This notebook is our final model which is running a Support Vector Machines Model with the best parameters (from support_vector_machine.ipynb) using Faizan's preprocessing tools excluding the augmented text function using back translation.

Accuracy: 0.7684041851258365

```
In [12]: import pandas as pd
         import numpy as np
         import re
         import string
         import nltk
         from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize
         from sklearn.model selection import train test split, GridSearchCV
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.svm import SVC
         from sklearn.pipeline import Pipeline
         from sklearn.metrics import classification_report, accuracy_score, confusion
         import plotly.figure_factory as ff
 In [3]: # Load the data
         path = './kaggle_sentiment_data.csv'
         df = pd.read_csv(path)
 In [4]: # Handle NaN values in the statement column
         df['statement'] = df['statement'].fillna('')
 In [5]: # Data Preprocessing
         def preprocess text(text):
             text = text.lower() # Lowercase text
             text = re.sub(r'\[.*?\]', '', text) # Remove text in square brackets
             text = re.sub(r'https?://\S+|www\.\S+', '', text) # Remove links
             text = re.sub(r'<.*?>+', '', text) # Remove HTML tags
             text = re.sub(r'[%s]' % re.escape(string.punctuation), '', text) # Remo
             text = re.sub(r'\n', '', text) # Remove newlines
             text = re.sub(r'\w*\d\w*', '', text) # Remove words containing numbers
             return text
 In [6]: # Tokenization and Stopwords Removal
         stop words = set(stopwords.words('english'))
         def remove_stopwords(text):
             tokens = word tokenize(text)
             tokens = [word for word in tokens if word not in stop_words]
             return ' '.join(tokens)
 In [7]: # Preprocess the text data
         df['cleaned_statement'] = df['statement'].apply(preprocess_text).apply(remov
         # Ensure no NaN values
         df['cleaned_statement'] = df['cleaned_statement'].fillna('')
```

```
# Splitting the data
         X = df['cleaned statement']
         v = df['status']
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, rar
 In [8]: pipeline = Pipeline([
             ('tfidf', TfidfVectorizer(max_features=2500, ngram_range=(1, 1))),
             ('svm', SVC(C=10, kernel='rbf', gamma='scale'))
         ])
 In [9]: # Train the pipeline
         pipeline.fit(X_train, y_train)
 Out[9]:
                  Pipeline
              TfidfVectorizer
                    SVC
In [10]: # Evaluate on the test set
         y_pred = pipeline.predict(X_test)
In [11]: print("Accuracy:", accuracy_score(y_test, y_pred))
         print("Classification Report:\n", classification_report(y_test, y_pred))
        Accuracy: 0.7684041851258365
        Classification Report:
                               precision
                                            recall f1-score
                                                               support
                                             0.76
                                                       0.80
                                                                  778
                     Anxiety
                                   0.84
                                                       0.77
                     Bipolar
                                   0.89
                                             0.69
                                                                  575
                  Depression
                                   0.69
                                             0.75
                                                       0.72
                                                                 3081
                      Normal
                                   0.84
                                             0.95
                                                       0.90
                                                                 3270
        Personality disorder
                                   0.91
                                             0.49
                                                       0.64
                                                                  240
                      Stress
                                   0.71
                                             0.48
                                                       0.57
                                                                  534
                    Suicidal
                                   0.70
                                             0.64
                                                       0.67
                                                                 2131
                                                       0.77
                                                                10609
                    accuracy
                   macro avg
                                   0.80
                                             0.68
                                                       0.72
                                                                10609
                weighted avg
                                   0.77
                                             0.77
                                                       0.76
                                                                 10609
In [27]: # Confusion Matrix
         cm = confusion_matrix(y_test, y_pred)
         cm_fig = ff.create_annotated_heatmap(
             z=cm.
             x=list(("Anxiety", "Bipolar", "Depression", "Normal", "Personality Disor
             y=list(("Anxiety", "Bipolar", "Depression", "Normal", "Personality Disor
             annotation text=cm,
             colorscale='Viridis'
```

```
cm_fig.update_layout(title='Confusion Matrix')
cm_fig.update_layout(title='Confusion Matrix', width=800, height=600)
cm_fig.show()
```