```
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.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report
import matplotlib.pyplot as plt
data = pd.read_csv('spam.csv', encoding='latin-1')
data = data[['v1', 'v2']]
data.columns = ['label', 'message']
data['label'] = data['label'].map({'ham': 0, 'spam': 1})
X_train, X_test, y_train, y_test = train_test_split(data['message'], data['label'], test_size=0.2, random_state=42)
tfidf_vectorizer = TfidfVectorizer(stop_words='english')
X_train_tfidf = tfidf_vectorizer.fit_transform(X_train)
X_test_tfidf = tfidf_vectorizer.transform(X_test)
classifiers = {
    'Naive Bayes': MultinomialNB(),
    'Logistic Regression': LogisticRegression(),
    'Support Vector Machine': SVC()
}
for name, classifier in classifiers.items():
   classifier.fit(X_train_tfidf, y_train)
    y_pred = classifier.predict(X_test_tfidf)
   accuracy = accuracy_score(y_test, y_pred)
   print(f"{name} Accuracy: {accuracy}")
   print(classification_report(y_test, y_pred))
best_model = LogisticRegression()
best_model.fit(X_train_tfidf, y_train)
def classify_sms(message):
    message_tfidf = tfidf_vectorizer.transform([message])
    prediction = best_model.predict(message_tfidf)
    return "spam" if prediction[0] == 1 else "legitimate"
spam_count = data['label'].value_counts()[1]
legitimate_count = data['label'].value_counts()[0]
plt.figure(figsize=(8, 6))
plt.bar(['Spam', 'Legitimate'], [spam_count, legitimate_count], color=['red', 'green'])
plt.title('Distribution of Spam vs Legitimate Messages')
plt.xlabel('Message Type')
plt.ylabel('Count')
plt.show()
```

| Naive Bayes Accuracy: 0.9668161434977578            |           |        |          |         |
|---|-----------|--------|----------|---------|
| -   | precision | recall | f1-score | support |
|   |           | 4 00   |          | 0.55    |
| 0   | 0.96      |        | 0.98     | 965     |
| 1   | 1.00      | 0.75   | 0.86     | 150     |
| accuracy  |           |        | 0.97     | 1115    |
| macro avg   | 0.98      | 0.88   | 0.92     | 1115    |
| weighted avg  | 0.97      | 0.97   | 0.96     | 1115    |
| weighted avg  | 0.57      | 0.57   | 0.50     | 1113    |
| Logistic Regression Accuracy: 0.9524663677130045    |           |        |          |         |
|   | precision | recall | f1-score | support |
|   |           |        |          |         |
| 0   | 0.95      | 1.00   | 0.97     | 965     |
| 1   | 0.97      | 0.67   | 0.79     | 150     |
|   |           |        |          |         |
| accuracy  |           |        | 0.95     | 1115    |
| macro avg   | 0.96      | 0.83   | 0.88     | 1115    |
| weighted avg  | 0.95      | 0.95   | 0.95     | 1115    |
| Support Vector Machine Accuracy: 0.9766816143497757 |           |        |          |         |
| Support Vect  |           | -      |          |         |
|   | precision | recall | f1-score | support |
| 0   | 0.97      | 1.00   | 0.99     | 965     |
| 1   | 0.99      | 0.83   | 0.91     | 150     |
| 1   | 0.55      | 0.03   | 0.51     | 130     |
| accuracy  |           |        | 0.98     | 1115    |
| macro avg   | 0.98      | 0.92   | 0.95     | 1115    |
| weighted avg  | 0.98      | 0.98   | 0.98     | 1115    |
| biicca avg  | 0.50      | 0.50   | 0.50     | 1113    |

