# emotion-detection-nlp

January 6, 2024

```
[38]: #import necessary library
      import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      import spacy
      from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LogisticRegression
      from sklearn.neighbors import KNeighborsClassifier
      from sklearn.naive_bayes import MultinomialNB
      from sklearn.ensemble import RandomForestClassifier
      from sklearn.feature extraction.text import TfidfVectorizer
      from sklearn.pipeline import Pipeline
      from sklearn.metrics import classification_report
      from sklearn.metrics import confusion_matrix
      import warnings
      warnings.filterwarnings('ignore')
```

### 1 Load The Dataset

```
[40]: Emotion

joy 5362

sadness 4666

anger 2159

fear 1937

love 1304

surprise 572

Name: count, dtype: int64
```

### 1.1 Map emotions to numerical values

```
[41]:

Description Emotion num

i didnt feel humiliated sadness

i i can go from feeling so hopeless to so damned... sadness

i m grabbing a minute to post i feel greedy wrong anger

i am ever feeling nostalgic about the fireplac... love

i am feeling grouchy anger

2
```

```
[42]: #Bar plot for emotion distribution

plt.figure(figsize=(8, 6))

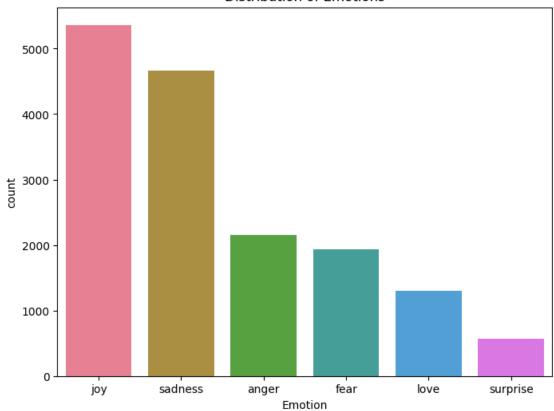
sns.countplot(x='Emotion', data=df, order=df['Emotion'].value_counts().index,

→palette="husl")

plt.title('Distribution of Emotions')

plt.show()
```

### Distribution of Emotions



```
[43]: # Histogram plot for emotion distribution with KDE

plt.figure(figsize=(10, 6))

sns.histplot(x='Emotion', data=df, kde=True, palette='Set2', element='bars',

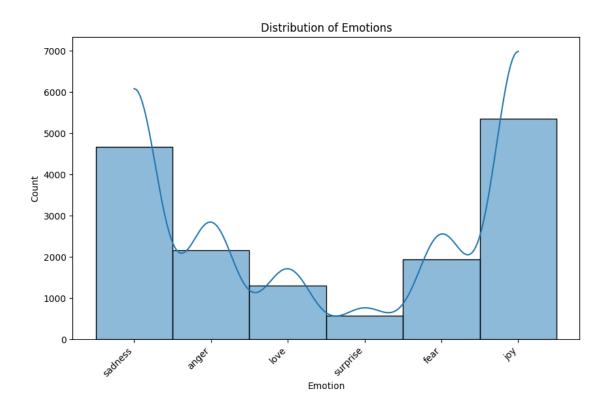
stat='count',

common_norm=False)

plt.title('Distribution of Emotions')

plt.xticks(rotation=45, ha='right')

plt.show()
```



```
[44]: # Load spaCy English model
nlp = spacy.load("en_core_web_sm")

# Function to preprocess text using spaCy
def preprocess(text):
    doc = nlp(text)
    filtered_tokens = []
    for token in doc:
        if token.is_stop or token.is_punct:
            continue
        else:
            filtered_tokens.append(token.lemma_)
        return " ".join(filtered_tokens)
```

```
[45]: df['processed_text'] = df["Description"].apply(preprocess)
df
```

```
[45]:
                                                    Description
                                                                  Emotion
                                        i didnt feel humiliated
      0
                                                                  sadness
      1
             i can go from feeling so hopeless to so damned... sadness
      2
              im grabbing a minute to post i feel greedy wrong
                                                                    anger
      3
             i am ever feeling nostalgic about the fireplac...
                                                                   love
                                           i am feeling grouchy
                                                                    anger
```

i just had a very brief time in the beanbag an... sadness 15995 15996 i am now turning and i feel pathetic that i am... 15997 i feel strong and good overall 15998 i feel like this was such a rude comment and i... anger 15999 i know a lot but i feel so stupid because i ca... sadness Emotion\_num processed\_text 0 not feel humiliate 1 1 feel hopeless damned hopeful care awake 2 m grab minute post feel greedy wrong 3 feel nostalgic fireplace know property 2 feel grouchy 15995 1 brief time beanbag say anna feel like beat 15996 turn feel pathetic wait table sub teaching degree feel strong good overall 15997 0 2 feel like rude comment m glad t 15998 15999 know lot feel stupid portray [16000 rows x 4 columns]

### 2 Train-test split

```
[46]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(df['Description'],

odf['Emotion_num'], test_size=0.2, random_state=42)
```

### 3 KNN

[48]: print("Classification Report:\n", classification\_report(y\_test, knn\_y\_pred))

#### Classification Report:

	precision	recall	f1-score	support
0	0.66	0.87	0.75	1021
1	0.73	0.82	0.77	946

2	0.76	0.61	0.68	427
3	0.77	0.52	0.62	397
4	0.77	0.35	0.48	296
5	0.64	0.30	0.41	113
accuracy			0.71	3200
macro avg	0.72	0.58	0.62	3200
weighted avg	0.72	0.71	0.69	3200

### 4 Logistic Regression

```
[50]: print("Classification Report:\n", classification_report(y_test, lr_y_pred))
```

### Classification Report:

	precision	recall	f1-score	support
0	0.79	0.96	0.87	1021
1	0.87	0.94	0.90	946
2	0.90	0.77	0.83	427
3	0.85	0.70	0.77	397
4	0.89	0.55	0.68	296
5	0.88	0.43	0.58	113
accuracy			0.84	3200
macro avg	0.86	0.73	0.77	3200
weighted avg	0.85	0.84	0.83	3200

# 5 Multinomial Naive Bayes

```
[52]: print("Classification Report:\n", classification_report(y_test, nb_y_pred))
```

Classification Report:

		precision	recall	f1-score	support
	0	0.55	0.99	0.71	1021
	1	0.69	0.90	0.78	946
	2	0.91	0.17	0.29	427
	3	0.93	0.11	0.19	397
	4	1.00	0.01	0.01	296
	5	0.00	0.00	0.00	113
accura	асу			0.62	3200
macro a	avg	0.68	0.36	0.33	3200
weighted a	avg	0.71	0.62	0.52	3200

### 6 Random Forest

## [54]: print("Classification Report:\n", classification\_report(y\_test, rfc\_y\_pred))

Classification Report:

	precision	recall	f1-score	support
0	0.79	0.94	0.86	1021
1	0.91	0.89	0.90	946
2	0.90	0.80	0.85	427
3	0.86	0.80	0.83	397
4	0.87	0.66	0.75	296
5	0.87	0.66	0.75	113
accuracy			0.85	3200
macro avg	0.87	0.79	0.82	3200
${\tt weighted} \ {\tt avg}$	0.86	0.85	0.85	3200

## 7 Confusion Matrix Heatmap for Random Forest

```
[56]: cm = confusion_matrix(y_test, rfc_y_pred)
[56]: array([[960,
                     26,
                            4,
                                 8,
                                            2],
                                     21,
              [ 67, 844,
                          17,
                                11,
                                       6,
                                            1],
                     30, 342,
              [ 43,
                               12,
                                            0],
                                       Ο,
              [ 44,
                     13,
                           12, 319,
                                       1,
                                            8],
              [ 88,
                      6,
                            6,
                                 1, 195,
                                            0],
              [ 12,
                      4,
                            Ο,
                                22,
                                           75]])
                                       Ο,
[58]: # Plot confusion matrix
          plt.figure(figsize=(10, 7))
          sns.heatmap(cm, annot=True, fmt='d', cmap='coolwarm')
          plt.xlabel('Prediction')
          plt.ylabel('Truth')
          plt.show()
                   960
             0 -
                                                                                     - 800
                                                                                     - 600
                              30
                                        342
                                         12
                                                   319
                                                                                     - 400
                                                                                     - 200
                    ó
                              i
                                                              4
                                                                         5
```

Prediction