```
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 [4]: data = pd.read_csv('SlowRateSDN/LR-HR DDoS 2024 Dataset for SDN-Based Networks.csv')
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=['object']).columns.tolist(),'\n')
        # Print the first 5 lines
        data.head()
        0 null entries have been found in the dataset
        0 duplicate entries have been found in the dataset
        All duplicates have been removed
        Categorical columns: []
Out[6]:
            flow_duration protocol srcport dstport byte_count packet_count Tot Bwd Pkts TotLen Bwd Pkts TotLen Bwd Pkts Fwd Pkt Len Std ... Fwd Header Len Bwd Header Len Std Flag Cnt SYN Flag Cnt RST Flag Cnt Pkt Size Avg Init Fwd Win Byts Init Bwd Win Byts Label
                   274
                                                 26656
                                                               272
                                                                            2
                                                                                         31
                                                                                                                21.92031 ...
                                                                                                                                                         541.8000
                                                                                                                                                                                                 0 30.750000
                                                                                                                                                                                                                                      64240
                   274
                                                 26656
                                                               272
                                                                            2
                                                                                         31
                                                                                                                21.92031 ...
                                                                                                                                      40
                                                                                                                                                    40
                                                                                                                                                          541.8000
                                                                                                                                                                                                 0 30.750000
                                                                                                                                                                                                                          -1
                                                                                                                                                                                                                                      64240
                                                                                                                                                                          0
                                                               29
                                                                                                                                      40
                    29
                                                  2842
                                                                                                                 0.00000 ...
                                                                                                                                                    20
                                                                                                                                                           0.0000
                                                                                                                                                                          0
                                                                                                                                                                                                      0.000000
                                                                                                                                                                                                                          -1
                                                                                                                                                                                                                                      64240
                    29
                                                  2842
                                                               29
                                                                            1
                                                                                                        0
                                                                                                                 0.00000 ...
                                                                                                                                      40
                                                                                                                                                    20
                                                                                                                                                           0.0000
                                                                                                                                                                          0
                                                                                                                                                                                                      0.000000
                                                                                                                                                                                                                          -1
                                                                                                                                                                                                                                      64240
                                   0
                                                             29823
                                                                                        2556
                                                                                                     5188
                                                                                                                                     100
                                                                                                                                                                                                 0 693.692308
                                                                                                                                                                                                                                      64240
                   218
                             6 46004
                                          80 246238678
                                                                                                               700.40217 ...
                                                                                                                                                   160 441000.7473
                                                                                                                                                                                                                          -1
                                                                                                                                                                          0
        5 rows × 25 columns
In [7]: data.columns
Out[7]: Index(['flow_duration', 'protocol', 'srcport', 'dstport', 'byte_count',
                'packet_count', 'Tot Bwd Pkts', 'TotLen Fwd Pkts', 'TotLen Bwd Pkts',
                'Fwd Pkt Len Std', 'Flow Pkts/s', 'Fwd PSH Flags', 'Bwd PSH Flags',
```

In [8]: data['Label'].value\_counts()

dtype='object')

Out[8]: 0 42890

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Name: Label, dtype: int64

In [9]: # changing attack labels to their respective attack class

'Fwd URG Flags', 'Bwd URG Flags', 'Fwd Header Len', 'Bwd Header Len',

'Pkt Len Var', 'FIN Flag Cnt', 'SYN Flag Cnt', 'RST Flag Cnt', 'Pkt Size Avg', 'Init Fwd Win Byts', 'Init Bwd Win Byts', 'Label'],

#def change\_label(df):

#df['Label'].replace(['Hulk-Reqtimeout','Hulk-NoDefense','Hulk-Security2','Hulk-Evasive','TCPFlood-Reqtimeout','TCPFlood-Security2','TCPFlood-NoDefense','Slowloris-Reqtimeout','Slowhttptest-Security2',

# 'Slowhttptest-Evasive','Slowhttptest-NoDefense','Slowhttptest-Reqtimeout','Slowloris-Evasive','Slowloris-Security2'],'Attack',inplace=True)

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

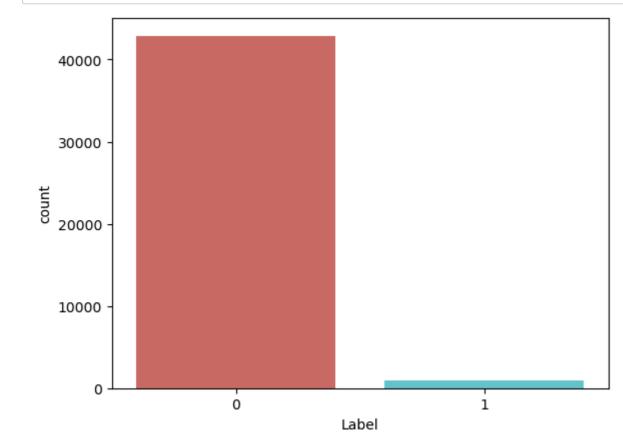
## In [10]: #change\_label(data)

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 43877 entries, 0 to 43876
Data columns (total 25 columns):
# Column
                      Non-Null Count Dtype
--- -----
                     -----
                     43877 non-null int64
0 flow_duration
    protocol
                     43877 non-null int64
1
                     43877 non-null int64
2 srcport
3 dstport
                     43877 non-null int64
                     43877 non-null int64
4 byte_count
5 packet_count
                     43877 non-null int64
                     43877 non-null int64
6 Tot Bwd Pkts
    TotLen Fwd Pkts
                    43877 non-null int64
 8 TotLen Bwd Pkts
                     43877 non-null int64
 9 Fwd Pkt Len Std
                    43877 non-null float64
                     43877 non-null float64
 10 Flow Pkts/s
11 Fwd PSH Flags
                     43877 non-null int64
                     43877 non-null int64
 12 Bwd PSH Flags
13 Fwd URG Flags
                     43877 non-null int64
14 Bwd URG Flags
                     43877 non-null int64
                     43877 non-null int64
 15 Fwd Header Len
                     43877 non-null int64
 16 Bwd Header Len
17 Pkt Len Var
                     43877 non-null float64
 18 FIN Flag Cnt
                     43877 non-null int64
                     43877 non-null int64
 19 SYN Flag Cnt
                     43877 non-null int64
 20 RST Flag Cnt
21 Pkt Size Avg
                     43877 non-null float64
22 Init Fwd Win Byts 43877 non-null int64
23 Init Bwd Win Byts 43877 non-null int64
                     43877 non-null int64
24 Label
dtypes: float64(4), int64(21)
memory usage: 8.4 MB
```

# 

plt.show()
#plt.savefig('count\_plot') Labeling traffic as normal (0) or malicious (1).



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```
In [13]: plt.figure(figsize = (10,5))
          sns.heatmap(data.corr(), annot = True, cmap="rainbow")
          plt.show()
               - 1.0
                                                                                      0.120.10.04<del>0</del>.00<mark>.26</mark>.010.07
                                                                                     -0.070106010768.149.40.0202.13
                                                                                    0.0302062005.859.18.0104033
                                                                                                                 -0.049.24
                                                                                                                                   - 0.8
                                                                                    -0.030401030103.064.202.0103029 0.0506078
                 byte_count-0.0-2060-0-202-0.03 1 0-010-0006-00.2010106-004-0-0.570.0087
packet_count-0.0-4080-0020-0602-0.71 1 0.04.0-2080-0.20-07-011 0.019
                                                                                   0.0590095
                                                                                    0.0404040101-10.010.102.00031056
                                                                                                                  0.10.018
                                                                                                                                   0.6
                 Tot Bwd Pkts -0.130.0060394-00305000664 1 0.350.86.0-0030074 0.016
                                                                                       870.99.0908.0110.001092
              TotLen Fwd Pkts -0.0-10.001.0006.000.0010.002012.03.39 10 0006.20-10.00210.0018
                                                                                        30.4090 906 00 003 9003 12 0 00 5063 2 0 .0 40 70 0 3 2
              TotLen Bwd Pkts 0.0801.0480206.002.80001.63 0.86003 00.000.000550,0011
                                                                                        80.840.107.009.04950010526
                                                                                                                                   0.4
              0.020302<del>0</del>.420.0305063005055 0.000067032
                   Flow Pkts/s-0.04020605.140.107.0905.090.104070400201905.50191 -0.023
                                                                                    -0.0907.1900.104089999.49430901.3047 0.0-840048
                                                                                                                                   0.2
               Fwd PSH Flags -
                                                                                    0.020201060905.0365.140.040.06 0.366.058
               Bwd PSH Flags - 0.4 0.3 0.1-0.09 008.00 10900 1.60 001.00 304 02 3
               Fwd URG Flags -
                                                                                                                                  - 0.0
              Bwd URG Flags -
             Fwd Header Len -0.1-0.0010302.03040007640.870.
                                                                                       1 0.90.0902.0104102.001011.9
                                                                                      0.92 10.0905.0101.101.00107.2
             Bwd Header Len -0.1-0.0610302-0300003040.990.490.89.0204007 0.016
                                                                                                                 0.0706011
                                                                                                                                   - -0.2
                  Pkt Len Var 0.0404007.8005.8003004.8010209.800601170.402.0089 0.005
                                                                                    0.09020951-0.0106101.00357
                                                                                                                  0.0503014
                  FIN Flag Cnt -0.0-70.1040509.0-6401-01.002.001.20-0036609.4903.50094-0.036
                                                                                    -0.0-D40-D101<mark>61-0</mark>.08.D04.4089
                                                                                                                  0.10.025
                                                                                     0.120.110.1-10.0811-0.010336
                 SYN Flag Cnt -0.2 (0.4 (0.180.202.060.120.10.030209050608.048 -0.19
                                                                                                                  0.50.073
                                                                                                                                   -0.4
                 RST Flag Cnt-0.0-D10 D20 194.00.30 01600300.0000560156055013 -0.01
                                                                                    -0.0.020.0010702564.401<u>31-</u>0.013 0.0209004
                  Pkt Size Avg -0.070.1080308029.09.0560.20.030.260.50.047 -0.06
                                                                                     0.190.20.70.08936.01
                                                                                                                  0.08.082
                                                                                                                                    -0.6
             Init Fwd Win Byts -
            Init Bwd Win Byts -0.410.70.0490506059.10.07050107006000006084
                                                            TotLen Bwd Pkts -
Fwd Pkt Len Std -
Flow Pkts/s -
                                                                      Fwd PSH Flags -
Bwd PSH Flags -
Fwd URG Flags -
Bwd URG Flags -
Fwd Header Len -
Bwd Header Len -
                                              byte_count
packet_count
Tot Bwd Pkts
                                                                                                FIN Flag Cnt
SYN Flag Cnt
                                                                                                        Flag Cnt
                                           dstport
In [14]: | data['Label'].value_counts()
Out[14]: 0 42890
          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"]
          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]: ['protocol', 'srcport', 'dstport', 'Bwd Header Len']
In [21]: len(selected_columns)
Out[21]: 4
In [22]: df = data[['protocol', 'srcport', 'dstport', 'Bwd Header Len', 'Label']]
In [23]: df.columns
Out[23]: Index(['protocol', 'srcport', 'dstport', 'Bwd Header Len', '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)
```

# BernoulliNB

accuracy.append(round(a, 3))
precision.append(round(b, 3))
recall.append(round(c, 3))
f1score.append(round(d, 3))

```
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)

In [31]: storeResults('PassiveAggressive',pa\_acc,pa\_prec,pa\_f1)

# **Passive Aggressive**

# **SGDClassifier**

In [33]: storeResults('SGDClassifier',sgd\_acc,sgd\_prec,sgd\_rec,sgd\_f1)

In [35]: storeResults('MLPClassifier',mlp\_acc,mlp\_prec,mlp\_f1)

## **MLP Classifier**

## Ensemble

# Extension

## In [39]: storeResults('Extension',ml\_acc,ml\_prec,ml\_rec,ml\_f1)

## Comparison

In [41]: result
Out[41]: ML Model Ac

```
ML Model Accuracy Precision Recall F1_score
        BernoulliNB
                                0.994 0.994
                      0.979
                                1.000 0.979
                                                0.989
1 PassiveAggressive
      SGDClassifier
                      0.994
                                0.994 0.994
                                                0.993
                      0.994
                                0.994 0.994
                                                0.993
      MLPClassifier
                      0.994
                                0.994 0.994
                                                0.993
         Ensemble
                      0.994
                                0.994 0.994
                                                0.993
         Extension
```

# Modelling

```
In [43]: import joblib
filename = 'models/model_slowratesdn.sav'
joblib.dump(ext, filename)

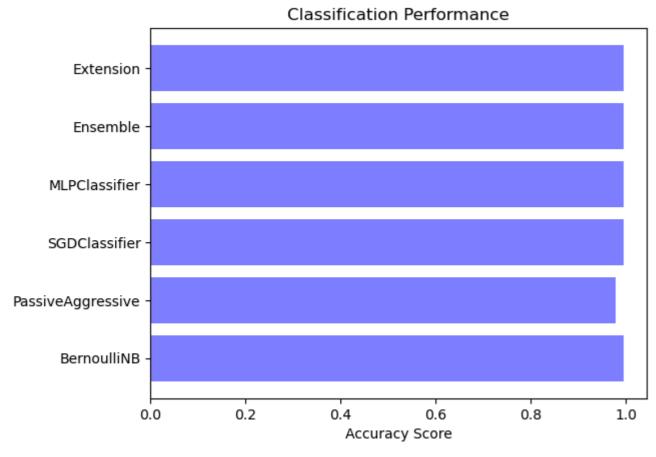
Out[43]: ['models/model_slowratesdn.sav']
```

# Graph

```
In [44]: classifier = ML_Model
y_pos = np.arange(len(classifier))
```

# Accuracy

```
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()
```

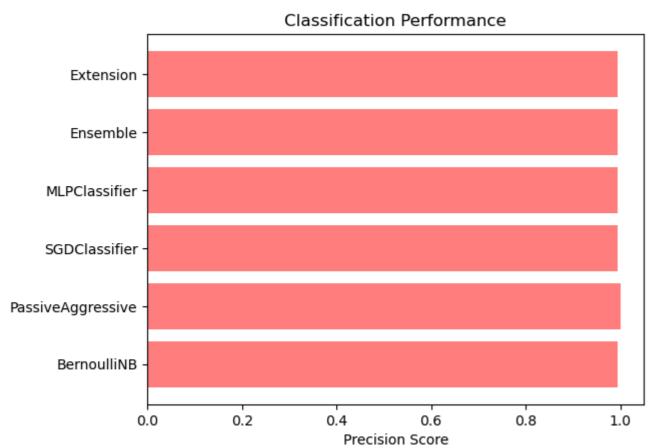


# **Precision**

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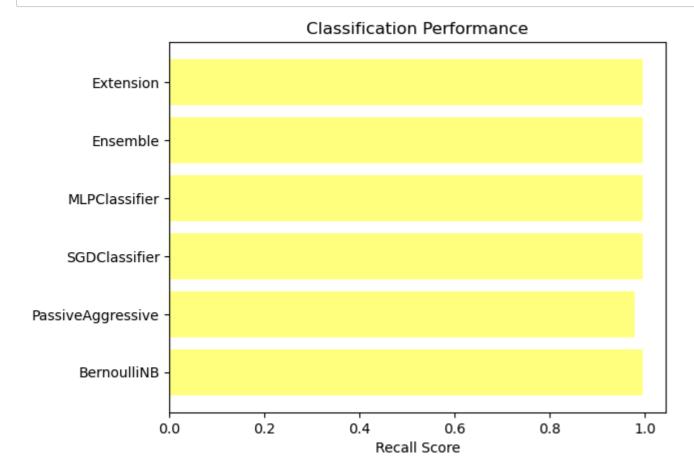
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In [46]: plt2.barh(y\_pos, precision, align='center', alpha=0.5,color='red')
 plt2.yticks(y\_pos, classifier)
 plt2.xlabel('Precision Score')
 plt2.title('Classification Performance')
 plt2.show()



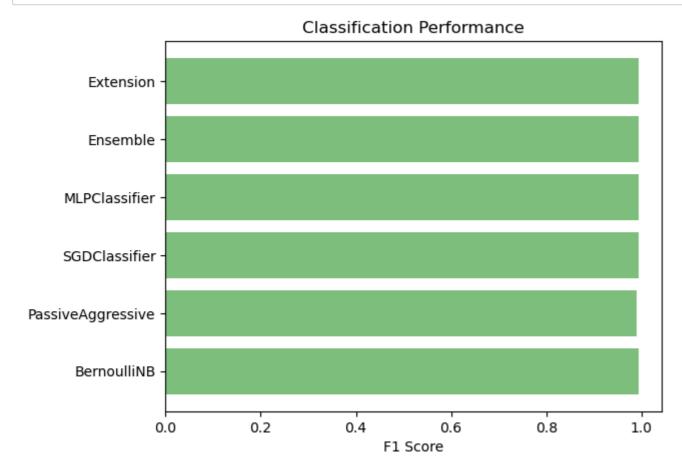
## Recall

```
In [47]: 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 [48]: 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 [ ]: