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```
In [1]: import warnings
        warnings.filterwarnings('ignore')
In [2]: import numpy as np
        import pandas as pd
        import os
        import matplotlib.pyplot as plt
        import seaborn as sns
In [3]: data = pd.read_parquet('OwnData/data.parquet')
In [6]: null_counts = data.isnull().sum()
        # Print the number of null values
        print(f"{null_counts.sum()} null entries have been found in the dataset\n")
        # Drop null values
                                          # or df_data = df_data.dropna()
        data.dropna(inplace=True)
        # Find and handle duplicates
        duplicate_count = data.duplicated().sum()
        # Print the number of duplicate entries
        print(f"{duplicate_count} duplicate entries have been found in the dataset\n")
        # Remove duplicates
        data.drop_duplicates(inplace=True) # or df_data = df_data.drop_duplicates()
        # Display relative message
        print(f"All duplicates have been removed\n")
        # Reset the indexes
        data.reset_index(drop=True, inplace=True)
        # Inspect the dataset for categorical columns
        print("Categorical columns:",data.select_dtypes(include=['category']).columns.tolist(),'\n')
        # Print the first 5 lines
        data.head()
```

0 null entries have been found in the dataset

O duplicate entries have been found in the dataset

All duplicates have been removed

Categorical columns: ['Label']

Out[6]:	Prot	tocol	Flow Duration	Total Fwd Packet	Total Bwd packets	Total Length of Fwd Packet	Total Length of Bwd Packet	Fwd Packet Length Max	Fwd Packet Length Min	Fwd Packet Length Mean	Fwd Packet Length Std ^{···}	Fwd Seg Size Min	Active Mean	Active Std	Active Max	Active Min	ldle Mean	ldle Std	ldle Max	ldle Min	Label
	0	6	22545	22	20	336.0	0.0	32.0	0.0	15.272727	9.207563	32	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	TCPFlood- Evasive
	1	6	1203699	7	5	413.0	11192.0	413.0	0.0	59.000000	156.099335	32	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	Hulk-NoDefense
	2	6	776558	8	6	365.0	11192.0	365.0	0.0	45.625000	129.046982	32	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	Hulk-NoDefense
	3	6	239234	11	6	225.0	4256.0	77.0	0.0	20.454546	35.041016	32	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	Benign
	4	6	432672	8	6	376.0	11192.0	376.0	0.0	47.000000	132.936081	32	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	Hulk-NoDefense

5 rows × 78 columns

```
In [7]: data.columns
```

```
Out[7]: Index(['Protocol', 'Flow Duration', 'Total Fwd Packet', 'Total Bwd packets',
                'Total Length of Fwd Packet', 'Total Length of Bwd Packet',
                'Fwd Packet Length Max', 'Fwd Packet Length Min',
                'Fwd Packet Length Mean', 'Fwd Packet Length Std',
                'Bwd Packet Length Max', 'Bwd Packet Length Min',
                 'Bwd Packet Length Mean', 'Bwd Packet Length Std', 'Flow Bytes/s',
                'Flow Packets/s', 'Flow IAT Mean', 'Flow IAT Std', 'Flow IAT Max',
                'Flow IAT Min', 'Fwd IAT Total', 'Fwd IAT Mean', 'Fwd IAT Std',
                'Fwd IAT Max', 'Fwd IAT Min', 'Bwd IAT Total', 'Bwd IAT Mean',
                'Bwd IAT Std', 'Bwd IAT Max', 'Bwd IAT Min', 'Fwd PSH Flags',
                'Bwd PSH Flags', 'Fwd URG Flags', 'Bwd URG Flags', 'Fwd Header Length',
                'Bwd Header Length', 'Fwd Packets/s', 'Bwd Packets/s',
                'Packet Length Min', 'Packet Length Max', 'Packet Length Mean',
                'Packet Length Std', 'Packet Length Variance', 'FIN Flag Count',
                'SYN Flag Count', 'RST Flag Count', 'PSH Flag Count', 'ACK Flag Count',
                'URG Flag Count', 'CWR Flag Count', 'ECE Flag Count', 'Down/Up Ratio',
                'Avg Packet Size', 'Fwd Segment Size Avg', 'Bwd Segment Size Avg',
                'Fwd Bytes/Bulk Avg', 'Fwd Packet/Bulk Avg', 'Fwd Bulk Rate Avg',
                'Bwd Bytes/Bulk Avg', 'Bwd Packet/Bulk Avg', 'Bwd Bulk Rate Avg',
                'Subflow Fwd Packets', 'Subflow Fwd Bytes', 'Subflow Bwd Packets', 'Subflow Bwd Bytes', 'FwD Init Win Bytes', 'Bwd Init Win Bytes',
                'Fwd Act Data Packets', 'Fwd Seg Size Min', 'Active Mean', 'Active Std',
                'Active Max', 'Active Min', 'Idle Mean', 'Idle Std', 'Idle Max',
                'Idle Min', 'Label'],
```

In [8]: data['Label'].value_counts()

dtype='object')

```
Out[8]: Hulk-Reqtimeout
                                 128346
        Hulk-NoDefense
                                 127671
        Hulk-Security2
                                 121956
                                 113552
        Hulk-Evasive
        Benign
                                  45708
        TCPFlood-Reqtimeout
                                  45405
        TCPFlood-Evasive
                                  45390
        TCPFlood-Security2
                                  44547
        TCPFlood-NoDefense
                                  43762
        Slowloris-Reqtimeout
                                   1017
        Slowhttptest-Security2
                                   1003
                                   1002
        Slowhttptest-Evasive
        Slowhttptest-NoDefense
                                    998
                                    852
        Slowhttptest-Reqtimeout
                                    267
        Slowloris-Evasive
                                    267
        Slowloris-NoDefense
        Slowloris-Security2
                                    267
        Name: Label, dtype: int64
```

In [9]: # changing attack labels to their respective attack class def change_label(df):

df['Label'].replace(['Hulk-Reqtimeout','Hulk-NoDefense','Hulk-Security2','Hulk-Evasive','TCPFlood-Reqtimeout','TCPFlood-Security2','TCPFlood-NoDefense','Slowhttptest-Reqtimeout','Slowhttptest-NoDefense','Slowhttptest-NoDefense','Slowhttptest-NoDefense','Slowloris-NoDefense','Slowloris-Security2'],'Attack',inplace=True)

df['Label'].replace(['Benign'],'Normal',inplace=True)

In [10]: change_label(data)

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```
In [11]: data.info()
```

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```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 722010 entries, 0 to 722009
Data columns (total 78 columns):
# Column
                               Non-Null Count Dtype
---
                               -----
                               722010 non-null int8
0 Protocol
    Flow Duration
                               722010 non-null int32
                               722010 non-null int16
    Total Fwd Packet
    Total Bwd packets
                               722010 non-null int16
    Total Length of Fwd Packet 722010 non-null float32
    Total Length of Bwd Packet 722010 non-null float32
6 Fwd Packet Length Max
                               722010 non-null float32
    Fwd Packet Length Min
                               722010 non-null float32
8 Fwd Packet Length Mean
                               722010 non-null float32
                               722010 non-null float32
9 Fwd Packet Length Std
10 Bwd Packet Length Max
                               722010 non-null float32
11 Bwd Packet Length Min
                               722010 non-null float32
                               722010 non-null float32
12 Bwd Packet Length Mean
13 Bwd Packet Length Std
                               722010 non-null float32
14 Flow Bytes/s
                               722010 non-null float32
                               722010 non-null float32
15 Flow Packets/s
                               722010 non-null float32
16 Flow IAT Mean
                               722010 non-null float32
17 Flow IAT Std
18 Flow IAT Max
                               722010 non-null float32
19 Flow IAT Min
                               722010 non-null float32
20 Fwd IAT Total
                               722010 non-null float32
                               722010 non-null float32
21 Fwd IAT Mean
22 Fwd IAT Std
                               722010 non-null float32
23 Fwd IAT Max
                               722010 non-null float32
24 Fwd IAT Min
                               722010 non-null float32
                               722010 non-null float32
25 Bwd IAT Total
26 Bwd IAT Mean
                               722010 non-null float32
27 Bwd IAT Std
                               722010 non-null float32
28 Bwd IAT Max
                               722010 non-null float32
                               722010 non-null float32
29 Bwd IAT Min
30 Fwd PSH Flags
                               722010 non-null int8
                               722010 non-null int16
31 Bwd PSH Flags
                               722010 non-null int8
32 Fwd URG Flags
33 Bwd URG Flags
                               722010 non-null int8
                               722010 non-null int32
34 Fwd Header Length
                               722010 non-null int32
35 Bwd Header Length
36 Fwd Packets/s
                               722010 non-null float32
37 Bwd Packets/s
                               722010 non-null float32
38 Packet Length Min
                               722010 non-null float32
39 Packet Length Max
                               722010 non-null float32
40 Packet Length Mean
                               722010 non-null float32
                               722010 non-null float32
41 Packet Length Std
                               722010 non-null float32
42 Packet Length Variance
43 FIN Flag Count
                               722010 non-null int8
44 SYN Flag Count
                               722010 non-null int8
45 RST Flag Count
                               722010 non-null int8
46 PSH Flag Count
                               722010 non-null int16
                               722010 non-null int16
47 ACK Flag Count
48 URG Flag Count
                               722010 non-null int8
49 CWR Flag Count
                               722010 non-null int8
50 ECE Flag Count
                               722010 non-null int8
51 Down/Up Ratio
                               722010 non-null float32
                               722010 non-null float32
52 Avg Packet Size
                               722010 non-null float32
53 Fwd Segment Size Avg
                               722010 non-null float32
54 Bwd Segment Size Avg
55 Fwd Bytes/Bulk Avg
                               722010 non-null int32
56 Fwd Packet/Bulk Avg
                               722010 non-null int8
57 Fwd Bulk Rate Avg
                               722010 non-null int32
58 Bwd Bytes/Bulk Avg
                               722010 non-null int32
                               722010 non-null int16
59 Bwd Packet/Bulk Avg
60 Bwd Bulk Rate Avg
                               722010 non-null int32
61 Subflow Fwd Packets
                               722010 non-null int8
62 Subflow Fwd Bytes
                               722010 non-null int16
                               722010 non-null int8
63 Subflow Bwd Packets
64 Subflow Bwd Bytes
                               722010 non-null int16
65 FWD Init Win Bytes
                               722010 non-null int16
66 Bwd Init Win Bytes
                               722010 non-null int16
67 Fwd Act Data Packets
                               722010 non-null int8
68 Fwd Seg Size Min
                               722010 non-null int8
69 Active Mean
                               722010 non-null float32
                               722010 non-null float32
70 Active Std
71 Active Max
                               722010 non-null float32
72 Active Min
                               722010 non-null float32
```

722010 non-null category 77 Label dtypes: category(1), float32(45), int16(10), int32(7), int8(15) memory usage: 168.0 MB

In [12]: sns.countplot(x='Label',data=data, palette='hls')

73 Idle Mean 74 Idle Std

75 Idle Max

76 Idle Min

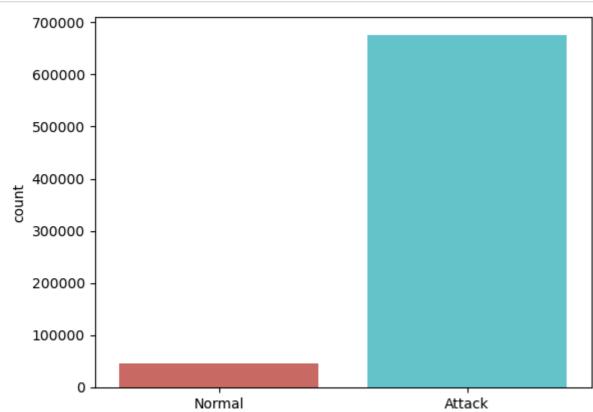
#plt.savefig('count_plot') mal: the nodule malignancy, 0: benign, 1: malignant

722010 non-null float32

722010 non-null float32

722010 non-null float32

722010 non-null float32



Label

In [13]: plt.figure(figsize = (10,5)) sns.heatmap(data.corr(), annot = True, cmap="rainbow")

plt.show()

1.00 Protocol Total Bwd packet Fwd Packet Length Max Fwd Packet Length St - 0.75 Bwd Packet Length Mean Flow Packets Flow IAT Max 0.50 Fwd IAT Mear Fwd IAT Mig Bwd IAT Sta Fwd PSH Flage 0.25 Bwd URG Flags (TST) Fwd Packets/8 Packet Length Mako 0.00 Packet Length Varianc RST Flag Coun URG Flag Coun Down/Up Ratie Bwd Segment Size Av Fwd Bulk Rate Ay Bwd Bulk Rate Avo Subflow Bwd Packets -0.50Bwd Init Win Byte Active Mean -0.75Active Min

In [14]: data['Label'].value_counts()

Out[14]: Attack 676302

Normal 45708 Name: Label, dtype: int64

In [16]: # Import Label encoder

from sklearn import preprocessing

label_encoder object knows # how to understand word labels.

label_encoder = preprocessing.LabelEncoder()

Encode labels in column 'species'. data['Label']= label_encoder.fit_transform(data['Label'])

In [17]: X = data.drop(["Label"],axis =1) y = data["Label"]

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```
FS
In [18]: from sklearn.feature_selection import SelectKBest, SelectPercentile, mutual_info_classif
In [19]: selector = SelectPercentile(mutual_info_classif, percentile=15)
         X_reduced = selector.fit_transform(X, y)
         #X_reduced.shape
In [20]: | cols = selector.get_support(indices=True)
         selected_columns = X.iloc[:,cols].columns.tolist()
         selected_columns
Out[20]: ['Total Length of Bwd Packet',
           'Fwd Packet Length Max',
          'Fwd Packet Length Mean',
          'Fwd Packet Length Std',
          'Bwd Packet Length Mean',
          'Fwd IAT Min',
           'Bwd PSH Flags',
          'Packet Length Mean',
          'Avg Packet Size',
          'Fwd Segment Size Avg',
           'Bwd Segment Size Avg',
          'FWD Init Win Bytes']
In [21]: len(selected_columns)
Out[21]: 12
In [22]: df = data[['Total Length of Bwd Packet',
          'Fwd Packet Length Max',
          'Fwd Packet Length Mean',
          'Fwd Packet Length Std',
          'Bwd Packet Length Mean',
          'Fwd IAT Min',
          'Bwd PSH Flags',
          'Packet Length Mean',
          'Avg Packet Size',
          'Fwd Segment Size Avg',
           'Bwd Segment Size Avg',
          'FWD Init Win Bytes','Label']]
In [23]: df.columns
Out[23]: Index(['Total Length of Bwd Packet', 'Fwd Packet Length Max',
                 'Fwd Packet Length Mean', 'Fwd Packet Length Std',
                 'Bwd Packet Length Mean', 'Fwd IAT Min', 'Bwd PSH Flags',
                'Packet Length Mean', 'Avg Packet Size', 'Fwd Segment Size Avg',
                 'Bwd Segment Size Avg', 'FWD Init Win Bytes', 'Label'],
               dtype='object')
In [24]: | X = df.drop(["Label"],axis =1)
         y = df["Label"]
In [25]: from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.20, random_state = 42)
         #X_train.shape, y_train.shape, X_test.shape, y_test.shape
In [26]: from sklearn.metrics import accuracy_score # for calculating accuracy of model
         from sklearn.metrics import precision_score
         from sklearn.metrics import recall_score
         from sklearn.metrics import f1_score
In [27]: ML_Model = []
         accuracy = []
         precision = []
         recall = []
         f1score = []
         #function to call for storing the results
         def storeResults(model, a,b,c,d):
             ML_Model.append(model)
             accuracy.append(round(a, 3))
             precision.append(round(b, 3))
             recall.append(round(c, 3))
             f1score.append(round(d, 3))
         BernoulliNB
In [28]: from sklearn.naive_bayes import BernoulliNB
         bnb = BernoulliNB(alpha=1.0, binarize=0.0, fit_prior=True, class_prior=None)
         bnb.fit(X_train, y_train)
         y_pred = bnb.predict(X_test)
```

```
bnb_acc = accuracy_score(y_pred, y_test)
         bnb_prec = precision_score(y_pred, y_test,average='weighted')
         bnb_rec = recall_score(y_pred, y_test,average='weighted')
         bnb_f1 = f1_score(y_pred, y_test,average='weighted')
In [29]: | storeResults('BernoulliNB', bnb_acc, bnb_prec, bnb_rec, bnb_f1)
```

Passive Aggressive

```
In [30]: from sklearn.linear_model import PassiveAggressiveClassifier
         pa = PassiveAggressiveClassifier(C=1.0, fit_intercept=True, max_iter=1000, tol=0.001, early_stopping=False,
                                          validation_fraction=0.1, n_iter_no_change=5, shuffle=True, verbose=0,
                                          loss='hinge', n_jobs=None, random_state=None, warm_start=False,
                                          class_weight=None, average=False)
         pa.fit(X_train, y_train)
         y_pred = pa.predict(X_test)
         pa_acc = accuracy_score(y_pred, y_test)
         pa_prec = precision_score(y_pred, y_test,average='weighted')
         pa_rec = recall_score(y_pred, y_test,average='weighted')
         pa_f1 = f1_score(y_pred, y_test,average='weighted')
```

SGDClassifier

```
In [32]: from sklearn.linear_model import SGDClassifier
         sgd = SGDClassifier(loss='hinge', penalty='12', alpha=0.0001, l1_ratio=0.15, fit_intercept=True,
                            max_iter=1000, tol=0.001, shuffle=True, verbose=0, epsilon=0.1, n_jobs=None,
                            random_state=None, learning_rate='optimal', eta0=0.0, power_t=0.5, early_stopping=False,
                            validation_fraction=0.1, n_iter_no_change=5, class_weight=None, warm_start=False, average=False)
         sgd.fit(X_train, y_train)
         y_pred = sgd.predict(X_test)
         sgd_acc = accuracy_score(y_pred, y_test)
         sgd_prec = precision_score(y_pred, y_test,average='weighted')
         sgd_rec = recall_score(y_pred, y_test,average='weighted')
         sgd_f1 = f1_score(y_pred, y_test,average='weighted')
```

In [33]: storeResults('SGDClassifier',sgd_acc,sgd_prec,sgd_rec,sgd_f1)

In [31]: storeResults('PassiveAggressive',pa_acc,pa_prec,pa_rec,pa_f1)

MLP Classifier

```
In [34]: from sklearn.neural_network import MLPClassifier
         mlp = MLPClassifier(hidden_layer_sizes=(100,), activation='relu', solver='adam', alpha=0.0001, batch_size='auto',
                             learning_rate='constant', learning_rate_init=0.001, power_t=0.5, max_iter=200, shuffle=True,
                             random_state=None, tol=0.0001, verbose=False, warm_start=False, momentum=0.9, nesterovs_momentum=True,
                             early_stopping=False, validation_fraction=0.1, beta_1=0.9, beta_2=0.999, epsilon=1e-08,
                             n iter no change=10, max fun=15000)
         mlp.fit(X_train, y_train)
         y_pred = mlp.predict(X_test)
         mlp_acc = accuracy_score(y_pred, y_test)
         mlp_prec = precision_score(y_pred, y_test,average='weighted')
         mlp_rec = recall_score(y_pred, y_test,average='weighted')
         mlp_f1 = f1_score(y_pred, y_test,average='weighted')
In [35]: storeResults('MLPClassifier',mlp_acc,mlp_prec,mlp_f1)
```

Ensemble

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Extension

Comparison

In [43]: result

```
Out[43]:
                    ML Model Accuracy Precision Recall F1_score
                   BernoulliNB
                                                          0.676
                                          0.887 0.618
           1 PassiveAggressive
                                          0.991 0.991
                                 0.991
                                                          0.991
                 SGDClassifier
                                 0.980
                                          0.980 0.980
                                                          0.980
                                           0.989 0.989
                  MLPClassifier
                                 0.989
                                 0.992
                                          0.992 0.992
                                                          0.992
                     Ensemble
                                 1.000
                                           1.000 1.000
                                                          1.000
                     Extension
```

In [41]: | storeResults('Extension',ml_acc,ml_prec,ml_rec,ml_f1)

Modelling

Out[45]: ['models/model_owndata.sav']

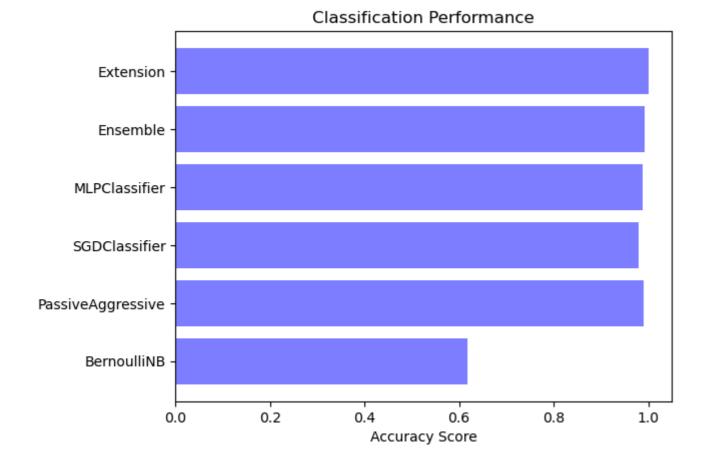
```
In [45]: import joblib
filename = 'models/model_owndata.sav'
joblib.dump(ext, filename)
```

Graph

```
In [46]: classifier = ML_Model
y_pos = np.arange(len(classifier))
```

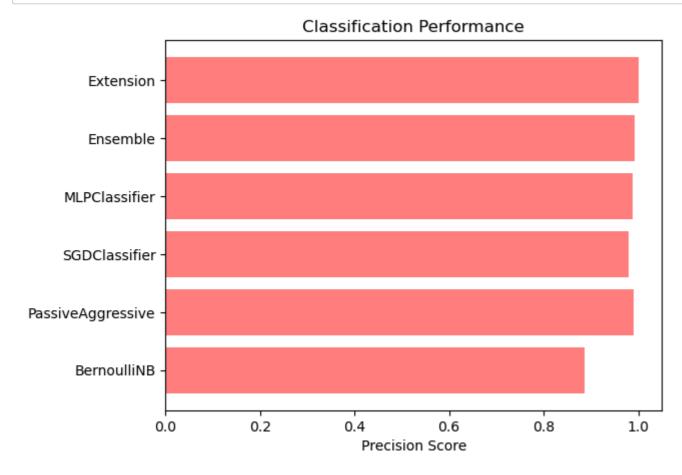
Accuracy

```
In [47]: import matplotlib.pyplot as plt2
    plt2.barh(y_pos, accuracy, align='center', alpha=0.5,color='blue')
    plt2.yticks(y_pos, classifier)
    plt2.xlabel('Accuracy Score')
    plt2.title('Classification Performance')
    plt2.show()
```



Precision

```
In [48]:
    plt2.barh(y_pos, precision, align='center', alpha=0.5,color='red')
    plt2.yticks(y_pos, classifier)
    plt2.xlabel('Precision Score')
    plt2.title('Classification Performance')
    plt2.show()
```

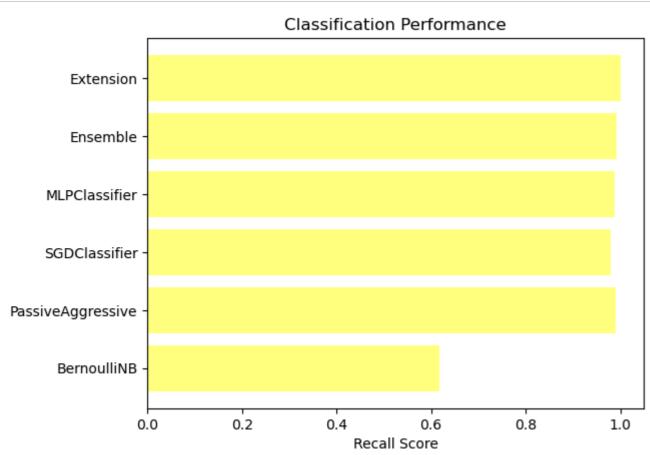


Recall

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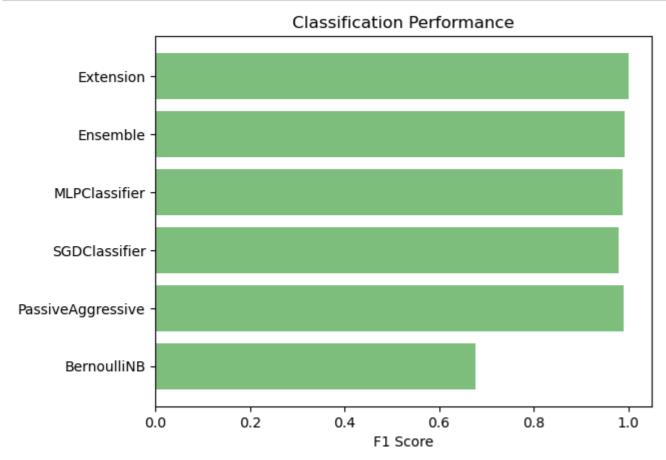
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In [49]: plt2.barh(y_pos, recall, align='center', alpha=0.5,color='yellow')
 plt2.yticks(y_pos, classifier)
 plt2.xlabel('Recall Score')
 plt2.title('Classification Performance')
 plt2.show()



F1 Score

In [50]: plt2.barh(y_pos, f1score, align='center', alpha=0.5,color='green')
 plt2.yticks(y_pos, classifier)
 plt2.xlabel('F1 Score')
 plt2.title('Classification Performance')
 plt2.show()



In []: