

```
!pip install scikit-plot
```

```
Requirement already satisfied: scikit-plot in /usr/local/lib/python3.10/dist-packages (0.3.7)
Requirement already satisfied: matplotlib>=1.4.0 in /usr/local/lib/python3.10/dist-packages (from scikit-plot) (3.7.1)
Requirement already satisfied: scikit-learn>=0.18 in /usr/local/lib/python3.10/dist-packages (from scikit-plot) (1.3.2)
Requirement already satisfied: scipy>=0.9 in /usr/local/lib/python3.10/dist-packages (from scikit-plot) (1.13.1)
Requirement already satisfied: joblib>=0.10 in /usr/local/lib/python3.10/dist-packages (from scikit-plot) (1.4.2)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (1.1.1)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (4.53.0)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (1.4.5)
Requirement already satisfied: numpy>=1.20 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (1.26.4)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (24.1)
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (9.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (3.1.2)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=1.4.0->scikit-plot) (2.8.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=0.18->scikit-plot) (3.2.0)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.7->matplotlib>=1.4.0->scikit-plot) (1.16.0)
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
sns.set()
```

```
import re
import string
from wordcloud import WordCloud
from collections import Counter
import warnings
warnings.filterwarnings('ignore')
```

```
from nltk import sent_tokenize, word_tokenize
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
```

```
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.ensemble import RandomForestClassifier
```

```
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
from sklearn.metrics import roc_curve, roc_auc_score
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
```

```
data = pd.read_excel ("/content/dataset.xlsx")
data.head()
```

```


```

	text	label
0	oh my gosh	1.0
1	trouble sleeping, confused mind, restless hear...	1.0
2	All wrong, back off dear, forward doubt. Stay ...	1.0
3	I've shifted my focus to something else but I'...	1.0
4	I'm restless and restless, it's been a month a...	1.0

```
print(data.shape)
```

```
(6982, 2)
```

```
data=data.dropna(how='any')
```

```
data['label'].value_counts()
```

```


```

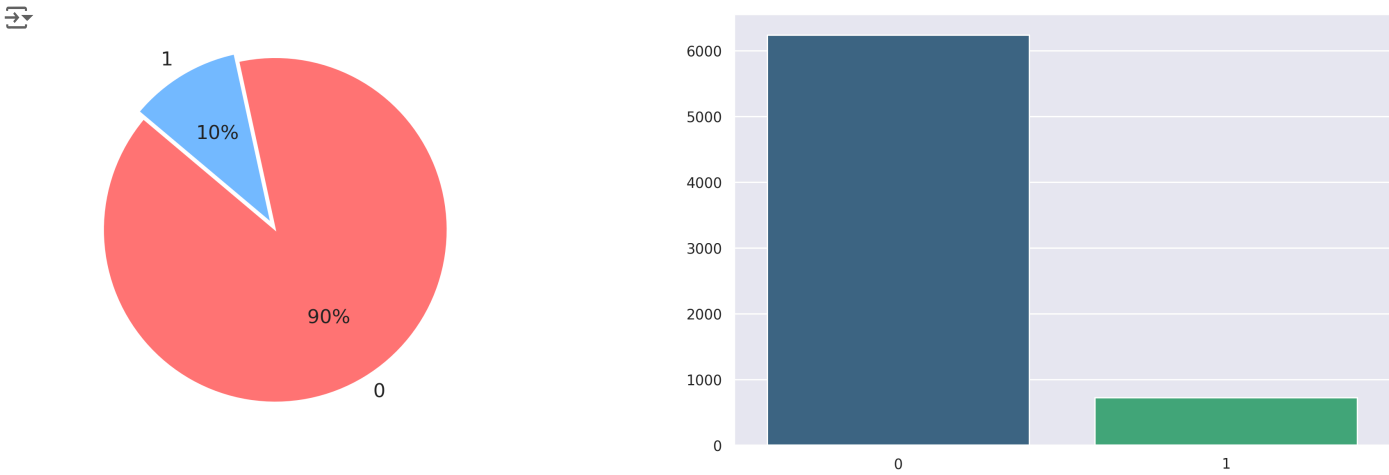
	count
label	
0.0	6240
1.0	730

```
labels = [0,1]
sizes = [6240, 730]
custom_colours = ['#ff7675', '#74b9ff']

plt.figure(figsize=(20, 6), dpi=227)
plt.subplot(1, 2, 1)
plt.pie(sizes, labels = labels, textprops={'fontsize': 15}, startangle=140,
        autopct='%1.0f%%', colors=custom_colours, explode=[0, 0.05])

plt.subplot(1, 2, 2)
sns.barplot(x=labels,y = sizes, palette= 'viridis')

plt.show()
```



```
data['Total Words'] = data['text'].apply(lambda x: len(x.split()))

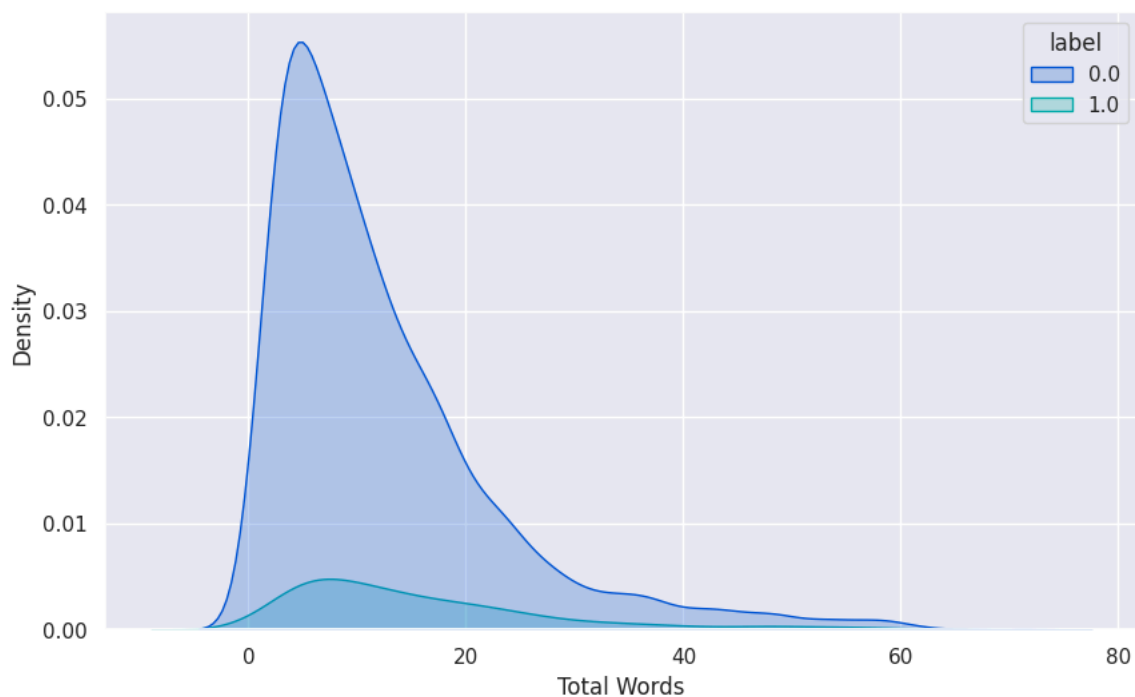
def count_total_words(text):
    char = 0
    for word in text.split():
        char += len(word)
    return char

data['Total Chars'] = data["text"].apply(count_total_words)

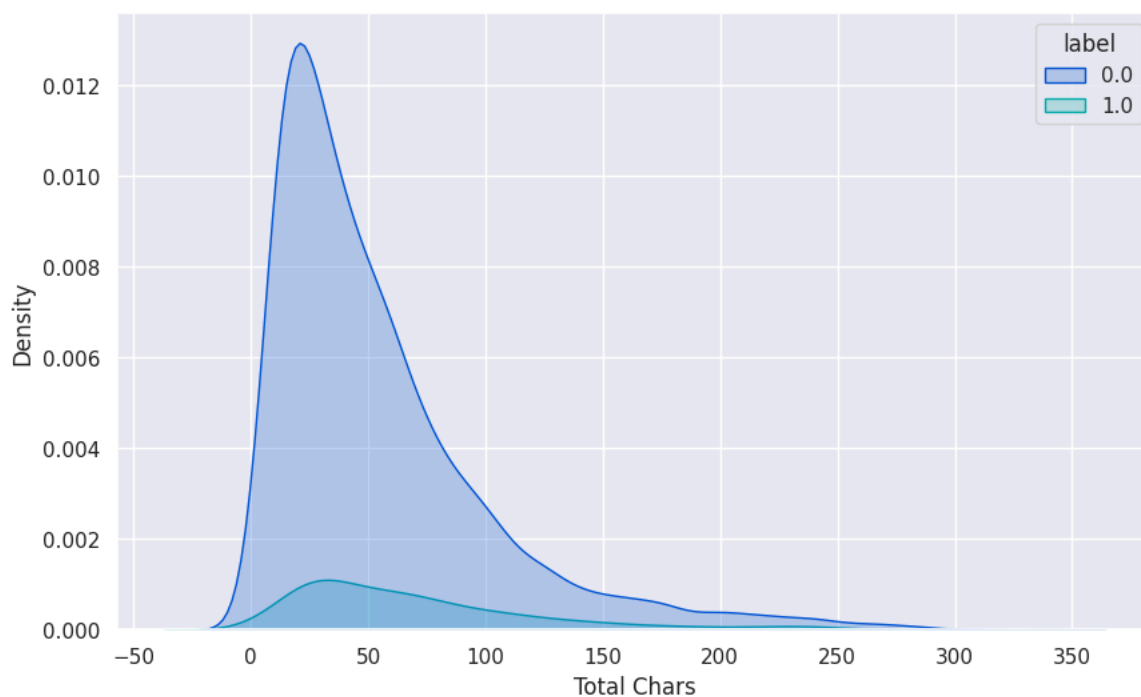
data.head()
```

	text	label	Total Words	Total Chars
0	oh my gosh	1.0	3	8
1	trouble sleeping, confused mind, restless hear...	1.0	10	55
2	All wrong, back off dear, forward doubt. Stay ...	1.0	14	65
3	I've shifted my focus to something else but I'...	1.0	11	51
4	I'm restless and restless, it's been a month n...	1.0	14	59

```
plt.figure(figsize = (10, 6))
sns.kdeplot(x = data['Total Words'], hue= data['label'], palette= 'winter', shade = True)
plt.show()
```



```
plt.figure(figsize = (10, 6))
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plt.show()
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```
data.head()
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```
def convert_lowercase(text):
    text = text.lower()
    return text

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

def remove_url(text):
    re_url = re.compile('https?://\S+|www\.\S+')
    return re_url.sub('', text)

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

exclude = string.punctuation

def remove_punc(text):
    return text.translate(str.maketrans('', '', exclude))

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

import nltk
nltk.download('punkt')
def remove_stopwords(text):
    new_list = []
    words = word_tokenize(text)
    stopwrds = stopwords.words('english')
    for word in words:
        if word not in stopwrds:
            new_list.append(word)
    return ' '.join(new_list)

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

[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data] Unzipping tokenizers/punkt.zip.

def perform_stemming(text):
    stemmer = PorterStemmer()
    new_list = []
    words = word_tokenize(text)
    for word in words:
        new_list.append(stemmer.stem(word))

    return " ".join(new_list)

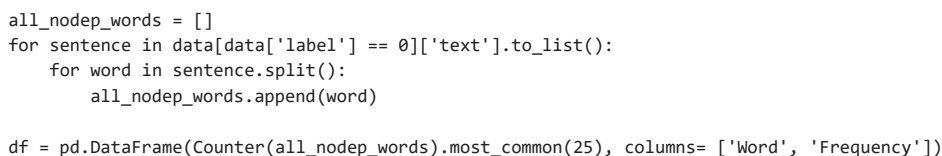
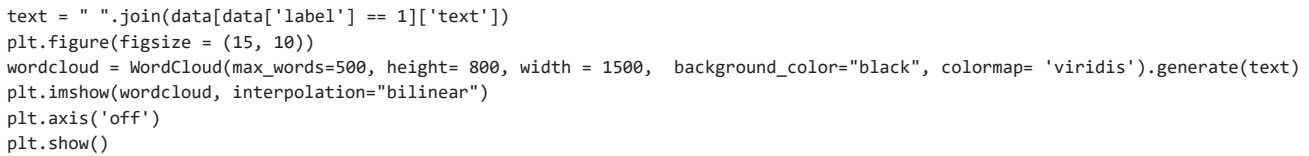
data['text'] = data['text'].apply(perform_stemming)

data['Total Words After Transformation'] = data['text'].apply(lambda x: np.log(len(x.split()))))

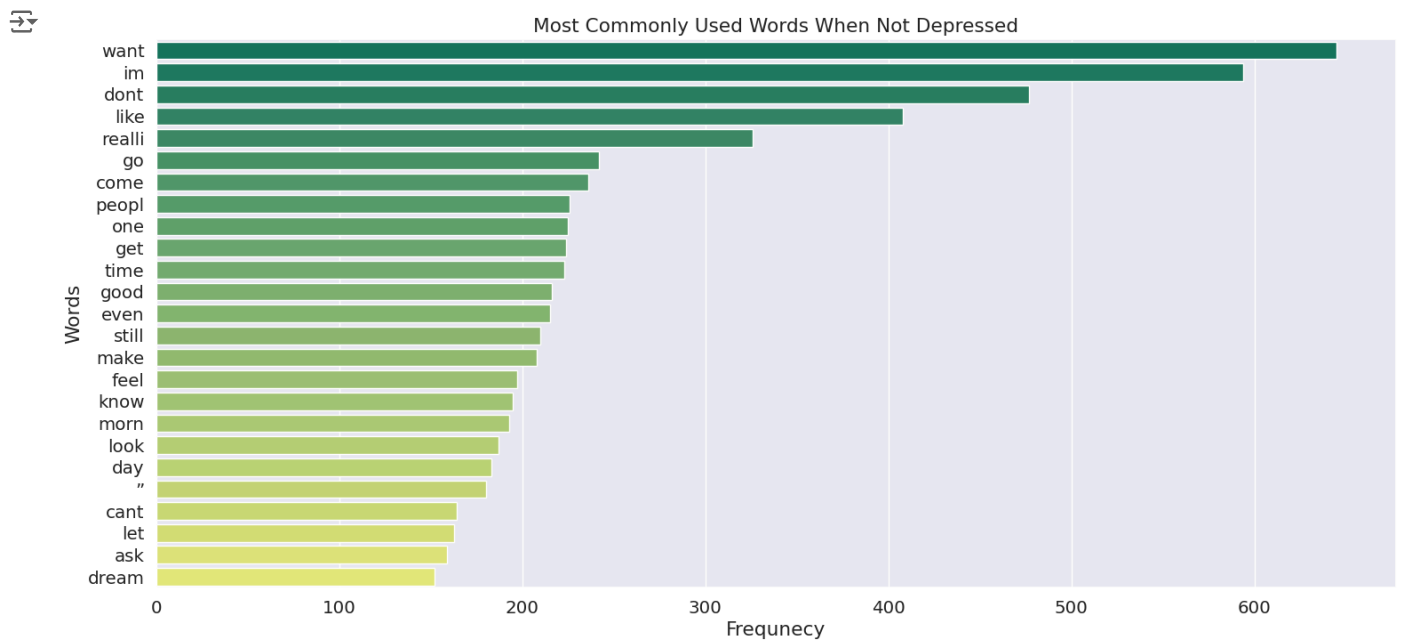
data.head()
```

	text	label	Total Words	Total Chars	Total Words After Transformation
0	oh gosh	1.0	3	8	0.693147
1	troubl sleep confus mind restless heart tune	1.0	10	55	1.945910
2	wrong back dear forward doubt stay restless re...	1.0	14	65	2.197225
3	ive shift focu someth els im still worri	1.0	11	51	2.079442
4	im restless restless month boy mean	1.0	14	59	1.791759

```
text = " ".join(data[data['label'] == 0]['text'])
plt.figure(figsize = (15, 10))
wordcloud = WordCloud(max_words=500, height= 800, width = 1500, background_color="black", colormap= 'viridis').generate(text)
plt.imshow(wordcloud, interpolation="bilinear")
plt.axis('off')
plt.show()
```



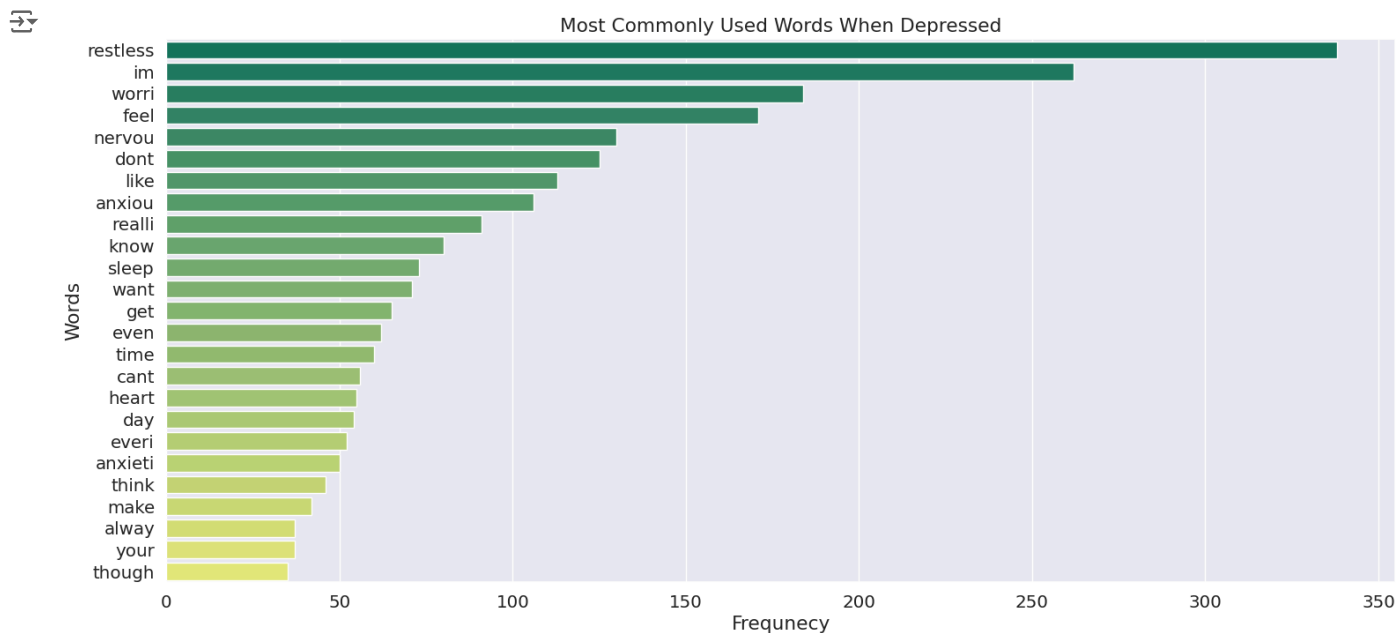
```
sns.set_context('notebook', font_scale= 1.3)
plt.figure(figsize=(18,8))
sns.barplot(y = df['Word'], x= df['Frequency'], palette= 'summer')
plt.title("Most Commonly Used Words When Not Depressed")
plt.xlabel("Frequency")
plt.ylabel("Words")
plt.show()
```



```
all_dep_words = []
for sentence in data[data['label'] == 1]['text'].to_list():
    for word in sentence.split():
        all_dep_words.append(word)

df = pd.DataFrame(Counter(all_dep_words).most_common(25), columns= ['Word', 'Frequency'])

sns.set_context('notebook', font_scale= 1.3)
plt.figure(figsize=(18,8))
sns.barplot(y = df['Word'], x= df['Frequency'], palette= 'summer')
plt.title("Most Commonly Used Words When Depressed")
plt.xlabel("Frequency")
plt.ylabel("Words")
plt.show()
```



```
X = data["text"]
y = data['label'].values

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size= 0.2, random_state= 42, stratify = y)

tfidf = TfidfVectorizer(max_features= 2500, min_df= 2)
X_train = tfidf.fit_transform(X_train).toarray()
X_test = tfidf.transform(X_test).toarray()
```

```
def train_model(model):
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    y_prob = model.predict_proba(X_test)
    accuracy = round(accuracy_score(y_test, y_pred), 3)
    precision = round(precision_score(y_test, y_pred), 3)
    recall = round(recall_score(y_test, y_pred), 3)

    print(f'Accuracy of the model: {accuracy}')
    print(f'Precision Score of the model: {precision}')
    print(f'Recall Score of the model: {recall}')

    sns.set_context('notebook', font_scale= 1.3)
    fig, ax = plt.subplots(1, 2, figsize = (25, 8))

    # Import necessary function for confusion matrix plotting
    from sklearn.metrics import ConfusionMatrixDisplay

    cm = confusion_matrix(y_test, y_pred)
    disp = ConfusionMatrixDisplay(confusion_matrix=cm)
    disp.plot(cmap=plt.cm.Blues, ax=ax[0]) # Plot on the first subplot

    from sklearn.metrics import roc_curve, roc_auc_score

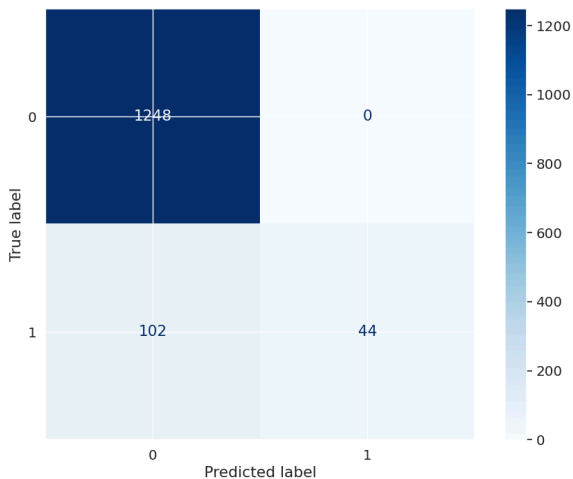
    # Assuming you have y_true (true labels) and y_score (predicted probabilities)
    fpr, tpr, _ = roc_curve(y_test, y_prob[:, 1]) # Use y_prob for class 1
    roc_auc = roc_auc_score(y_test, y_prob[:, 1])

    # Plot ROC on the second subplot
    ax[1].plot(fpr, tpr, color='darkorange', lw=2, label='ROC curve (area = %0.2f)' % roc_auc)
    ax[1].plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')
    ax[1].set_xlim([0.0, 1.0])
    ax[1].set_ylim([0.0, 1.05])
    ax[1].set_xlabel('False Positive Rate')
    ax[1].set_ylabel('True Positive Rate')
    ax[1].set_title('Receiver Operating Characteristic')
    ax[1].legend(loc="lower right")

    plt.show() # Dedent plt.show() to be outside the function

nb = MultinomialNB()
train_model(nb)
```

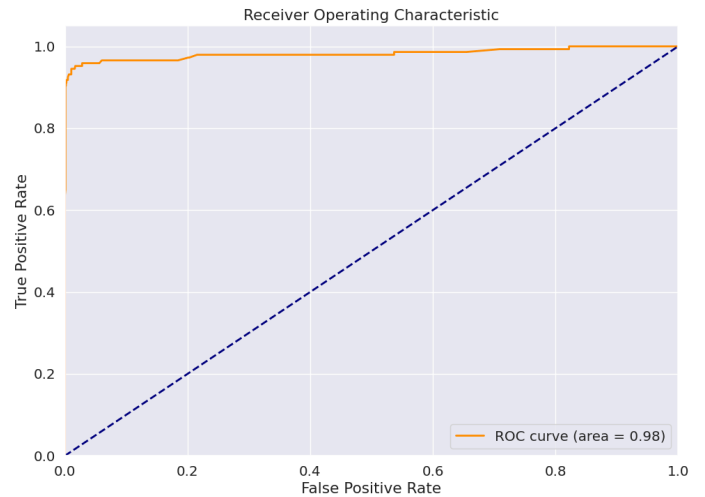
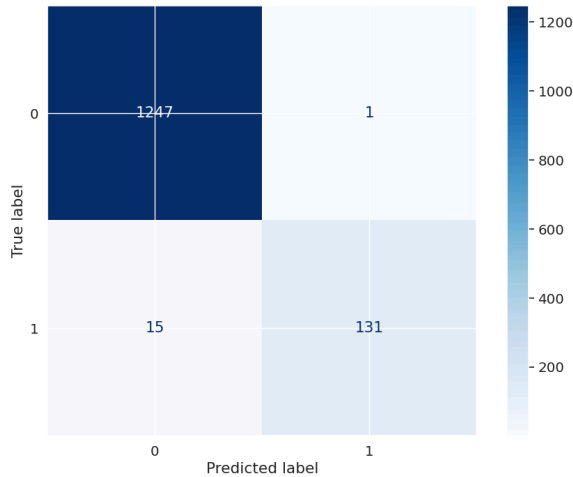
↗ Accuracy of the model: 0.927  
 Precision Score of the model: 1.0  
 Recall Score of the model: 0.301



```
rf = RandomForestClassifier(n_estimators= 300)
train_model(rf)
```



Accuracy of the model: 0.989  
 Precision Score of the model: 0.992  
 Recall Score of the model: 0.897



```
!pip install lime
# Import the LimeTabularExplainer module
from lime.lime_tabular import LimeTabularExplainer

# Get the class names
class_names = ['depressed', 'Not depressed']

# Get the feature names
feature_names = tfidf.get_feature_names_out()

# X_train is already a dense array, no need to call toarray()
X_train_dense = X_train

# Fit the Explainer on the training data set using the LimeTabularExplainer
explainer = LimeTabularExplainer(X_train_dense,
                                  feature_names=feature_names,
                                  class_names=class_names,
                                  mode='classification')
```

Requirement already satisfied: lime in /usr/local/lib/python3.10/dist-packages (0.2.0.1)  
 Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packages (from lime) (3.7.1)  
 Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from lime) (1.26.4)  
 Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (from lime) (1.13.1)  
 Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages (from lime) (4.66.5)  
 Requirement already satisfied: scikit-learn>=0.18 in /usr/local/lib/python3.10/dist-packages (from lime) (1.3.2)  
 Requirement already satisfied: scikit-image>=0.12 in /usr/local/lib/python3.10/dist-packages (from lime) (0.23.2)  
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 Requirement already satisfied: pillow>=9.1 in /usr/local/lib/python3.10/dist-packages (from scikit-image>=0.12->lime) (9.4.0)  
 Requirement already satisfied: imageio>=2.33 in /usr/local/lib/python3.10/dist-packages (from scikit-image>=0.12->lime) (2.34.2)  
 Requirement already satisfied: tifffile>=2022.8.12 in /usr/local/lib/python3.10/dist-packages (from scikit-image>=0.12->lime) (2024.12.1)  
 Requirement already satisfied: packaging>=21 in /usr/local/lib/python3.10/dist-packages (from scikit-image>=0.12->lime) (24.1)  
 Requirement already satisfied: lazy-loader>=0.4 in /usr/local/lib/python3.10/dist-packages (from scikit-image>=0.12->lime) (0.4)  
 Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=0.18->lime) (1.4.2)  
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 Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.7.1->lime) (1.2.1)  
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 Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.7->matplotlib>=3.7.1->lime) (1.16.0)

```
idx = 0 # For example, the first instance in your test set
instance = X_test[idx]

# Generate an explanation
explanation = explainer.explain_instance(instance, rf.predict_proba, num_features=10)

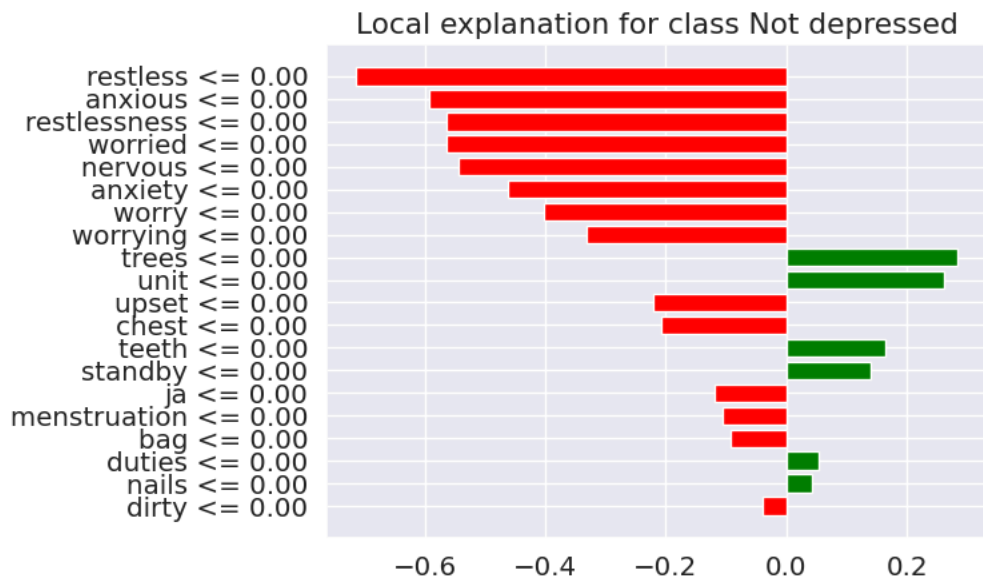
# Now you can plot the explanation
fig = explanation.as_pyplot_figure()
plt.show()
```



```
idx = 0 # For example, the first instance in your test set
instance = X_test[idx]

# Generate an explanation
explanation = explainer.explain_instance(instance, rf.predict_proba, num_features=20)

# Now you can plot the explanation
fig = explanation.as_pyplot_figure()
plt.show()
```

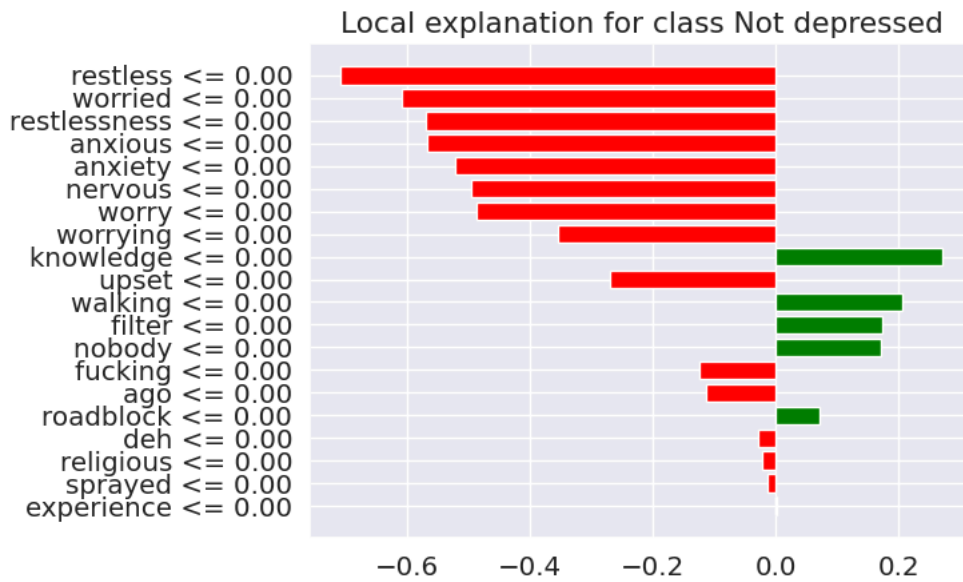


Start coding or [generate](#) with AI.

```
idx = 0 # For example, the first instance in your test set
instance = X_test[idx]

# Generate an explanation
explanation = explainer.explain_instance(instance, rf.predict_proba, num_features=20)

# Now you can plot the explanation
fig = explanation.as_pyplot_figure()
plt.show()
```



Start coding or [generate](#) with AI.

```
!pip install lime
import lime.lime_text # Import the correct module for text data
import numpy as np

# Assuming 'clf' is your trained text classifier (e.g., text_clf from previous examples)
explainer = lime.lime_text.LimeTextExplainer(
    class_names=['not depressed', 'depressed'] # Replace with your actual class names
)
idx = 2 # Index of the instance to explain

# **Make sure 'vectorizer' is defined here,
# it should be the same one used during training**
vectorizer = tfidf

exp = explainer.explain_instance(X.iloc[idx],
                                lambda texts: rf.predict_proba(vectorizer.transform(texts)), # Vectorize the text before prediction
                                num_features=5)

# Visualize the explanation
exp.as_pyplot_figure()
```