In [1]: pip install pandas scikit-learn nltk

Requirement already satisfied: pandas in ./anaconda3/lib/python3.11/sit e-packages (1.5.3) Requirement already satisfied: scikit-learn in ./anaconda3/lib/python3. 11/site-packages (1.3.2) Requirement already satisfied: nltk in ./anaconda3/lib/python3.11/sitepackages (3.8.1) Requirement already satisfied: python-dateutil>=2.8.1 in ./anaconda3/li b/python3.11/site-packages (from pandas) (2.8.2) Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3. 11/site-packages (from pandas) (2022.7) Requirement already satisfied: numpy>=1.21.0 in ./anaconda3/lib/python 3.11/site-packages (from pandas) (1.24.3) Requirement already satisfied: scipy>=1.5.0 in ./anaconda3/lib/python3. 11/site-packages (from scikit-learn) (1.10.1) Requirement already satisfied: joblib>=1.1.1 in ./anaconda3/lib/python 3.11/site-packages (from scikit-learn) (1.2.0) Requirement already satisfied: threadpoolctl>=2.0.0 in ./anaconda3/lib/ python3.11/site-packages (from scikit-learn) (2.2.0) Requirement already satisfied: click in ./anaconda3/lib/python3.11/site

-packages (from nltk) (8.0.4)
Requirement already satisfied: regex>=2021.8.3 in ./anaconda3/lib/pytho

n3.11/site-packages (from nltk) (2022.7.9)
Requirement already satisfied: tqdm in ./anaconda3/lib/python3.11/site-

packages (from nltk) (4.65.0)

Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/s ite-packages (from python-dateutil>=2.8.1->pandas) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

In [2]: import pandas as pd from sklearn.feature_extraction.text import TfidfVectorizer from sklearn.model_selection import train_test_split from sklearn.svm import SVC from sklearn.metrics import classification_report, accuracy_score import nltk #Natural Langunage Tool Kit from nltk.corpus import stopwords #StopWords are commonly used words from nltk.stem import SnowballStemmer

In [3]: | nltk.download('stopwords')

import re

[nltk_data] Downloading package stopwords to
[nltk_data] /Users/venkatasrideepthisrikotapeetamabaram/nltk_data.
[nltk_data] ...
[nltk_data] Package stopwords is already up-to-date!

Out[3]: True

```
In [4]: df = pd.read_csv("spam_ham_dataset.csv")
df
```

Out [4]:

	Unnamed: 0	label	text	label_num
0	605	ham	Subject: enron methanol ; meter # : 988291\r\n	0
1	2349	ham	Subject: hpl nom for january 9 , 2001\r\n(see	0
2	3624	ham	Subject: neon retreat\r\nho ho ho , we ' re ar	0
3	4685	spam	Subject: photoshop , windows , office . cheap \dots	1
4	2030	ham	Subject: re : indian springs\r\nthis deal is t	0
5166	1518	ham	Subject: put the 10 on the ft\r\nthe transport	0
5167	404	ham	Subject: 3 / 4 / 2000 and following noms\r\nhp	0
5168	2933	ham	Subject: calpine daily gas nomination\r\n>\r\n	0
5169	1409	ham	Subject: industrial worksheets for august 2000	0
5170	4807	spam	Subject: important online banking alert\r\ndea	1

5171 rows × 4 columns

In [5]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5171 entries, 0 to 5170
Data columns (total 4 columns):
                 Non-Null Count Dtype
     Column
     Unnamed: 0 5171 non-null
 0
                                 int64
 1
     label
                 5171 non-null
                                 object
 2
                 5171 non-null
     text
                                 object
 3
     label_num
                 5171 non-null
                                 int64
dtypes: int64(2), object(2)
memory usage: 161.7+ KB
```

```
In [6]: #Data Preprocessing
```

```
def preprocessingText(text):
    text = re.sub('[^a-zA-Z]',' ',text)
    text.lower()
    words = text.split()
    words = [word for word in words if word not in stopwords.words('engl stemmer = SnowballStemmer('english')
    words = [stemmer.stem(word) for word in words]
    return ''.join(words)
```

```
In [7]: # Apply preprocessing to the text data
         df['text'] = df['text'].apply(preprocessingText)
         df['text']
Out[7]: 0
                 subjectenronmethanolmeterfollownotegavemondayp...
                 subjecthplnomjanuariseeattachfilehplnolxlshpln...
         1
         2
                 subject neon retreath ohohoar oundwonder time yearne...
         3
                 subjectphotoshopwindowofficcheapmaintrendabasd...
         4
                 subjectindianspringdealbooktecopyrrevenuunders...
                 subjectputfttransportvolumdecreascontractthank...
         5166
         5167
                 subjectfollownomhpltakeextrammcfweekendtrinext...
         5168
                 subjectcalpindailigasnominjulimentionearlierho...
         5169
                 subjectindustriworksheetaugustactivattachworks...
                 subjectimportonlinbankalertdearvalucitizensrba...
         5170
         Name: text, Length: 5171, dtype: object
In [8]: # Feature extraction using TF-IDF
         vectorizer = TfidfVectorizer(max features=3000)
         X = vectorizer.fit transform(df['text'])
         Χ
Out[8]: <5171x3000 sparse matrix of type '<class 'numpy.float64'>'
                 with 3554 stored elements in Compressed Sparse Row format>
In [9]: # Convert labels to binary (0 for ham, 1 for spam)
         y = df['label'].apply(lambda x: 1 if x == 'spam' else 0)
Out[9]: 0
                 0
         1
                 0
         2
                 0
         3
                 1
                 0
                 . .
         5166
                 0
         5167
                 0
         5168
                 0
         5169
                 0
         5170
         Name: label, Length: 5171, dtype: int64
In [10]: # Split data into training and testing sets
         X train, X test, y train, y test = train test split(X, y, test size=0.2,
         # Train SVM model
         svm = SVC(kernel='linear')
         svm.fit(X_train, y_train)
Out[10]:
                   SVC
          SVC(kernel='linear')
```

```
In [11]: # Make predictions
         y_pred = svm.predict(X_test)
         y_pred
Out[11]: array([0, 0, 0, ..., 0, 0, 0])
In [12]: # Evaluate the model
         print("Accuracy:", accuracy_score(y_test, y_pred))
         print("Classification Report:")
         print(classification_report(y_test, y_pred))
         Accuracy: 0.7265700483091787
         Classification Report:
                        precision
                                     recall f1-score
                                                         support
                     0
                             0.72
                                       1.00
                                                  0.84
                                                             742
                     1
                                                  0.07
                                                             293
                             1.00
                                       0.03
                                                  0.73
                                                            1035
             accuracy
            macro avg
                             0.86
                                       0.52
                                                  0.45
                                                            1035
         weighted avg
                             0.80
                                       0.73
                                                  0.62
                                                            1035
 In []:
```