

STATS 415 Homework 4

Due Thursday Oct 10, 2019

Turn in a printout of your homework at the classroom before the lecture begins.

1. Suppose you have one continuous predictor X and a binary categorical response Y , which can take values 1 or 2. Suppose you collected training data from the two classes and obtained class-specific sample means $\hat{\mu}_1 = -1$ and $\hat{\mu}_2 = 3$, along with the pooled variance estimate over the two classes, $\hat{\sigma}^2 = 1$. (40pt total, 8pt for each question)
 - (a) Assume equal class priors and derive the LDA classification rule for this problem. Sketch the estimated class-conditional densities and show your decision boundary on the plot. Make sure you label the axes and indicate the numerical value for the boundary; let's call it c .
 - (b) Suppose the estimates were in fact obtained from 100 training points, among which 40 were from class 1 and 60 were from class 2. Suppose now you will estimate class priors from data, repeat all the calculations in part (a) and obtain a new boundary value, let's call it \tilde{c} . Without actually doing this, would you be able to tell whether \tilde{c} will be the same as, less than, or greater than c , or is there no way to tell? Explain your answer without calculating \tilde{c} . Note: It's ok to recheck your answer once you have actually calculate \tilde{c} in part (c), but your explanation must not involve the numerical value.
 - (c) Now calculate the new boundary value \tilde{c} described in part (b).
 - (d) Suppose in addition to the pooled covariance value $\hat{\sigma}^2$ I now tell you the individual class specific covariances were estimated as $\hat{\sigma}_1^2 = 0.25$ and $\hat{\sigma}_2^2 = 1.5$. Based on this new information, would you recommend using LDA or QDA, and why?
 - (e) Derive the QDA rule for part (d), assuming equal class priors.

2. In this problem, you will develop a model to predict whether a given car will be classified as having high or low gas mileage based on the `Auto` data set in the `ISLR` package. (60pt total, 10pt each question)
- (a) Create a binary variable, `mpg01`, that is equal to 1 if the value of `mpg` for that car is above 25, and 0 otherwise. You may then want to use the `data.frame()` function to create a single data set containing both `mpg01` and the other `Auto` variables.
 - (b) Make some exploratory plots to investigate the association between `mpg01` and other variables. Describe your findings. Which of the features seem most likely to be useful in predicting `mpg01`? Scatterplots and boxplots may be useful tools to answer this question. (Note: do not use the `mpg` variable that was used to create `mpg01`).
 - (c) Split the data into a training set and a test set: fix the random seed to the value 123, and randomly select 80% of the observations (round down to the nearest integer) from *each* class to be the training data. Use the rest as test data.
 - (d) Perform LDA on the training data in order to predict `mpg01` using four quantitative variables that seem most associated with `mpg01` based on (b). Report the training and test errors. Make a plot of the training data points, using two variables which appear to be most associated with the class as your axes. Using different colors to show the true values of `mpg01`, and different plotting symbols to show predicted values.
 - (e) Perform QDA on the training data in order to predict `mpg01` using the same variables you used for LDA. Report the training and test errors. Make a plot analogous to the one you made for LDA.
 - (f) Compare and contrast the performance of LDA and QDA. What do your results suggest about the class-specific covariances?

Please limit your answer to Q2 to 8 pages, organized into a coherent typed data analysis report. Answers to Q1 may be either typed or handwritten. Please staple everything together and clearly write your name, your UMID, and your GSI/lab number on the homework.