```
import pandas as pd
from sklearn.model selection import train test split
from sklearn.feature extraction.text import CountVectorizer,
TfidfVectorizer
from sklearn.svm import SVC
from sklearn.naive bayes import MultinomialNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.neural network import MLPClassifier
from sklearn.metrics import classification report, accuracy score
d = pd.read csv('/content/data.csv')
t = d['Sentence'].values
l = d['Sentiment'].values
tr d, tt d, tr l, tt l = train test split(t, l, test size=0.2)
def evaluate model(model, train features, train labels, test features,
test labels):
    model.fit(train_features, train labels)
    predictions = model.predict(test features)
    accuracy = accuracy score(test labels, predictions)
    print(classification report(test labels, predictions))
    print(f"Accuracy: {accuracy}")
    return accuracy
sample sentences = ['This is an outstanding Movie. Don\'t dare to beat
Vijay Setupati. Once it was dared by Hrithik Roshan, not equivalent to
his nail.'l
models = {
    'SVM': SVC(),
    'Naive Bayes': MultinomialNB(),
    'Decision Tree': DecisionTreeClassifier(),
    'Random Forest': RandomForestClassifier(),
```

```
'MLP Classifier': MLPClassifier(max iter=300)
}
vectorizers = {
    'CountVectorizer': CountVectorizer(),
    'TfidfVectorizer': TfidfVectorizer()
}
results = {}
for vec_name, vectorizer in vectorizers.items():
    print(f"\nUsing {vec name}:\n")
    fe tr = vectorizer.fit transform(tr d)
    fe tt = vectorizer.transform(tt d)
    results[vec name] = {}
    for model name, model in models.items():
        print(f"\n{model name}:\n")
        accuracy = evaluate model(model, fe tr, tr l, fe tt, tt l)
        results[vec name][model name] = accuracy
        nfe = vectorizer.transform(sample sentences)
        nop = model.predict(nfe)
        print(f"Prediction for sample sentences: {nop}")
print("\nAccuracy Results:")
for vec name in results:
    print(f"\n{vec name}:")
    for model name in results[vec name]:
        print(f"{model name}: {results[vec name][model name]}")
Using CountVectorizer:
SVM:
                           recall f1-score
              precision
                                              support
                   0.40
                             0.10
                                       0.16
                                                   172
    negative
                             0.90
                                       0.77
                   0.68
                                                   624
    neutral
                                       0.65
                   0.72
                             0.59
                                                   373
    positive
```

accuracy			0.68	1169
macro avg	0.60	0.53	0.53	1169
weighted avg	0.65	0.68	0.64	1169

Accuracy: 0.6809238665526091

Prediction for sample sentences: ['neutral']

Naive Bayes:

	precision	recall	f1-score	support
negative neutral positive	0.48 0.72 0.72	0.34 0.84 0.61	0.39 0.78 0.66	172 624 373
accuracy macro avg weighted avg	0.64 0.68	0.60 0.69	0.69 0.61 0.68	1169 1169 1169

Accuracy: 0.6937553464499572

Prediction for sample sentences: ['positive']

Decision Tree:

	precision	recall	f1-score	support
negative neutral positive	0.22 0.66 0.60	0.21 0.64 0.65	0.22 0.65 0.63	172 624 373
accuracy macro avg weighted avg	0.50 0.58	0.50 0.58	0.58 0.50 0.58	1169 1169 1169

Accuracy: 0.5808383233532934

Prediction for sample sentences: ['neutral']

Random Forest:

	precision	recall	f1-score	support
negative neutral positive	0.17 0.66 0.75	0.10 0.83 0.58	0.13 0.73 0.65	172 624 373
accuracy macro avg weighted avg	0.53 0.62	0.50 0.64	0.64 0.50 0.62	1169 1169 1169

Accuracy: 0.6398631308810949

Prediction for sample sentences: ['neutral']

MLP Classifier:

precision	recall	f1-score	support
0.25	0.21	0.23	172
0.70	0.76	0.73	624
0.75	0.69	0.72	373
		0.66	1169
0.57	0.55	0.56	1169
0.65	0.66	0.65	1169
	0.25 0.70 0.75	0.25 0.21 0.70 0.76 0.75 0.69 0.57 0.55	0.25 0.21 0.23 0.70 0.76 0.73 0.75 0.69 0.72 0.66 0.57 0.55 0.56

Accuracy: 0.6578272027373824

Prediction for sample sentences: ['positive']

Using TfidfVectorizer:

SVM:

	precision	recall	f1-score	support
negative neutral positive	0.23 0.68 0.78	0.08 0.88 0.62	0.12 0.77 0.69	172 624 373
accuracy macro avg weighted avg	0.56 0.64	0.53 0.68	0.68 0.53 0.65	1169 1169 1169

Accuracy: 0.6809238665526091

Prediction for sample sentences: ['neutral']

Naive Bayes:

	precision	recall	f1-score	support
negative neutral positive	0.67 0.63 0.71	0.02 0.97 0.37	0.04 0.76 0.49	172 624 373
accuracy macro avg weighted avg	0.67 0.66	0.46 0.64	0.64 0.43 0.57	1169 1169 1169

Accuracy: 0.6415739948674081

Prediction for sample sentences: ['neutral']

Decision Tree:

	precision	recall	f1-score	support
negative neutral positive	0.21 0.65 0.59	0.24 0.62 0.60	0.23 0.64 0.60	172 624 373
accuracy macro avg weighted avg	0.49 0.57	0.49 0.56	0.56 0.49 0.56	1169 1169 1169

Accuracy: 0.5585970915312233

Prediction for sample sentences: ['neutral']

Random Forest:

	precision	recall	f1-score	support
negative	0.17	0.09	0.12	172
neutral	0.66	0.83	0.74	624
positive	0.77	0.58	0.66	373
accuracy			0.64	1169
macro avg	0.53	0.50	0.51	1169
weighted avg	0.62	0.64	0.62	1169

Accuracy: 0.6449957228400343

Prediction for sample sentences: ['neutral']

MLP Classifier:

	precision	recall	f1-score	support
negative	0.25	0.20	0.23	172
neutral	0.70	0.76	0.73	624
positive	0.73	0.69	0.71	373
accuracy	0.75	0.00	0.65	1169
macro avg	0.56	0.55	0.55	1169
weighted avg	0.64	0.65	0.65	1169

Accuracy: 0.6544054747647562

Prediction for sample sentences: ['positive']

Accuracy Results:

CountVectorizer:

SVM: 0.6809238665526091

Naive Bayes: 0.6937553464499572 Decision Tree: 0.5808383233532934 Random Forest: 0.6398631308810949 MLP Classifier: 0.6578272027373824 TfidfVectorizer:

SVM: 0.6809238665526091

Naive Bayes: 0.6415739948674081 Decision Tree: 0.5585970915312233 Random Forest: 0.6449957228400343 MLP Classifier: 0.6544054747647562