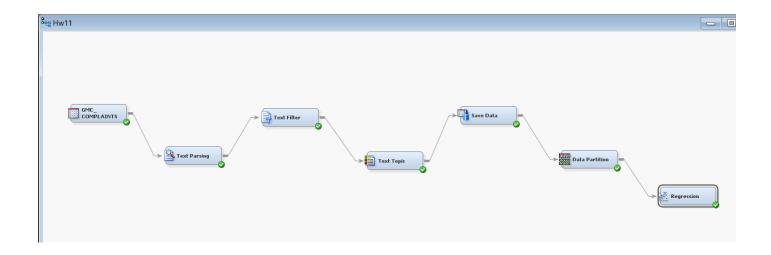
STAT 656 Week 11 Assignment

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PART 1 SAS EM

1) Project Diagram & Property Window



Text Parsing Property

Property	Value
General	
Node ID	TextParsing
Imported Data	
Exported Data	
Notes	
Train	
Variables	
⊟Parse	
Parse Variable	Description
^{1.} Language	English
Detect	
Different Parts of Speech	Yes
Noun Groups	Yes
Multi-word Terms	SASHELP.ENG_MULTI
Find Entities	None
Custom Entities	
☐Ignore	
Ignore Parts of Speech	'Aux' 'Conj' 'Det' 'Interj' 'Part' 'Prep' 'Pi
Ignore Types of Entities	
^L Ignore Types of Attributes	'Num' 'Punct'
□Synonyms	
Stem Terms	Yes
L-Synonyms	SASHELP.ENGSYNMS
□Filter	
-Start List	
Stop List	SASHELP.ENGSTOP
- Select Languages	
Report	
Number of Terms to Display	20000
Status	
Create Time	4/10/19 1:11 PM
Run ID	680b8b80-9570-45b7-a14c-fe48c1e8887
Last Error	
Last Status	Complete
Last Run Time	4/10/19 1:15 PM
Run Duration	0 Hr. 0 Min. 14.32 Sec.
Grid Host	
User-Added Node	No

Text Filter Property

Property	Value
General	
Node ID	TextFilter
Imported Data	
Exported Data	
Notes	
Train	
Variables	
☐ Spelling	
Check Spelling	No
Dictionary	
■Weightings	
Frequency Weighting	None
Term Weight	Inverse Document Frequency
Term Filters	
Minimum Number of Documents	10
Maximum Number of Terms	
Import Synonyms	
□ Document Filters	
Search Expression	
Subset Documents	
Results	
Filter Viewer	
Spell-Checking Results	
Exported Synonyms	
Report	
Terms to View	All
Number of Terms to Display	20000
Status	
Create Time	4/10/19 1:12 PM
Run ID	dd51f76a-4b2e-42f1-8d1f-9d4fa4005c41
Last Error	
Last Status	Complete
Last Run Time	4/10/19 1:21 PM
Run Duration	0 Hr. 0 Min. 3.99 Sec.
Grid Host	
User-Added Node	No

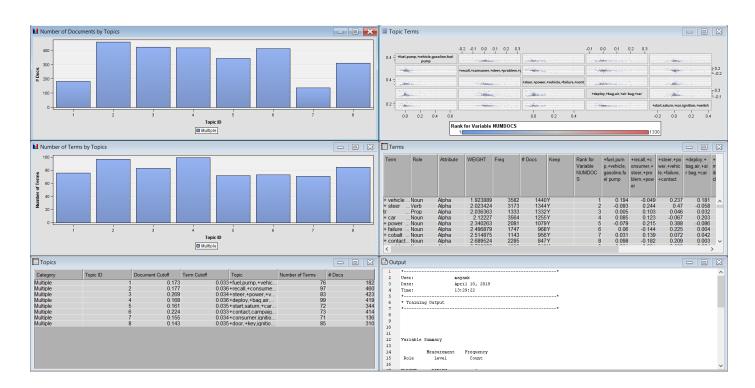
Text Topic Property

Property	Value	
General		
Node ID	TextTopic	
Imported Data		
Exported Data		
Notes		
Train		
Variables		
User Topics		
Term Topics		
Number of Single-term Topics	0	
Learned Topics		
Number of Multi-term Topics	8	
Correlated Topics	No	
Results		
Topic Viewer		
Status		
Create Time	4/10/19 1:12 PM	
Run ID	0bdf209b-4623-4842-bc15-fd30a91d3ae3	
Last Error		
Last Status	Complete	
Last Run Time	4/10/19 1:23 PM	
Run Duration	0 Hr. 0 Min. 4.46 Sec.	
Grid Host		
User-Added Node	No	

Regression Property

Property	Value
General	
Node ID	Reg
Imported Data	
Exported Data	
Notes	
Train	
Variables	
■Equation	
Main Effects	Yes
Two-Factor Interactions	No
Polynomial Terms	No
Polynomial Degree	2
User Terms	No
Term Editor	
⊟Class Targets	
Regression Type	Logistic Regression
Link Function	Logit
■Model Options	
Suppress Intercept	No
Input Coding	Deviation
Model Selection	
Selection Model	Stepwise
Selection Criterion	Default
Use Selection Defaults	Yes
Use Selection Defaults Selection Options	
Optimization Options	
Technique	Default
Default Ontimization	Yes
Max Iterations	0
Max Function Calls	0
Maximum Time	1 Hour
Convergence Criteria	
Uses Defaults	Yes
L. Options	
Output Options	
Confidence Limits	No
Save Covariance	No
Covariance	No
Correlation	No
Statistics	No
Suppress Output	No
Details	No
Design Matrix	No
Score	
Excluded Variables	Reject
Status	,,
Create Time	4/10/19 1:14 PM
Run ID	9ff5887a-7ff7-4ab9-8ae4-75a324813ft
Last Error	5.1007.0 7.17 7009 0004 75052401510
Last Status	Complete
Last Run Time	4/10/19 1:29 PM
Education Fillio	1/10/15 1.25 1111

2)



3) Confusion Matrix

		Predicted	
		Negative	Positive
	Negative	582	19
Actual	Positive	93	127

4) Results

Accuracy	0.863581
Precision	0.869863
Recall	0.577273
F1	0.693989

1) PYTHON CODE

```
# -*- coding: utf-8 -*-
Created on Wed Apr 10 10:41:13 2019
@author: mayank
import pandas as pd
import string
import nltk
import numpy as np
from nltk import pos tag
from nltk.tokenize import word tokenize
from nltk.stem.snowball import SnowballStemmer
from nltk.stem import WordNetLemmatizer
from nltk.corpus import wordnet as wn
from nltk.corpus import stopwords
from sklearn.feature extraction.text import CountVectorizer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.decomposition import LatentDirichletAllocation
#For regression
from AdvancedAnalytics import ReplaceImputeEncode
from AdvancedAnalytics import logreg
from sklearn.model_selection import cross_validate
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
#Analyzer
def analyzer(s):
    # Synonym List
    syns = {'veh': 'vehicle', 'car': 'vehicle', 'chev':'cheverolet', \
               'chevy':'cheverolet', 'air bag': 'airbag', \
              'seat belt':'seatbelt', "n't":'not', 'to30':'to 30', \
               'wont':'would not', 'cant':'can not', 'cannot':'can not', \
               'couldnt':'could not', 'shouldnt':'should not', \
'wouldnt':'would not', 'air':'airbag', 'bag':'airbag'}
    s = s.lower()
    s = s.replace(',', '. ')
    # Tokenize
    tokens = word_tokenize(s)
    tokens = [word.replace(',','') for word in tokens ]
    tokens = [word for word in tokens if ('*' not in word) and \
```

```
("''" != word) and ("``" != word) and \
              (word!='description') and (word !='dtype') \
              and (word != 'object') and (word!="'s")]
    # Map synonyms
    for i in range(len(tokens)):
        if tokens[i] in syns:
            tokens[i] = syns[tokens[i]]
    # Remove stop words
    punctuation = list(string.punctuation)+['..', '...']
    pronouns = ['i', 'he', 'she', 'it', 'him', 'they', 'we', 'us', 'them']
    # For adding extra stop words
    other = ['own','go','get','seem','say','would','regard','report','involve'\
             ,'do','anoth','consumer',"'ve",'happen','try','either','come',]
    stop = stopwords.words('english') + punctuation + pronouns + other
    filtered terms = [word for word in tokens if (word not in stop) and \
                  (len(word)>1) and (not word.replace('.','',1).isnumeric()) \
                  and (not word.replace("'",'',2).isnumeric())]
    # Stemming
    tagged_words = pos_tag(filtered_terms, lang='eng')
    stemmer = SnowballStemmer("english")
    wn_tags = {'N':wn.NOUN, 'J':wn.ADJ, 'V':wn.VERB, 'R':wn.ADV}
    wnl = WordNetLemmatizer()
    stemmed tokens = []
    for tagged_token in tagged_words:
        term = tagged token[0]
        pos = tagged_token[1]
        pos = pos[0]
        try:
                  = wn_tags[pos]
            stemmed_tokens.append(wnl.lemmatize(term, pos=pos))
            stemmed_tokens.append(stemmer.stem(term))
    return stemmed_tokens
# Stopping and Stemming using NLTK
def preprocessor(s):
    #Vectorizer sends one string at a time
    s = s.lower()
    s = s.replace(',', '. ')
    print("preprocessor")
    return(s)
def tokenizer(s):
   # Tokenize
    print("Tokenizer")
    tokens = word_tokenize(s)
    tokens = [word.replace(',','') for word in tokens ]
```

```
tokens = [word for word in tokens if word.find('*')!=True and \
              word != "''" and word !="``" and word!='description' \
              and word !='dtype']
    return tokens
# Increase Pandas column width
pd.set option('max colwidth', 32000)
# California Cabernet Reviews
df = pd.read excel("GMC Complaints.xlsx")
# Setup simple constants
n_docs = len(df['description'])
n_samples = n_docs
m features = 1000
s_words = 'english'
ngram = (1,2)
max df=0.8
discussions = []
for i in range(n samples):
    discussions.append(("%s" %df['description'].iloc[i]))
# Create Word Frequency
cv = CountVectorizer(max df=max df, min df=2, max features=m features,\
                     analyzer=analyzer, ngram_range=ngram)
tf = cv.fit_transform(discussions)
n topics
                = 8
max_iter
learning offset = 20.
learning_method = 'online'
tf idf = TfidfTransformer()
print("\nTF-IDF Parameters\n", tf idf.get params(),"\n")
tf_idf = tf_idf.fit_transform(tf)
tfidf_vect = TfidfVectorizer(max_df=max_df, min_df=2, max_features=m_features,\
                             analyzer=analyzer, ngram_range=ngram)
tf_idf = tfidf_vect.fit_transform(discussions)
print("\nTF_IDF Vectorizer Parameters\n", tfidf_vect, "\n")
lda = LatentDirichletAllocation(n components=n topics, max iter=max iter,\
                                learning method=learning method, \
                                learning offset=learning offset, \
                                random state=12345)
lda.fit_transform(tf idf)
print('{:..<22s}{:>6d}'.format("Number of Reviews", tf.shape[0]))
print('{:.<22s}{:>6d}'.format("Number of Terms",
                                                    tf.shape[1]))
print("\nTopics Identified using LDA with TF IDF")
```

```
tf features = cv.get feature names()
\max \text{ words} = 15
desc = []
for topic idx, topic in enumerate(lda.components ):
        message = "Topic #%d: " % topic_idx
        message += " ".join([tf_features[i]
                             for i in topic.argsort()[:-max_words - 1:-1]])
        print(message)
        print()
        desc.append([tf features[i] for i in topic.argsort()[:-max words - 1:-1]])
#Extract topic probablities
topics = pd.DataFrame(lda.fit_transform(tf_idf))
preds = ['Year', 'make', 'model', 'crashed', 'abs', 'mileage']
reg_df = pd.concat([df[preds],topics], axis=1, ignore_index=True)
reg df.columns = ['Year',
'make', 'model', 'crashed', 'abs', 'mileage', '0', '1', '2', '3', '4', '5', '6', '7']
# logistics regression model
attribute_map = {
        'Year'
                  :['I',(2003,2011),[0,0]],
                  :['N',('CHEVROLET','PONTIAC','SATURN'),[0,0]],
        'make'
        'model'
                  :['N',('COBALT','G5','HHR','ION','SKY','SOLSTICE'),[0,0]],
        'crashed' :['B',('N','Y'),[0,0]],
                  :['B',('N','Y'),[0,0]],
        'abs'
        'mileage' :['I',(0,200000),[0,0]],
                  :['I',(0,1),[0,0]],
        '1'
                  :['I',(0,1),[0,0]],
        '2'
                  :['I',(0,1),[0,0]],
                  :['I',(0,1),[0,0]],
        '3'
        '4'
                  :['I',(0,1),[0,0]],
        '5'
                  :['I',(0,1),[0,0]],
        '6'
                  :['I',(0,1),[0,0]],
        '7'
                  :['I',(0,1),[0,0]],
}
varlist = ['crashed']
rie = ReplaceImputeEncode(data_map=attribute_map, \
                                nominal_encoding='one-hot',
                           interval scale = None, drop=False, display=False)
encoded_df = rie.fit_transform(reg_df)
X = encoded_df.drop(varlist, axis=1)
y = encoded df[varlist]
np y=np.ravel(y)
\max f1 = 0
C_list=[.1,1,10,100]
score_list = ['accuracy', 'recall', 'precision', 'f1']
for c in C_list:
    print("\nRegularization Parameter: ", c)
```

```
lgr = LogisticRegression(C=c, tol=1e-8, max iter=1000)
    lgr.fit(X, np y)
    scores = cross_validate(lgr, X, np_y,\
                            scoring=score_list, return_train_score=False, \
                            cv=10)
    print("{:.<13s}{:>6s}{:>13s}".format("Metric", "Mean", "Std. Dev."))
    for s in score list:
       var = "test_"+s
       mean = scores[var].mean()
        std = scores[var].std()
        print("{:.<13s}{:>7.4f}{:>10.4f}".format(s, mean, std))
       if mean > max_f1:
            max f1 = mean
            best predictor = c
print("\nBest based on F1-Score")
print("Best Regularization Parameter = ", best predictor)
X_train, X_valid, y_train, y_valid= \
train test split(X,y,test size = 0.3, random state=7)
np_y_train = np.ravel(y_train)
np y valid = np.ravel(y valid)
reg = LogisticRegression(C=best_predictor, tol=1e-8, max_iter=1000)
reg.fit(X train,np y train)
logreg.display_coef(reg,21,2,X_train.columns)
logreg.display binary split metrics(reg,X_train,np_y_train,X_valid,np_y_valid)
```

OUTPUT

2) MODEL METRICS

Model Metrics	Training	Validation
Observations	1913	821
Coefficients	21	21
DF Error	1892	800
Mean Absolute Error	0.2712	0.2967
Avg Squared Error	0.1317	0.1516
Accuracy	0.8191	0.7881
Precision	0.8818	0.8554
Recall (Sensitivity)	0.3573	0.3047
F1-score	0.5085	0.4494
MISC (Misclassification)	18.1%	21.2%

class 0	1.7%	2.0%
class 1	64.3%	69.5%

3) CONFUSION MATRIX

Training

Confusion Matrix	Class 0	Class 1
Class 0	1388	24
Class 1	322	179

Validation

Confusion Matrix	Class 0	Class 1
Class 0	576	12
Class 1	162	71