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.info()
        df.head(2)
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 765360 entries, 0 to 765359
        Data columns (total 16 columns):
                            Non-Null Count
         #
             Column
                                             Dtype
             -----
                            -----
        - - -
         0
             avg ipt
                            765360 non-null float64
         1
             bytes in
                            765360 non-null
                                             int64
         2
             bytes out
                            765360 non-null
                                             int64
         3
             dest ip
                            765360 non-null
                                             int64
         4
             dest port
                            740863 non-null
                                             float64
         5
             entropy
                            765360 non-null
                                             float64
         6
             num pkts out
                            765360 non-null
                                             int64
         7
                                             int64
             num pkts in
                            765360 non-null
         8
             proto
                            765360 non-null
                                             int64
```

10 src_port 740863 non-null float64 11 time_end 765360 non-null int64 12 time start 765360 non-null int64

765360 non-null

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

src ip

Out[2]:

9

	avg_ipt	bytes_in	bytes_out	dest_ip	dest_port	entropy	num_pkts_out	num_pkts_in	protc
0	7.5	342	3679	786	9200.0	5.436687	2	2	6
1	0.0	0	0	786	55972.0	0.000000	1	1	E

int64

```
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
                          0
        num pkts out
                          0
        num pkts in
                          0
        proto
        src ip
                          0
        src port
                          0
        time end
                          0
        time start
                          0
        total_entropy
                          0
        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'l
         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, t
         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, 'Va
         Training set: X:
                             (592690, 15) y: (592690,) Validation set: X: (14
         8173, 15) printy:
                             (148173,)
In [45]: from sklearn.ensemble import RandomForestClassifier
         rfc = RandomForestClassifier(n estimators=1000,
                                           max depth=30,
                                           n jobs=-1,
                                           min samples leaf=4,
                                           min samples split=4)
         rfc.fit(X_train, y train)
Out[45]:
                                   RandomForestClassifier
          RandomForestClassifier(max depth=30, min samples leaf=4, min samples
          split=4,
                                 n estimators=1000, n jobs=-1)
In [46]: y pred = rfc.predict(X val)
```

```
In [57]: 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')
          dot: graph is too large for cairo-renderer bitmaps. Scaling by 0.03956
          05 to fit
Out[57]: 0
In [48]: # plot loss vs. accuracy (HOML p. 305)
          #import matplotlib.pyplot as plt
          #pd.DataFrame(history.history).plot()
          #plt.grid(True)
          #plt.gca().set ylim(0,1)
          #plt.show()
In [49]: from sklearn.metrics import *
         print("Predicted labels:\t", y_pred)
print("Actual labels:\t\t", y_val)
          print(classification report(y val, y pred))
          Predicted labels:
                                    [0 1 0 ... 2 2 1]
                                    [0 \ 1 \ 0 \ \dots \ 2 \ 2 \ 1]
          Actual labels:
                                     recall f1-score
                        precision
                                                           support
                                                             75511
                     0
                              1.00
                                        1.00
                                                   1.00
                     1
                              1.00
                                        0.99
                                                   0.99
                                                             24572
                     2
                              0.99
                                        1.00
                                                   1.00
                                                             48090
                                                   1.00
                                                            148173
              accuracy
                              1.00
                                        1.00
                                                   1.00
                                                            148173
             macro avg
          weighted avg
                              1.00
                                         1.00
                                                   1.00
                                                            148173
```

Validate Model with Data from June 2022

```
In [50]: # 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):

Daca	cocamiis (cocac	10 CO Camin 5 / 1				
#	Column	Non-Null Count	Dtype			
0	avg_ipt	770853 non-null	float64			
1	bytes_in	770853 non-null	int64			
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			
6	num_pkts_out	770853 non-null	int64			
7	num_pkts_in	770853 non-null	int64			
8	proto	770853 non-null	int64			
9	<pre>src_ip</pre>	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			
dtype	es: float64(4),	int64(11), object(1)				
memoi	ry usage: 94.1+	MB				

Out[50]:

	avg_ipt	bytes_in	bytes_out	dest_ip	dest_port	entropy	num_pkts_out	num_pkts_in	pro
(34.57143	34	29	786	5900	5.040459	7	10	
	37.00000	34	29	786	5900	5.127916	7	10	

```
In [51]: # clean data
         df2.dropna(inplace=True)
         df2.isna().sum()
Out[51]: avg ipt
                           0
         bytes in
                           0
         bytes out
                           0
         dest ip
                           0
         dest port
                           0
         entropy
                           0
         num pkts out
                           0
         num pkts in
                           0
                           0
         proto
         src ip
                           0
                           0
         src port
         time end
                           0
         time start
                           0
         total entropy
                           0
         label
                           0
         duration
                           0
         dtype: int64
In [52]: 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 [53]: 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 [54]: # change labels to numpy
         y 22 = y 22.to numpy()
In [55]: # test predictions
         X test new = X 22
         test_pred_22 = rfc.predict(X_test_new)
In [56]: print("Predicted labels:\t", test pred 22)
         print("Actual labels:\t\t", y_22)
         print(classification report(y 22, test pred 22))
                                   [2 2 2 ... 2 2 2]
         Predicted labels:
         Actual labels:
                                   [2 2 2 ... 1 1 2]
                                     recall f1-score
                        precision
                                                         support
                    0
                             1.00
                                       1.00
                                                 1.00
                                                          366310
                    1
                             0.27
                                       0.46
                                                 0.34
                                                           69389
                             0.87
                                       0.74
                                                 0.80
                                                          335154
                                                 0.84
                                                          770853
             accuracy
            macro avg
                             0.71
                                       0.74
                                                 0.71
                                                          770853
         weighted avg
                             0.88
                                       0.84
                                                 0.85
                                                          770853
In [ ]:
```