```
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, cross_val_score
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.preprocessing import LabelEncoder
from sklearn.linear_model import LogisticRegression
from sklearn.naive_bayes import MultinomialNB
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report, confusion_matrix, accuracy_s
from sklearn.pipeline import make_pipeline
import joblib
import warnings
warnings.filterwarnings('ignore')
from google.colab import drive
drive.mount('/content/drive')
→ Mounted at /content/drive
```

Importing the Dependencies

Data Collection & Pre-Processing

```
# loading the data from csv file to a pandas Dataframe
raw_mail_data = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/dataset/mai
```

print(raw\_mail\_data)

<b>→</b>	0 1 2 3 4	Category ham ham spam ham ham	Message Go until jurong point, crazy. Available only Ok lar Joking wif u oni Free entry in 2 a wkly comp to win FA Cup fina U dun say so early hor U c already then say Nah I don't think he goes to usf, he lives aro
	5567 5568 5569 5570 5571	spam ham ham ham ham	This is the 2nd time we have tried 2 contact u Will $\tilde{A}^{\frac{1}{4}}$ b going to esplanade fr home? Pity, * was in mood for that. Soany other s The guy did some bitching but I acted like i'd Rofl. Its true to its name

[5572 rows x 2 columns]

```
# replace the null values with a null string
  mail_data = raw_mail_data.where((pd.notnull(raw_mail_data)),'')
  # printing the first 5 rows of the dataframe
  mail_data.head()
\overline{\mathbf{x}}
         Category
                                                     Message
      0
                       Go until jurong point, crazy.. Available only ...
               ham
      1
                                        Ok lar... Joking wif u oni...
               ham
      2
                    Free entry in 2 a wkly comp to win FA Cup fina...
      3
               ham
                     U dun say so early hor... U c already then say...
      4
                       Nah I don't think he goes to usf, he lives aro...
               ham
 Next
         Generate code with mail_data
                                         View recommended plots
                                                                       New interactive sheet
 steps:
  # checking the number of rows and columns in the dataframe
  mail_data.shape
     (5572, 2)
Label Encoding
  # Encode labels
  label_encoder = LabelEncoder()
  mail_data['Category'] = label_encoder.fit_transform(mail_data['Category'])
spam - 0
ham - 1
# separating the data as texts and label
X = mail_data['Message']
y = mail_data['Category']
```

```
print(X)
             Go until jurong point, crazy.. Available only ...
    0
     1
                                   Ok lar... Joking wif u oni...
     2
             Free entry in 2 a wkly comp to win FA Cup fina...
     3
             U dun say so early hor... U c already then say...
     4
             Nah I don't think he goes to usf, he lives aro...
     5567
             This is the 2nd time we have tried 2 contact u...
     5568
                          Will \tilde{A}_{4}^{1} b going to esplanade fr home?
     5569
             Pity, * was in mood for that. So...any other s...
             The guy did some bitching but I acted like i'd...
     5570
                                      Rofl. Its true to its name
     5571
     Name: Message, Length: 5572, dtype: object
print(y)
             0
     1
             0
     2
             1
     3
             0
     4
             0
     5567
             1
     5568
             0
     5569
             0
     5570
             0
     5571
             0
     Name: Category, Length: 5572, dtype: int64
Splitting the data into training data & test data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, strati
print(X.shape)
print(X_train.shape)
print(X test.shape)
\rightarrow
    (5572,)
     (4457,)
```

**Feature Extraction** 

(1115,)

```
# Vectorize text using TF-IDF
vectorizer = TfidfVectorizer(min_df=1, stop_words='english', lowercase=True)
X_train_tfidf = vectorizer.fit_transform(X_train)
X_test_tfidf = vectorizer.transform(X_test)

# Save vectorizer
joblib.dump(vectorizer, 'tfidf_vectorizer.pkl')

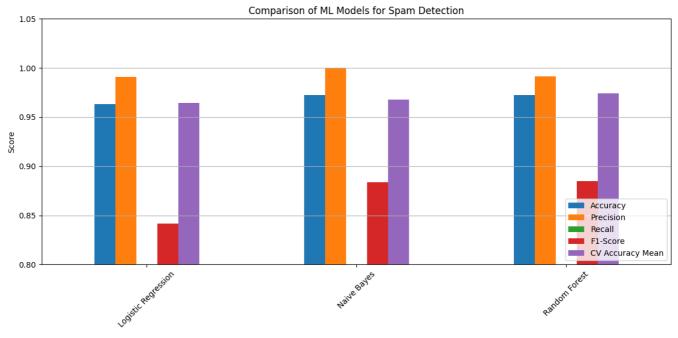
Tritialize models
models = {
    'Logistic Regression': LogisticRegression(),
    'Naive Bayes': MultinomialNB(),
    'Random Forest': RandomForestClassifier(n_estimators=100, random_state=42)
}
results = {}
```

```
# Train, cross-validate and evaluate models
for name, model in models.items():
    model.fit(X_train_tfidf, y_train)
    y_pred = model.predict(X_test_tfidf)
    acc = accuracy_score(y_test, y_pred)
    report = classification_report(y_test, y_pred, output_dict=True)
    cm = confusion_matrix(y_test, y_pred)
    roc_auc = None
    if hasattr(model, "predict_proba"):
        y_scores = model.predict_proba(X_test_tfidf)[:, 1]
    else:
        y_scores = model.decision_function(X_test_tfidf)
    fpr, tpr, _ = roc_curve(y_test, y_scores)
    roc_auc = auc(fpr, tpr)
    # Cross-validation
    pipeline = make_pipeline(TfidfVectorizer(stop_words='english', lowercase=Tr
    cv_scores = cross_val_score(pipeline, X, y, cv=5, scoring='accuracy')
    results[name] = {
        'Model' : model,
        'Accuracy': acc,
        'Precision': report['1']['precision'],
        'Recall': report['1']['recall'],
        'F1-Score': report['1']['f1-score'],
        'Confusion Matrix': cm,
        'ROC AUC': roc_auc,
        'CV Accuracy Mean': cv_scores.mean(),
        'FPR': fpr,
        'TPR': tpr
    }
# Save models
for name, data in results.items():
    joblib.dump(data['Model'], f"{name.replace(' ', '_').lower()}_model.pkl")
```

```
# Display evaluation results
eval df = pd.DataFrame({k: {
    'Accuracy': v['Accuracy'],
    'Precision': v['Precision'],
    'Recall': v['Recall'],
    'F1-Score': v['F1-Score'],
    'ROC AUC': v['ROC AUC'],
    'CV Accuracy Mean': v['CV Accuracy Mean']
} for k, v in results.items()}).T
print("Model Evaluation Comparison:")
print(eval_df)
→ Model Evaluation Comparison:
                         Accuracy Precision
                                                Recall F1-Score
                                                                   ROC AUC
    Logistic Regression
                         0.963229
                                    0.990909
                                              0.731544 0.841699
                                                                 0.991645
    Naive Bayes
                                              0.791946 0.883895
                         0.972197
                                    1.000000
                                                                 0.990520
    Random Forest
                         0.972197
                                    0.991667
                                              0.798658 0.884758
                                                                 0.988825
                         CV Accuracy Mean
    Logistic Regression
                                 0.964106
    Naive Bayes
                                 0.968054
    Random Forest
                                 0.974154
```

```
# Plot comparison
plt.figure(figsize=(10, 6))
eval_df[['Accuracy', 'Precision', 'Recall', 'F1-Score', 'CV Accuracy Mean']].pl
plt.title("Comparison of ML Models for Spam Detection")
plt.ylabel("Score")
plt.ylim(0.8, 1.05)
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.legend(loc='lower right')
plt.tight_layout()
plt.show()
```

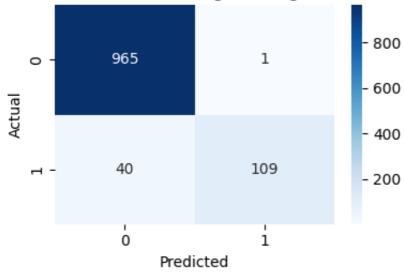
## → <Figure size 1000x600 with 0 Axes>



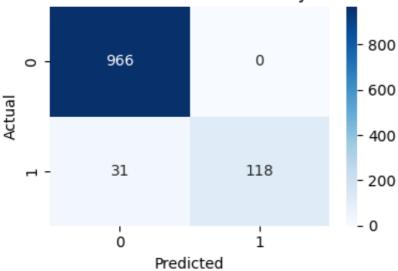
```
# Show confusion matrices
for name, metrics in results.items():
    plt.figure(figsize=(4, 3))
    sns.heatmap(metrics['Confusion Matrix'], annot=True, fmt='d', cmap='Blues')
    plt.title(f"Confusion Matrix - {name}")
    plt.xlabel("Predicted")
    plt.ylabel("Actual")
    plt.tight_layout()
```



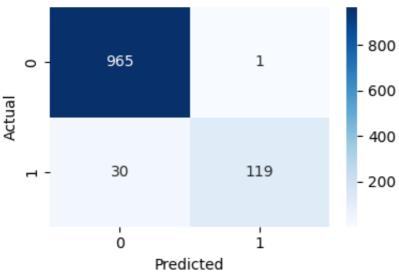




## Confusion Matrix - Naive Bayes

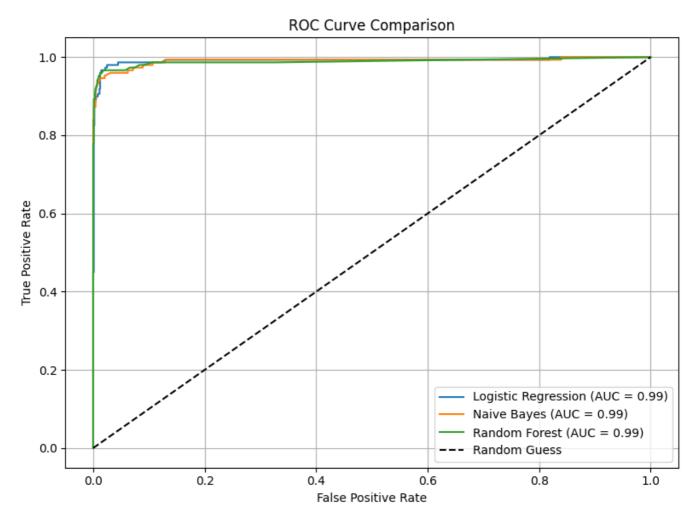


## Confusion Matrix - Random Forest



```
# Plot ROC Curves
plt.figure(figsize=(8, 6))
for name, metrics in results.items():
    plt.plot(metrics['FPR'], metrics['TPR'], label=f"{name} (AUC = {metrics['RC']})
plt.plot([0, 1], [0, 1], 'k--', label="Random Guess")
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ROC Curve Comparison")
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.show()
```





```
# Support for real-time prediction
def predict_email_category(model_path, vectorizer_path, new_message):
    model = joblib.load(model_path)
    vectorizer = joblib.load(vectorizer_path)
    new_vector = vectorizer.transform([new_message])
    prediction = model.predict(new_vector)
    category = "Spam" if prediction[0] == 1 else "Ham"
    return category

# Example usage
example_text = "Congratulations! You've won a $1000 Walmart gift card. Go to h
predicted_label = predict_email_category("./logistic_regression_model.pkl", "t
print(f"Predicted label: {predicted_label}")
```

→ Predicted label: Spam