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<pre>5 10 6 11 7 12 8 13 9 1 10 2 11 3 12 4 13 5 #To difeatur X = df y = df #To ms X_trai # perf from s classi classi classi</pre>	1 0 1 0 0 1 0 1 0 1 2 2 0 1 0 0 1 2 2 0 0 1 0 0 1 2 2 0 0 1 1 0 0 1 1 2 2 1 1 1 1
#To di featur X = df y = df #To di from s X_trai # perf from s classi classi	vide our data into attribute set and Labol: e. cols = ['Dutlook', 'Temprature', 'Humidity', 'Wind'] [[feature_cols]
<pre># perf from s classi classi Decisio</pre>	<pre>klearn.model_selection import train_test_split n, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30) form training klearn.tree import DecisionTreeClassifier fiter =DecisionTreeClassifier(criterion="entropy", random_state=100) fiter.fit(X_train, y_train) fiter.fit(X_train, y_train) fiter.epeciasifier(criterion='entropy', random_state=100) ct the response for test dataset = classifier.predict(X_test) fiter.fiter.predict(X_test) fiter.fiter.predict(X_test)</pre> Accuracy, how often is the classifier correct? klearn.metrics import accuracy_score
#Predi	ct the response for test dataset = classifier.predict(X_test) Accuracy, how often is the classifier correct? Accuracy import accuracy_score
# Mode. from s	
Actu 1 11 10 5	val Predicted 0 0 1 0 1 1 0 1
print(tklearn.metrics import classification_report, confusion_matrix confusion_matrix(y_test, y_pred)) classification_report(y_test, y_pred)) precision recall f1-score support 0 0.50 0.50 0.50 2
macr weighte	1 0.67 0.67 0.67 3 Suracy 1 0.58 0.58 0.58 0.58 5 Solid avg 0.60 0.60 0.60 5 Using disaster-tweet.csv, Write a code to classify tweets as Relevant or not Relevant using TF-IDF vectorization method Out utility libraries
import import import import %matpl # Impo from s	pandas as pd numpy as np os matplotlib.pyplot as plt seaborn as sn ootlib inline not libraries for text manipulation klearn.feature_extraction.text import TfidfVectorizer, CountVectorizer klearn.model_selection import train_test_split klearn.naive_bayes import MultinomialNB not modules for evaluation purposes not libraries for predcton klearn import metrics klearn.maive for predcton klearn import metrics klearn.maive for predcton klearn import confusion_matrix,accuracy_score,roc_auc_score,roc_curve,auc,fi_score oordcloud import WordCloud
0 1 2 3	id keyword location text target 0 ablaze NaN Communal violence in Bhainsa, Telangana. "Ston 1.0 1 ablaze NaN Telangana: Section 144 has been imposed in Bha 1.0 2 ablaze New York City Arsonist sets cars ablaze at dealership https: 1.0 3 ablaze Morgantown, WV Arsonist sets cars ablaze at dealership https: 1.0
11366 1 11367 1 11368 1	A ablaze NaN "Lord Jesus, your love brings freedom and pard 0.0
#Relev	ws × 5 columns vant columns OLUMN = 'text' COLUMN = 'target'
data = data.h O Comm Telang A	text only the text and target columns from our dataframe adata[[TEXT_COLUMN, TARGET_COLUMN]] text target munal violence in Bhainsa, Telangana. "Ston 1.0 data section 144 has been imposed in Bha 1.0 data sonist sets cars ablaze at dealership https: 1.0 data sonist sets cars ablaze at dealership https: 1.0 data sonist sets cars ablaze at dealership https: 1.0
#check data.c ['targe # Coun	t total NaN at each column in a DataFrame
Count t text target dtype: # Remo	
1 T 2 3	text target Communal violence in Bhainsa, Telangana. "Ston 1.0 Telangana: Section 144 has been imposed in Bha 1.0 Arsonist sets cars ablaze at dealership https: 1.0 Arsonist sets cars ablaze at dealership https: 1.0 "Lord Jesus, your love brings freedom and pard 0.0 "
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print(Count t text target dtype:	
	True
# Show print(print(print(print(X Train X Test Y Train	n, X_test, y_train, y_test = train_test_split(data[TEXT_COLUMN], data[TARGET_COLUMN].values , test_size=0.20, random_state=0) 'the size of our datasets 'X Train Size:',X_train.shape) 'X Test Size:',X_test.shape) 'Y Test Size:',y_train.shape) 'Y Test Size:',y_test.shape) 'Size: (6361,) Size: (1591,) Size: (6361,) Size: (1591,)
<pre>count_ # Appl # Befo train test = print(</pre>	te a Counter of tokens vectorizer = CountVectorizer(decode_error='ignore', lowercase=True, min_df=2) y it on the train data to get the vocabulary and the mapping. This vocab and mapping is then applied to the test set. ure, we convert to Unicode to avoid issues with CountVectorizer = count_vectorizer.fit_transform(X_train.values.astype('U')) count_vectorizer.transform(X_test.values.astype('U')) 'Train size: ',train.shape) 'Test size: ',test.shape)
<pre>vocab print([('hi', # try # crea</pre>	size: (6361, 6697)
<pre># Nume train # Nume test = print(print(Train s</pre>	= TfidfVectorizer(decode_error='ignore', lowercase = True, min_df=2) pricalize the train dataset = tfidf.fit_transform(X_train.values.astype('U')) pricalize the test dataset = tfidf.transform(X_test.values.astype('U')) 'Train size: ',train.shape) 'Test size: ',test.shape) size: (6361, 6697) tize: (1591, 6697)
id keyword locatio text target dtype:	object object float64 object object total NaN at each column in a DataFrame
	on 3418 0 1
<pre>data = print(2 3 5 6 7</pre>	all rows with any NaN and NaT values data.dropna() data) id keyword location \ 2 ablaze New York City 3 ablaze Morgantown, wv 5 ablaze Condon, England 7 ablaze Bharat 11362 wrecked feuille d'érable
11366 11367 11368 2 3 5 6 7 	11365 wrecked Blue State in a red sea 11366 wrecked arohanoces 11367 wrecked auroraborealis text target Arsonist sets cars ablaze at dealership https: 1.0 Arsonist sets cars ablaze at dealership https: 1.0 If this child was Chinese, this tweet would ha 0.0 Several houses have been set ablaze in Ngemsib 1.0 Asansol: A BJP office in Salanpur village was 1.0 Stell wrecked ako palagi sayo. Haha. #ALABTOPS 0.0
11365 11366 11367 11368 [7952 r data[' <ipytho A value</ipytho 	Media should have warned us well in advance. T 0.0 i feel directly attacked iconsider moonbin 0.0 i feel directly attacked iconsider moonbin 0.0 ok who remember "outcast" nd the "dora" au?? T 0.0 rows x 5 columns] target'] = data['target'].astype(int) on-input-59-b84843albd3a>:1: SettingWithCopyWarning: e is trying to be set on a copy of a slice from a DataFrame. and .loc[row_indexer,col_indexer] = value instead
See the data[# Extr vocab print(e caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy "target'] = data['target'].astype(int) act the vocabulary as a list of (word, frequency) = list(count_vectorizer.vocabularyitems()) vocab[:10]) 3604), ('kevin', 4240), ('very', 8198), ('sorry', 7131), ('about', 260), ('this', 7727), ('freight', 3144), ('train', 7890), ('has', 3509), ('derailed', 2194)]
<pre>from s from s # Defi parame 'C 'g } # Tune</pre>	kklearn.model_selection import GridSearchCV kklearn.svm import SVC ne the parameters to tune tters = { '': [1.0, 10],
<pre>from s from s text_c text_c</pre>	
model. print(print(train s test sc	m = X_train.replace((np.inf, -np.inf, np.nan), 0).reset_index(drop=True) = MultinomialNB() fit(train, y_train) "train score:", model.score(train, y_train)) "test score: ", model.score(test, y_test)) score: 0.8888679138500236 score: 0.8485229415461973
<pre># Pred y_pred print(plot_c</pre>	<pre>ficting the Test set results = model.predict(test) metrics.classification_report(y_test, y_pred, digits=5)) confussion_matrix(y_test, y_pred) coc_curve(y_test, y_pred) coc_curve(y_test, y_pred)</pre>
	1 0.94565 0.26935 0.41928 323 Suracy
1.0	- 600 - 400 - 200 0 1 Receiver Operating Characteristic
True Positive Rate	—— AUC = 0.63
cm # df	O 0.2 0.4 0.6 0.8 1.0 False Positive Rate Out_confussion_matrix(y_test, y_pred): ' Plot the confussion matrix for the target labels and predictions ''' a confusion_matrix(y_test, y_pred) Create a dataframe with the confussion matrix values [cm = pd.DataFrame(cm, range(cm.shape[0]),
#p. # sn sn pl # ROC # plot # Calc def pl	range(cm.shape[1])) tlt.figure(figsize = (10,7)) Plot the confussion matrix u.set(font_scale=1.4) #for label size u.heatmap(df_cm, annot=True, fmt='.0f', annot_kws={"size": 10})# font size t.t.show() Curve un os kill vulate the points in the ROC curve uot_roo_curve(y_test, y_pred):
fp ro pl pl pl pl pl pl	' Plot the ROC curve for the target labels and predictions''' rr, tpr, thresholds = roc_curve(y_test, y_pred, pos_label=1) rc_auc= auc(fpr, tpr) t.title('Receiver Operating Characteristic') t.plot(fpr, tpr, 'b', label = 'AUC = %0.2f' % roc_auc) t.legend(loc = 'lower right') t.plot([0, 1], [0, 1], 'r') t.xlim([0, 1]) t.ylim([0, 1]) t.ylam([0, 1]) t.ylabel('True Positive Rate') t.xlabel('False Positive Rate')
	t.xlabel('False Positive Rate') t.show()