

```
In [28]: !pip install pandas numpy matplotlib seaborn scikit-learn

Requirement already satisfied: pandas in /home/sargam/.conda/envs/notebook
s/lib/python3.10/site-packages (2.3.3)
Requirement already satisfied: numpy in /home/sargam/.conda/envs/notebook
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da/envs/notebooks/lib/python3.10/site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /home/sargam/.conda/envs/no
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Requirement already satisfied: contourpy>=1.0.1 in /home/sargam/.conda/en
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Requirement already satisfied: packaging>=20.0 in /home/sargam/.conda/en
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Requirement already satisfied: joblib>=1.2.0 in /home/sargam/.conda/envs/n
otebooks/lib/python3.10/site-packages (from scikit-learn) (1.5.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in /home/sargam/.cond
a/envs/notebooks/lib/python3.10/site-packages (from scikit-learn) (3.6.0)
Requirement already satisfied: six>=1.5 in /home/sargam/.conda/envs/note
books/lib/python3.10/site-packages (from python-dateutil>=2.8.2->pandas) (1.
17.0)
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In [29]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix, classification_report, accu
```

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In [30]: data = pd.read_csv('/home/sargam/things/college/daa/lp3/ml/emails/emails.
```

```
In [31]: print("Dataset shape:", data.shape)
print(data.info())
print(data.head())
```

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Dataset shape: (5172, 3002)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5172 entries, 0 to 5171
Columns: 3002 entries, Email No. to Prediction
dtypes: int64(3001), object(1)
memory usage: 118.5+ MB
None
   Email No. the to ect and for of a you hou ... connevey jay
\ 
0 Email 1 0 0 1 0 0 0 2 0 0 ... 0 0
1 Email 2 8 13 24 6 6 2 102 1 27 ... 0 0
2 Email 3 0 0 1 0 0 0 8 0 0 ... 0 0
3 Email 4 0 5 22 0 5 1 51 2 10 ... 0 0
4 Email 5 7 6 17 1 5 2 57 0 9 ... 0 0

   valued lay infrastructure military allowing ff dry Prediction
0 0 0 0 0 0 0 0 0
1 0 0 0 0 0 0 1 0
2 0 0 0 0 0 0 0 0
3 0 0 0 0 0 0 0 0
4 0 0 0 0 0 0 1 0

[5 rows x 3002 columns]

```

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In [32]: print("Missing values in each column:\n", data.isnull().sum())
```

```

Missing values in each column:
Email No. 0
the 0
to 0
ect 0
and 0
...
military 0
allowing 0
ff 0
dry 0
Prediction 0
Length: 3002, dtype: int64

```

```
In [33]: X = data.iloc[:, 1:-1].values # all rows, all columns except first and last
y = data.iloc[:, -1].values # last column: Prediction (1=spam, 0=not spam)
```

```
In [34]: # Step 5: Train-test split (70% train, 30% test)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
```

```
In [35]: # Step 6: Scaling
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
```

```
In [36]: # Step 7: K-Nearest Neighbors (KNN) Classification
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X_train_scaled, y_train)
y_pred_knn = knn.predict(X_test_scaled)
print("KNN Accuracy:", accuracy_score(y_test, y_pred_knn))
print("KNN Classification Report:\n", classification_report(y_test, y_pred_knn))
print("KNN Confusion Matrix:\n", confusion_matrix(y_test, y_pred_knn))
```

```

KNN Accuracy: 0.821520618556701
KNN Classification Report:
      precision    recall   f1-score   support
0         0.97     0.78     0.86     1102
1         0.63     0.93     0.75      450

   accuracy          0.82     1552
macro avg       0.80     0.85     0.81     1552
weighted avg    0.87     0.82     0.83     1552

KNN Confusion Matrix:
[[856 246]
 [ 31 419]]

```

```
In [37]: # Step 8: SVM Classification
svm = SVC(kernel='linear', random_state=42)
svm.fit(X_train_scaled, y_train)
y_pred_svm = svm.predict(X_test_scaled)
print("SVM Accuracy:", accuracy_score(y_test, y_pred_svm))
print("SVM Classification Report:\n", classification_report(y_test, y_pred_svm))
print("SVM Confusion Matrix:\n", confusion_matrix(y_test, y_pred_svm))
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SVM Accuracy: 0.9516752577319587
SVM Classification Report:
      precision    recall   f1-score   support
0         0.97     0.97     0.97     1102
1         0.92     0.92     0.92      450

   accuracy          0.95     1552
macro avg       0.94     0.94     0.94     1552
weighted avg    0.95     0.95     0.95     1552

```

```

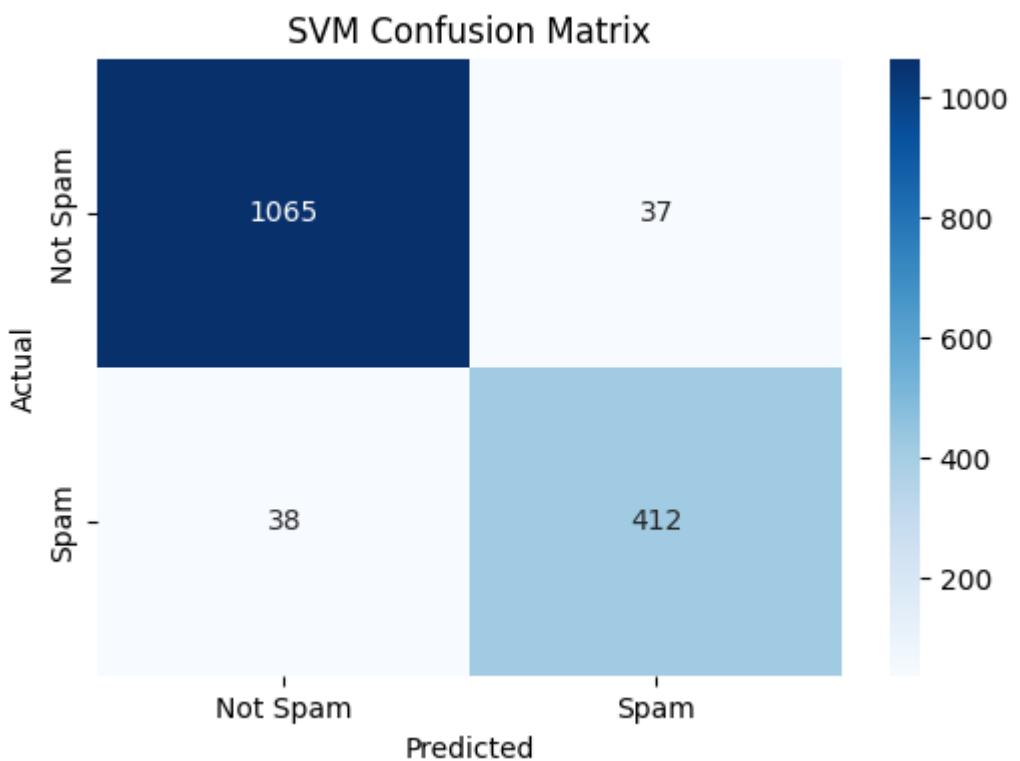
SVM Confusion Matrix:
[[1065  37]
 [ 38 412]]

```

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In [38]: # Step 9: Compare performance
if accuracy_score(y_test, y_pred_svm) > accuracy_score(y_test, y_pred_knn):
    print("SVM performed better.")
else:
    print("KNN performed better.")
```

SVM performed better.

```
In [39]: # Step 10: (Optional) Visualize confusion matrix for SVM
plt.figure(figsize=(6, 4))
sns.heatmap(confusion_matrix(y_test, y_pred_svm), annot=True, fmt='d',
            xticklabels=['Not Spam', 'Spam'], yticklabels=['Not Spam', 'Spam'])
plt.title('SVM Confusion Matrix')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
```



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In [ ]:
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