This notebook runs the gradient boosting machine with the best parameters (from gradient_boosting_machine.ipynb) using Faizan's preprocessing tools excluding the augmented text function using back translation

Accuracy: 0.752945612216043

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In [1]: 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.ensemble import GradientBoostingClassifier
        from sklearn.metrics import classification report, accuracy score
In [2]: # Load the data
        path = './kaggle_sentiment_data.csv'
        df = pd.read csv(path)
In [3]: # Handle NaN values in the statement column
        df['statement'] = df['statement'].fillna('')
In [4]: # 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 [5]: # 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 [6]: # 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']
```

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y = df['status']
        # Vectorization
        vectorizer = TfidfVectorizer(max_features=10000)
        X_vec = vectorizer.fit_transform(X)
        X_train, X_test, y_train, y_test = train_test_split(X_vec, y, test_size=0.2,
In [7]: # Gradient Boosting Model
        gbm = GradientBoostingClassifier(learning_rate=0.1, max_depth=5, n_estimator
In [8]:
        gbm.fit(X train, y train)
Out[8]:
                        GradientBoostingClassifier
        GradientBoostingClassifier(max_depth=5, n_estimators=300)
In [9]: y_pred = gbm.predict(X_test)
        # Evaluate performance
        print("Accuracy:", accuracy_score(y_test, y_pred))
        print("Classification Report:\n", classification_report(y_test, y_pred))
       Accuracy: 0.752945612216043
       Classification Report:
                              precision
                                           recall f1-score
                                                               support
                    Anxiety
                                  0.81
                                             0.72
                                                       0.76
                                                                  778
                                                       0.77
                                                                  575
                    Bipolar
                                  0.85
                                             0.70
                 Depression
                                  0.69
                                             0.73
                                                       0.71
                                                                 3081
                                  0.81
                     Normal
                                             0.95
                                                       0.88
                                                                 3270
       Personality disorder
                                  0.80
                                             0.46
                                                      0.58
                                                                  240
                     Stress
                                             0.50
                                                       0.57
                                  0.67
                                                                  534
                   Suicidal
                                  0.70
                                             0.61
                                                      0.65
                                                                 2131
                                                       0.75
                                                                10609
                   accuracy
                  macro avg
                                  0.76
                                             0.67
                                                       0.70
                                                                10609
                                             0.75
                                                       0.75
               weighted avg
                                  0.75
                                                                10609
```