### Fake news detection

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```
import pandas as pd
import numpy as np
import.matplotlib.pyplot.as.plt
import.seaborn.as.sns.
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn import feature_extraction, linear_model, model_selection, preprocessing
from sklearn.metrics import accuracy_score
from sklearn.model_selection import train_test_split
from sklearn.pipeline import Pipeline
```

#### Read datasets

## Data cleaning and preparation

```
# Add flag to track fake and real
fake['target'] = 'fake'
true['target'] = 'true'

# Concatenate dataframes
data = pd.concat([fake, true]).reset_index(drop = True)
data.shape
```

https://colab.research.google.com/drive/1pYufl50bkTleIgV5ds1N6_UwKRC2yPCD#scrollTo=Qxwy6f3gHAdZ&printMode=true 1/11 FakeNewsDetection.ipynb - Colaboratory	

```
(44898, 5)
# Shuffle the data
from sklearn.utils import shuffle
data = shuffle(data)
data = data.reset_index(drop=True)
# Check the data
data.head()
                                                                    title text subject date target
                              Trump looks to deliver
                                                      would
                                                                               October 13,
                                                       For the moment, U.S.
                                                                               2017
       nln a sudden flurry,
                                                         NATIONS (Reuters)
                               1Trump budget cut bid
                                                          true
       for...
       WASHINGTON (Reuters)
                                                          24,
2017true
                                make it 'impossible...
                                Pr...politicsNews UNITED
                                                       March 19,
                    - U.S. President Dona...
                    politicsNews May Lowest L...
                                                           News
       2Trump's Approval Rating Since taking the oath
                                                           2017 fake
       TANKS To The
                                 ofoffice, alleged presi...
# Removing the date (we won't use it for the analysis)
data.drop(["date"],axis=1,inplace=True)
data.head()
                                                                          title text subject target
                                                  WASHINGTON (Reuters) - For the
                                                  moment, U.S. Pr...politicsNews true UNITED
       oln a sudden flurry, Trump looks to deliver
                                                  NATIONS (Reuters) - U.S. President Dona...
       for... 1 Trump budget cut bid would make it
                                                  politicsNews true Since taking the oath of
                                                  office,
       'impossible...
                                                  alleged presi... News fake
       2Trump's Approval Rating TANKS To The
       Lowest L...
                                    Donald Trump Campaign CEO Republicans are great at creating _{\mbox{N f k}}
# Removing the title (we will only use the text)
```

data.drop(["title"],axis=1,inplace=True)

data.head()

#### text subject target

**0** WASHINGTON (Reuters) - For the moment, U.S. Pr... politicsNews true **1** UNITED NATIONS (Reuters) - U.S. President Dona... politicsNews true **2** Since taking the oath of office, alleged presi... News fake **3** Republicans are great at creating controversy ... News fake **4** The video below is a much watch! A young Donal... politics fake

 $https://colab.research.google.com/drive/1pYufl50bkTlelgV5ds1N6\_UwKRC2yPCD\#scrollTo=Qxwy6f3gHAdZ\&printMode=true~2/11~FakeNewsDetection.ipynb-Colaboratory$ 

```
# Convert to lowercase

data['text'] = data['text'].apply(lambda x: x.lower())
data.head()
```

#### text subject target

- 0 washington (reuters) for the moment, u.s. pr... politicsNews true
- 1 united nations (reuters) u.s. president dona... politicsNews true
- 2 since taking the oath of office, alleged presi... News fake
- 3 republicans are great at creating controversy ... News fake
- 4 the video below is a much watch! a young donal... politics fake

```
# Remove punctuation
import string

def punctuation_removal(text):
    all_list = [char for char in text if char not in string.punctuation] clean_str = ''.join(all_list) return clean_str

data['text'] = data['text'].apply(punctuation_removal)

# Check
data.head()
```

#### text subject target

- **0** washington reuters for the moment us presiden... politicsNews true
- 1 united nations reuters us president donald tr... politicsNews true
- 2 since taking the oath of office alleged presid... News fake
- 3 republicans are great at creating controversy ... News fake
- 4 the video below is a much watch a young donald... politics fake

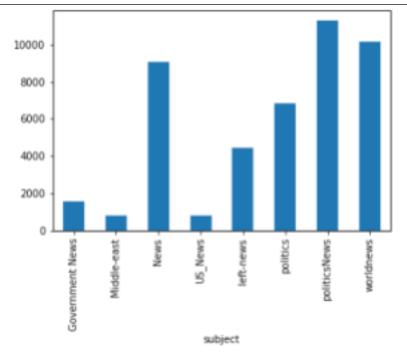
text subject target

**0** washington reuters moment us president donald ... politicsNews true **1** united nations reuters us president donald tru... politicsNews true **2** since taking oath office alleged president don... News fake **3** republicans great creating controversy none sh... News fake **4** video much watch young donald j trump speaks l... politics fake

## Basic data exploration

```
# How many articles per subject?
print(data.groupby(['subject'])['text'].count())
data.groupby(['subject'])['text'].count().plot(kind="bar")
plt.show()

subject
Government News 1570
Middle-east 778
News 9050
US_News 783
left-news 4459
politics 6841
politicsNews 11272
worldnews 10145
Name: text, dtype: int64
```



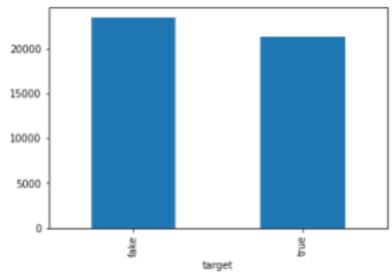
# How many fake and real articles?
print(data.groupby(['target'])['text'].count())

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```
data.groupby(['target'])['text'].count().plot(kind="bar")
plt.show()
```

target fake 23481 true 21417

Name: text, dtype: int64



# Word cloud for fake news
from wordcloud import WordCloud

interpolation='bilinear')

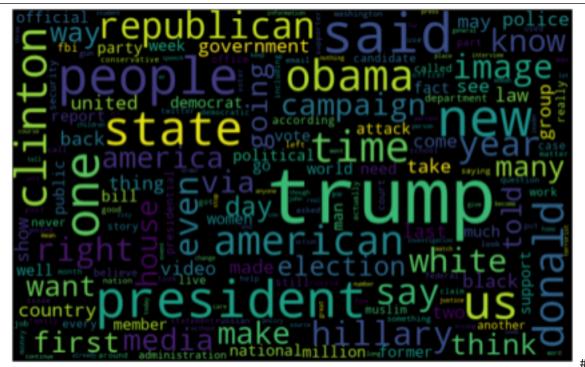
```
plt.axis("off")
plt.show()
```

```
american selection say week and on the country selection say white reports week and the country selection say white reports week and the country selection say white reports week and the country selection say white reports white reports week and the country selection say white reports white reports week and the country selection say white reports week and the country selection say white reports week and the country selection say white reports the country selection say white reports week and the country selection say white reports week and the country selection say white reports white say white says of the country selection say white reports white says of the country selection say white says of the country selection sa
```

HAdZ&printMode=true

# Word cloud for real
news
from wordcloud
import
WordCloud

real\_data =

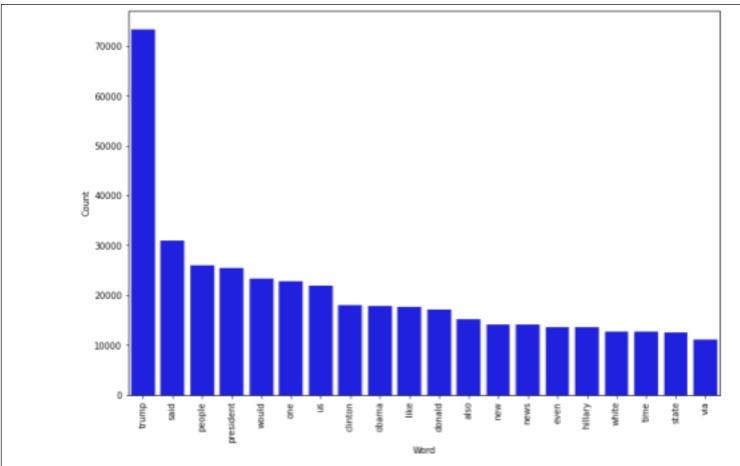


# Most

frequent words counter (Code adapted from https://www.kaggle.com/rodolfoluna/fake-n from nltk import tokenize

 $https://colab.research.google.com/drive/1pYufl50bkTlelgV5ds1N6\_UwKRC2yPCD\#scrollTo=Qxwy6f3gHAdZ\&printMode=true\\ 6/11\ FakeNewsDetection.ipynb-Colaboratory$ 

```
# Most frequent words in fake news
counter(data[data["target"] == "fake"], "text", 20)
```

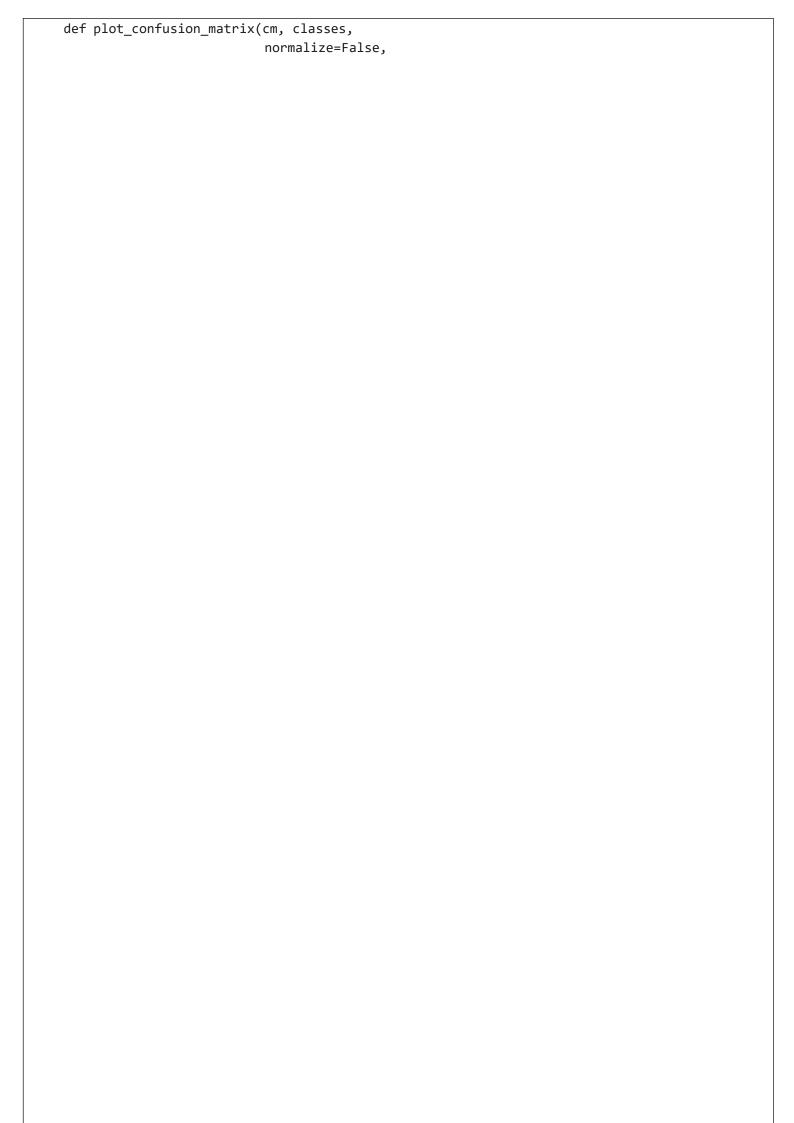


# Most frequent words in real news
counter(data[data["target"] == "true"], "text", 20)

 $https://colab.research.google.com/drive/1pYufl50bkTlelgV5ds1N6\_UwKRC2yPCD\#scrollTo=Qxwy6f3gHAdZ\&printMode=true\\ 7/11\ FakeNewsDetection.ipynb-Colaboratory$ 

## Modeling

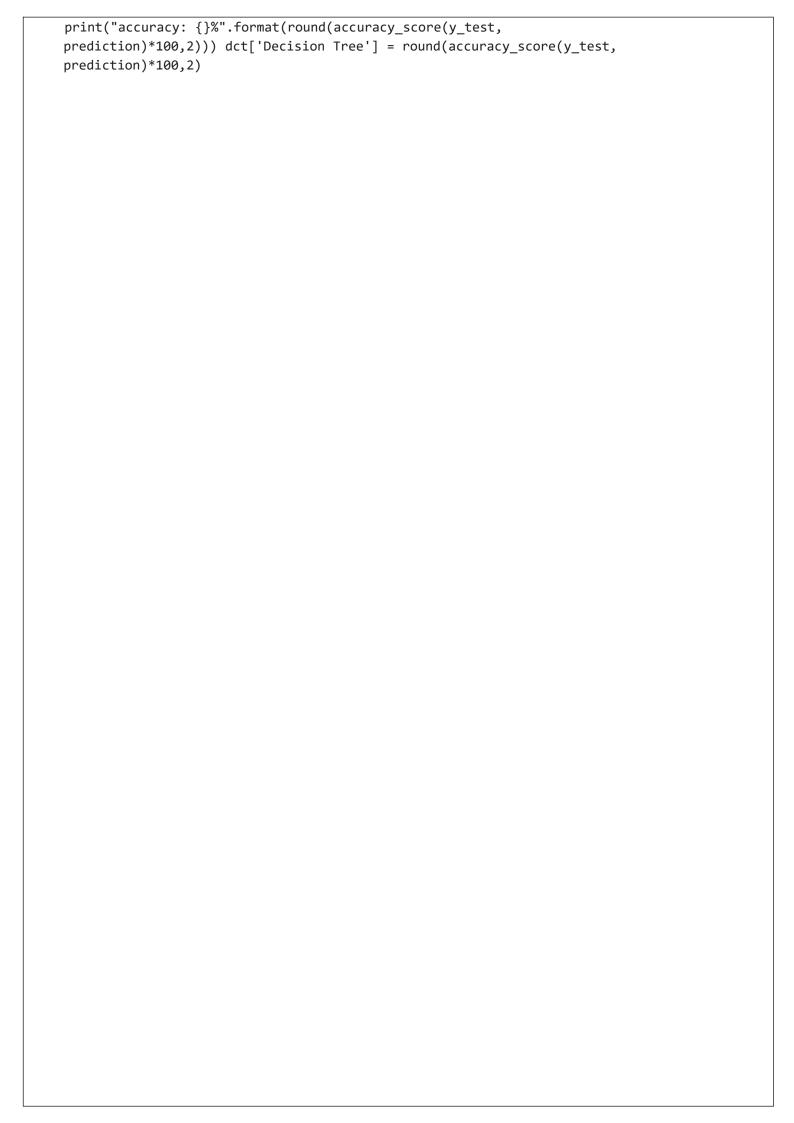
# Function to plot the confusion matrix (code from https://scikit-learn.org/stable/auto\_ex
from sklearn import metrics
import itertools



```
title='Confusion matrix',
                      cmap=plt.cm.Blues):
plt.imshow(cm, interpolation='nearest', cmap=cmap)
plt.title(title)
plt.colorbar()
tick_marks = np.arange(len(classes))
plt.xticks(tick_marks, classes, rotation=45)
plt.yticks(tick_marks, classes)
if normalize:
    cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
    print("Normalized confusion matrix")
else:
    print('Confusion matrix, without normalization')
thresh = cm.max() / 2.
for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
    plt.text(j, i, cm[i, j],
             horizontalalignment="center",
             color="white" if cm[i, j] > thresh else "black")
plt.tight_layout()
plt.ylabel('True label')
plt.xlabel('Predicted label')
```

### Preparing the data

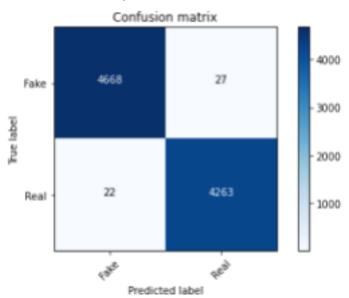
## Decision Tree



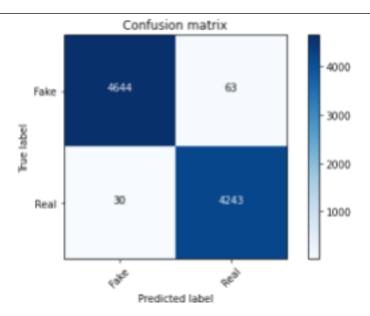
```
accuracy: 99.45%
```

```
cm = metrics.confusion_matrix(y_test, prediction)
plot_confusion_matrix(cm, classes=['Fake', 'Real'])
```

Confusion matrix, without normalization



### Random Forest



# Comparing Different Models

```
import matplotlib.pyplot as plt
plt.figure(figsize=(8,7))
plt.bar(list(dct.keys()),list(dct.values()))
plt.ylim(90,100)
plt.yticks((91, 92, 93, 94, 95, 96, 97, 98, 99, 100))
    https://colab.research.google.com/drive/1pYufl50bkTlelgV5ds1N6_UwKRC2yPCD#scrollTo=Qxwy6f3gHAdZ&printMode=true 10/11
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```

