Accuracy: 0.85

Classification Report:

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	precision	recall	f1-score	support
negative	0.83	0.88	0.85	4961
positive	0.87	0.82	0.85	5039
accuracy			0.85	10000
macro avg	0.85	0.85	0.85	10000
weighted avg	0.85	0.85	0.85	10000

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score, classification_report
# Sample dataset with labeled documents (positive or negative)
data = pd.read_csv('G:/Games/preseed/sample.csv')
# List of polarity words
positive_words = ["good", "happy", "excellent", "awesome", "amazing"]
negative_words = ["bad", "sad", "terrible", "awful", "disappointing"]
# Function to classify documents based on polarity words
def classify_document(document):
  pos_count = sum(1 for word in document.split() if word in positive_words)
  neg_count = sum(1 for word in document.split() if word in negative_words)
  if pos_count > neg_count:
    return 'positive'
  elif neg_count > pos_count:
    return 'negative'
  else:
    return 'neutral'
# Apply classification to the dataset
data['predicted_sentiment'] = data['review'].apply(classify_document)
# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(data['review'], data['sentiment'], test_size=0.2,
random state=42)
```

```
# Create a CountVectorizer to convert text data to numerical features
vectorizer = CountVectorizer()

X_train_counts = vectorizer.fit_transform(X_train)

X_test_counts = vectorizer.transform(X_test)

# Train a Multinomial Naive Bayes classifier
classifier = MultinomialNB()
classifier.fit(X_train_counts, y_train)

# Predict document sentiment
y_pred = classifier.predict(X_test_counts)

# Evaluate the model
accuracy = accuracy_score(y_test, y_pred)

report = classification_report(y_test, y_pred)

print(f'Accuracy: {accuracy:.2f}')

print('Classification Report:\n', report)
```