

Sarcasm_the_secret_language

May 12, 2023

With the NLP toolkit we will be attempting to detect sarcasm in text. I will be using practice text data with '/' at the end of the text to indicate if text was sarcastic. Later I will create a prediction model from this to determine if future text is sarcastic

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[1]: #Import the needed librairies
%matplotlib inline

import bz2
import json
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns

from collections import Counter
from lime.lime_text import LimeTextExplainer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import ENGLISH_STOP_WORDS
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, f1_score
from sklearn.pipeline import make_pipeline
from sklearn.dummy import DummyClassifier
from sklearn.ensemble import RandomForestClassifier

[2]: # random seed is to ensure the reproducibility of my model
RANDOM_SEED = 655

[3]: #loading practice data
train_df = pd.read_csv('assets/sarcasm.train.tsv.gz', sep='\t',
    ↳compression='gzip').dropna()
test_imb_df = pd.read_csv('assets/sarcasm.test-imb.tsv.gz', sep='\t',
    ↳compression='gzip').dropna()
test_bal_df = pd.read_csv('assets/sarcasm.test-bal.tsv.gz', sep='\t',
    ↳compression='gzip').dropna()

[4]: #TfidfVec is for featurizing the corpus, to speed this process up I will use a
    ↳min document frequent of 100
def data_convert(df):
    X = df.text
```

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    y = list(df.label)
    return X,y

#init function
X_train, y_train = data_convert(train_df)
X_test_bal, y_test_bal = data_convert(test_bal_df)
X_test_imb, y_test_imb = data_convert(test_imb_df)

#create vec
vectorizer = TfidfVectorizer(min_df = 100, stop_words = 'english',ngram_range_
↳=(1,2) )

#X's
X_train = vectorizer.fit_transform(X_train)
X_test_bal = vectorizer.transform(X_test_bal)
X_test_imb = vectorizer.transform(X_test_imb)

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[5]: # Lets check the label count to ensure everything under the hood is working_
↳according to plan
label_count = Counter(y_train)
label_count

```

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[5]: Counter({1: 128540, 0: 128541})

```

Now it is time to train my model I will use a Logistic Regression classifier and a Random Forest Classifier for comparison. I will also use a dummy classifier why you ask? I want to be certain that my model is predicting higher than random chance. Consider my dummy classifier a baseline for performance since I am expecting to predict higher than random chance.

```

[6]: lr_clf = LogisticRegression(solver = 'lbfgs', multi_class = 'auto')
rf_clf = RandomForestClassifier(n_estimators = 50, max_depth = 15, random_state_
↳= RANDOM_SEED)
random_clf = DummyClassifier(strategy='uniform',random_state = RANDOM_SEED).
↳fit(X_train, y_train)
# can't forget to fit our model--
lr_clf.fit(X_train, y_train)
rf_clf.fit(X_train, y_train)

```

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[6]: RandomForestClassifier(max_depth=15, n_estimators=50, random_state=655)

```

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[7]: y_pred_bal = lr_clf.predict(X_test_bal)
y_pred_imb = lr_clf.predict(X_test_imb)
rf_y_pred_bal = rf_clf.predict(X_test_bal)
rf_y_pred_imb = rf_clf.predict(X_test_imb)
random_y_pred_bal = random_clf.predict(X_test_bal)
random_y_pred_imb = random_clf.predict(X_test_imb)

```

```
[8]: #Now let us evaluate and see if sarcasm detection is possible
      # I will be using f1 scores since this problem is binary (sarcastic or not). I
      ↪will use binary for average
```

```
def score(y,pred):
    f1 = f1_score(y, pred, average='binary')
    return f1

lr_imb_f1 = score(y_test_imb,y_pred_imb)
lr_bal_f1 = score(y_test_bal,y_pred_bal)

rf_imb_f1 = score(y_test_imb,rf_y_pred_imb)
rf_bal_f1 = score(y_test_bal,rf_y_pred_bal)

rand_imb_f1 = score(y_test_imb,random_y_pred_imb)
rand_bal_f1 = score(y_test_bal,random_y_pred_bal)
```

```
[9]: print('LogisticRegression_pred_scores',lr_bal_f1)
      print('RandomForrest_pred_scores',rf_bal_f1,)
      print('RandomForrest_pred_scores',rand_bal_f1)
```

```
LogisticRegression_pred_scores 0.6045505212772933
RandomForrest_pred_scores 0.4345472123751208
RandomForrest_pred_scores 0.502941629447097
```

60% accuracy not bad, but let us see it at work, shall we? Before we start let me give a warning, I am pulling comments at random. The problem with this is the comment are random so please excuse the inappropriate language. Now that we got that out the way let begin!

```
[10]: class_names = ['sarcastic','not-sarcastic']

lr_pipe = make_pipeline(vectorizer,lr_clf)
rf_pipe = make_pipeline(vectorizer,rf_clf)

explainer = LimeTextExplainer(class_names = class_names)

test_row_df = test_imb_df.sample(1, random_state=RANDOM_SEED)
_, inst_text, label = next(test_row_df.itertuples())

print('Comment has text "%s"' %(inst_text))
# print('True label: %d' % label)
print('LogisticRegression Probability(Sarcastic) =', lr_pipe.
      ↪predict_proba([inst_text])[0, 1])
print('RandomForest Probability(Sarcastic) =', rf_pipe.
      ↪predict_proba([inst_text])[0, 1])
```

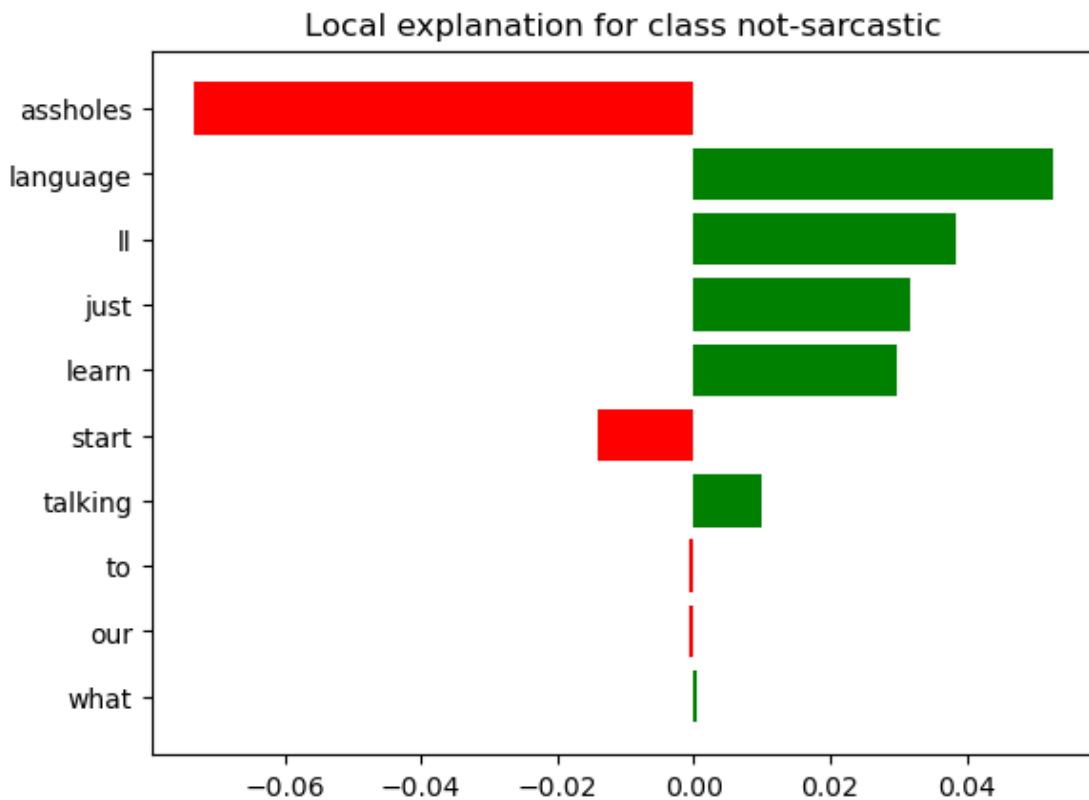
Comment has text "Now just get a supercomputer to learn the language and they'll start talking with dolphins behind our backs about what assholes we are."

LogisticRegression Probability(Sarcastic) = 0.5316511260593914
RandomForest Probability(Sarcastic) = 0.4930586390484008

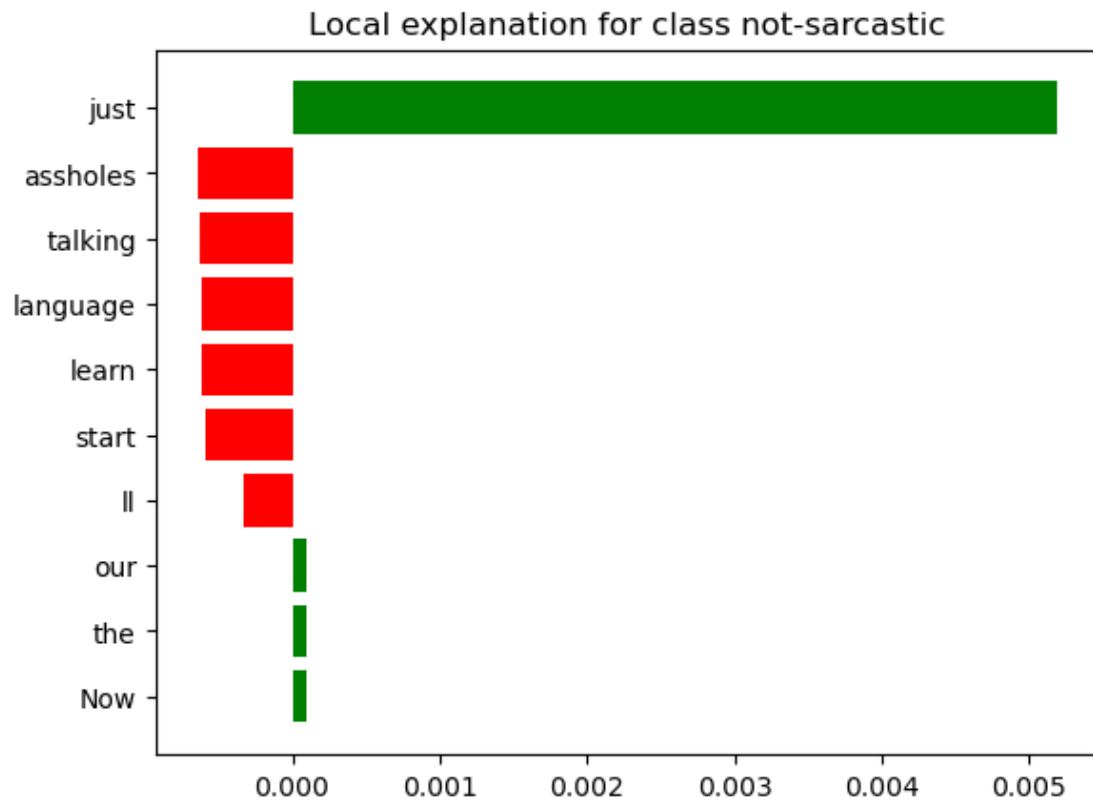
The funny part is I did not detect any sarcasm at all, this all seem plausible to me

```
[11]: #I wonder what part of speech most likely lead to the high sarcasm score.

rf_explanation = explainer.explain_instance(inst_text, rf_pipe.predict_proba,
    ↪num_features=10)
lr_explanation = explainer.explain_instance(inst_text, lr_pipe.predict_proba,
    ↪num_features=10)
fig = lr_explanation.as_pyplot_figure()
```



```
[12]: fig = rf_explanation.as_pyplot_figure()
```



I hope you enjoyed this exercise with me also... HIRE ME. I mean if you got this far you must see something you like right, and I am not being sarcastic. LinkedIn is in the bio... Have a great Day!!!

[]: