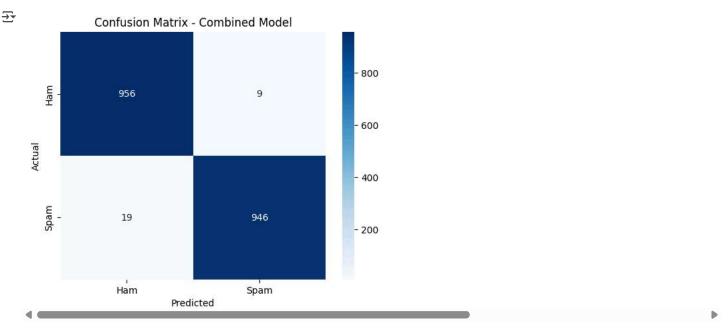
```
import pandas as pd
import re
import nltk
import matplotlib.pyplot as plt
import seaborn as sns
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix, roc_curve, auc
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.utils import resample
from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score, precision_score
# Download required nltk data
nltk.download('stopwords')
     [nltk data] Downloading package stopwords to /root/nltk data...
     [nltk data]
                  Package stopwords is already up-to-date!
import pandas as pd
# Load the dataset
df = pd.read_csv("/content/spam.csv", encoding="latin1")
# Display first few rows
print(df.head())
\rightarrow
          v1
                                                                 v2 Unnamed: 2
         ham
              Go until jurong point, crazy.. Available only \dots
                                                                            NaN
        ham
                                    Ok lar... Joking wif u oni...
     2 spam Free entry in 2 a wkly comp to win FA Cup fina...
                                                                            NaN
     3
        ham U dun say so early hor... U c already then say...
                                                                            NaN
         ham Nah I don't think he goes to usf, he lives aro...
       Unnamed: 3 Unnamed: 4
     0
              NaN
                          NaN
               NaN
                          NaN
               NaN
                          NaN
     2
     3
               NaN
                          NaN
     4
               NaN
                          NaN
# Keep only necessary columns (adjust based on dataset structure)
df = df.iloc[:, [0, 1]] # Keeping only first two columns
df.columns = ['label', 'message'] # Rename columns for clarity
# Convert labels to binary (spam \rightarrow 1, ham \rightarrow 0)
df['label'] = df['label'].map({'spam': 1, 'ham': 0})
print(df.head())
₹
        label
     a
               Go until jurong point, crazy.. Available only ...
          NaN
     1
          NaN
                                     Ok lar... Joking wif u oni...
          NaN U dun say so early hor... U c already then say...
          NaN
               Nah I don't think he goes to usf, he lives aro...
               Even my brother is not like to speak with me. ...
     <ipython-input-38-0b0b86c6e2b6>:6: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-cc">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-cc</a>
       df['label'] = df['label'].map({'spam': 1, 'ham': 0})
# Check for NaN values
df = df.dropna(subset=['label'])
# Ensure dataset has both spam & ham
if df['label'].nunique() < 2:</pre>
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raise ValueError("Dataset does not contain both spam and ham messages. Check preprocessing!")
# Class distribution before resampling
print("\nClass distribution before resampling:\n", df['label'].value_counts())
     Class distribution before resampling:
      label
     0
          4825
     1
           747
     Name: count, dtype: int64
# Separate spam and ham messages
ham = df[df['label'] == 0]
spam = df[df['label'] == 1]
# Resampling: Balance the dataset
if len(spam) < len(ham):</pre>
    spam_upsampled = resample(spam, replace=True, n_samples=len(ham), random_state=42)
    df = pd.concat([ham, spam_upsampled])
# Class distribution after resampling
print("\nClass Distribution After Resampling:")
print(df['label'].value_counts())
\overline{2}
     Class Distribution After Resampling:
     label
     0 4825
          4825
     Name: count, dtype: int64
# Text Preprocessing
stemmer = PorterStemmer()
stop_words = set(stopwords.words("english"))
# Initialize stemmer and stop words
stemmer = PorterStemmer()
stop_words = set(stopwords.words("english"))
def preprocess_text(text):
    text = text.lower() # Convert to lowercase
    \label{text} \begin{array}{l} \text{text = re.sub(r'\W', '', text)} & \text{\# Remove special characters} \\ \text{text = re.sub(r'\s+', '', text).strip()} & \text{\# Remove extra spaces} \end{array}
    words = text.split() # Tokenization
    words = [stemmer.stem(word) for word in words if word not in stop_words] # Stemming & stopword removal
    return ' '.join(words)
# Example usage:
text = "This is a SAMPLE message to preprocess!?"
print(preprocess_text(text))
⇒ sampl messag preprocess
# Apply preprocessing to all messages
df['cleaned_message'] = df['message'].apply(preprocess_text)
# Feature Extraction using TF-IDF
vectorizer = TfidfVectorizer(max_features=5000)
X = vectorizer.fit_transform(df['cleaned_message']).toarray()
y = df['label'].values
# Train-test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42, stratify=y)
```

```
# Models
models = {
    "Random Forest": RandomForestClassifier(n_estimators=100,max_depth=10, random_state=42),
    "Gradient Boosting": GradientBoostingClassifier(n_estimators=100, random_state=42),
    "MultinomialNB": MultinomialNB()
}
# Train & evaluate models
results = {}
for name, model in models.items():
   model.fit(X train, y train)
   y_pred = model.predict(X_test)
   results[name] = {
        'Accuracy': accuracy_score(y_test, y_pred),
        'Precision': precision_score(y_test, y_pred)
   print(f"{name} - Accuracy: {results[name]['Accuracy']:.4f}, Precision: {results[name]['Precision']:.4f}")
Random Forest - Accuracy: 0.9409, Precision: 0.9965
     Gradient Boosting - Accuracy: 0.9528, Precision: 0.9659
     MultinomialNB - Accuracy: 0.9839, Precision: 0.9844
# Model names and their accuracy
model names = list(results.keys()) + ["Combined Model"]
accuracies = [results[name]['Accuracy'] for name in results] + ['accuracy_combined']
# Ensemble (Combining Random Forest & MultinomialNB)
rf model = models["Random Forest"]
nb_model = models["MultinomialNB"]
rf_preds = rf_model.predict_proba(X_test)[:, 1] # Probabilities from Random Forest
nb_preds = nb_model.predict_proba(X_test)[:, 1] # Probabilities from MultinomialNB
combined_preds = (rf_preds + nb_preds) / 2 # Average probabilities
\label{final_preds} \mbox{final\_preds = (combined\_preds > 0.5).astype(int) \mbox{ \# Convert to binary classification} \\
# Evaluation for Combined Model
accuracy_combined = accuracy_score(y_test, final_preds)
precision_combined = precision_score(y_test, final_preds)
print(f"\nCombined Model - Accuracy: {accuracy_combined:.4f}, Precision: {precision_combined:.4f}")
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     Combined Model - Accuracy: 0.9855, Precision: 0.9906
# Function to Predict User Input SMS
def predict_sms(message):
   processed_msg = preprocess_text(message)
   vectorized_msg = vectorizer.transform([processed_msg]).toarray()
   # Predict using both models
   rf prob = rf model.predict proba(vectorized msg)[:, 1]
   nb_prob = nb_model.predict_proba(vectorized_msg)[:, 1]
   combined prob = (rf prob + nb prob) / 2
   prediction = "Spam" if combined_prob > 0.5 else "Ham"
    return prediction
# Confusion Matrix for the Combined Model
y_pred_combined = final_preds
cm = confusion_matrix(y_test, y_pred_combined)
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['Ham', 'Spam'], yticklabels=['Ham', 'Spam'])
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.title("Confusion Matrix - Combined Model")
plt.show()
```

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# Take user input
user\_message = input("\nEnter an SMS message to classify: ")
print(f"\nOutput: {predict\_sms(user\_message)}")

Enter an SMS message to classify: FreeMsg Hey there darling it's been 3 week's now and no word back! I'd like some fun you up for it s..

Output: Spam