This notebook runs Naive Bayes Classifier with Faizan's preprocessing tools without the augmented text function.

Accuracy: 0.678386275803563

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In [4]: import pandas as pd
         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, Stratifi
         from sklearn.feature extraction.text import TfidfVectorizer
         from sklearn.metrics import classification report, accuracy score, confusion
         from sklearn.naive bayes import MultinomialNB
         import plotly.figure factory as ff
 In [5]: # Load the data
         path = './kaggle_sentiment_data.csv'
         df = pd.read csv(path)
 In [6]: # Handle NaN values in the statement column
         df['statement'] = df['statement'].fillna('')
 In [7]: # Data Preprocessing
         def preprocess text(text):
             text = text.lower() # Lowercase text
             text = re.sub(r'\setminus[.*?\setminus]', '', 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 [8]: # 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 [17]: # 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']
         y = df['status']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, rar
In [18]: # Vectorization
         vectorizer = TfidfVectorizer(max features=10000)
         X_train_tfidf = vectorizer.fit_transform(X_train)
         X_test_tfidf = vectorizer.transform(X_test)
In [19]: # Define parameter grid for MultinomialNB
         param_grid = {
             'alpha': [0, 0.0000001, 0.1, 0.5, 1.0, 2.0, 5.0],
         cv = StratifiedKFold(n splits=5, shuffle=True, random state=42)
In [20]: # GridSearchCV with MultinomialNB
         grid search = GridSearchCV(
             MultinomialNB(),
             param_grid,
             scoring='accuracy',
             cv=cv,
             verbose=1,
             n_{jobs}=-1
         grid_search.fit(X_train_tfidf, y_train)
        Fitting 5 folds for each of 7 candidates, totalling 35 fits
        /opt/anaconda3/lib/python3.12/site-packages/sklearn/naive_bayes.py:890: Runt
        imeWarning: divide by zero encountered in log
          self.feature log prob = np.log(smoothed fc) - np.log(
        /opt/anaconda3/lib/python3.12/site-packages/sklearn/naive bayes.py:890: Runt
        imeWarning: divide by zero encountered in log
          self.feature log prob = np.log(smoothed fc) - np.log(
        /opt/anaconda3/lib/python3.12/site-packages/sklearn/naive_bayes.py:890: Runt
        imeWarning: divide by zero encountered in log
          self.feature_log_prob_ = np.log(smoothed_fc) - np.log(
        /opt/anaconda3/lib/python3.12/site-packages/sklearn/naive bayes.py:890: Runt
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        /opt/anaconda3/lib/python3.12/site-packages/sklearn/naive_bayes.py:890: Runt
        imeWarning: divide by zero encountered in log
          self.feature_log_prob_ = np.log(smoothed_fc) - np.log(
Out[20]:
                 GridSearchCV
          ▶ estimator: MultinomialNB
                MultinomialNB
In [21]: # Get best model and hyperparameters
         best_model = grid_search.best_estimator_
         best_params = grid_search.best_params_
```

```
print("Best Parameters:", best_params)
        Best Parameters: {'alpha': 0.1}
In [22]: # Make predictions
         y_pred = best_model.predict(X_test_tfidf)
         # Evaluate the model
         print("Accuracy:", accuracy_score(y_test, y_pred))
         print(classification_report(y_test, y_pred))
        Accuracy: 0.6736732962578943
                              precision
                                           recall f1-score
                                                               support
                                              0.64
                     Anxiety
                                   0.78
                                                        0.71
                                                                   778
                     Bipolar
                                   0.80
                                              0.52
                                                        0.63
                                                                   575
                                                        0.63
                  Depression
                                   0.53
                                              0.79
                                                                  3081
                      Normal
                                   0.85
                                              0.80
                                                        0.82
                                                                  3270
        Personality disorder
                                   0.93
                                              0.21
                                                        0.35
                                                                   240
                      Stress
                                   0.76
                                              0.23
                                                        0.35
                                                                   534
                    Suicidal
                                   0.67
                                              0.54
                                                        0.60
                                                                  2131
                                                        0.67
                                                                 10609
                    accuracy
                                   0.76
                                              0.53
                                                        0.58
                                                                 10609
                   macro avg
                weighted avg
                                   0.71
                                              0.67
                                                        0.67
                                                                 10609
In [15]: # Confusion Matrix
         cm = confusion_matrix(y_test, y_pred)
         cm fig = ff.create annotated heatmap(
             z=cm,
             x=list(set(y_test)),
             y=list(set(y_test)),
             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()