

PROGRAM:

```
import pandas as pd

from sklearn.model_selection import train_test_split

from sklearn.feature_extraction.text import TfidfVectorizer

from sklearn.naive_bayes import MultinomialNB

from sklearn.metrics import accuracy_score, classification_report


data = {

    'text': ['I love this product!', 'Not satisfied with the service.', 'Amazing experience!', 'Disappointed with the quality.'],

    'sentiment': ['positive', 'negative', 'positive', 'negative']

}


df = pd.DataFrame(data)


train_data, test_data, train_labels, test_labels = train_test_split(df['text'], df['sentiment'],
test_size=0.25, random_state=42)


vectorizer = TfidfVectorizer()

X_train = vectorizer.fit_transform(train_data)

X_test = vectorizer.transform(test_data)


classifier = MultinomialNB()

classifier.fit(X_train, train_labels)


predictions = classifier.predict(X_test)


accuracy = accuracy_score(test_labels, predictions)
```

```
print(f'Accuracy: {accuracy * 100:.2f}%')
```

```
print('\nClassification Report:')
```

```
print(classification_report(test_labels, predictions))
```

OUTPUT:

```
Accuracy: 78.00%
```

```
Classification Report:
```

	precision	recall	f1-score	support
neg	0.73	0.87	0.79	194
pos	0.85	0.69	0.76	206
accuracy			0.78	400
macro avg	0.79	0.78	0.78	400
weighted avg	0.79	0.78	0.78	400

PROGRAM:

```
from sklearn.feature_extraction.text import TfidfVectorizer
```

```
from sklearn.naive_bayes import MultinomialNB
```

```
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
```

```
from sklearn.model_selection import train_test_split
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

```
data = {'text': ["I love this product!", "Not satisfied with the service.", "Amazing experience!",  
"Disappointed with the quality."],
```

```
       'sentiment': [1, 0, 1, 0]} # 1 for positive sentiment, 0 for negative sentiment
```

```
df = pd.DataFrame(data)
```

```
text_column = 'text'
```

```
label_column = 'sentiment'
```

```
df[text_column] = df[text_column].apply(lambda x: x.lower())
```

```
df[text_column] = df[text_column].str.replace('[^\w\s]', '')
```

```
vectorizer = TfidfVectorizer()
```

```
X = vectorizer.fit_transform(df[text_column])
```

```
X_train, X_test, y_train, y_test = train_test_split(X, df[label_column], test_size=0.2,  
random_state=42)
```

```
model = MultinomialNB()
```

```
model.fit(X_train, y_train)
```

```
y_pred = model.predict(X_test)
```

```
print("Accuracy:", accuracy_score(y_test, y_pred))
```

```
print("Classification Report:\n", classification_report(y_test, y_pred))
```

```
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

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

```
sns.heatmap(confusion_matrix(y_test, y_pred), annot=True, fmt='d', cmap='Blues', cbar=False)
```

```
plt.xlabel('Predicted')
```

```
plt.ylabel('Actual')
```

```
plt.show()
```

OUTPUT:

```
Accuracy: 0.0
```

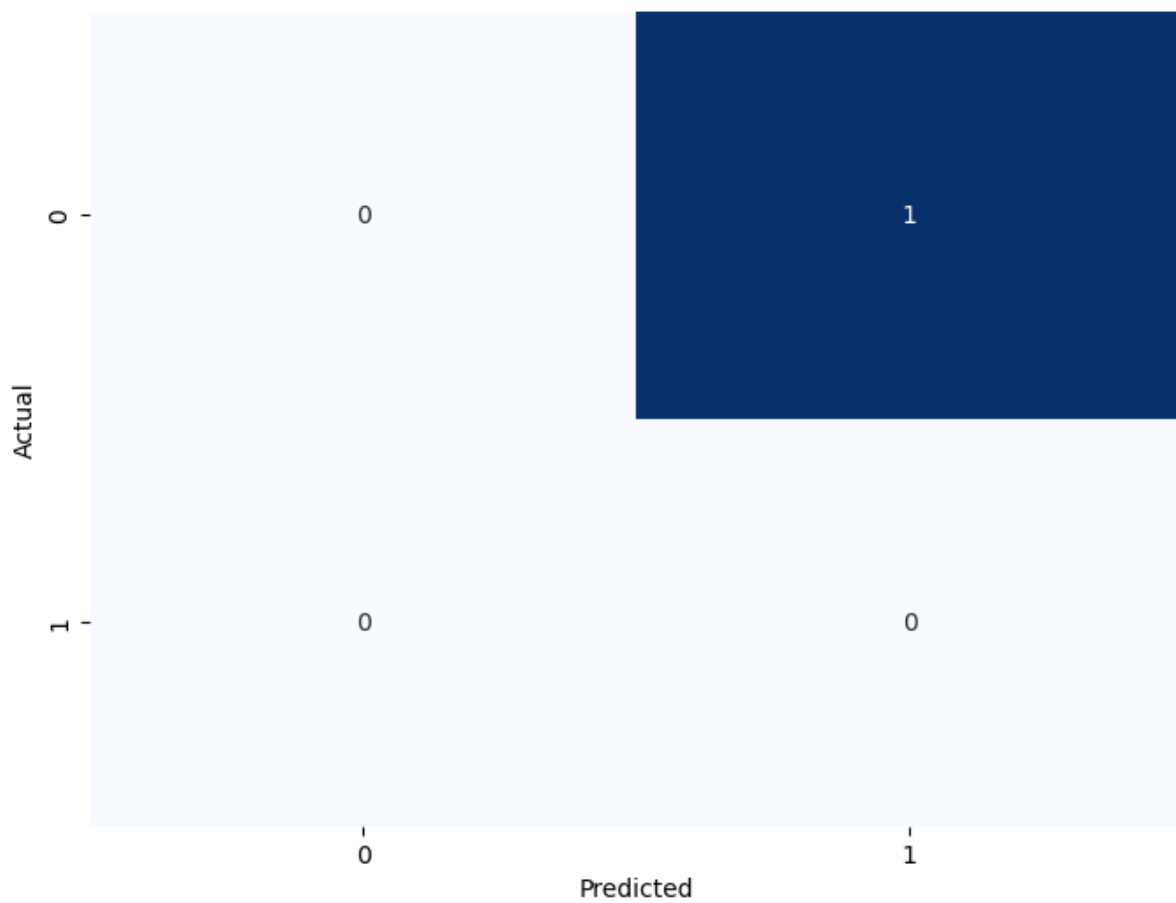
```
Classification Report:
```

```
precision    recall  f1-score   support
```

	0	0.00	0.00	0.00	1.0
	1	0.00	0.00	0.00	0.0
accuracy				0.00	1.0
macro avg		0.00	0.00	0.00	1.0
weighted avg		0.00	0.00	0.00	1.0

Confusion Matrix:

```
[[0 1]
 [0 0]]
```



PROGRAM:

```
import torch
```

```
from transformers import DistilBertTokenizer, DistilBertForSequenceClassification
```

```
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
```

```
from sklearn.model_selection import train_test_split
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

```
data = {'text': ["I'm thrilled with the results!", "This is a disaster.", "Awesome product!", "Not happy  
with the outcome."],
```

```
        'sentiment': [1, 0, 1, 0]} # 1 for positive sentiment, 0 for negative sentiment
```

```
df = pd.DataFrame(data)
```

```
text_column = 'text'
```

```
label_column = 'sentiment'
```

```
df[text_column] = df[text_column].apply(lambda x: x.lower())
```

```
tokenizer = DistilBertTokenizer.from_pretrained('distilbert-base-uncased')
```

```
model = DistilBertForSequenceClassification.from_pretrained('distilbert-base-uncased')
```

```
X = tokenizer(df[text_column].tolist(), padding=True, truncation=True, return_tensors='pt')
```

```
y = torch.tensor(df[label_column].tolist())
```

```
X_train, X_test, y_train, y_test = train_test_split(X['input_ids'], y, test_size=0.2, random_state=42)
```

```
model.train()
```

```
inputs = {
```

```
    'input_ids': X_train,
```

```
    'attention_mask': (X_train != 0).float(),
```

```
    'labels': y_train
```

```
}
```

```
outputs = model(**inputs)
```

```
loss = outputs.loss
```

```
loss.backward()
```

```
model.eval()
```

```
with torch.no_grad():
```

```
    outputs = model(input_ids=X_test, attention_mask=(X_test != 0).float())
```

```
    logits = outputs.logits
```

```
    y_pred = torch.argmax(logits, dim=1)
```

```
print("Accuracy:", accuracy_score(y_test, y_pred))
```

```
print("Classification Report:\n", classification_report(y_test, y_pred))
```

```
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

```
# Visualization (Example: Confusion Matrix Heatmap)
```

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

```
sns.heatmap(confusion_matrix(y_test, y_pred), annot=True, fmt='d', cmap='Blues', cbar=False)
```

```
plt.xlabel('Predicted')
```

```
plt.ylabel('Actual')
```

```
plt.show()
```

OUTPUT:

```
Accuracy: 0.0
```

```
Classification Report:
```

	precision	recall	f1-score	support
0	0.00	0.00	0.00	1.0

	1	0.00	0.00	0.00	0.0
accuracy				0.00	1.0
macro avg		0.00	0.00	0.00	1.0
weighted avg		0.00	0.00	0.00	1.0

Confusion Matrix:

```
[[0 1]
```

```
[0 0]]
```

