

☀️ Tamizhan Skills SE RISE Internship – Machine Learning & AI

Task 1: Email Spam Detection using Naive Bayes

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Project Summary

>Objective:

To classify emails as "Spam" or "Not Spam" using a supervised machine learning algorithm like Naive Bayes or SVM.

>Tools Used:

Google Colab

Python

Scikit-learn (sklearn)

Natural Language Toolkit (nltk)

Email dataset (CSV)

>Approach:

1. Loaded and cleaned the email dataset
2. Preprocessed text using NLP techniques like tokenization, stopword removal, and stemming
3. Converted text into numerical format using TF-IDF or CountVectorizer
4. Trained a Naive Bayes classifier or SVM on the data
5. Evaluated performance using accuracy, precision, recall, and confusion matrix

>Result:

The final model accurately detects spam emails with good accuracy and can be used for email filtering systems in real-world applications.



Model Evaluation Screenshot:

```
... Accuracy: 0.9659192825112107
Precision: 1.0
Recall: 0.7466666666666667

Classification Report:
      precision    recall  f1-score   support

     0       0.96       1.00       0.98        965
     1       1.00       0.75       0.85        150

 accuracy          0.97        1115
 macro avg          0.98        1115
 weighted avg       0.97        1115
```



Code Used (Google Colab):

- Import libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import string
import nltk
from nltk.corpus import stopwords
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score, precision_score,
recall_score, classification_report

nltk.download('stopwords')
```

- Load and clean dataset

```
# Load the CSV file
data = pd.read_csv('spam.csv', encoding='latin-1')
```

```

# Keep only important columns
data = data[['v1', 'v2']]
data.columns = ['label', 'text']

# Display first 5 rows
data.head()

stop_words = set(stopwords.words('english'))

def preprocess_text(text):
    text = text.lower() # Lowercase
    text = "".join([char for char in text if char not in string.punctuation])
    # Remove punctuation
    words = text.split()
    words = [word for word in words if word not in stop_words] #
    Remove stopwords
    return ' '.join(words)

data['clean_text'] = data['text'].apply(preprocess_text)

# Show cleaned data
data[['text', 'clean_text']].head()

```

- **Preprocess text**

```

vectorizer = TfidfVectorizer()
X = vectorizer.fit_transform(data['clean_text'])

# Convert labels: ham = 0, spam = 1
y = data['label'].map({'ham': 0, 'spam': 1})

# Split into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)

```

- **Train model**

```
model = MultinomialNB()  
model.fit(X_train, y_train)  
  
y_pred = model.predict(X_test)
```

- **Evaluate model**

```
print("Accuracy:", accuracy_score(y_test, y_pred))  
print("Precision:", precision_score(y_test, y_pred))  
print("Recall:", recall_score(y_test, y_pred))  
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

✨ This model can be extended using deep learning (LSTM) or deployed using Streamlit for live spam detection.