Could not connect to the reCAPTCHA service. Please check your internet connection and reload to get a reCAPTCHA challenge.

1. IMPORT LIBRARIES & SETUP

```
!pip install nltk scikit-learn matplotlib seaborn joblib
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import nltk
import string
import joblib
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.ensemble import RandomForestClassifier
from sklearn.svm import LinearSVC
from sklearn.metrics import (
   accuracy_score, precision_score, recall_score, f1_score,
   confusion matrix, classification report
)
# Download NLTK resources
nltk.download('stopwords')
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
ps = PorterStemmer()
Requirement already satisfied: nltk in /usr/local/lib/python3.12/dist-packages (3.9.1)
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.12/dist-packages (1.6.1)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.12/dist-packages (3.10.0)
Requirement already satisfied: seaborn in /usr/local/lib/python3.12/dist-packages (0.13.2)
Requirement already satisfied: joblib in /usr/local/lib/python3.12/dist-packages (1.5.2)
Requirement already satisfied: click in /usr/local/lib/python3.12/dist-packages (from nltk) (8.3.0)
Requirement already satisfied: regex>=2021.8.3 in /usr/local/lib/python3.12/dist-packages (from nltk) (2024.11.6)
Requirement already satisfied: tqdm in /usr/local/lib/python3.12/dist-packages (from nltk) (4.67.1)
Requirement already satisfied: numpy>=1.19.5 in /usr/local/lib/python3.12/dist-packages (from scikit-learn) (2.0.2)
Requirement already satisfied: scipy>=1.6.0 in /usr/local/lib/python3.12/dist-packages (from scikit-learn) (1.16.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.12/dist-packages (from scikit-learn) (3.6.0)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (1.3.3)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (4.60.1)
Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (1.4.9)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (25.0)
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (11.3.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (3.2.5)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.12/dist-packages (from matplotlib) (2.9.0.post0)
Requirement already satisfied: pandas>=1.2 in /usr/local/lib/python3.12/dist-packages (from seaborn) (2.2.2)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.2->seaborn) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.2->seaborn) (2025.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-packages (from python-dateutil>=2.7->matplotlib) (1.17.0)
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Unzipping corpora/stopwords.zip.
# ensure these were executed before saving
best_model_name = results_df['Accuracy'].idxmax()
best_model = models[best_model_name]
```

```
import joblib
joblib.dump(best_model, "best_spam_model.pkl")
joblib.dump(tfidf, "tfidf_vectorizer.pkl")
['tfidf_vectorizer.pkl']
```

2. DATASET SUMMARY (FROM ONLINE SOURCE)

```
# Dataset Source: https://archive.ics.uci.edu/ml/datasets/sms+spam+collection
url = "https://raw.githubusercontent.com/justmarkham/pycon-2016-tutorial/master/data/sms.tsv"
```

```
aτ = pa.reaa_csv(uri, sep='\t', neader=None, names=['iabei', 'message'])
    df.head()
    ✓ Dataset Loaded Successfully!
        label
                                                           \blacksquare
                                                message
        ham
                  Go until jurong point, crazy.. Available only ...
                                                           16
         ham
                                  Ok lar... Joking wif u oni...
               Free entry in 2 a wkly comp to win FA Cup fina...
        spam
    3
         ham
                U dun say so early hor... U c already then say...
         ham
                 Nah I don't think he goes to usf, he lives aro...
Next steps: ( Generate code with df
                                     New interactive sheet
```

3. EXPLORATORY DATA ANALYSIS (EDA)

```
print("\nDataset Info:")
print(df.info())

print("\nClass Distribution:")
print(df['label'].value_counts())

# Visualize class distribution
plt.figure(figsize=(5,4))
sns.countplot(x='label', data=df, palette='viridis')
plt.title("Distribution of Spam vs Ham Messages")
plt.xlabel("Message Type")
plt.ylabel("Count")
plt.show()
```

```
Dataset Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 2 columns):
# Column Non-Null Count Dtype
    -----
             -----
             5572 non-null object
0 label
   message 5572 non-null object
dtypes: object(2)
memory usage: 87.2+ KB
Class Distribution:
label
       4825
ham
spam
        747
Name: count, dtype: int64
/tmp/ipython-input-429945280.py:9: FutureWarning:
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le
  sns.countplot(x='label', data=df, palette='viridis')
             Distribution of Spam vs Ham Messages
   5000
    4000
   3000
   2000
    1000
       0
                   ham
                                           spam
                           Message Type
```

4. TEXT PREPROCESSING

```
def clean_text(text):
        text = text.lower()
        text = ''.join([ch for ch in text if ch not in string.punctuation])
        words = text.split()
        words = [ps.stem(word) for word in words if word not in stopwords.words('english')]
        return ' '.join(words)
    df['cleaned'] = df['message'].apply(clean_text)
    df['label_num'] = df['label'].map({'ham':0, 'spam':1})
    df.head()
        label
                                                                                                                         \blacksquare
                                                                                                 cleaned label_num
                                                    message
     0
         ham
                   Go until jurong point, crazy.. Available only ... go jurong point crazi avail bugi n great world...
                                                                                                                    0
                                     Ok lar... Joking wif u oni...
                                                                                       ok lar joke wif u oni
                                                                                                                    0
     1
          ham
        spam
               Free entry in 2 a wkly comp to win FA Cup fina... free entri 2 wkli comp win fa cup final tkt 21...
                                                                                                                    1
     3
                 U dun say so early hor... U c already then say...
                                                                          u dun say earli hor u c alreadi say
                                                                                                                    0
          ham
                  Nah I don't think he goes to usf, he lives aro...
                                                                   nah dont think goe usf live around though
          ham
Next steps: ( Generate code with df )
                                        New interactive sheet
```

5. FEATURE EXTRACTION USING TF-IDF

6. MODEL TRAINING (MULTIPLE MODELS)

```
models = {
    "Naive Bayes": MultinomialNB(),
    "Logistic Regression": LogisticRegression(max_iter=1000),
    "Random Forest": RandomForestClassifier(n_estimators=100),
    "SVM": LinearSVC()
}
results = {}
for name, model in models.items():
   model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    acc = accuracy_score(y_test, y_pred)
    prec = precision_score(y_test, y_pred)
    rec = recall_score(y_test, y_pred)
    f1 = f1_score(y_test, y_pred)
    results[name] = [acc, prec, rec, f1]
    print(f"\n===== {name} =====")
    print(classification_report(y_test, y_pred))
==== Naive Bayes =====
                           recall f1-score
              precision
                                              support
           0
                   0.98
                             1.00
                                       0.99
                                                   966
                   1.00
                                                  149
                             0.86
                                       0.92
           1
   accuracy
                                       0.98
                                                 1115
                   0.99
                             0.93
                                       0.96
                                                 1115
  macro avg
weighted avg
                   0.98
                             0.98
                                       0.98
                                                 1115
==== Logistic Regression =====
              precision
                           recall f1-score
                                              support
           0
                   0.96
                             1.00
                                       0.98
                                                   966
          1
                   0.99
                             0.72
                                       0.84
                                                  149
                                       0.96
                                                 1115
   accuracy
                   0.98
                             0.86
                                       0.91
  macro avg
                                                 1115
weighted avg
                   0.96
                             0.96
                                       0.96
                                                 1115
==== Random Forest =====
              precision
                           recall f1-score
                                              support
           0
                   0.98
                             1.00
                                       0.99
                                                   966
                   1.00
                             0.86
                                       0.92
                                                  149
   accuracy
                                       0.98
                                                 1115
                             0.93
  macro avg
                   0.99
                                       0.96
                                                 1115
weighted avg
                   0.98
                             0.98
                                       0.98
                                                 1115
==== SVM =====
             precision
                           recall f1-score
                                              support
           0
                   0.99
                             1.00
                                       0.99
                                                   966
                   0.99
                                       0.94
                                                  149
                                       0.99
   accuracy
                                                 1115
                   0.99
                             0.95
                                       0.97
  macro avg
```

```
weighted avg 0.99 0.99 0.99 1115
```

7. MODEL PERFORMANCE COMPARISON

```
results_df = pd.DataFrame(results, index=['Accuracy','Precision','Recall','F1']).T
   print("\nModel Comparison:")
   display(results_df)
   # Visualization
   plt.figure(figsize=(8,5))
   sns.barplot(x=results_df.index, y=results_df['Accuracy'], palette='mako')
   plt.title("Model Accuracy Comparison")
   plt.ylabel("Accuracy")
   plt.xticks(rotation=15)
   plt.show()
   Model Comparison:
                                                                  丽
                       Accuracy Precision
                                              Recall
                                                            F1
       Naive Bayes
                        0.981166
                                   1.000000 0.859060 0.924188
    Logistic Regression
                        0.962332
                                   0.990826 0.724832 0.837209
      Random Forest
                        0.981166
                                   1.000000 0.859060 0.924188
           SVM
                        0.985650
                                   0.985401 0.906040 0.944056
   /tmp/ipython-input-2983375317.py:7: FutureWarning:
   Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le
     sns.barplot(x=results_df.index, y=results_df['Accuracy'], palette='mako')
                                     Model Accuracy Comparison
       1.0
       0.8
       0.6
    Accuracy
       0.4
       0.2
       0.0
                                 Logistic Regression
                                                       Random Forest
                Naive Bayes
                                                                                 SVM
                                                  None
Next steps: (
            Generate code with results_df
                                            New interactive sheet
```

8. CONFUSION MATRIX (BEST MODEL)

```
best_model_name = results_df['Accuracy'].idxmax()
best_model = models[best_model_name]
y_pred_best = best_model.predict(X_test)
cm = confusion_matrix(y_test, y_pred_best)

plt.figure(figsize=(4,3))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
plt.title(f'{best_model_name} - Confusion Matrix')
```

```
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
print(f" Best Model Identified: {best_model_name}")
          SVM - Confusion Matrix
                                            800
             964
                              2
   0
                                            600
Actual
                                            400
              14
                              135
                                            200
                   Predicted

✓ Best Model Identified: SVM
```

9. TEST WITH NEW SAMPLE MESSAGES

```
sample_msgs = [
                "Congratulations! You've won a $500 Amazon gift card. Click the link to claim now!",
                "Hey, are we still going to class tomorrow?",
                "URGENT! Your account will be locked. Verify immediately!",
                "Lunch at 1 pm sound good?"
            ]
            sample_cleaned = [clean_text(msg) for msg in sample_msgs]
            sample_tfidf = tfidf.transform(sample_cleaned).toarray()
            predictions = best_model.predict(sample_tfidf)
            for msg, pred in zip(sample_msgs, predictions):
                label = "SPAM" if pred == 1 else "HAM"
                print(f"\\nMessage: \{msg\}\\nPrediction: \{label\}")
            Message: Congratulations! You've won a $500 Amazon gift card. Click the link to claim now!
Prediction: SPAM Could not connect to the reCAPTCHA service. Please check your internet connection and reload to get a reCAPTCHA challenge.
```