BUS 41201 Homework 3 Assignment

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Amazon Reviews

The dataset consists of 13 319 reviews for selected products on Amazon from Jan-Oct 2012. Reviews include product information, ratings, and a plain text review. The data consists of three tables:

##Review subset.csv is a table containing, for each review, its

- ProductId: Amazon ASIN product code
- UserId: ID of the reviewer
- Score: numeric 1-5 (the number of stars)
- Time: date of the review
- Summary: review summary in words
- Nrev: number of reviews by the user
- Length: number of words in the review
- Prod Category: Amazon product category
- Prod Group: Amazon product group

Word freq.csv

is a simple triplet matrix of word counts from the review text including

- Review ID: the row index of Review subset.csv
- Word ID: the row index of words.csv
- Times Word: how many times the word occurred in the review

Words.csv

contains 1125 alphabetically ordered words that occur in the reviews.

```
library(knitr) # library for nice R markdown output

# READ REVIEWS

data<-read.table("Review_subset.csv",header=TRUE)
dim(data)</pre>
```

[1] 13319 9

```
# 13319 reviews
# ProductID: Amazon ASIN product code
# UserID: id of the reviewer
# Score: numeric from 1 to 5
# Time: date of the review
# Summary: text review
# nrev: number of reviews by this user
# Length: length of the review (number of words)
# READ WORDS
words<-read.table("words.csv")
words<-words[,1]
length(words)</pre>
```

[1] 1125

```
#1125 unique words

# READ text-word pairings file

doc_word<-read.table("word_freq.csv")
names(doc_word)<-c("Review ID","Word ID","Times Word")

# Review ID: row of the file Review_subset

# Word ID: index of the word

# Times Word: number of times this word occurred in the text</pre>
```

Question 1

source("naref.R")

We want to build a predictor of customer ratings from product reviews and product attributes. For these questions, you will fit a LASSO path of logistic regression using a binary outcome:

$$Y = 1$$
 for 5 stars (1)

$$Y = 0$$
 for less than 5 stars. (2)

Fit a LASSO model with only product categories. The start code prepares a sparse design matrix of 142 product categories. What is the in-sample R2 for the AICc slice of the LASSO path? Why did we use standardize FALSE? (1 point)

```
# Let's define the binary outcome

# Y=1 if the rating was 5 stars

# Y=0 otherwise

Y<-as.numeric(data$Score==5)

# (a) Use only product category as a predictor

library(gamlr)

## Loading required package: Matrix</pre>
```

```
# Cast the product category as a factor
data$Prod_Category<-as.factor(data$Prod_Category)</pre>
class(data$Prod_Category)
[1] "factor"
# Since product category is a factor, we want to relevel it for the LASSO.
# We want each coefficient to be an intercept for each factor level rather than a contrast.
# Check the extra slides at the end of the lecture.
# look inside naref.R. This function relevels the factors for us.
data$Prod_Category<-naref(data$Prod_Category)</pre>
# Create a design matrix using only products
products<-data.frame(data$Prod_Category)</pre>
x_cat<-sparse.model.matrix(~., data=products)[,-1]</pre>
# Sparse matrix, storing 0's as .'s
# Remember that we removed intercept so that each category
# is standalone, not a contrast relative to the baseline category
colnames(x cat)<-levels(data$Prod Category)[-1]</pre>
# let's call the columns of the sparse design matrix as the product categories
# Let's fit the LASSO with just the product categories
lasso1<- gamlr(x_cat, y=Y, standardize=FALSE,family="binomial",</pre>
lambda.min.ratio=1e-3)
```

Question 2

Fit a LASSO model with both product categories and the review content (i.e. the frequency of occurrence of words). Use AICc to select lambda. How many words were selected as predictive of a 5 star review? Which 10 words have the most positive effect on odds of a 5 star review? What is the interpretation of the coefficient for the word 'discount'? (3 points)

```
[1] 13319 1125
x_cat2<-cbind(x_cat,spm)
lasso2 <- gamlr(x_cat2, y=Y,lambda.min.ratio=1e-3,family="binomial")</pre>
```

Question 3

Continue with the model from Question 2. Run cross-validation to obtain the best lambda value that minimizes OOS deviance. How many coefficients are nonzero then? How many are nonzero under the 1se rule? (1 point)

fold 1,2,3,4,5,done.