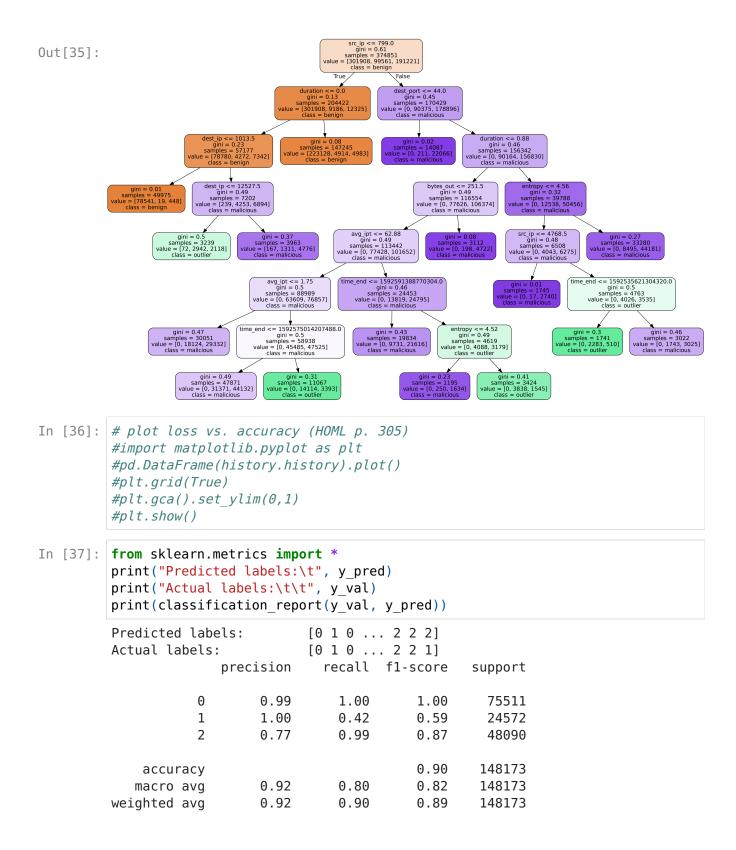
CS3315 Project: Random Forest Classifier

```
In [1]:
        import numpy as np
        import pandas as pd
        from sklearn.linear model import Perceptron
In [2]:
        # import data
        filename = 'data/2020.06.19.csv'
        df = pd.read csv(filename)
        # sample small subset
        # df = df.sample(500000, random state=78)
        df.head(2)
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 765360 entries, 0 to 765359
        Data columns (total 16 columns):
                            Non-Null Count
           Column
                                             Dtype
                            765360 non-null float64
         0
             avg ipt
         1 bytes in
                            765360 non-null int64
         2
           bytes out
                            765360 non-null int64
         3
           dest ip
                            765360 non-null int64
                           740863 non-null float64
           dest port
         5
                            765360 non-null float64
            entropy
          num pkts out 765360 non-null int64
         7
           num pkts in
                           765360 non-null int64
         8
           proto
                            765360 non-null int64
            src ip
                            765360 non-null int64
         10 src port
                           740863 non-null float64
         11 time end
                           765360 non-null int64
                            765360 non-null int64
         12 time start
         13 total entropy 765360 non-null float64
         14 label
                            765360 non-null object
         15 duration
                            765360 non-null float64
        dtypes: float64(6), int64(9), object(1)
        memory usage: 93.4+ MB
          avg_ipt bytes_in bytes_out dest_ip dest_port entropy num_pkts_out num_pkts_in proto
Out[2]:
        0
             7.5
                     342
                             3679
                                    786
                                           9200.0 5.436687
                                                                  2
                                                                             2
                                                                                   6
              0.0
                      0
                                    786
                                          55972.0 0.000000
                                                                  1
                                                                                   6
In [3]:
        # clean data
        df.dropna(inplace=True)
        df.isna().sum()
        # need to clean for features that are 0 and don't make sense (bytes = 0?)
```

```
Out[3]: avg_ipt
                          0
        bytes in
                          0
        bytes out
                          0
        dest ip
                          0
        dest port
                          0
        entropy
        num pkts out
        num pkts in
                          0
        proto
                          0
                          0
        src ip
        src port
                          0
        time end
                          0
        time start
                          0
                          0
        total entropy
        label
                          0
        duration
                          0
        dtype: int64
In [4]: print('label values:', df['label'].unique())
        def ordinal encoder(category):
            dict = {'benign':0, 'outlier':1, 'malicious':2}
            return dict[category]
        print('benign', ordinal encoder('benign'))
        print('outlier', ordinal_encoder('outlier'))
        print('malicious', ordinal encoder('malicious'))
        df['label'] = df['label'].apply(ordinal encoder)
        label values: ['benign' 'outlier' 'malicious']
        benign 0
        outlier 1
        malicious 2
In [5]: features = ['avg ipt',
                     'bytes in',
                     'bytes out',
                     'dest ip',
                     'dest port',
                     'entropy',
                     'num pkts in',
                     'num pkts out',
                     'proto',
                     'src ip',
                     'src port',
                     'time end',
                     'time start',
                     'total entropy',
                     'duration']
        X = df.loc[:, features]
        y = df.loc[:,'label']
```

```
In [13]: # test/train split
         from sklearn.model selection import train test split
         # 80/20 training/validation split
         X train, X val, y train, y val = train test split(X,y, train size=.8, test s
         y train = y train.to numpy()
         y val = y val.to numpy()
         # should print number of shape: (num features, num entries)
         print('Training set: ', 'X: ', X train.shape, 'y: ', y train.shape, 'Validat
         Training set: X: (592690, 15) y: (592690,) Validation set: X: (148173,
         15) printy: (148173,)
In [33]: from sklearn.ensemble import RandomForestClassifier
         rfc = RandomForestClassifier(n estimators=1000,
                                          max depth=30,
                                          max leaf nodes=16,
                                          n jobs=-1,
                                          min samples leaf=100,
                                          min samples split=100)
         rfc.fit(X train, y train)
Out[33]: ▼
                                  RandomForestClassifier
         RandomForestClassifier(max depth=30, max leaf nodes=16, min samples
         leaf=100,
                                 min samples split=100, n estimators=1000, n
         jobs=-1)
In [34]: y pred = rfc.predict(X val)
In [35]: from sklearn.tree import export graphviz
         export graphviz(rfc.estimators [900], out file='rfctree.dot',
                         feature names = features,
                         class_names = ['benign', 'outlier', 'malicious'],
                         rounded = True, proportion = False,
                         precision = 2, filled = True)
         from subprocess import call
         call(['dot', '-Tpng', 'rfctree.dot', '-o', 'tree.png', '-Gdpi=600'])
         # Display in jupyter notebook
         from IPython.display import Image
         Image(filename = 'tree.png')
```



Validate Model with Data from June 2022

```
In [38]: # import data
         filename = 'data/2020.06.20.csv'
         df2 = pd.read csv(filename)
         # sample small subset
         \#df2 = df2.sample(n=100000, random state=78)
         df2.info()
         df2.head(2)
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 770853 entries, 0 to 770852
         Data columns (total 16 columns):
              Column
                             Non-Null Count
                                              Dtype
                             -----
          0
              avg ipt
                             770853 non-null float64
                             770853 non-null int64
          1
              bytes in
          2
              bytes out
                             770853 non-null int64
          3
              dest ip
                             770853 non-null int64
          4
              dest port
                             770853 non-null int64
          5
              entropy
                             770853 non-null float64
              num pkts out
                             770853 non-null int64
          6
          7
                             770853 non-null int64
              num pkts in
          8
             proto
                             770853 non-null int64
          9
              src ip
                             770853 non-null int64
          10 src port
                             770853 non-null int64
          11 time end
                             770853 non-null int64
          12 time start
                             770853 non-null int64
          13 total entropy
                             770853 non-null float64
          14
             label
                             770853 non-null object
          15 duration
                             770853 non-null float64
         dtypes: float64(4), int64(11), object(1)
         memory usage: 94.1+ MB
             avg_ipt bytes_in bytes_out dest_ip dest_port entropy num_pkts_out num_pkts_in pro
Out[38]:
         0 34.57143
                        34
                                 29
                                       786
                                               5900 5.040459
                                                                     7
                                                                                10
         1 37.00000
                        34
                                 29
                                       786
                                               5900 5.127916
                                                                                10
In [39]: # clean data
         df2.dropna(inplace=True)
```

df2.isna().sum()

```
Out[39]: avg_ipt
                          0
         bytes in
                          0
         bytes out
                          0
         dest ip
                          0
         dest port
                          0
         entropy
         num pkts out
         num pkts in
                          0
                          0
         proto
         src ip
                          0
         src port
                          0
         time end
         time start
                          0
         total entropy
                          0
         label
                          0
         duration
                          0
         dtype: int64
In [40]: print('label values:', df2['label'].unique())
         def ordinal encoder(category):
             dict = {'benign':0, 'outlier':1, 'malicious':2}
             return dict[category]
         print('benign', ordinal encoder('benign'))
         print('outlier', ordinal encoder('outlier'))
         print('malicious', ordinal encoder('malicious'))
         df2['label'] = df2['label'].apply(ordinal encoder)
         label values: ['malicious' 'benign' 'outlier']
         benign 0
         outlier 1
         malicious 2
In [41]: features = ['avg ipt',
                      'bytes in',
                      'bytes_out',
                      'dest ip',
                      'dest port',
                      'entropy',
                      'num pkts in',
                      'num pkts out',
                      'proto',
                      'src_ip',
                      'src_port',
                     'time end',
                      'time start',
                      'total entropy',
                      'duration']
         X 22 = df2.loc[:, features]
         y 22 = df2.loc[:,'label']
In [42]: # change labels to numpy
         y 22 = y 22.to numpy()
```

```
In [43]: # test predictions
         X \text{ test new} = X 22
         test_pred_22 = rfc.predict(X_test_new)
In [44]: print("Predicted labels:\t", test_pred_22)
         print("Actual labels:\t\t", y 22)
         print(classification_report(y_22, test_pred_22))
         Predicted labels:
                                 [2 2 2 ... 2 2 2]
                                 [2 2 2 ... 1 1 2]
         Actual labels:
                      precision recall f1-score support
                                                    366310
                   0
                           0.99
                                     1.00
                                              0.99
                   1
                           0.30
                                     0.13
                                              0.18
                                                      69389
                                              0.88
                   2
                           0.84
                                     0.93
                                                      335154
                                              0.89
                                                      770853
             accuracy
            macro avg
                           0.71
                                     0.69
                                              0.69
                                                      770853
         weighted avg
                           0.86
                                     0.89
                                              0.87
                                                      770853
 In [ ]:
```