Exam 2 A - MATH 4322

Cathy Poliak

Fall 2022

Name:	PSID:	

Instructions

- Allow one sheet of notes front and back to be turned in for extra credit.
- Allow calculator.
- Total possible points 100.
- For multiple choice circle your answer on this test paper.
- For short answer questions answer fully on this test paper, partial credit will be given.

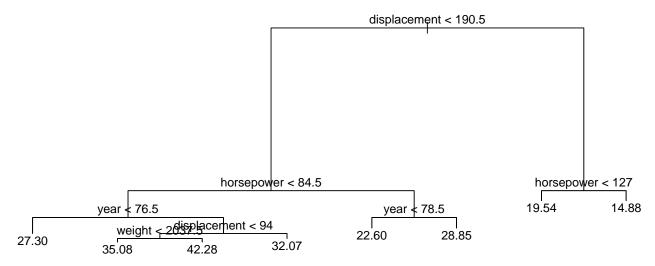
Part 1

We would like to predict the mpg based on some features. Here are the variables.

- mpg: miles per gallon (response variable)
- cylinders: Number of cylinders between 4 and 8
- displacement: Engine displacement (cu. inches)
- horsepower: Engine horsepower
- weight: Vehicle weight (lbs.)
- acceleration: Time to accelerate from 0 to 60 mph (sec.)
- year: Model year (modulo 100)
- origin: Origin of car (1. American, 2. European, 3. Japanese)
- 1. (3 Possible Points) Is this a regression or classification problem? Give the reason for your answer.

This is a regression problem. Since the response variable "mpg" is quantitative.

2. (8 Possible Points) The following is the output based on the decision tree to predict the average mpg based on the predictors listed above.



a. Write down the full R code to produce this tree. Including the package that we need to call.

Answer

3 points

```
library(tree)
tree.auto = tree(mpg ~ . -name, data = Auto2, subset = train)
plot(tree.auto)
text(tree.auto,pretty = 0)
```

b. List all the predictors that appear in the tree.

Answer

2 points - displacement, horsepower, year, and weight

c. How do we interpret the following terminal nodes? Interpret the numbers at the bottom.



Answer 3 points - If the horsepower is less than 127, then the mpg is predicted to be 19.54. If the horsepower is at least 127, then the mpg is predicted to be 14.88.

- 3. (10 Possible Points) The following are the mean square errors based on the single tree, random forest and bagging.
 - a. (3 points) Give the formula for the mean squared error (MSE).

Answer

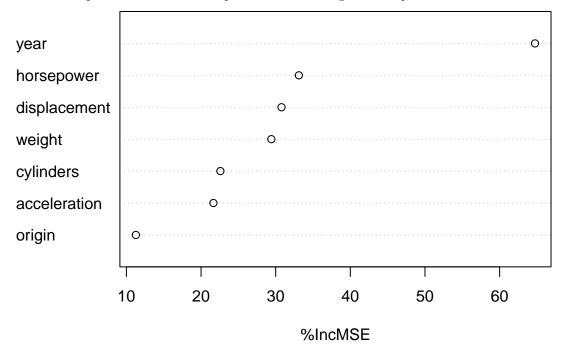
$$MSE = \frac{1}{n} \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$

- b. (2 points) Match the correct MSE with: a. Single tree or b. Random Forest.
 - i. 6.759509 **Random Forest**
 - ii. 12.33831 **Single Tree**
- c. (5 points) Interpret the MSE value of 12.33831?

Answer

With this tree produced, the predicted mpg will be off by $\sqrt{12.33831} = 3.512594$ on average.

4. (3 Possible Points) The variable of importance plot is below from the bagging method. What are the three most important variables? Compare that to the single tree in problem 1.



Answer

The three most important features are year, horsepower and displacement. This does not really match up with the original tree because displacement is at the top of the tree.

5. (3 Possible Points) Give the full R code to get the plot above. Including the code to get the model.

Answer

Need to call the library.

Have mpg as the response variable and make sure that importance = TRUE is in the function. Call the plot.

Part 2

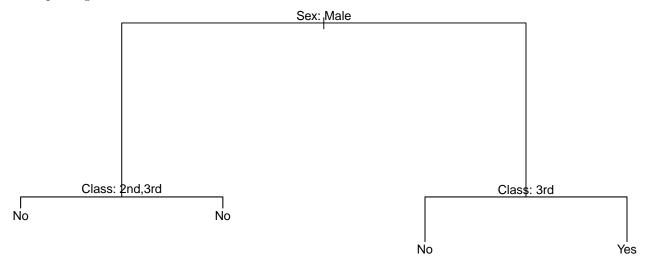
We are wanting to predict survival on the Titantic and determine which features help us determine if a person survived or not. The following features are used.

- Class: 1st, 2nd, 3rd, or Crew
- Sex: Male or Female
- Age: Child or Adult
- Survived: No or Yes (response variable)
- 1. (3 Possible Points) Is this a regression or classification problem? Give the reason for your answer.

Answer

This is a classification problem. Our response variable Survived is categorical.

2. (8 Possible Points) The following is the output based on the decision tree to predict the Survival of a passenger.



a. Write down the full R code to produce this tree, including what package we need to call.

Answer

2 points

```
library(tree)
tree.titanic = tree(Survived ~ . , data = titanic, subset = train)
plot(tree.titanic)
text(tree.titanic,pretty = 0)
```

b. List all the predictors that appear in the tree.

Answer

1 point

Sex and Class.

c. Describe from the tree what type of passenger is predicted to survive.

Answer

1 point

A female who is not in 3rd class is the only type of passenger that is predicted to survive.

- d. (1 points for each question) The following is the output of the decision tree. Interpret node 3) by answering the following questions.
 - i. How many passengers are in this node? 339
 - ii. How is this node separated? Sex: Female

- iii. What is the proportion of passengers that survived (Yes) in this node? 74.63%
- iv. What is the overall Survival prediction for this node Yes or No? Yes

```
node), split, n, deviance, yval, (yprob)
```

- * denotes terminal node
- 1) root 1541 1974.0 No (0.6606 0.3394)
 - 2) Sex: Male 1202 1281.0 No (0.7754 0.2246)
 - 4) Class: 2nd,3rd 487 428.5 No (0.8398 0.1602) \ast
 - 5) Class: 1st,Crew 715 832.0 No (0.7315 0.2685) *
 - 3) Sex: Female 339 384.0 Yes (0.2537 0.7463)
 - 6) Class: 3rd 139 192.7 No (0.5036 0.4964) *
 - 7) Class: 1st,2nd,Crew 200 111.5 Yes (0.0800 0.9200) *
 - 3. (3 Possible Points) Below is the confusion matrix for the tree. What is the test error rate?

		Observed Class	
		No	Yes
Predicted	1	468	118
Class	2	4	70

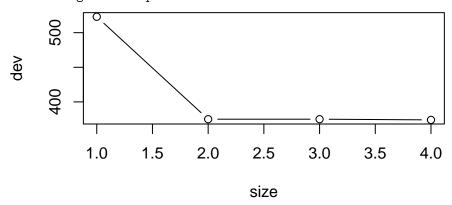
Answer

$$\frac{122}{660} = 0.1848$$

- 4. (8 Possible Points) The following is a plot of the cross-validation error based on the number of nodes.
 - a. (4 points) Write down the full R code to get this plot.

Answer

b. (4 points) According to this plot what should be the number of nodes we can prune for the tree?



Answer It looks like 2 nodes would be the best number of terminal nodes.

5. (3 Possible Points) Now we can prune the tree. What is the full R code to get a pruned tree?

```
prune.titanic = prune.misclass(tree.titanic, best = 2)
plot(prune.titanic)
text(prune.titanic, pretty = 0)
```

6. (3 Possible Points) The following is the confusion matrix based on the pruned tree. What is the test error rate? Compare this to the test error rate of the unpruned tree.

		Observed Class	
		No	Yes
Predicted	No	432	97
Class	Yes	40	91

Answer

$$\frac{137}{660} = 0.2076$$

This is a little bit higher than the miss classified error from the full tree.

Part 3

Given for a linear neural network model with input variables x_1, x_2, x_3 , and two hidden layers h_1, h_2 you are given the following parameter values:

input & hidden layer:

Input \ Hidden	h_1	h_2
1 (bias)	0.1	0.2
x_1	-0.2	0.3
x_2	0.2	-0.2
x_3	-0.3	0.75

hidden & output layer:

Hidden \ Output	y
1 (bias)	0.6
h_1	0.1
h_2	-0.5

1. (10 points) Draw a mathematical model of this linear neural network.

Answer

Make sure there is nodes for each x, hidden node and 2 biases and y - 5 points The weights are on the graph - 5 points

2. (10 points) Calculate the output for the case of $x_1 = 2$, $x_2 = -4$, and $x_3 = 5$. Show your work.

$$\hat{h}_1 = 0.1 - 0.2(2) + 0.2(-4) - 0.3(5) = -2.6$$

$$\hat{h}_2 = 0.2 + 0.3(2) - 0.2(-4) + 0.75(5) = 5.35$$

$$\hat{h}_2 = 0.2 + 0.3(2) - 0.2(-4) + 0.75(5) = 5.35$$

$$\hat{y} = 0.6 + 0.1(-2.6) - 0.5(5.35) = -2.335$$

Part 4: Multiple Choice

Circle the best answer. Each question is worth 5 points, for a total of 25 points for this part.

- 1. When a given method yields a small training MSE but a large test MSE, we are said to be
 - a. Underfitting the data
 - b. Overfitting the data
 - c. Exactly right
 - d. Biased of the data
 - e. All of these are true.
- 2. What method is to select a subtree from a very large tree that leads to a best lowest test error rate?
 - a. Random Forest
 - b. Bagging
 - c. Boosting
 - d. Pruning
 - e. Cross-Validation
- 3. Below is an output for estimating the median based on the bootstrap method. What is the estimate of the median based on all of the bootstrap samples.
 - a. 24.3
 - b. -0.02
 - c. 0.6719
 - d. **24.28**
 - e. 24.97

ORDINARY NONPARAMETRIC BOOTSTRAP

```
Call:
boot(data = s.data, statistic = median.fun, R = 1000)

Bootstrap Statistics :
    original bias std. error
t1* 24.29888 -0.01934575 0.6718975
```

4.	Suppose we have a data frame with $n = 200$ observations.	. We want to do a 10-fold Cross-validation to
	determine the test MSE, how many observations are used	l in the test set each time?

- a. **20**
- b. 100
- $c.\ 150$
- d. 200
- e. 2000

5. In the case of the neural networks, the sigmoid function, $g(z) = \frac{1}{1+e^{-z}}$ is an example of:

- a. activation function
- b. tree function
- c. input funciton
- d. linear function
- e. random forest function