DETECTION OF FAKE NEWS USING MACHINE LEARNING

IMPORTING THE LIBRARIES

```
In [1]:
```

```
import pandas as pd
import numpy as np
import re
import string
import nltk
nltk.download('punkt')
from nltk.tokenize import word tokenize
nltk.download('wordnet')
from nltk.corpus import wordnet as wn
from nltk.stem.wordnet import WordNetLemmatizer
from nltk.stem import WordNetLemmatizer
nltk.download('stopwords')
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import GridSearchCV
\textbf{from} \  \, \textbf{sklearn.neighbors} \  \, \textbf{import} \  \, \textbf{KNeighborsClassifier}
{\bf from} \  \, {\bf sklearn.linear\_model} \  \, {\bf import} \  \, {\bf LogisticRegression}
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import AdaBoostClassifier
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.ensemble import VotingClassifier
from sklearn.pipeline import Pipeline
from mlxtend.classifier import StackingClassifier
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import cross_val_score
import matplotlib.pyplot as plt
import itertools
from sklearn.metrics import classification_report
[nltk_data] Downloading package punkt to C:\Users\SAKSHI
[nltk_data]
                 NEERAJ\AppData\Roaming\nltk_data...
[nltk_data]
               Package punkt is already up-to-date!
```

IMPORTING THE DATASET

[nltk_data] Downloading package wordnet to C:\Users\SAKSHI

[nltk_data] Downloading package stopwords to C:\Users\SAKSHI

NEERAJ\AppData\Roaming\nltk_data...
Package wordnet is already up-to-date!

NEERAJ\AppData\Roaming\nltk_data...

Package stopwords is already up-to-date!

In [2]:

[nltk_data]

[nltk_data]

[nltk_data]

[nltk_data]

```
df_fake=pd.read_csv("Fake.csv")
df_true=pd.read_csv("True.csv")
```

In [3]:

```
df_fake.tail(10)
```

Out[3]:

	title	text	subject	date
23471	Seven Iranians freed in the prisoner swap have	21st Century Wire says This week, the historic	Middle-east	January 20, 2016
23472	#Hashtag Hell & The Fake Left	By Dady Chery and Gilbert MercierAll writers	Middle-east	January 19, 2016
23473	Astroturfing: Journalist Reveals Brainwashing	Vic Bishop Waking TimesOur reality is carefull	Middle-east	January 19, 2016
23474	The New American Century: An Era of Fraud	Paul Craig RobertsIn the last years of the 20t	Middle-east	January 19, 2016
23475	Hillary Clinton: 'Israel First' (and no peace	Robert Fantina CounterpunchAlthough the United	Middle-east	January 18, 2016
23476	McPain: John McCain Furious That Iran Treated	21st Century Wire says As 21WIRE reported earl	Middle-east	January 16, 2016
23477	JUSTICE? Yahoo Settles E-mail Privacy Class-ac	21st Century Wire says It s a familiar theme	Middle-east	January 16, 2016
23478	Sunnistan: US and Allied 'Safe Zone' Plan to T	Patrick Henningsen 21st Century WireRemember	Middle-east	January 15, 2016
23479	How to Blow \$700 Million: Al Jazeera America F	21st Century Wire says Al Jazeera America will	Middle-east	January 14, 2016
23480	10 U.S. Navy Sailors Held by Iranian Military	21st Century Wire says As 21WIRE predicted in	Middle-east	January 12, 2016

In [4]:

df_fake.shape

Out[4]:

(23481, 4)

```
In [5]:
```

```
df_true.tail(10)
```

Out[5]:

	title	text	subject	date
21407	Mata Pires, owner of embattled Brazil builder	SAO PAULO (Reuters) - Cesar Mata Pires, the ow	worldnews	August 22, 2017
21408	U.S., North Korea clash at U.N. forum over nuc	GENEVA (Reuters) - North Korea and the United	worldnews	August 22, 2017
21409	U.S., North Korea clash at U.N. arms forum on	GENEVA (Reuters) - North Korea and the United	worldnews	August 22, 2017
21410	Headless torso could belong to submarine journ	COPENHAGEN (Reuters) - Danish police said on T	worldnews	August 22, 2017
21411	North Korea shipments to Syria chemical arms a	UNITED NATIONS (Reuters) - Two North Korean sh	worldnews	August 21, 2017
21412	'Fully committed' NATO backs new U.S. approach	BRUSSELS (Reuters) - NATO allies on Tuesday we	worldnews	August 22, 2017
21413	LexisNexis withdrew two products from Chinese	LONDON (Reuters) - LexisNexis, a provider of I	worldnews	August 22, 2017
21414	Minsk cultural hub becomes haven from authorities	MINSK (Reuters) - In the shadow of disused Sov	worldnews	August 22, 2017
21415	Vatican upbeat on possibility of Pope Francis	MOSCOW (Reuters) - Vatican Secretary of State	worldnews	August 22, 2017
21416	Indonesia to buy \$1.14 billion worth of Russia	JAKARTA (Reuters) - Indonesia will buy 11 Sukh	worldnews	August 22, 2017

In [6]:

```
df_true.shape
```

Out[6]:

(21417, 4)

In [7]:

```
df_fake["class"]=0
df_true["class"]=1
```

In [8]:

```
df_fake_manual_testing=df_fake.tail(10)
df_fake.drop([23470,23480],axis=0,inplace=True)
df_true_manual_testing=df_true.tail(10)
df_true.drop([21406,21416],axis=0,inplace=True)
df_manual_testing=pd.concat([df_fake_manual_testing,df_true_manual_testing],axis=0)
df_manual_testing.to_csv("manual_testing.csv")
```

In [9]:

```
df_merge=pd.concat([df_fake,df_true],axis=0)
```

In [10]:

```
df=df_merge.drop(["subject","date"],axis=1)
df = df.sample(frac = 1)
df.head()
```

Out[10]:

	title	text	class
10231	AUDIT: Obama's IRS 'Misled' Americans to Get T	Soooo the IRS lied to Americans to prod them	0
13470	Kremlin: U.S. sanctions aimed at turning busin	MOSCOW (Reuters) - The Kremlin said on Thursda	1
22875	SYRIA: British and American Presence Directly	US paratrooper on security duty during a miss	0
2240	Watch NBC's Andrea Mitchell Get BULLIED Out O	If one thing has become abundantly clear, it s	0
17190	FAMILY THREATENED AT GUNPOINT FOR DISPLAYING C	Nothing says tolerance like putting a loaded g	0

In [11]:

```
df.isnull().sum()
```

Out[11]:

title 0
text 0
class 0
dtype: int64

DATA PREPROCCESING

```
In [12]:
```

```
def conversion(title):
 title = title.lower()
 return title
```

```
In [13]:
```

```
df["title"] = df["title"].apply(conversion)
```

In [14]:

```
def tokenization(title):
    title = word_tokenize(title)
return title
```

In [15]:

```
df["title"] = df["title"].apply(tokenization)
```

In [16]:

```
df.head()
```

Out[16]:

	title	text	class
10231	[audit, obama, s, irs, misled, americans, to,	Soooo the IRS lied to Americans to prod them	0
13470	[kremlin, u, s, sanctions, aimed, at, turning,	MOSCOW (Reuters) - The Kremlin said on Thursda	1
22875	[syria, british, and, american, presence, dire	US paratrooper on security duty during a miss	0
2240	[watch, nbc, s, andrea, mitchell, get, bullied	If one thing has become abundantly clear, it s	0
17190	[family, threatened, at, gunpoint, for, displa	Nothing says tolerance like putting a loaded g	0

In [17]:

```
lmtzr=WordNetLemmatizer()
def lemmetization(title):
    title = ' '.join([lmtzr.lemmatize(w,wn.NOUN) for w in title])
     return title
```

In [18]:

```
df["title"] = df["title"].apply(lemmetization)
```

In [19]:

df.head() Out[19]:

	title	text	class
10231	audit obama s irs misled american to get them	Soooo the IRS lied to Americans to prod them	0
13470	kremlin u s sanction aimed at turning business	MOSCOW (Reuters) - The Kremlin said on Thursda	1
22875	syria british and american presence directly e	US paratrooper on security duty during a miss	0
2240	watch nbc s andrea mitchell get bullied out of	If one thing has become abundantly clear, it s	0
17190	family threatened at gunpoint for displaying c	Nothing says tolerance like putting a loaded g	0

In [20]:

```
\textbf{from nltk.corpus } \textbf{import} \ \textbf{stopwords}
stop = stopwords.words('english')
df["title"] = df["title"].apply(lambda x: ' '.join([word for word in x.split() if word not in (stop)]))
```

```
In [21]:
```

df.head()

Out[21]:

	title	text	class
10231	audit obama irs misled american get sign obama	Soooo the IRS lied to Americans to prod them	0
13470	kremlin u sanction aimed turning business elit	MOSCOW (Reuters) - The Kremlin said on Thursda	1
22875	syria british american presence directly escal	US paratrooper on security duty during a miss	0
2240	watch nbc andrea mitchell get bullied state de	If one thing has become abundantly clear, it s	0
17190	family threatened gunpoint displaying confeder	Nothing says tolerance like putting a loaded g	0

SPLITTING DATA INTO TRAINING AND TESTING DATA

```
In [22]:
```

```
x = df.iloc[0:5000,0]
y = df.iloc[0:5000,-1]
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2)
```

VECTORIZATION

```
In [23]:
```

```
vectorization = TfidfVectorizer()
xv_train = vectorization.fit_transform(x_train)
xv_test = vectorization.transform(x_test)
```

In [24]:

```
xv = vectorization.fit_transform(x)
```

CODE TO PLOT CONFUSION MATRIX

In [25]:

```
def plot_confusion_matrix(cm, classes,
    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)
    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')
```

KNN

```
In [26]:
```

```
Fitting 3 folds for each of 192 candidates, totalling 576 fits
```

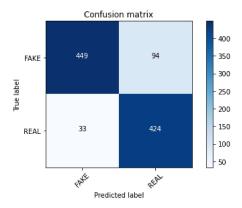
Out[26]:

```
{'metric': 'minkowski', 'n_neighbors': 25, 'weights': 'distance'}
```

In [27]:

```
knn=KNeighborsClassifier(n_neighbors=25)
knn.fit(xv_train, y_train)
pred_train = knn.predict(xv_train)
pred_test = knn.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(knn, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.11350000000000005
Variance is: 0.127
Accuracy is: 0.873
Cross Validation result is: 0.8726
Confusion \ {\tt matrix}, \ {\tt without} \ {\tt normalization}
                precision
                               recall f1-score
                                                      support
             0
                      0.93
                                  0.83
                                             0.88
                                                          543
                      0.82
                                  0.93
                                             0.87
                                                          457
                                             0.87
                                                         1000
    accuracy
   macro avg
                      0.88
                                  0.88
                                             0.87
                                                         1000
weighted avg
                      0.88
                                  0.87
                                             0.87
                                                         1000
```

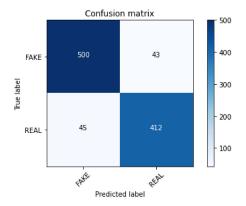


LOGISTIC REGRESSION

In [28]:

```
LR = LogisticRegression()
LR.fit(xv_train,y_train)
pred_train = LR.predict(xv_train)
pred_test = LR.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(LR, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.03274999999999946
Variance is: 0.0879999999999997
Accuracy is: 0.912
Cross Validation result is: 0.9152000000000001
Confusion \ {\tt matrix}, \ {\tt without} \ {\tt normalization}
                precision
                               recall f1-score
                                                     support
            0
                      0.92
                                 0.92
                                            0.92
                                                         543
                      0.91
                                 0.90
                                            0.90
                                                         457
                                            0.91
                                                        1000
    accuracy
   macro avg
                      0.91
                                 0.91
                                            0.91
                                                        1000
weighted avg
                      0.91
                                 0.91
                                            0.91
                                                        1000
```

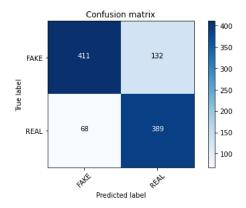


DECISION TREE

In [29]:

```
DT = DecisionTreeClassifier()
DT.fit(xv_train, y_train)
pred_train = DT.predict(xv_train)
pred_test = DT.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(DT, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.0
Variance is: 0.199999999999999999996
Accuracy is: 0.8
Cross Validation result is: 0.8472
Confusion \ {\tt matrix}, \ {\tt without} \ {\tt normalization}
                precision
                               recall f1-score
                                                      support
             0
                      0.86
                                  0.76
                                             0.80
                                                           543
                      0.75
                                  0.85
                                             0.80
                                                          457
                                             0.80
                                                         1000
    accuracy
   macro avg
                      0.80
                                  0.80
                                             0.80
                                                         1000
weighted avg
                      0.81
                                  0.80
                                             0.80
                                                         1000
```

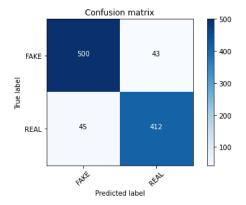


SUPPORT VECTOR CLASSIFIER

In [34]:

```
svc = SVC()
svc.fit(xv_train,y_train)
pred_train = svc.predict(xv_train)
pred_test = svc.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(svc, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.03274999999999946
Variance is: 0.0879999999999997
Accuracy is: 0.912
Cross Validation result is: 0.9152000000000001
Confusion\ {\tt matrix},\ {\tt without}\ {\tt normalization}
               precision
                              recall f1-score
                                                    support
            0
                     0.92
                                 0.92
                                            0.92
                                                         543
                     0.91
                                 0.90
                                            0.90
                                                         457
                                            0.91
                                                        1000
    accuracy
   macro avg
                     0.91
                                 0.91
                                            0.91
                                                        1000
weighted avg
                     0.91
                                 0.91
                                            0.91
                                                        1000
```

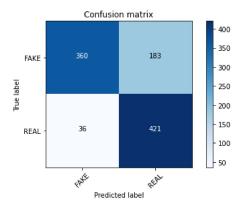


ADA BOOST CLASSIFIER (INBUILT ENSEMBLING)

In [35]:

```
ada = AdaBoostClassifier()
ada.fit(xv_train, y_train)
pred_train = ada.predict(xv_train)
pred_test = ada.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(ada, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.16825
Cross Validation result is: 0.8106
Confusion\ {\tt matrix},\ {\tt without}\ {\tt normalization}
             precision
                          recall f1-score
                                            support
          0
                  0.91
                            0.66
                                     0.77
                                                543
                  0.70
                            0.92
                                     0.79
                                                457
                                     0.78
                                               1000
   accuracy
   macro avg
                  0.80
                            0.79
                                     0.78
                                               1000
weighted avg
                  0.81
                            0.78
                                     0.78
                                               1000
```

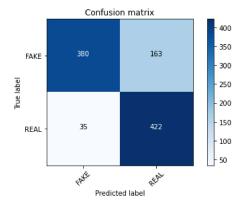


GRADIENT BOOSTING CLASSIFIER (IN BUILT ENSEMBLING)

In [36]:

```
GBC = GradientBoostingClassifier()
GBC.fit(xv_train, y_train)
pred_train = GBC.predict(xv_train)
pred_test = GBC.predict(xv_test)
print("Bais is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(GBC, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.142000000000000002
Variance is: 0.19799999999999995
Accuracy is: 0.802
Cross Validation result is: 0.825399999999999
Confusion \ {\tt matrix}, \ {\tt without} \ {\tt normalization}
                precision
                               recall f1-score
                                                     support
             0
                      0.92
                                 0.70
                                             0.79
                                                          543
                      0.72
                                 0.92
                                             0.81
                                                         457
                                             0.80
                                                         1000
    accuracy
   macro avg
                      0.82
                                 0.81
                                             0.80
                                                         1000
weighted avg
                      0.83
                                 0.80
                                             0.80
                                                         1000
```

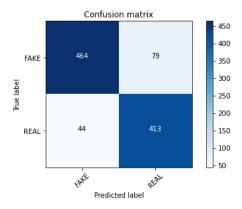


RANDOM FOREST CLASSIFIER (INBUILT ENSEMBLING)

In [37]:

```
RFC = RandomForestClassifier()
RFC.fit(xv_train, y_train)
pred_train = RFC.predict(xv_train)
pred_test = RFC.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(RFC, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.0
Variance is: 0.123
Accuracy is: 0.877
Cross Validation result is: 0.9002000000000001
Confusion \ {\tt matrix}, \ {\tt without} \ {\tt normalization}
                precision
                               recall f1-score
                                                      support
             0
                      0.91
                                  0.85
                                             0.88
                                                          543
                      0.84
                                  0.90
                                             0.87
                                                          457
                                             0.88
                                                         1000
    accuracy
   macro avg
                      0.88
                                  0.88
                                             0.88
                                                         1000
weighted avg
                      0.88
                                  0.88
                                             0.88
                                                         1000
```

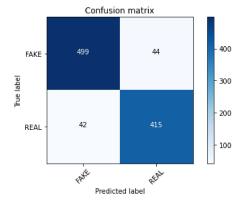


MAXIMUM VOTING CLASSIFIER (CUSTOM ENSEMBLING)

In [38]:

```
knn = KNeighborsClassifier (n\_neighbors = 25)
knn.fit(xv_train, y_train)
LR = LogisticRegression()
LR.fit(xv_train,y_train)
svc = SVC()
svc.fit(xv_train, y_train)
models = list()
logistic_regression = Pipeline([('m', LogisticRegression())])
models.append(('logistic', logistic_regression))
svc = Pipeline([('m', SVC())])
models.append(('svc', svc))
k_n_n = Pipeline([('m', KNeighborsClassifier(n_neighbors=3))])
models.append(('knn', k_n_n))
maxvoting = VotingClassifier(estimators=models, voting='hard')
maxvoting.fit(xv_train,y_train)
pred_train = maxvoting.predict(xv_train)
pred_test = maxvoting.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(maxvoting, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.014000000000000012
Cross Validation result is: 0.918800000000001
Confusion matrix, without normalization
            precision
                        recall f1-score
                                          support
                 0.92
                          0.92
                                   0.92
          0
                                             543
                 0.90
                          0.91
                                             457
                                   0.91
   accuracy
                                   0.91
                                            1000
                 0.91
                          0.91
                                   0.91
                                            1000
   macro avg
                 0.91
                          0.91
                                   0.91
                                            1000
weighted avg
```



STACKING (CUSTOM ENSEMBLING)

In [39]:

```
base1=SVC()
base2=KNeighborsClassifier(n_neighbors=25)
meta_model=LogisticRegression()
stack=StackingClassifier(classifiers=[base1,base2],meta_classifier=meta_model)
stack.fit(xv_train,y_train)
pred_train = stack.predict(xv_train)
pred_test = stack.predict(xv_test)
print("Bias is : ",1-accuracy_score(pred_train,y_train))
print("Variance is: ",1-accuracy_score(pred_test,y_test))
print("Accuracy is: ",accuracy_score(pred_test,y_test))
print("Cross Validation result is: ",cross_val_score(stack, xv, y, cv=10, scoring ='accuracy').mean())
cm=confusion_matrix(y_test,pred_test)
plot_confusion_matrix(cm, classes=['FAKE', 'REAL'])
print(classification_report(y_test, pred_test))
```

```
Bias is : 0.0010000000000000000
Variance is: 0.0849999999999996
Accuracy is: 0.915
Cross Validation result is: 0.9200000000000002
Confusion matrix, without normalization
             precision
                           recall f1-score
                                              support
           0
                   0.92
                             0.92
                                       0.92
                                                  543
           1
                   0.90
                             0.91
                                       0.91
                                                  457
   accuracy
                                       0.92
                                                 1000
                   0.91
                             0.91
                                       0.91
                                                 1000
   macro avg
                             0.92
                                       0.92
                                                 1000
weighted avg
                   0.92
```

