## **Importing Modules**

```
In [1]: 1 import numpy as np
2 import pandas as pd
3 import itertools
4 from sklearn.model_selection import train_test_split
5 from sklearn.feature_extraction.text import TfidfVectorizer
6 from sklearn.linear_model import PassiveAggressiveClassifier
7 from sklearn.metrics import accuracy_score, confusion_matrix
8 import matplotlib.pyplot as plt
```

## **Data Wrangling**

### **DataFrame**

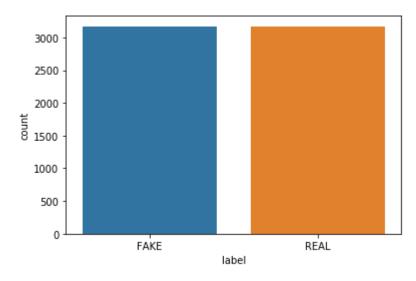
#### Out[3]:

	Unnamed: 0	title	text	label
0	8476	You Can Smell Hillary's Fear	Daniel Greenfield, a Shillman Journalism Fello	FAKE
1	10294	Watch The Exact Moment Paul Ryan Committed Pol	Google Pinterest Digg Linkedin Reddit Stumbleu	FAKE
2	3608	Kerry to go to Paris in gesture of sympathy	U.S. Secretary of State John F. Kerry said Mon	REAL
3	10142	Bernie supporters on Twitter erupt in anger ag	— Kaydee King (@KaydeeKing) November 9, 2016 T	FAKE
4	875	The Battle of New York: Why This Primary Matters	It's primary day in New York and front-runners	REAL

```
In [8]: 1 import seaborn as sns
```

```
In [10]: 1 sns.countplot(x= "label", data = df)
```

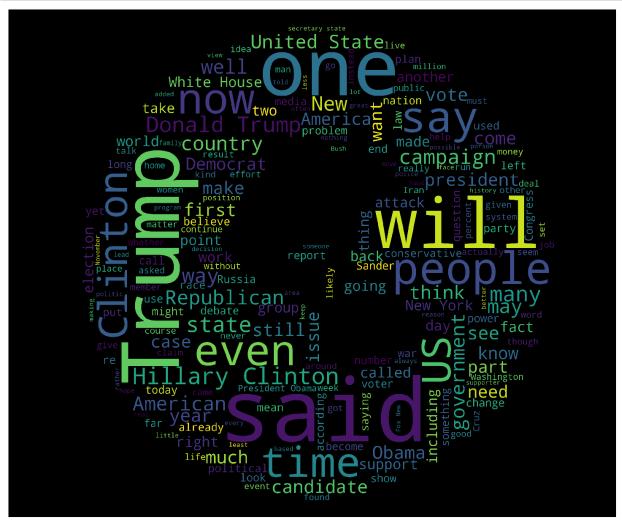
Out[10]: <matplotlib.axes.\_subplots.AxesSubplot at 0x19ac6771240>



## Information of the Data Set

```
In [4]:
             df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 6335 entries, 0 to 6334
         Data columns (total 4 columns):
         Unnamed: 0
                       6335 non-null int64
         title
                       6335 non-null object
                       6335 non-null object
         text
         label
                       6335 non-null object
         dtypes: int64(1), object(3)
         memory usage: 198.1+ KB
In [12]:
              all_words = ' '.join([text for text in df['text']])
In [13]:
           1 from wordcloud import WordCloud
             from PIL import Image
             import requests
In [14]:
             mask = np.array(Image.open(requests.get('http://clipart-library.com/images/L
```

```
In [16]:
           1
              def generate wordcloud(all words, mask):
                  word_cloud = WordCloud(width = 900, height = 600, background_color='blac
           2
           3
                  plt.figure(figsize=(20,18),facecolor = 'white', edgecolor='blue')
           4
                  plt.figure
           5
                  plt.imshow(word_cloud)
           6
                  plt.axis('off')
           7
                  plt.tight_layout(pad=0)
                  plt.show()
           8
              generate_wordcloud(all_words, mask)
           9
```



```
In [ ]: 1
```

#### Labels from the DataFrame.

### Split the dataset into training and testing sets.

Let's initialize a TfidfVectorizer with stop words from the English language and a maximum document frequency of 0.7 (terms with a higher document frequency will be discarded). Stop words are the most common words in a language that are to be filtered out before processing the natural language data. And a TfidfVectorizer turns a collection of raw documents into a matrix of TF-IDF features. Now, fit and transform the vectorizer on the train set, and transform the vectorizer on the test set.

We'll initialize a PassiveAggressiveClassifier. This is. We'll fit this on tfidf train and y train.

Then, we'll predict on the test set from the TfidfVectorizer and calculate the accuracy with accuracy\_score() from sklearn.metrics

Accuracy: 92.82%

We got an accuracy of 92.66% with this model. Finally, let's print out a confusion matrix to gain insight into the number of false and true negatives and positives.

So with this model, we have 586 true positives,52 true negatives,41 false positives, and 588 false negatives.

# **Summary**

Today, we learned to detect fake news with Python. We took a political dataset, implemented a TfidfVectorizer, initialized a PassiveAggressiveClassifier, and fit our model. We ended up obtaining an accuracy of 93.05% in magnitude.