NLP Based Question & Answers for Ecommerce

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1. Pre-Introduction

Project was done in

• Operating System: - Windows10

• **Processor**: - Intel core i7

RAM: -16 GB

System Type: -64-bit operating System, x64-based processor

Software Used:- R Version 3.6.1 (architecture x86_64).

Please install the required library if not already installed in R as shown below

```
install.packages(c("XML", "sqldf", "dplyr", "tm", "tidyverse", "tidytext", "textclean", "
qdapRegex", "hunspell", "textstem", "DataCombine", "stringr", "sentimentr", "magrittr", "
NLP", "lubridate", "caret", "rlist", "shiny", "DT", "ggplot2", "wordcloud", "reshape2", "
reshape2", "ggraph", "ggforce", "igraph", "corrplot", "PerformanceAnalytics"))
```

Sequence of Code file to run

- 01.EDA.R
- 02.EDA.R
- 03.Visualization.R
- 04.Function for Text Analysis.R
- 05.Model.R
- 06.Shiny.R

Please change file path present in starting line of code to the path where you would be storing our folders. Please modify only highlighted path shown below only on files 01.EDA.r, 02.EDA.r, 05.Model.r:

```
library("tidyverse")
library("tidytext")
library("tidytext")
library("lubridate")

data3<-read.csv(

E:/Data Science/00.Excelr/PROJECT/Final Data with Text mining in r/00. Latest Final Script/
Data/data4_withNumbers.csv', stringsAsFactors = summary(data3)
```

2. Business Objective

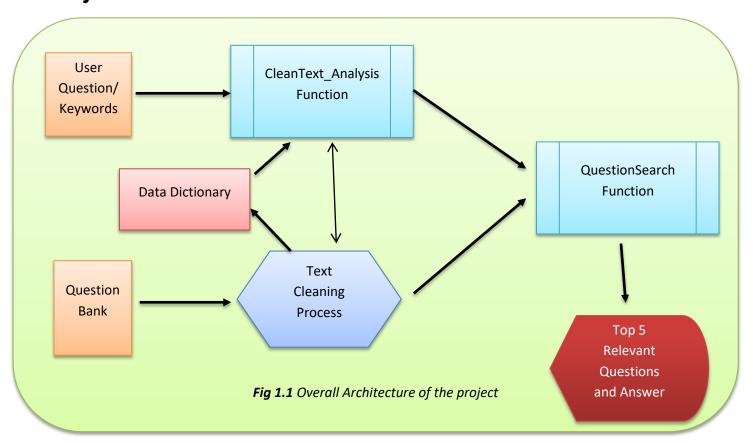
In any website or in ecommerce business there are very few FAQ or limited user review. All the questions cannot be shown in website as website will look clumsy. Due to limited exposure of the questions, there is a possibility that user might not find question and the answer He/ She is searching for or the question is not present or never have been asked.

Considering different scenarios, we are proposing a model which will take user question and will bring out relevant top five question and answer from Question Bank.

Our model ensures that

- i. User will get relevant answer for the question he/ she is searching for
- ii. Website will not be clumsy
- iii. Since all the process is happening at server side and only top 5 result is achieved, website will not be clumsy.

3. Project Architecture



3.1 Input Layer

User Question and Question bank both are considered as input.

- **User question/Keywords** is the text or question that user will provide as an input.
- Question bank means the existing question and answer set that is already
 present. Later User question will be compared with this question bank to find
 relevant Question and answer.

3.2 **Preprocessing**

Before considering taking text as input, text needs to be cleaned. Process of cleaning text includes:

- Removal of Stop words (common words occur in a sentence)
- Elimination of URLs (with https or ftp [absolute URLs] and broken URLs (without https or ftp]).
- Exclusion of emails.
- Deletion unwanted words
- Word spell check and correction
- Word stemming

3.3 Model

Model is a function "QuestionSearch" which computes cosine similarity between user Questions/Keywords and each Question bank to find the similar questions.

3.4 Final Output

Final Output is the relevant question and answer shown based on the user input/keywords. Top five results will be shown with decreasing order of cosine similarity.

4. Question Bank Data

Data was provided in **NDJson** format.

```
'questionType': 'yes/no', 'asin': '0594033926', 'answerTime':
'Dec 27, 2013', 'unixTime': 1388131200, 'question': 'Is this
cover the one that fits the old nook color? Which I believe is
8x5.', 'answerType': 'Y', 'answer': 'Yes this fits both the
nook color and the same-shaped nook tablet'}
{'questionType': 'yes/no', 'asin': '0594033926', 'answerTime':
'Jan 5, 2015', 'unixTime': 1420444800, 'question': 'Does it
fit Nook GlowLight?', 'answerType': 'N', 'answer': 'No. The
nook color or color tablet'}
{'answer': "I don't think so. The nook color is 5 x 8 so not
sure anything smaller would stay locked in, but would be
close.", 'asin': '0594033926', 'answerTime': '2 days ago',
'question': 'Would it fit Nook 1st Edition? 4.9in x 7.7in ?',
'questionType': 'open-ended'}
{'questionType': 'yes/no', 'asin': '0594033926', 'answerTime':
'17 days ago', 'question': "Will this fit a Nook Color that's
5 x 8?", 'answerType': 'Y', 'answer': 'yes'}
{'questionType': 'yes/no', 'asin': '0594033926', 'answerTime':
'Feb 10, 2015', 'unixTime': 1423555200, 'question': 'will this fit the Samsung Galaxy Tab 4 Nook 10.1', 'answerType': 'N',
'answer': "No, the tab is smaller than the 'color'"}
{'questionType': 'yes/no', 'asin': '0594033926', 'answerTime':
'Jan 30, 2015', 'unixTime': 1422604800, 'question': 'does it
have a flip stand?', 'answerType': 'N', 'answer': 'No, there
is not a flip stand. It has a pocket in the front flap. It is
a very nice cover.'}
```

Fig 3.1 shows a portion of sample data in qa_electronics.json

Data was converted into csv format in python and stored as file bearing name "data_file.csv". Python code is present file "qa_Electronics to Data.csv.ipynb".

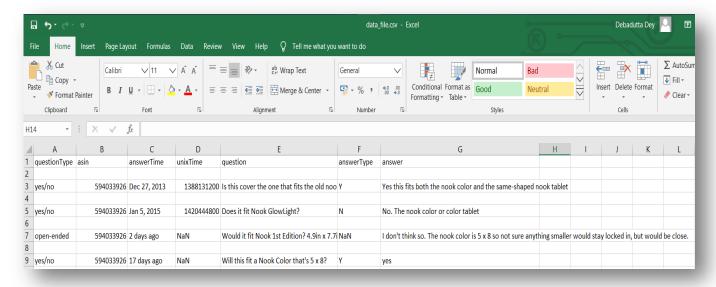


Fig 3.2 shows a portion of sample data in data file.csv

Dataset bearing name "data_file.csv" was loaded into R global environment.

```
data=read.csv('data_file.csv', stringsAsFactors = FALSE)
```

Checked and converted the data into correct data type.

- Converted **unixtime** from Posixct to GMT.
- Converted questiontype to factor datatype
- Converted **asin** to factor datatype
- Converted **answerType** to factor datatype

```
data$unixTime <- as.POSIXct(as.numeric(data$unixTime), origin = '1970-01-01', tz = 'GMT')
data$questionType<-as.factor(data$questionType)
data$asin <-as.factor(data$asin )
data$answerType<-as.factor(data$answerType)</pre>
```

Final Data in R

*	questionType	asin ‡	answerTime	unixTime ‡	question	answerType ‡	answer ‡
	yes/no	0594033926	Dec 27, 2013	2013-12-27 08:00:00	Is this cover the one that fits the old nook color? Which I bel	Υ	Yes this fits both the nook color and the same-shaped nook
	yes/no	0594033926	Jan 5, 2015	2015-01-05 08:00:00	Does it fit Nook GlowLight?	N	No. The nook color or color tablet
	open-ended	0594033926	2 days ago	NA	Would it fit Nook 1st Edition? 4.9in x 7.7in ?	Nan	I don't think so. The nook color is 5 x 8 so not sure anything
4	yes/no	0594033926	17 days ago	NA	Will this fit a Nook Color that's 5 x 8?	Υ	yes
	yes/no	0594033926	Feb 10, 2015	2015-02-10 08:00:00	will this fit the Samsung Galaxy Tab 4 Nook 10.1	N	No, the tab is smaller than the 'color'
	yes/no	0594033926	Jan 30, 2015	2015-01-30 08:00:00	does it have a flip stand?	N	No, there is not a flip stand. It has a pocket in the front flap
	yes/no	0594033926	Jan 30, 2015	2015-01-30 08:00:00	does this have a flip stand		Hi, no it doesn't
8	open-ended	0594033926	Dec 22, 2014	2014-12-22 08:00:00	also fits the HD+?	Nan	It should. They are the same size and the charging port is in
9	yes/no	0594033926	Nov 16, 2014	2014-11-16 08:00:00	Does it have 2 positions for the reader? Horizontal/vertical T	Υ	Yes

Columns Details with their datatype

Column Name	Column Description	Column Datatype
QuestionType	It conveys if the question is open ended or yes/No type of question	Character, Factor
Asin	Product id of a product. Uniquely identifies the product	Character, Factor
AnswerTime	It contains date and time. It records the answer date	Character
UnixTime	It contains date and time. It records the answer date in unix	
	format	
Question	Contains Questions	Character
AnswerType	Conveys if answer is yes or no and for open ended question it is	Character, Factor
	blank. Since it was converted from python null values are	
	represented by NaN	
Answer	Contains Answer	Character

5. Exploratory Data Analysis

i) One Hot Encoding

 Column QuestionType represents if questiontype is open ended or yes or no. It is factor datatype and hence performed one hot encoding on questionType and created a new column "QuestionTypeOE".

```
## One hot Encoding with QuestionType
data1<-sqldf(c("alter table data1 add column questiontype0E bit","select * from data1"))
data1<- sqldf(c("update data1 set questiontype0E=1 where questiontype='open-ended'","select * from data1"))
data1<- sqldf(c("update data1 set questiontype0E=0 where questiontype<>'open-ended'","select * from data1"))
```

ii) Removing similar columns

 UnixTime was provided in posixct format. After converting the unixtime to GMT, found that data in AnswerTime is same as data in UnixTime. Hence removed AnswerTime column.

iii) Imputing Null Values by Median Values

- Null Values were present in **UnixTime** Columns.
- o Imputing the null values with median values of that particular product (asin).

```
data1$unixTime<-as.Date(data1$unixTime , format = "%m/%d/%y") #mdy(data4$unixTime)

data2<-sqldf("select asin,median(unixTime) MedianDate from data1 group by asin")
data2$MedianDate<- as.Date(data2$MedianDate, origin='1970-01-01')

data1<- merge(data1,data2,by="asin")</pre>
```

 After calculating median values by product, still some records have null and had to impute with median value of total dataset.

```
sqldf("select * from data1 where asin in ('B0007XD4LC','B000VNJD1S')")
sqldf("select median(unixTime) from data1") #16126
as.Date(16126, origin='1970-01-01') #"2014-02-25"

data1<-sqldf(c("update data1 set MedianDate ='2014-02-25' where asin in ('B0007XD4LC','B000VNJD1S') and MedianDate is null", "select * from data1"))
data1<-sqldf(c("update data1 set unixTime=MedianDate where unixTime is null", "select * from data1"))
data1<-data1[, !(colnames(data1) %in% c("MedianDate"))]</pre>
```

iv) Creating new columns

 Created new columns which contain the Question length, Question Word Count, Answer length, Answer Word Count.

```
data1$QuestionLength<-nchar(data1$question)
data1$AnswerLength<-nchar(data1$answer)

library(stringr)
data1$questionWCount<-str_count(data1$question,'\\w+')
data1$answerWCount<-str_count(data1$answer,'\\w+')</pre>
```

 Creating new columns from the sentiment, average sentiment and standard deviation of question and answer column.

```
library(sentimentr)
summary(data1)

questionSentiment<-sentiment_by(data1$question, by = NULL)
answerSentiment<-sentiment_by(data1$answer, by = NULL)

data1$questionAve_Sentiment<-questionSentiment$ave_sentiment
data1$answerAve_Sentiment<-answerSentiment$ave_sentiment

data1$questionsd<-questionSentiment$sd
data1$answersd<-answerSentiment$sd
```

Replacing the na values with median

```
sqldf("select median(questionsd) from data1") #0.1020621
sqldf("select median(answersd) from data1") #0.1833333

#median of questionsd is 0.1020621. Imputing this with na values.
data1<-sqldf(c("update data1 set questionsd = 0.1020621 where questionsd is null", "select * from data1"))

#median of answersd is 0.1833333. Imputing this with na values.
data1<-sqldf(c("update data1 set answersd = 0.1833333 where answersd is null", "select * from data1"))</pre>
```

Creating new columns from bing sentiment.

```
data2<-data1 %>% unnest_tokens(word, question) %>%
  inner_join(get_sentiments("bing")) %>% # pull out only sentiment words
  count(rownum,sentiment) %>% # count the # of positive & negative words
  spread(sentiment, n, fill = 0) %>% # made data wide rather than narrow
  mutate(sentiment = positive - negative) # # of positive words - # of negative words
  colnames(data2)<-c("rownum","negativeQ","positiveQ","sentimentQ")</pre>
```

Na values in negetiveQ, PositiveQ, sentiment is replaced by 0

```
data3<-merge(x = data1, y = data2, by = "rownum", all.x = TRUE)

data3$negativeQ[is.na(data3$negativeQ)] <- 0
data3$positiveQ[is.na(data3$positiveQ)] <- 0
data3$sentimentQ[is.na(data3$sentimentQ)] <- 0</pre>
```

Same steps were performed for answerwords too.

```
data3<-data2 %>% unnest_tokens(word, answer) %>%
  inner_join(get_sentiments("bing")) %>% # pull out only sentiment words
  count(rownum,sentiment) %>% # count the # of positive & negative words
  spread(sentiment, n, fill = 0) %>% # made data wide rather than narrow
  mutate(sentiment = positive - negative) # # of positive words - # of negative words
  colnames(data3)<-c("rownum","negativeA","positiveA","sentimentA")

data4<-merge(x = data4, y = data3, by = "rownum", all.x = TRUE)

data4$negativeA[is.na(data4$negativeA)] <- 0
  data4$positiveA[is.na(data4$positiveA)] <- 0
  data4$sentimentA[is.na(data4$sentimentA)] <- 0</pre>
```

 Removing negetiveQ,positiveQ,negetiveA and positiveA as sum of negetiveQ, positiveQ equals SentimentQ and sum of negetiveA and positiveA equals SentimentA

v) Text Analysis

- Loaded stop words from already created stop words list.
- Removed "yes" and "No" from stop words as it needs to be present in text for further analysis.

```
#removing stopwords in dataframe
stopwd <- read.table('stop.txt')
stopwd <-as.character(stopwd$V1)

#stopwd<-as.character(stopwd$stopwords())
stopwd<-gsub('no',' ',stopwd)
stopwd<-gsub('yes',' ',stopwd)</pre>
```

- Created 2 new columns QuestionWords and AnswerWords in which preprocessing of text will be performed. This is done to retain the original question and answer in the data frame.
- Removing of all the URL's, email's, and words starting with @

```
## removing the Urls
datal$questionWords<-replace_url(datal$questionWords)
datal$answerWords<-replace_url(datal$answerWords)

datal$questionWords<-gsub("\\s*[^[:space:]/]+/[^[:space:]/]+","", datal$questionWords)
datal$answerWords<-gsub("\\s*[^[:space:]/]+/[^[:space:]/]+","", datal$answerWords)

##removing email
datal$questionWords<-gsub('\\S+@\\S+', ' ', datal$questionWords)
datal$answerWords<-gsub('\\S+@\\S+', ' ', datal$answerWords)

##removing words starting with @
datal$questionWords<-gsub('@\\S+', ' ', datal$questionWords)
datal$answerWords<-gsub('@\\S+', ' ', datal$answerWords)</pre>
```

o Removing the words which contain or end with .com, .org and .eu

```
##removing words containing .com in between and does not contain http or fpt
datal$questionWords<-gsub('\S+.com+\\s', ' ', datal$questionWords)
datal$questionWords<-gsub('\S+.eu+\\s', ' ', datal$questionWords)
datal$questionWords<-gsub('\S+.eu+\\s', ' ', datal$questionWords)

datal$answerWords<-gsub('\S+.com+\\s', ' ', datal$answerWords)
datal$answerWords<-gsub('\S+.org+\\s', ' ', datal$answerWords)
datal$answerWords<-gsub('\S+.eu+\\s', ' ', datal$answerWords)

##removing words ending with .com and does not contain http or fpt
datal$questionWords<-gsub('\S+\\.com', ' ', datal$questionWords)
datal$questionWords<-gsub('\S+\\.org', ' ', datal$questionWords)
datal$questionWords<-gsub('\S+\\.eu', ' ', datal$questionWords)
datal$answerWords<-gsub('\S+\\.com', ' ', datal$answerWords)
datal$answerWords<-gsub('\S+\\.org', ' ', datal$answerWords)
datal$answerWords<-gsub('\S+\\.org', ' ', datal$answerWords)
datal$answerWords<-gsub('\S+\\.org', ' ', datal$answerWords)
</pre>
```

Removing any words which come after or before /

```
##removing word/ and does not contain http or fpt
data1$questionWords<-gsub('\\S+/', ' ', data1$questionWords)
data1$answerWords<-gsub('\\S+/', ' ', data1$answerWords)

##removing /word and does not contain http or fpt
data1$questionWords<-gsub('/\\S+', ' ', data1$questionWords)
data1$answerWords<-gsub('/\\S+', ' ', data1$answerWords)</pre>
```

Removing the words who have combination of word&, word&word,
 &word

```
##removing word&word nd does not contain http or fpta
datal$questionWords<-gsub('\S+&amp;\\S+', ' ', datal$questionWords)
datal$answerWords<-gsub('\S+&amp;\\S+', ' ', datal$answerWords)

#replace_html(datal$questionWords,FALSE)

##removing word&amp; and does not contain http or fpt
datal$questionWords<-str_replace_all(datal$questionWords, "&amp;", " ")
datal$answerWords<-str_replace_all(datal$answerWords, "&amp;", " ")

##removing word&word
datal$questionWords<-gsub('\S+&\\S+', ' ', datal$questionWords)
datal$answerWords<-gsub('\S+&\\S+', ' ', datal$questionWords)

##removing &word;
datal$questionWords<-gsub('&\\S+;', ' ', datal$questionWords)
datal$answerWords<-gsub('\S\S+;', ' ', datal$questionWords)

datal$answerWords<-gsub('\S\S+;', ' ', datal$questionWords)

datal$answerWords<-gsub('\S\S+;', ' ', datal$questionWords)</pre>
```

 Removing any combination of character and number as a word. It is generally used in serial number or model name.

```
##removing combination of character and number as a word. It is genearlly serial number or model name.
datal$questionWords<-gsub('[a-z]+[0-9]+\\S+', ' ', datal$questionWords)
datal$answerWords<-gsub('[a-z]+[0-9]+\\S+', ' ', datal$answerWords)

##removing word-number
datal$questionWords<-gsub('\\S+\\-+[0-9]+', ' ', datal$questionWords)
datal$answerWords<-gsub('\\S+\\-+[0-9]+', ' ', datal$answerWords)

##removing number-word
datal$questionWords<-gsub('[0-9]+-\\S+', ' ', datal$questionWords)
datal$answerWords<-gsub('[0-9]+-\\S+', ' ', datal$questionWords)

##removing word#
datal$questionWords<-gsub('([a-z]+#)', ' ', datal$questionWords)
datal$answerWords<-gsub('([a-z]+#)', ' ', datal$questionWords)
datal$answerWords<-gsub('([a-z]+#)', ' ', datal$questionWords)</pre>
```

Removing the stop words, numbers and punctuations.

```
## Removing stop words
datal$questionWords <- removeWords(datal$questionWords, stopwd)
datal$answerWords <- removeWords(datal$answerWords, stopwd)

#Removing numbers
datal$questionWords <- removeNumbers(datal$questionWords)
datal$answerWords <- removeNumbers(datal$answerWords)

#Removing punctuation mark
datal$questionWords<-removePunctuation(datal$questionWords)
datal$answerWords<-removePunctuation(datal$answerWords)</pre>
```

 Removing single letter words, whitespace and certain word which were not removed by the stop words.

```
#remove single letter words
datal$questionWords <- rm_nchar_words(datal$questionWords, "1,1")
datal$answerWords <- rm_nchar_words(datal$answerWords, "1,1")

#removing whitespace
datal$questiDonWords <- stripWhitespace(datal$questionWords)
datal$answerWords <- stripWhitespace(datal$answerWords)

# replacing certain words
datal$questionWords<-str_replace_all(datal$questionWords, " wi fi ", " wifi ")
datal$questionWords<-str_replace_all(datal$questionWords, " aca ", " ")
datal$questionWords<-str_replace_all(datal$questionWords, " & ", "and")
datal$questionWords<-str_replace_all(datal$questionWords, " aaaaaarge ", " ")

datal$answerWords<-str_replace_all(datal$answerWords, " aca ", " ")
datal$answerWords<-str_replace_all(datal$answerWords, " aca ", " ")
datal$answerWords<-str_replace_all(datal$answerWords, " aca ", " ")
datal$answerWords<-str_replace_all(datal$answerWords, " aaaaaarge ", " ")
datal$answerWords<-str_replace_all(datal$answerWords, " aaaaaarge ", " ")</pre>
```

Word stemming using the "textstem" package

```
## word stemming
library(textstem)
data1$questionWords<-textstem::lemmatize_strings(data1$questionWords)
data1$answerWords<-textstem::lemmatize_strings(data1$answerWords)</pre>
```

- Created a word dictionary which contains wrong word and correct words using hunspell package.
 - Collecting list of words with incorrect spelling

```
# vector of words to replace
wrong <- unlist(hunspell(data1$questionWords))</pre>
```

 Creating a data frame of incorrect word with its first suggested word as correct word.

```
# vector of the first suggested words
correct <- data.frame(wrong,sapply(wrong, function(x) hunspell_suggest(x)[[1]][1]))
colnames(correct)<-c("wrong","correct")</pre>
```

Repeating the same for the answer column

```
# vector of words to replace
wrongA <- unlist(hunspell(data1$answerWords))

# vector of the first suggested words
correctA <- data.frame(wrongA,sapply(wrongA, function(x) hunspell_suggest(x)[[1]][1]))
colnames(correctA)<-c("wrong","correct")</pre>
```

Below mentioned words need not be replaced as either they are name of company or name of the product and also combining the dictionary formed from question and answer into 1 dictionary.

 Calculating the weight between two words by jarowinkler distance. It shows how much two words are close. If distance =1 it means wrong and correct word are same word.

```
library(RecordLinkage)
correct2<-mutate(correct2, weight = jarowinkler(correct,wrong))
summary(correct2)</pre>
```

 Manually correcting the incorrect words as they were provided with incorrect words and also increasing their weight to 91.

```
correct2<-sqldf(c("update correct2 set correct='sony' where wrong='soni'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set correct='netflix' where wrong='netflix'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set correct='of the' where wrong='soni'', "select * from correct2"))
correct2<-sqldf(c("update correct2 set correct=' the' where wrong='cosina'', "select * from correct2"))
correct2<-sqldf(c("update correct2 set correct=' ' where wrong='arrrgh'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='arrrgh'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='anaaaarge'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nrvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='nvida'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='isawrap'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='sawrap'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='sawrap'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set weight=91 where wrong='sawrap'", "select * from correct2"))
correct2<-sqldf(c("update correct2 set we
```

 Creating a cutoff value (0.971428571428571) to keep most significant incorrect and correct words in data dictionary and final dictionary is created and stored in correct3.csv.

```
## final table of wrong and correct words
correct3<-sqldf("select * from correct2 where weight>=0.971428571428571")
correct3$correct<-tolower(correct3$correct)</pre>
```

 Replacing the wrong words with correct words created in data dictionary for questionWords and answerwords using the **Findreplace** function in **DataCombine** package.

This whole text analysis is converted into a function (CleanText_Analysis)
 except for creating data dictionary. Data dictionary already created is used in
 this function.



answerWords	QuestionLength	Answerlength	questionWCount	* answerWCount	questionAve_Sentiment	answerAve_Sentiment	questioned '	answersd	sentimentQ *
yes fit nook color shape nook ttablet	75	65	16	13	0.00000000	0.2218800785	0.000000000	0.183333300	0
no nook color color ttablet	27	34	5	7	0.17888544	0.0000000000	0.102062100	0.000000000	0
nook color not small stay lock close	46	112	11	25	0.08911020	-0.3038635721	0.115470054	0.393750000	0
yes	40	3	11	1	0.14142136	0.8000000000	0.102062100	0.183333300	0
no tab small color	48	39	11	8	0.14142136	0.0000000000	0.102062100	0.183333300	0

vi) Finding the value of ? in AnswerType column

- All the next steps of processing the dataset will be done in "DATA3".
- Changing the Unix time format into YYYY/MM/DD, factorizing QuestionTypeOE, AnswerType and Asin.

```
data3$unixTime<-as.Date(data3$unixTime , format = "%Y-%m-%d")
data3$questiontype0E<-as.factor(data3$questiontype0E)
data3$answerType<-as.factor(data3$answerType)
data3$asin<-as.factor(data3$asin)</pre>
```

- Divided the dataset into 2 parts
 - Data220 containing answertype where "?" value is not present
 - Data2201 containing answertype where "?" value is present

Now creating a new dataset with only the columns that are needed.

Changing the variables of answer type: NO to N, YES to Y, NaN to NotApp.

```
data221<-sqldf(c("update data221 set answerType='No' where answerType='N'", "select * from data221"))
data221<-sqldf(c("update data221 set answerType='Yes' where answerType='Y'", "select * from data221"))
data221<-sqldf(c("update data221 set answerType='NotApp' where answerType='NaN'", "select * from data221"))
data221$answerType<-as.factor(as.character(data221$answerType))</pre>
```

Creating a normalize function and normalizing all the columns.

```
normalize <- function(x)</pre>
  return ((x - min(x)) / (max(x) - min(x)))
data221 $QuestionLength<-normalize(data221 $QuestionLength)
data221$AnswerLength<-normalize(data221$AnswerLength)
data221$questionWCount<-normalize(data221$questionWCount)
data221\sanswerWCount<-normalize(data221\sanswerWCount)
data221$questionAve_Sentiment<-normalize(data221$questionAve_Sentiment)</pre>
data221$answerAve_Sentiment<-normalize(data221$answerAve_Sentiment)</pre>
data221$questionsd <-normalize(data221$questionsd )</pre>
data221$answersd<-normalize(data221$answersd)</pre>
data221$negativeQ<-normalize(data221$negativeQ)</pre>
data221$positiveQ<-normalize(data221$positiveQ)</pre>
data221$sentimentQ <-normalize(data221$sentimentQ )</pre>
data221$negativeA<-normalize(data221$negativeA)</pre>
data221$positiveA<-normalize(data221$positiveA)</pre>
data221$sentimentA<-normalize(data221$sentimentA)</pre>
```

Checking the distribution in original and partitioned data

```
#Checking distibution in origanl data and partitioned data
prop.table(table(data220$answerType)) * 100
```

- Now we will do sampling using the Caret package.
- Dividing the data into train and test datasets on basis of AnswerType.

Here we will use all the different types of sampling: Normal, Down, Up and
 Smote and select the one which has the best output.

```
ctrl <- trainControl(method = "cv", number
                       summaryFunction = multiClassSummary,
                       classProbs = TRUE
model_rf1 <- caret::train(answerType ~ .,
                             data = train,
method = "rf",
metric="ROC",
                             trControl = ctrl
ctrl$sampling <- "down"
model_rf_down <- caret::train(answerType ~ .,
                                 data = train,
method = "rf",
metric="ROC",
                                  trControl = ctrl)
ctrl$sampling <- "up"
model_rf_up <- caret::train(answerType ~ .,
                               data = train,
                               method = "rf
                               metric="ROC"
                               trControl = ctrl)
ctrl$sampling <- "smote"
model_rf_smote <- caret::train(answerType ~ .,
                                        data = train,
method = "rf"
                                        metric="ROC",
                                        trControl = ctrl)
```

Preparing a confusion matrix of all the models

```
caret::confusionMatrix(predict(model_rf1,test[1:(length(test)-1)]), test$answerType)
caret::confusionMatrix(predict(model_rf_down,test[1:(length(test)-1)]),test$answerType)
caret::confusionMatrix(predict(model_rf_up,test[1:(length(test)-1)]),test$answerType)
caret::confusionMatrix(predict(model_rf_smote,test[1:(length(test)-1)]),test$answerType)
```

Printing all the values on which to select the model.

```
print(model_rf1)
print(model_rf_down)
print(model_rf_up)
print(model_rf_smote)
```

 Smote was selected because it had the most balanced accuracy (Comparison of data sampling.xlsx)

<u>Smote</u>					<u>Normal</u>					<u>Down</u>					up				
Reference					Reference					Reference					Reference				
Prediction No NotApp Yes					Prediction No NotApp Yes					Prediction No NotApp Yes					Prediction No NotApp Yes				
No 1047 0 649					No 833 0 149					No 984 0 462					No 829 0 167				
NotApp 0 895	3 0				NotApp 0 8953 0					NotApp 0 8953 0					NotApp 0 8953 0				
Yes 157	0 2552				Yes 371 0 3052					Yes 220 0 2739					Yes 375 0 3034				
Overall Statistics					Overall Statistics					Overall Statistics					Overall Statistics				
Accura	cy: 0.9397				Accuracy : 0.9611					Accuracy : 0.9489					Accuracy : 0.9594				
95%	CI : (0.9355,	0.9436)			95% (I: (0.9577	0.9643)			95% CI : (0.9451, 0.9526)					95% CI : (0.9559, 0.9627)				
No Information Ra	te : 0.6702				No Information Rate : 0.6702					No Information Ra	te : 0.6702				No Information Rate : 0.6702				
P-Value [Acc > NI	R] : < 2.2e-1	L6			P-Value [Acc > NIR] : < 2.2e-16					P-Value [Acc > NIR] : < 2.2e-16					P-Value [Acc > NIR] : < 2.2e-16				
Kap	pa : 0.877				Kappa : 0.9194					Kappa : 0.8954					Kappa : 0.916				
Mcnemar's Test P-Val	ue : NA				Mcnemar's Test P-Value : NA					Mcnemar's Test P-Value : NA					Mcnemar's Test P-Value : NA				
Statistics by Class:					Statistics by Class:					Statistics by Class:					Statistics by Class:				
		ass: NotApp C					uss: NotApp C				Class: No Cla						ass: NotApp C		
Sensitivity	0.86960	1.0000	0.7973		Sensitivity	0.69186	1.0000	0.9535		Sensitivity	0.81728	1.0000	0.8557		Sensitivity	0.68854	1.0000	0.9478	
Specificity	0.94660	1.0000	0.9845		Specificity	0.98774	1.0000	0.9635		Specificity	0.96199	1.0000	0.9783		Specificity	0.98626	1.0000	0.9631	
Pos Pred Value	0.61733	1.0000	0.9420		Pos Pred Value	0.84827	1.0000	0.8916		Pos Pred Value	0.68050	1.0000	0.9257		Pos Pred Value	0.83233	1.0000	0.8900	
Neg Pred Value	0.98654	1.0000	0.9391		Neg Pred Value	0.97002	1.0000	0.9850		Neg Pred Value	0.98153	1.0000	0.9556		Neg Pred Value	0.96967	1.0000	0.9832	
Prevalence	0.09013	0.6702	0.2396		Prevalence	0.09013	0.6702	0.2396		Prevalence	0.09013	0.6702	0.2396		Prevalence	0.09013	0.6702	0.2396	
Detection Rate	0.07838	0.6702	0.1910		Detection Rate	0.06236	0.6702	0.2285		Detection Rate	0.07366	0.6702	0.2050		Detection Rate	0.06206	0.6702	0.2271	
Detection Prevalence		0.6702	0.2028		Detection Prevalence	0.07351	0.6702	0.2563		Detection Prevalence	0.10825	0.6702	0.2215		Detection Prevalence	0.07456	0.6702	0.2552	
Balanced Accuracy	0.90810	1.0000	0.8909		Balanced Accuracy	0.83980	1.0000	0.9585		Balanced Accuracy	0.88963	1.0000	0.9170		Balanced Accuracy	0.83740	1.0000	0.9555	

Fig 4.1 shows the output of different data sampling

- Now time to select which model is accurate for the classification of AnswerType.
- Different models like GLMNET, SVM RADIAL, KNN, NAÏVE BAYES, CART, C5.0, BAGGED CART and RANDOM FOREST.

Storing the results

Model RANDOM FOREST was selected.

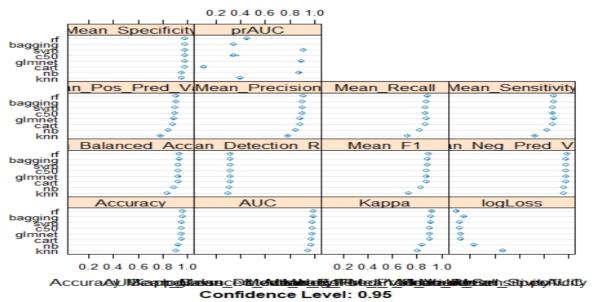


Fig 4.2 shows comparison of different methods

Variable importance of Random forest

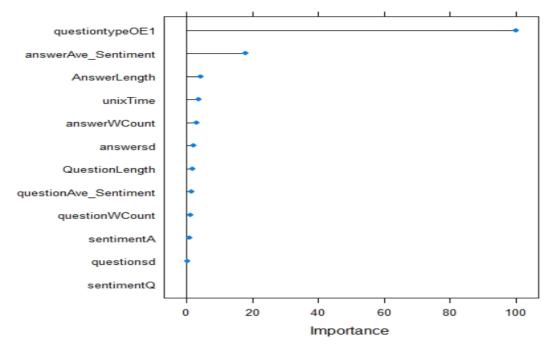


Fig 4.3 shows the columns and their importance

Got the accuracy of **0.9611** or **96.11%**

Now implementing these models on the test dataset

```
data2202
data2201
finalData1
- sqldf ("select rownum, unixTime, questiontype0E, question, answer, questionWords, answerWords, QuestionLength, AnswerLength, questionWCount, answerWCount, questionAve_Sentiment, answerAve_Sentiment, questionsd, answersd, negative0, positive0, sentiment0, negativeA, positiveA, sentimentA, answerType
from data220
union
select rownum, unixTime, questiontype0E, question, answer, questionWords, answerWords, QuestionLength, AnswerLength, questionWCount, answerWcount, questionAve_SentimentA, answerType
from data220
union
select rownum, unixTime, questiontype0E, question, answer, questionWords, answerWords, QuestionLength, AnswerLength, questionWCount, answerWcount, questionAve_Sentiment, answerAve_Sentiment, questionSd, answersd, negative0, positive0, sentiment0, negativeA, positiveA, sentimentA, answerType
from data2201

FinalData<-FinalData1
FinalData<-FinalData1
FinalData<-FinalData1
FinalData<-FinalData1
FinalData<-FinalData1
FinalData<-sqldf(c("update FinalData set answerType='No' where answerType='N'', "select * from FinalData"))
FinalData<-sqldf(c("update FinalData set answerType='NotApp' where answerType='NaN'', "select * from FinalData"))
FinalData<-sqldf(c("update FinalData set answerType='NotApp' where answerType='NaN'', "select * from FinalData"))
FinalData<-sqldf(c("update FinalData set answerType='NotApp' where answerType='NaN'', "select * from FinalData"))
FinalData<-sqldf(c("update FinalData set answerType='NotApp' where answerType='NaN'', "select * from FinalData"))
FinalData<-sqldf(c("update FinalData set answerType='NotApp' where answerType='NaN'', "select * from FinalData"))
```

- The FINAL DATA looks like this: FinalData.csv
- Here are some visualization of the dataset

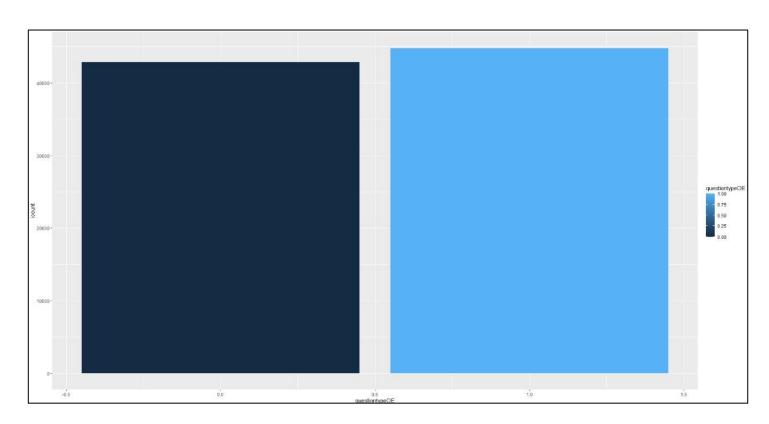


Fig 4.4 shows the columns and their importance

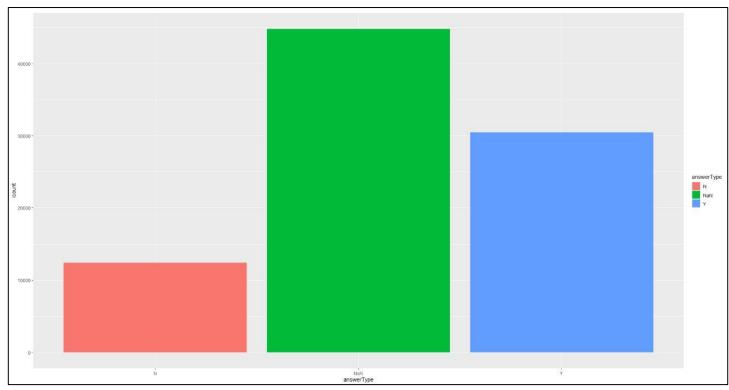
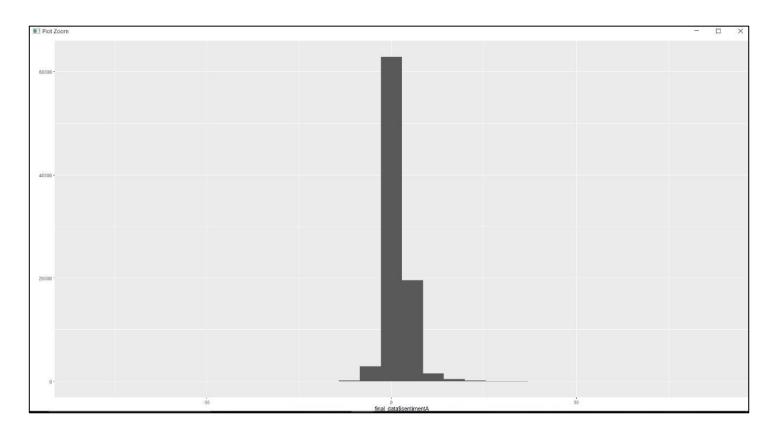


Fig 4.5 shows the histogram of the column answertype



 $\textit{Fig 4.6} \ \textit{shows the histogram of the column SENTIMENTA}$

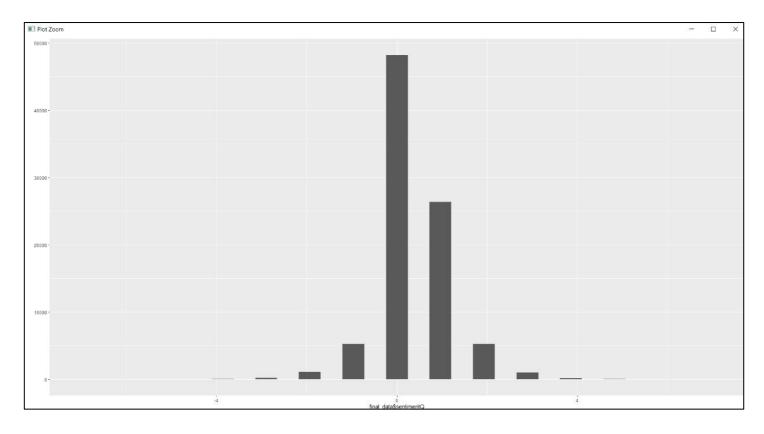


Fig 4.7 shows the histogram of the column SENTIMENTQ

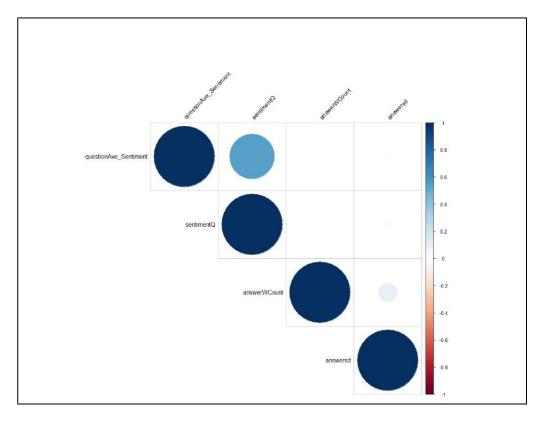


Fig 4.8 shows the sentimental analysis of question and answers

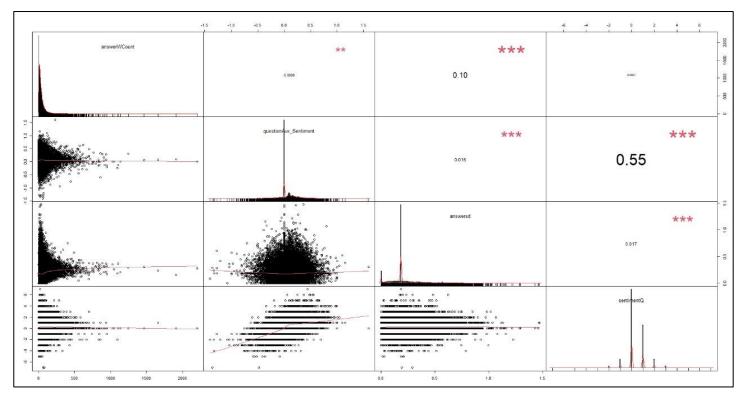


Fig 4.9 shows co-relation plot of all numeric columns

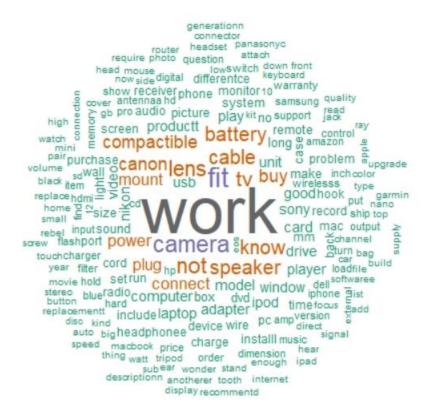


Fig 4.10 shows a word cloud of the question column

6. Model Building

Loading the necessary libraries and the Final Data.

```
library("XML")
library("sqldf")
library("dplyr")
library("tim")
library("tidyverse")
library("tidytext")
library("hunspell")

FinalData=read.csv('FinalData.csv', stringsAsFactors = FALSE)

str(FinalData)
summary(FinalData)

#FinalData$rownum<-as.numeric(FinalData$rownum)

FinalData$unixTime<-as.Date(FinalData$unixTime , format = "%Y-%m-%d")
#stopwords_en <- tibble("word" = stopwords("en", source = "smart"))

FinalData <- FinalData %>%
   mutate(rowIndex=as.numeric(row.names(.)))

str(FinalData)
```

 Loading stop words to correct user input and removing yes and no from the list

```
## Loading stopwords to correct user input
stopwd <- read.table("stop.txt")
stopwd <-as.character(stopwd$V1)

stopwd<-gsub('no',' ',stopwd)
stopwd<-gsub('yes',' ',stopwd)</pre>
```

Loading correct spellcheck for user input

```
## Loading correct spell check for user input
correct3=read.csv('correct3.csv', stringsAsFactors = FALSE)
```

- First the model will be explained individually and then the screenshot of the whole code is inserted
- The first thing to do is convert the Question Word column into list and get its length

```
QuestionList <- as.list(FinalData$questionWords)
QuestionLength <- length(QuestionList)</pre>
```

 As the model is created in the form of a function, it is necessary to define the function which is going to be based on the USER QUESTION

QuestionSearch <- function(UserQuestion)

The first thing the function is going to do is do a text analysis of the user question and do
a spell check of the same using the CleanText_analysis function that was created above
and the correct3 dataset as well

UserQuestion<-CleanText_Analysis(UserQuestion,correct3)</pre>

- Then it will combine the user question and question list into one called as total questions
- Then the function will create a corpus and remove the white space and stem the document

```
QuestionCorpus <- VCorpus(TotalQuestions) %>%
tm_map(stemDocument) %>%
tm_map(stripWhitespace)
```

Now it will create a term document matrix of the given corpus

- The next part of the code will first get converted into data frame by the tidy function and get grouped by the number of documents as per the term document matrix.
- The mutate function is used to create a new variable from a data set called vtrLen and it

```
TotalQuestions <- VectorSource(c(QuestionList, UserQuestion))
```

will select terms based on most occurring terms. This all process will get stored in a variable called term.doc.matrix.

```
term.doc.matrix <- tidy(term.doc.matrix.stm) %>%
  group_by(document) %>%
  mutate(vtrLen=sqrt(sum(count^2))) %>%
  mutate(count=count/vtrLen) %>%
  ungroup() %>%
  select(term:count)
```

The next code will take the term.doc.matrix mutate it, and filter it

```
docMatrix <- term.doc.matrix %>%
  mutate(document=as.numeric(document)) %>%
  filter(document<QuestionLength+1)
QuestionMatrix <- term.doc.matrix %>%
  mutate(document=as.numeric(document)) %>%
  filter(document>=QuestionLength+1)
```

• The output will be in QuestionMatrix will now find terms of question and answers based on cosine similarity and return the top 5 question and answer in Worsearch

• So, the whole code/model looks something like this

```
QuestionSearch <- function(UserQuestion) {</pre>
  UserQuestion<-CleanText_Analysis(UserQuestion,correct3)
 TotalQuestions <- VectorSource(c(QuestionList, UserQuestion))
QuestionCorpus <- VCorpus(TotalQuestions) %>%
   tm_map(stemDocument) %>%
   tm_map(stripWhitespace)
 term.doc.matrix.stm <- TermDocumentMatrix(QuestionCorpus,</pre>
                                                control=list(
                                                  weighting=function(x) weightSMART(x,spec="ltc"),
                                                  wordLengths=c(1,Inf)))
 # QuestionSearch <- function(UserQuestion)</pre>
 term.doc.matrix <- tidy(term.doc.matrix.stm) %>%
   group_by(document) %>%
mutate(vtrLen=sqrt(sum(count^2))) %>%
   mutate(count=count/vtrLen) %>%
   ungroup() %>%
   select(term:count)
 docMatrix <- term.doc.matrix %>%
   mutate(document=as.numeric(document)) %>%
   filter(document<QuestionLength+1)
QuestionMatrix <- term.doc.matrix %>%
  mutate(document=as.numeric(document)) %>%
   filter(document>=QuestionLength+1)
 Wordsearch <- docMatrix %>%
   suffix=c(".doc",".query")) %>%
mutate(termScore=round(count.doc*count.query,4)) %>%
   group_by(document.query,document.doc) %>%
   summarise(Score=sum(termScore)) %>%
filter(row_number(desc(Score))<=5) %>%
   arrange(desc(Score))%>%
   left_join(FinalData,by=c("document.doc"="rowIndex")) %>%
   ungroup() %>%
select(question,answer) %>%
data.frame()
 return(Wordsearch)
```

The output of the function is

QuestionSearch("sony")

```
[1] "Question: Sony MOR-7506"
[2] "Answer: i do not know what the question is, but, the straps worked well enough to hold my manfrotto tripod and manfrotto ball head to the bottom of my manfrotto camera back pack, which the items that it was holding was about 5 lbs. i do hope this answers your questions, if not, then do please feel free to contact me again."
[1] "Question: Anyone tried this with the Sony MXXOV?"
[1] "Answer: I use it with the Sony RXIOV and it fits perfectly. If the specs are similar, it should work well. The RXIOV doesn't have room to spare, though--it's a comfortable but snug fit. Hope this helps!"
[1] "Question: need help with my sony dun'90: Is there something off here?"
[1] "Answer: Yup I HAD that problem as well when I was using RCA inputs but when I changed to Digital optical, I found that the volume was way too loud and had to turn it down a lot just to fit viewing need. Well, the way I have my setup with the same HTIG(Home Theater in BOX) is: PS3>Reciever via Digital Optical for Audio PS3>TV via Component(Plan to upgrade to HDMI once the cable comes) TV>Reciever via RCA(Red and White plugs) For yours specifically C astellanos, I think we need a bit more info on the TV and the Toshiba, like what kind of Inputs and Outputs are on the back of both products or if not that way, give a link to see information on those items. Hopefully we can get this straighten out as with others in the same boat."
[1] "Question: A30 w/ a Sony KOL40V2SOO?"
[1] "Answer: TheA-30 is definitely a better bet. Just received one 2 weeks ago and no regrets at all. Even upconverts regular dvd's beautifully."
[1] "Answer: TheA-30 is definitely a better bet. Just received one 2 weeks ago and no regrets at all. Even upconverts regular dvd's beautifully."
[1] "Answer: TheA-30 is definitely a better bet. Just received one 2 weeks ago and no regrets at all. Even upconverts regular dvd's beautifully."
```

We have now built the model successfully.

7. Model Deployment

- o The model is going to be deployed in R Shiny
- o The libraries needed for deployment are

```
library(shiny)
library(rlist)
library(DT)
library(shinyWidgets)
library(shinythemes)
```

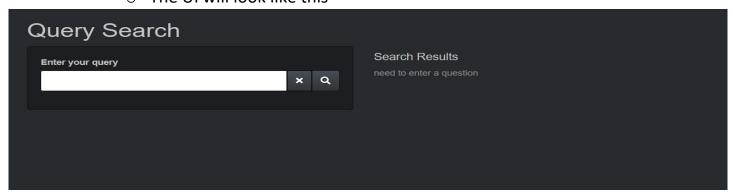
o The UI (User Interface) codes are

```
ui<- fluidPage(theme = shinytheme("slate"),

pageWithSidebar(
   headerPanel("Query Search"),
   sidebarPanel(
       searchInput("query",label ="Enter your query",btnSearch = icon("search"),btnReset = icon("remove"),width = "450px")

),
   mainPanel(
   h4("Search Results"),
       #verbatimTextOutput("results")
   #textOutput("selected_var")
   dataTableOutput("results")
)
)</pre>
```

The UI will look like this



Shiny codes are

```
server<-function(input, output) {</pre>
  x<-reactive({
                           \begin{array}{l} y < - \text{QuestionSearch(input} \\ \text{if(nrow(y)!=0)} \end{array} \} 
                                                     op<-list()
                                                     for (i in 1:nrow(y)){
    op<-list.append(op,paste0(y[i,1]))
    op<-list.append(op,paste0(y[i,2]))
                                                         i=i+1
                                                     op<-as.character(op)
                                                     iter=length(op)
output <- matrix(ncol=2, nrow = iter)</pre>
                                                     for (i in 1:iter){
  if((i %% 2) == 0) {
                                                           \begin{array}{ll} \text{output[i,1]<- paste("Answer",i/2)} \\ \text{output[i,2]<- paste(op[i])} \end{array}
                                                       output[i,1]<- paste("Question",round(i/3)+1)
output[i,2]<- paste(op[i])</pre>
                                                           else{
                                                              output[i,1]<- paste("Question",i)
output[i,2]<- paste(op[i])</pre>
                                                        i=i+1
                                      else{
                                         output <- matrix(ncol=2, nrow = 1)
output[1,1]<- paste(".")
output[1,2]<- paste("No Result Found")</pre>
                                       output<-data.frame(output)
                                       datatable(head(output), rownames = FALSE)
                                      return(output)
     output$results<-renderDataTable({
                                                        validate(
                                                                       need(input$query, 'need to enter a question')
                                                        datatable(x(),rownames = FALSE,colnames = "",
                                                                       options = list(dom = 't',autoWidth = TRUE,
columnDefs = list(list(width = '90%', targets = 1)))
```

```
}
runApp(
launch.browser= T,shinyApp(ui = ui, server = server)
)
```

The output looks like

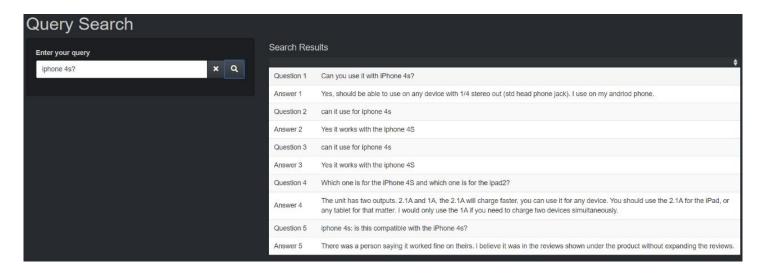


Fig 7.1 shows the output of putting in a single keyword

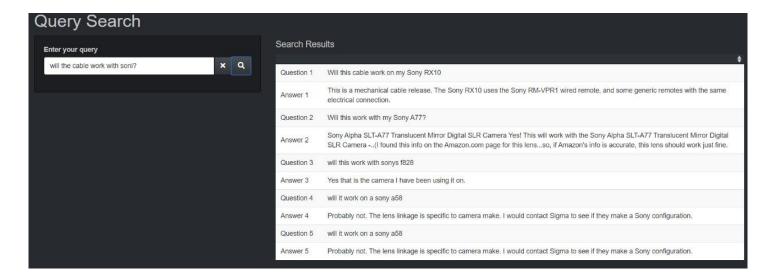


Fig 7.2 shows an incorrect word in input getting corrected and then giving appropriate ouput



Fig 7.3 shows if there is no related word in the dataset then it will show NO RESULTS FOUND

