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Homework-1

- Q1. For each of parts (a) through (d), indicate whether i. or ii. is correct, and explain your answer. In general, do we expect the performance of a flexible statistical learning method to perform better or worse than an inflexible method when:
- a. The sample size n is extremely large, and the number of predictors p is small?
- -> Better. A flexible method will fit the data closer and with the large sample size, would perform better than an inflexible approach.
- b. The number of predictors p is extremely large, and the number of observations n is small?
- -> Worse. A flexible method would overfit the small number of observations.
- c. The relationship between the predictors and response is highly non-linear?
- -> Better. With more degrees of freedom, a flexible method would fit better than an inflexible one.
- d. The variance of the error terms, i.e. $\sigma 2 = Var(\epsilon)\sigma 2 = Var(\epsilon)$, is extremely high?
- -> Worse. A flexible method would fit to the noise in the error terms and increase variance.
- Q2. Explain whether each scenario is a classification or regression problem, and indicate whether we are most interested in inference or prediction. Finally, provide n and p.
- a. We collect a set of data on the top 500 firms in the US. For each firm we record profit, number of employees, industry and the CEO salary. We are interested in understanding which factors affect CEO salary.
- -> Regression and inference with n=500 and p=3
- b. We are considering launching a new product and wish to know whether it will be a success or a failure. We collect data on 20 similar products that were previously launched. For each product we have recorded whether it was a success or failure, price charged for the product, marketing budget, competition price, and ten other variables.
- -> Classification and prediction with n=20 and p=13
- c. We are interested in predicting the % change in the US dollar in relation to the weekly changes in the world stock markets. Hence we collect weekly data for all of 2012. For each week we record the % change in the dollar, the % change in the US market, the % change in the British market, and the % change in the German market.
- -> Regression and prediction with n=52 and p=3

- 4. You will now think of some real-life applications for statistical learning.
- (a) Describe three real-life applications in which *classification* might be useful. Describe the response, as well as the predictors. Is the goal of each application inference or prediction? Explain your answer.
 - Classification example 1- Is the email a spam or not

Response- Yes/No

Predictors-Promotions/newsletters/subscriptions/updates/type of mail/sender's address/type of account

Goal- Prediction

> Classification example 2- Should a product be launched in the market or not

Response-Yes/No

Predictors- Features/Price/Market/Buyer/Effectiveness in market/ability to sustain/competitors/type of product

Goal-Prediction

> Classification example 3- Polio Vaccine test

Response- Successful/Not-successful

Predictors- Geography, health condition, age, test group, mental health Goal-Prediction

- (b) Describe three real-life applications in which *regression* might be useful. Describe the response, as well as the predictors. Is the goal of each application inference or prediction? Explain your answer.
- ➤ Regression example 1- Gas mileage that a new car will give Response-The mileage of this car will be XX Predictors-Model, make, Engine, Company Goal-Inference
- ➤ Regression example 2- GDP growth in an economy Response- The countries GDp growth for this year is ... Predictors- GDP, Employment, Education, Trade, Relations, Goal- Inference
- ➤ Regression example 3- Increase in house prices over next 3 years Response- The price of houses have increased by this amount Predictors- Parks, Schools, Average size of family, Average Income of Family, Crime Rate Goal- Inference
- (c) Describe three real-life applications in which *cluster analysis* might be useful.
 - Clustering example 1- Division of income group in USD-5000-10000, 10000-15000, 15000 and above

Response- This % of people fall under this income group

Predictors- Countries, social life, employment, political status, economic status Goal-Prediction

> Clustering example 2- Ratings given to cars as good/average/bad

Response-This car is rated good

Predictors-Companies, model, make, popularity, Advertisement

Goal-Prediction

> Clustering example 3- Division of countries into developed, developing and others Response- Asian countries are developing while European are developed

Predictors- GDP, Employment, Education, Trade, Relations,

- 5. What are the advantages and disadvantages of a very flexible (versus a less flexible) approach for regression or classification? Under what circumstances might a more flexible approach be preferred to a less flexible approach? When might a less flexible approach be preferred?
- -> The advantages of a very flexible approach are that it may give a better fit for non-linear models and it decreases the bias. The disadvantages of a very flexible approach are that it requires estimating a greater number of parameters, it follows the noise too closely (overfit) and it increases the variance.

A more flexible approach would be preferred to a less flexible approach when we are interested in prediction and not the interpretability of the results. A less flexible approach would be preferred to a more flexible approach when we are interested in inference and the interpretability of the results.

- 8. This exercise relates to the College data set, which can be found in the file College.csv. It contains a number of variables for 777 different universities and colleges in the US. The variables are
- Private : Public/private indicator
- Apps: Number of applications received
- Accept: Number of applicants accepted
- Enroll: Number of new students enrolled
- Top10perc : New students from top 10% of high school class
- Top25perc : New students from top 25% of high school class
- F.Undergrad: Number of full-time undergraduates
- P.Undergrad : Number of part-time undergraduates
- Outstate: Out-of-state tuition
- Room.Board: Room and board costs.
- Books : Estimated book costs
- Personal: Estimated personal spending
- PhD: Percent of faculty with Ph.D.'s
- Terminal: Percent of faculty with terminal degree
- S.F.Ratio: Student/faculty ratio
- perc.alumni : Percent of alumni who donate
- Expend: Instructional expenditure per student
- Grad.Rate: Graduation rate

Before reading the data into R, it can be viewed in Excel or a text

- (a) Use the read.csv() function to read the data into R. Call the loaded data college. Make sure that you have the directory set to the correct location for the data.
- > library(ISLR)
- > data("College")
- > college<- read.csv("College.csv")</pre>
- (b) Look at the data using the fix() function. You should notice that the first column is just the name of each university. We don't really want R to treat this as data. However, it may be handy tohave these names for later.

> head(college[,1:5])

Private Apps Accept Enroll Top10perc Abilene Christian University Yes 1660 1232 721 23 Adelphi University Yes 2186 1924 512 16 Adrian College Yes 1428 1097 22 336 Agnes Scott College Yes 417 349 137 60 Alaska Pacific University Yes 193 146 55 16 Albertson College Yes 587 479 158 38

(c) i. Use the summary() function to produce a numerical summary of the variables in the data set

> summary(College)

Private Apps Accept Enroll Top10perc Top25perc F.Undergrad
No:212 Min.: 81 Min.: 72 Min.: 35 Min.: 1.00 Min.: 9.0 Min.: 139
Yes:565 1st Qu.: 776 1st Qu.: 604 1st Qu.: 242 1st Qu.:15.00 1st Qu.: 41.0 1st Qu.: 992
Median: 1558 Median: 1110 Median: 434 Median: 23.00 Median: 54.0 Median: 17
07

Mean : 3002 Mean : 2019 Mean : 780 Mean : 27.56 Mean : 55.8 Mean : 3700 3rd Qu.: 3624 3rd Qu.: 2424 3rd Qu.: 902 3rd Qu.:35.00 3rd Qu.: 69.0 3rd Qu.: 4005 Max. :48094 Max. :26330 Max. :6392 Max. :96.00 Max. :100.0 Max. :31643 P.Undergrad Outstate Room.Board Books Personal PhD Terminal Min. : 1.0 Min. : 2340 Min. :1780 Min. : 96.0 Min. : 250 Min. : 8.00 Min. : 24.0 1st Qu.: 95.0 1st Qu.: 7320 1st Qu.:3597 1st Qu.: 470.0 1st Qu.: 850 1st Qu.: 62.00 1st Qu : 71.0

Median: 353.0 Median: 9990 Median: 4200 Median: 500.0 Median: 1200 Median: 75.00 Median: 82.0

Mean : 855.3 Mean :10441 Mean :4358 Mean : 549.4 Mean :1341 Mean : 72.66 Me an : 79.7

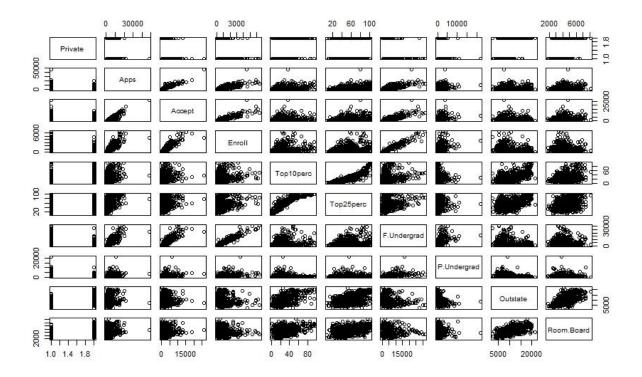
3rd Qu.: 967.0 3rd Qu.:12925 3rd Qu.:5050 3rd Qu.: 600.0 3rd Qu.:1700 3rd Qu.: 85.00 3rd Qu.: 92.0

Max. :21836.0 Max. :21700 Max. :8124 Max. :2340.0 Max. :6800 Max. :103.00 Max. :100.0

S.F.Ratio perc.alumni Expend Grad.Rate
Min.: 2.50 Min.: 0.00 Min.: 3186 Min.: 10.00
1st Qu.:11.50 1st Qu.:13.00 1st Qu.: 6751 1st Qu.: 53.00
Median: 13.60 Median: 21.00 Median: 8377 Median: 65.00
Mean: 14.09 Mean: 22.74 Mean: 9660 Mean: 65.46
3rd Qu.:16.50 3rd Qu.:31.00 3rd Qu.:10830 3rd Qu.: 78.00
Max.: 39.80 Max.: 64.00 Max.: 56233 Max.: 118.00

ii. Use the pairs() function to produce a scatterplot matrix of the first ten columns or variables of the data. Recall that you can reference the first ten columns of a matrix A using A[,1:10].

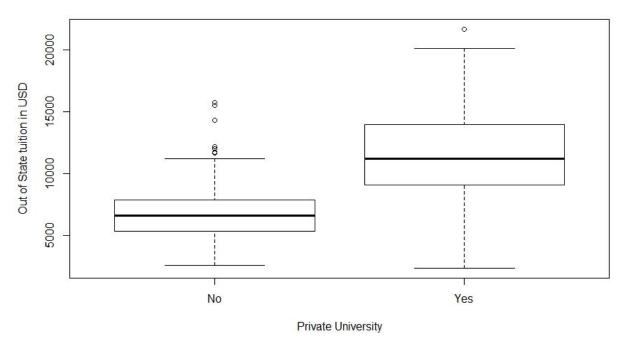
> pairs(College[1:10])



iii. Use the plot() function to produce side-by-side boxplots of Outstate versus Private.

> plot(College\$Private, College\$Outstate, xlab = "Private University", ylab = "Out of State tuition in USD", main = "Outstate Tuition Plot")

Outstate Tuition Plot



iv. Create a new qualitative variable, called **Elite**, by *binning* the **Top10perc** variable. We are going to divide universities into two groups based on whether or not the proportion of students coming from the top 10% of their high school classes exceeds 50%.

```
> Elite =rep ("No",nrow(college ))
```

- > Elite [college\$Top10perc >50]=" Yes"
- > Elite =as.factor (Elite)
- > college =data.frame(college,Elite)

Use the summary() function to see how many elite universities there are. Now use the plot() function to produce side-by-side boxplots of Outstate versus Elite.

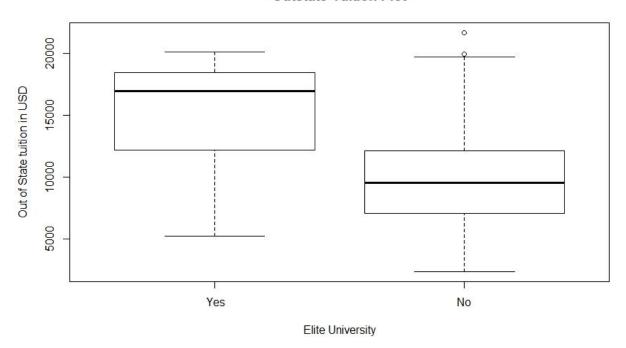
```
> Elite =rep ("No",nrow(college ))
```

- > Elite [college\$Top10perc >50]=" Yes"
- > Elite =as.factor (Elite)
- > college\$Elite <- Elite
- > summary(college\$Elite)

Yes No 78 699

> plot(college\$Elite, college\$Outstate, xlab = "Elite University", ylab = "Out of State tuition in US D", main = "Outstate Tuition Plot")

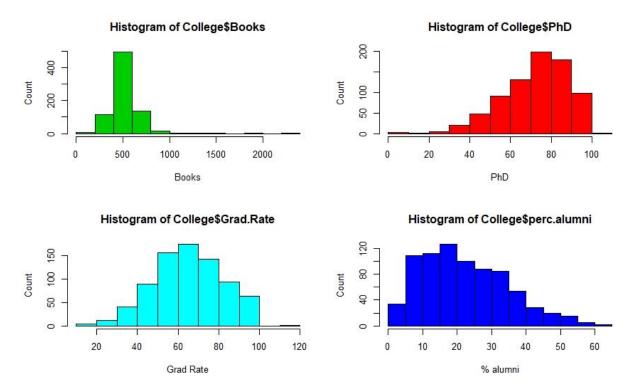
Outstate Tuition Plot



v. Use the hist() function to produce some histograms with differing numbers of bins for a few of the quantitative variables. You may find the command par(mfrow=c(2,2)) useful: it will divide the print window into four regions so that four plots can be made simultaneously. Modifying the arguments to this function will divide the screen in other ways.

```
> par(mfrow = c(2,2))
```

- > hist(college\$Books, col = 3, xlab = "Books", ylab = "Count")
- > hist(college\$PhD, col = 2, xlab = "PhD", ylab = "Count")
- > hist(college\$Grad.Rate, col = 5, xlab = "Grad Rate", ylab = "Count")
- > hist(college\$perc.alumni, col = 4, xlab = "% alumni", ylab = "Count")



vi. Continue exploring the data, and provide a brief summary of what you discover.

> summary(college\$Books)

Min. 1st Qu. Median Mean 3rd Qu. Max. 96.0 470.0 500.0 549.4 600.0 2340.0

- > weird.books <- college[College\$Books == 96,]
- > nrow(weird.books)

[1] 1

- > row.names(weird.books)
- [1] "Appalachian State University"

It is weird that the university has a min of only 96 books, Appalachian State University is one of them . A university having only 96 books is weird because there are several streams and the number of books per stream is a big number, so this little number is rare.