

Machine Learning Lab 8: Spam classification with Naive bayes

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```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

# import libraries
```

```
In [45]: df = pd.read_csv('data\\spam.csv')
df.head()

# import dataset
```

```
Out[45]:
```

	Category	Message
0	ham	Go until jurong point, crazy.. Available only ...
1	ham	Ok lar... Joking wif u oni...
2	spam	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	ham	Nah I don't think he goes to usf, he lives aro...

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```
In [3]: df.info()

# check for null values

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 2 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Category    5572 non-null   object
1   Message     5572 non-null   object
dtypes: object(2)
memory usage: 87.2+ KB
```

```
In [4]: df.isna().sum()
```

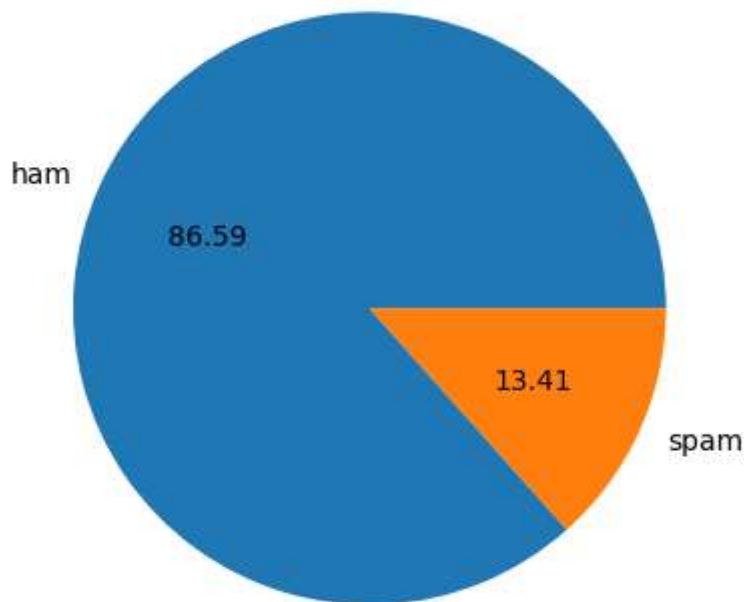
```
# check for null values
```

```
Out[4]: Category    0
        Message    0
        dtype: int64
```

```
In [46]: plt.pie(df['Category'].value_counts(), labels=['ham', 'spam'], autopct="%0.2f", lab
plt.plot()
```

```
# plot pie chart for ham and spam (class distribution)
```

```
Out[46]: []
```



```
In [47]: import nltk
import spacy
import re

nlp = spacy.load('en_core_web_lg')

# import libraries and load spacy model
```

```
In [7]: stopwords = spacy.lang.en.stop_words.STOP_WORDS

df['Message'] = df['Message'].apply(lambda x: re.sub(r'^\w\s', '', x))
df['Message'] = df['Message'].apply(lambda x: ' '.join([word.lower() for word in x.
df['Message'] = df['Message'].apply(lambda x: ' '.join([word.lemma_ for word in nlp

df['Message'].head()

# remove stopwords, punctuations, Lemmatize and convert to lower case
```

```
Out[7]: 0    go jurong point crazy available bugis n great ...
        1                                ok lar joke wif u oni
        2    free entry 2 wkly comp win fa cup final tkts 2...
        3                                u dun early hor u c
        4                                nah I do not think go usf life
        Name: Message, dtype: object
```

```
In [48]: df.Message[0]
```

```
# checking the above process
```

```
Out[48]: 'Go until jurong point, crazy.. Available only in bugis n great world la e buffe
t... Cine there got amore wat...'
```

```
In [21]: from sklearn.feature_extraction.text import TfidfVectorizer
        vectorizer = TfidfVectorizer()
```

```
embeddings = vectorizer.fit_transform(df['Message'])
target = df['Category'].map({'ham': 0, 'spam': 1})
```

```
# apply tfidf vectorizer to dataset
```

```
In [22]: from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test = train_test_split(embeddings, target, test_size=0
```

```
# split dataset into train and test
```

```
In [23]: from sklearn.naive_bayes import MultinomialNB
        naive_bayes = MultinomialNB(alpha=1.0, force_alpha=True)
```

```
# instantiate MultinomialNB
```

```
In [24]: naive_bayes.fit(X_train, y_train)
```

```
# fit MultinomialNB to train set
```

```
Out[24]: ▼ MultinomialNB ⓘ ?
```

```
MultinomialNB()
```

```
In [65]: print(naive_bayes.score(X_train, y_train))
        print(naive_bayes.score(X_test, y_test))
```

```
# check train and test scores
```

```
0.9748709894547902
```

```
0.9614349775784753
```

```
In [26]: y_pred = naive_bayes.predict(X_test)
```

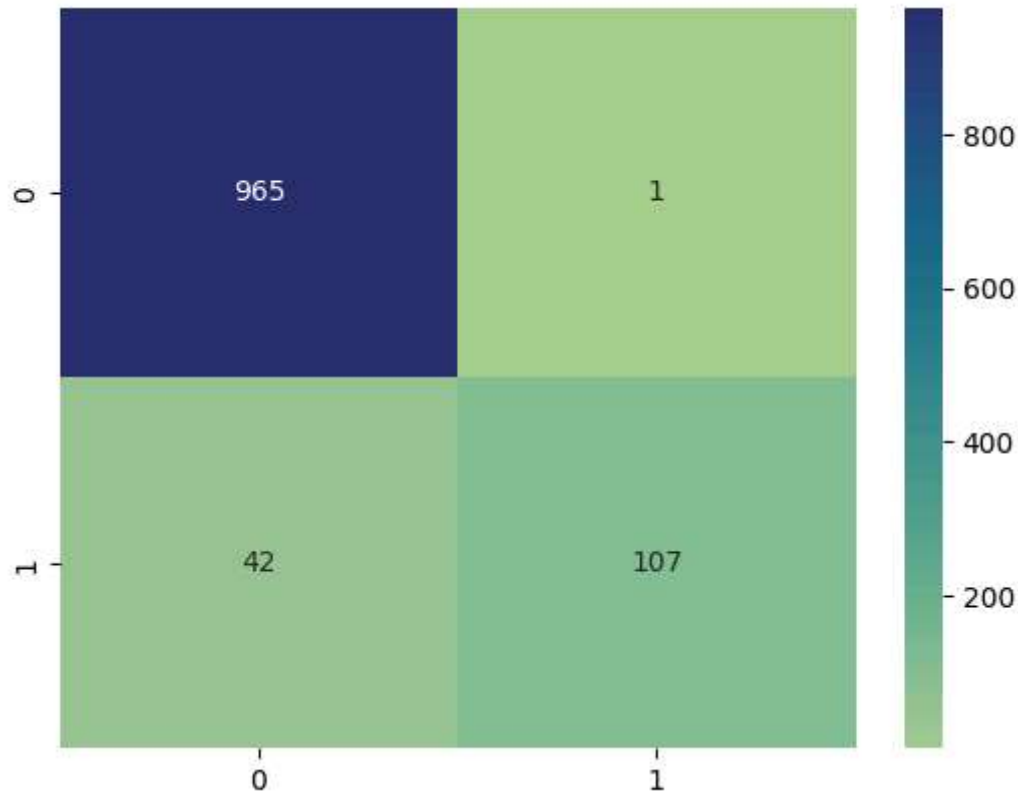
```
# predict on test set
```

```
In [73]: from sklearn.metrics import confusion_matrix, classification_report

sns.heatmap(confusion_matrix(y_test, y_pred), annot=True, cmap='crest', fmt='.0f')

# plotting confusion matrix
```

Out[73]: <Axes: >



```
In [74]: print(classification_report(y_test, y_pred))

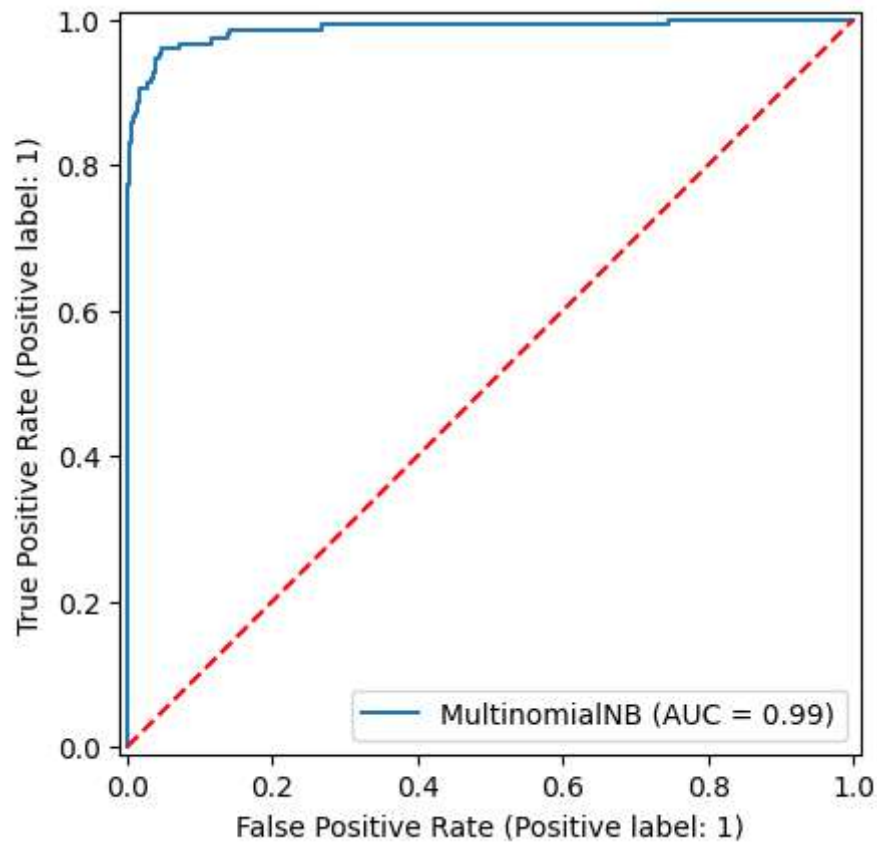
# print classification report
```

	precision	recall	f1-score	support
0	0.96	1.00	0.98	966
1	0.99	0.72	0.83	149
accuracy			0.96	1115
macro avg	0.97	0.86	0.91	1115
weighted avg	0.96	0.96	0.96	1115

```
In [80]: from sklearn.metrics import RocCurveDisplay, PrecisionRecallDisplay, precision_reca
RocCurveDisplay.from_estimator(naive_bayes, X_test, y_test)
plt.plot([0,1], [0,1], 'r--')

# plotting roc curve
```

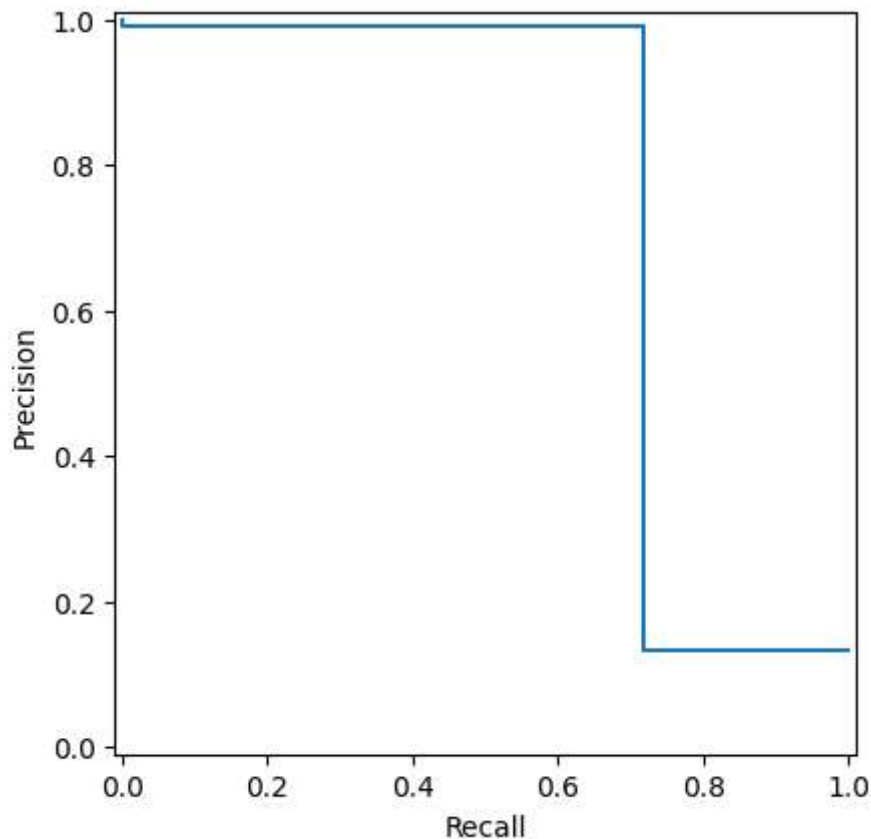
Out[80]: [<matplotlib.lines.Line2D at 0x140b85090d0>]



```
In [81]: precision, recall, threshold = precision_recall_curve(y_test, y_pred)
PrecisionRecallDisplay(precision, recall).plot()

# plotting precision recall curve
```

```
Out[81]: <sklearn.metrics._plot.precision_recall_curve.PrecisionRecallDisplay at 0x140b8281b10>
```



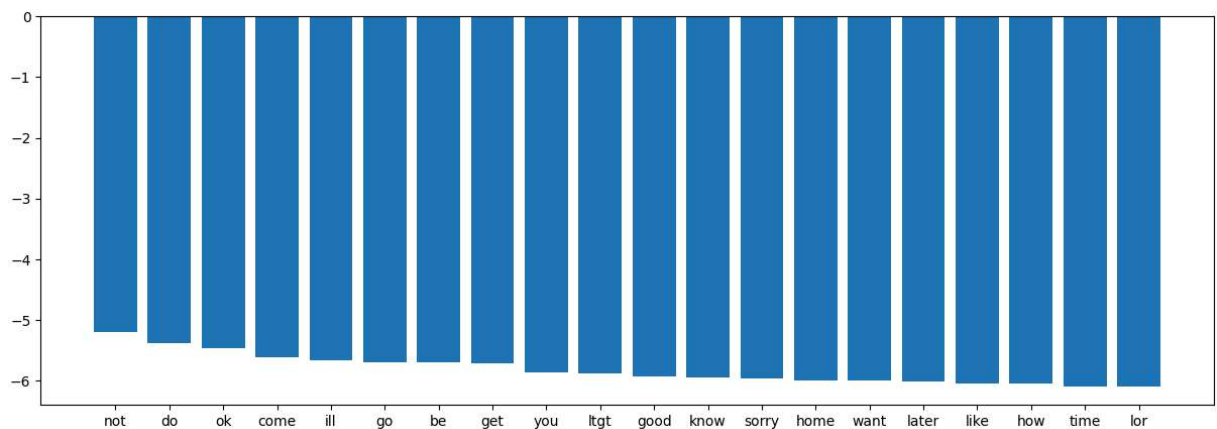
```
In [82]: feature_log_prob = naive_bayes.feature_log_prob_
features = vectorizer.get_feature_names_out()

top20 = sorted(zip(naive_bayes.feature_log_prob_[0], features), reverse=True)[:20]

plt.figure(figsize=(15, 5))
plt.bar([x[1] for x in top20], [x[0] for x in top20])

# plot top 20 features and their weights
```

Out[82]: <BarContainer object of 20 artists>

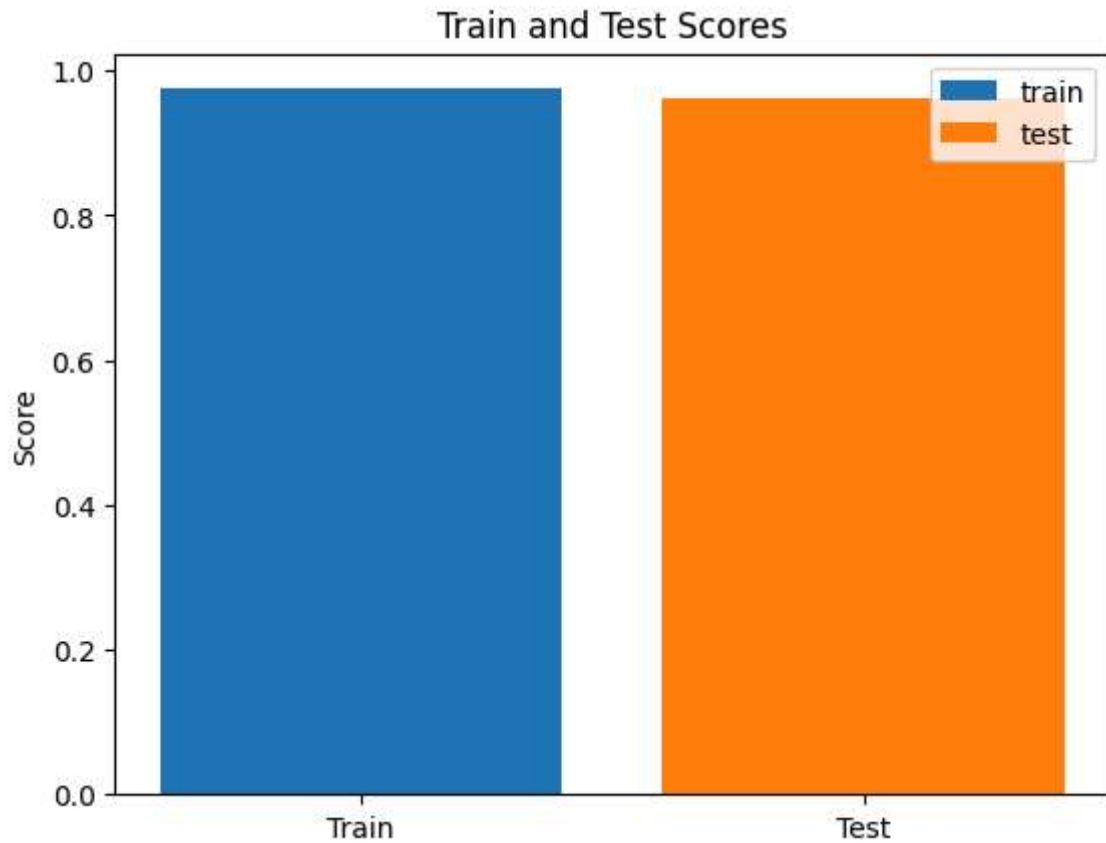


```
In [83]: plt.bar(1, naive_bayes.score(X_train, y_train), label='train')
plt.bar(2, naive_bayes.score(X_test, y_test), label='test')
plt.ylabel('Score')
plt.legend()
```

```
plt.xticks([1, 2], ['Train', 'Test'])  
plt.title('Train and Test Scores')  
plt.plot()
```

```
# plot train and test scores
```

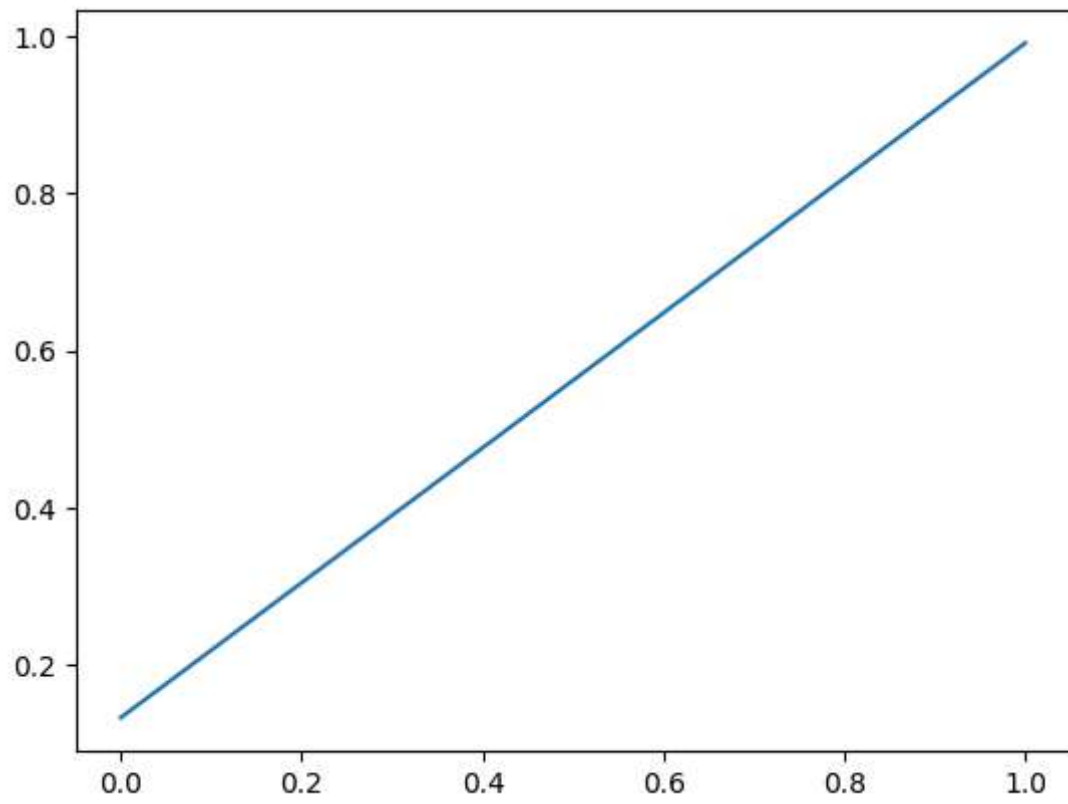
Out[83]: []



```
In [87]: plt.plot(threshold, precision[:-1], label='precision')
```

```
# plot precision and threshold
```

Out[87]: [



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