In [1]:	<pre>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</pre>
	<pre>from sklearn.pipeline import Pipeline fake = pd.read_csv("Fake.csv") true = pd.read_csv("True.csv") fake.shape (23481, 4)</pre>
In [5]:	true.shape (21417, 4) Data cleaning and preparation
	<pre># 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</pre>
In [8]:	<pre>from sklearn.utils import shuffle data = shuffle(data) data = data.reset_index(drop=True)</pre>
In [9]: Out[9]:	# Check the data data head() title text subject date target Pathetic: Trump Actually Needs A General To B With Reince Priebus out as White House chief o News August 5, 2017 fake SHOCKER! IS MITT ROMNEY Being Considered For A politics Nov 17, 2016 fake Bridgewater executive McCormick declines Defen NEW YORK (Reuters) - David McCormick has decli politicsNews January 10, 2017 true
In [10]:	Week of clashes in eastern Ethiopia kill 50, d ADDIS ABABA (Reuters) - Clashes along the bord worldnews September 17, 2017 true U.S. Senate number two Democrat calls Trump's WASHINGTON (Reuters) - The U.S. Senate's numbe politicsNews May 15, 2017 true # Removing the date (we won't use it for the analysis) data.drop(["date"], axis=1, inplace=True) data.head()
Out[10]:	title text subject target Pathetic: Trump Actually Needs A General To B With Reince Priebus out as White House chief o News fake SHOCKER! IS MITT ROMNEY Being Considered For A politics fake Bridgewater executive McCormick declines Defen NEW YORK (Reuters) - David McCormick has decli politicsNews true Week of clashes in eastern Ethiopia kill 50, d ADDIS ABABA (Reuters) - Clashes along the bord worldnews true
In [11]: Out[11]:	U.S. Senate number two Democrat calls Trump's WASHINGTON (Reuters) - The U.S. Senate's numbe politicsNews true # Removing the title (we will only use the text) data.drop(["title"], axis=1, inplace=True) data.head() text subject target 0 With Reince Priebus out as White House chief o News fake
	politics fake NEW YORK (Reuters) - David McCormick has decli politicsNews true ADDIS ABABA (Reuters) - Clashes along the bord worldnews true WASHINGTON (Reuters) - The U.S. Senate's numbe politicsNews true
In [12]: Out[12]:	<pre>data['text'] = data['text'].apply(lambda x: x.lower()) data.head()</pre>
In [13]:	3 addis ababa (reuters) - clashes along the bord worldnews true 4 washington (reuters) - the u.s. senate's numbe politicsNews true # Remove punctuation import string
In [14]:	<pre>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)</pre> data.head()
Out[14]:	text subject target with reince priebus out as white house chief o News fake new york reuters david mccormick has declined politics News true addis ababa reuters clashes along the border worldnews true washington reuters the us senate's number two politicsNews true
In [18]:	<pre># Removing stopwords import nltk nltk.download('stopwords') from nltk.corpus import stopwords stop = stopwords.words('english') data['text'] = data['text'].apply(lambda x: ' '.join([word for word in x.split() if word not in (stop)]))</pre>
In [17]: Out[17]:	<pre>[nltk_data] Downloading package stopwords to [nltk_data]</pre>
	politics fake new york reuters david mccormick declined offe politicsNews true addis ababa reuters clashes along border ethio worldnews true washington reuters us senate's number two demo politicsNews true
In [19]:	# 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
	10000 - 8000 - 6000 -
	Middle-east - News - Solitics News - Worldnews - World
In [20]:	# How many fake and real articles? print(data.groupby(['target'])['text'].count()) data.groupby(['target'])['text'].count().plot(kind="bar") plt.show() target
	fake 23481 true 21417 Name: text, dtype: int64 20000 -
	15000 - 10000 - 5000 - 0
In [23]:	# Most frequent words counter (Code adapted from https://www.kaggle.com/rodolfoluna/fake-news-detector) from nltk import tokenize token_space = tokenize.WhitespaceTokenizer() def counter(text, column_text, quantity):
	<pre>all_words = ' '.join([text for text in text[column_text]]) token_phrase = token_space.tokenize(all_words) frequency = nltk.FreqDist(token_phrase) df_frequency = pd.DataFrame({"word": list(frequency.keys()),</pre>
In [45]:	plt.xticks(rotation='vertical') plt.show() # Most frequent words in fake news counter(data[data["target"] == "fake"], "text", 20)
	60000 - 50000 - ± 40000 -
	30000 -
	trump People as a said and a sai
In [25]:	# Most frequent words in real news counter(data[data["target"] == "true"], "text", 20) 100000 -
	80000 - 60000 -
	2000 -
	trump would would washington washington washington washington washington washington washington washington washington
In [26]:	<pre>from sklearn import metrics import itertools def plot_confusion_matrix(cm, classes,</pre>
	<pre>normalize=False, 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)</pre>
	<pre>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])):</pre>
In [27]:	<pre>plt.text(j, i, cm[i, j],</pre>
In [28]:	<pre># Split the data X_train, X_test, y_train, y_test = train_test_split(data['text'], data.target, test_size=0.2, random_state=42)</pre> Naive Bayes dct = dict()
	<pre>from sklearn.naive_bayes import MultinomialNB NB_classifier = MultinomialNB() pipe = Pipeline([('vect', CountVectorizer()),</pre>
In [29]:	<pre>dct['Naive Bayes'] = round(accuracy_score(y_test, prediction)*100,2) accuracy: 95.13% cm = metrics.confusion_matrix(y_test, prediction) plot_confusion_matrix(cm, classes=['Fake', 'Real']) Confusion matrix, without normalization</pre>
	Fake - 4407 288 - 3500 - 3000 - 2500 - 2000
	Real - 149 4136 - 1500 - 1000 - 500 Predicted label
In [30]:	<pre># Vectorizing and applying TF-IDF from sklearn.linear_model import LogisticRegression pipe = Pipeline([('vect', CountVectorizer()),</pre>
	<pre># Fitting the model model = pipe.fit(X_train, y_train) # Accuracy prediction = model.predict(X_test) print("accuracy: {}%".format(round(accuracy_score(y_test, prediction)*100,2))) dct['Logistic Regression'] = round(accuracy_score(y_test, prediction)*100,2)</pre>
In [31]:	accuracy: 98.83% cm = metrics.confusion_matrix(y_test, prediction) plot_confusion_matrix(cm, classes=['Fake', 'Real']) Confusion matrix, without normalization Confusion matrix -4000
	Fake - 4637 58 - 3000 - 3000 - 2000 - 1000
	Predicted label Decision Tree
In [37]:	<pre>from sklearn.tree import DecisionTreeClassifier # Vectorizing and applying TF-IDF pipe = Pipeline([('vect', CountVectorizer()),</pre>
	<pre># Fitting the model model = pipe.fit(X_train, y_train) # Accuracy prediction = model.predict(X_test) print("accuracy: {}%".format(round(accuracy_score(y_test, prediction)*100,2))) dct['Decision Tree'] = round(accuracy_score(y_test, prediction)*100,2) accuracy: 99.62%</pre>
In [33]:	cm = metrics.confusion_matrix(y_test, prediction) plot_confusion_matrix(cm, classes=['Fake', 'Real']) Confusion matrix, without normalization Confusion matrix Fake - 4680 15
	Real - 19 4266 - 1000
In [36]:	Random Forest from sklearn.ensemble import RandomForestClassifier
	<pre>pipe = Pipeline([('vect', CountVectorizer()),</pre>
In [38]:	accuracy: 99.01% cm = metrics.confusion_matrix(y_test, prediction) plot_confusion_matrix(cm, classes=['Fake', 'Real']) Confusion matrix, without normalization Confusion matrix
	Fake - 4680 15 - 3000 - 3000 - 2000 - 2000 - 1000
	Predicted label SVM
In [39]:	<pre>from sklearn import svm #Create a svm Classifier clf = svm.SVC(kernel='linear') # Linear Kernel pipe = Pipeline([('vect', CountVectorizer()),</pre>
In [40]:	<pre>model = pipe.fit(X_train, y_train) prediction = model.predict(X_test) print("accuracy: {}%".format(round(accuracy_score(y_test, prediction)*100,2))) dct['SVM'] = round(accuracy_score(y_test, prediction)*100,2) accuracy: 99.57% cm = metrics.confusion_matrix(y_test, prediction) plot_confusion_matrix(cm, classes=['Fake', 'Real'])</pre> Confusion_matrixwithout_parmalization
	Confusion matrix, without normalization Confusion matrix Fake - 4673 22 -4000 -3000
	Real - 17 4268 - 1000 - 1000 - Predicted label
In [41]:	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)
Out[41]:	plt.ylim(90,100) plt.yticks((91, 92, 93, 94, 95, 96, 97, 98, 99, 100)) ([<matplotlib.axis.ytick 0x26abce30310="" at="">,</matplotlib.axis.ytick>
	<pre><matplotlib.axis.ytick 0x26ac18d2df0="" at="">, <matplotlib.axis.ytick 0x26ad551f580="" at="">, <matplotlib.axis.ytick 0x26ad551fd00="" at="">], [Text(0, 0, ''), Text(0, 0, ''</matplotlib.axis.ytick></matplotlib.axis.ytick></matplotlib.axis.ytick></pre>
	Text(0, 0, ''), Text(0, 0, ''), Text(0, 0, '')]) 100 99 98
	97 - 96 - 95 - 94 -
	93 - 92 - 91 - Naive Bayes Logistic Regression Decision Tree Random Forest SVM
In []:	