\frac{1}{253}$ |
| $\frac{347}{290}$ |
| |

Table 18: Weekly MclustDA Confusion Matrix

| | Pred Down | Pred Up |
|----------|-----------|---------|
| Act Down | 284 | 157 |
| Act Up | 166 | 378 |

Summary of MclustDA Model shows that the model off the training data with the lowest BIC uses a diagonal, varying volume and shape model with 14 groups for the Down class and 12 for the Up class. The BIC is -47,784.29. This is lower than the previous model with just the Lag variables and less of a fit according to BIC, but better of a fit according to class error.

Created the MclustDA model setting the model type to 'EDDA' on the training data with the new variables using the same method as in the first model.

Table 19: Summary of Mclust EDDA Model 2 (Train)

| | 0 | 1 |
|------------|------|------|
| n | 441 | 544 |
| Proportion | 0.45 | 0.55 |
| Model | VVV | VVV |
| G | 1 | 1 |

Table 20: Metric Summary of Mclust EDDA Model (Train)

| Metrics | Values |
|-------------|---------------|
| Class Error | 0.451 |
| Brier Score | 0.431 0.273 |
| Log Like | -35773.906 |
| BIC | -72443.856 |

Table 21: Weekly EDDA MclustDA Confusion Matrix

| | Pred Down | Pred Up |
|--------|-----------|------------|
| Act Up | 45 48 | 396 496 |
| Act Up | 48 | 496 |

Summary of EDDA MclustDA Model shows that the model off the training data with the lowest BIC uses a ellipsoidal, varying volume, shape, and orientation with 1 group for the Down and Up class. The BIC is -72,443.856. This is lower than the previous model with just the Lag variables and less of a fit according to BIC. It is also much worse than the Mclust model without the

EDDA model type.

I built all the models based on the methodology from the prior homework and built a data frame with a summary of the Test Error, Test TPR, and Test TNR and compared the results.

Table 22: Model Summaries for Weekly Data Select Lag Variables

| Metrics | GLM | LDA | QDA | KNN | Mclust | Mclust_EDDA |
|------------|----------------------|-------|-------|-------|--------|-------------|
| Test Error | 0.558 | 0.433 | 0.490 | 0.481 | 0.567 | 0.490 |
| Test TPR | 0.098 | 0.869 | 0.721 | 0.557 | 0.525 | 0.721 |
| Test TNR | 0.930 | 0.140 | 0.209 | 0.465 | 0.302 | 0.209 |

Comparing the 2 set of variables, the GLM model did not change. THe LDA model had a slightly better Test Error Rate, with a better TPR and worse TNR. QDA had a better Rest Error, with a better TPR and worse TNR. KNN Test Error stayed the same, with slightly better TPR and worse TNR. The Mclust DA model had a worse metrics for all 3 from the other model, while the EDDA model had a worse Test Error and TNR with a better TNR.

In addition, it looks like the Mclust model overfit the data as the traing error was lower than the first model, but the test error was higher than the first model.

Overall, similar to Homework 4, the KNN model seems to preform better with the second lowest Test Error, with a more balance TPR and TNR.

Source: An Introduction to Statistical Learning, MclustDA Part 1 Lecture, Dr. Saunders, Dakota State University, and Package 'mclust'