```
In [0]:
from google.colab import drive
drive.mount('/content/drive')
Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client id=947318989803-6bn6
qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect uri=urn%3aietf%3awg%3aoauth%3a2.0%
b&response type=code&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2
www.googleapis.com%2fauth%2fdrive%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly
ttps%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly
Enter your authorization code:
Mounted at /content/drive
                                                                                                  F
In [0]:
import pandas as pd
df = pd.read csv('/content/drive/My Drive/Train data.csv')
print(df)
        duration protocol_type ... dst_host_srv_rerror_rate xAttack
Ω
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                        icmp ...
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                                                               normal
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                                                                normal
125971
                                                         0.00
              Ω
                         icmp ...
                                                                 dos
125972
              0
                                                         0.00
                          icmp ...
                                                               normal
[125973 rows x 42 columns]
In [0]:
X = df.iloc[:, :-1].values
print(X)
[[0 'icmp' 20 ... 0.0 0.05 0.0]
[0 'udp' 45 ... 0.0 0.0 0.0]
 [0 'icmp' 50 ... 1.0 0.0 0.0]
 [0 'icmp' 55 ... 0.0 0.01 0.0]
 [0 'icmp' 31 ... 1.0 0.0 0.0]
 [0 'icmp' 20 ... 0.0 0.0 0.0]]
In [0]:
y = df.iloc[:, -1].values
print(y)
['normal' 'normal' 'dos' ... 'normal' 'dos' 'normal']
In [0]:
import pandas as pd
from sklearn.preprocessing import LabelEncoder, OneHotEncoder
labelencoder_X = LabelEncoder()
X[:,0] = labelencoder X.fit transform(X[:,0])
X[:,1] = labelencoder X.fit transform(X[:,1])
onehotencoder = OneHotEncoder(categorical_features=[0,1])
X = onehotencoder.fit transform(X).toarray()
print(X)
```

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/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/_encoders.py:415: FutureWarning: The
handling of integer data will change in version 0.22. Currently, the categories are determined
based on the range [0, max(values)], while in the future they will be determined based on the
unique values.
If you want the future behaviour and silence this warning, you can specify "categories='auto'".
In case you used a LabelEncoder before this OneHotEncoder to convert the categories to integers, t
hen you can now use the OneHotEncoder directly.
 warnings.warn(msg, FutureWarning)
/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/ encoders.py:451: DeprecationWarning:
The 'categorical features' keyword is deprecated in version 0.20 and will be removed in 0.22. You
can use the ColumnTransformer instead.
  "use the ColumnTransformer instead.", DeprecationWarning)
     0. 0. ... 0.
                        0.05 0. ]
[[1.
               ... 0.
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0. 0. ]]
 ſ1.
           0.
 [1.
      0.
           0.
                ... 0.
In [0]:
labelencoder y = LabelEncoder()
y = labelencoder y.fit transform(y)
print(y)
[1 1 0 ... 1 0 1]
In [0]:
X.shape
Out[0]:
(125973, 3023)
In [0]:
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA, IncrementalPCA
X st = StandardScaler().fit transform(X)
In [0]:
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X_st, y, test_size=0.2, random_state=0)
In [0]:
from sklearn.neighbors import KNeighborsClassifier
#create new a knn model
knn = KNeighborsClassifier()
knn.fit(X train, y train)
y pred = knn.predict(X test)
In [33]:
y_pred = knn.predict(X_test)
KeyboardInterrupt
                                          Traceback (most recent call last)
<ipython-input-33-08dc72747141> in <module>()
----> 1 y pred = knn.predict(X test)
/usr/local/lib/python3.6/dist-packages/sklearn/neighbors/classification.py in predict(self, X)
    147
               X = check array(X, accept sparse='csr')
    148
--> 149
                neigh dist. neigh ind = self.kneighbors(X)
```

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nergn aree, nergn rna
                                        UULL . ....ULGIWULU (/1/
   150
                classes = self.classes
   151
                y = self. y
/usr/local/lib/python3.6/dist-packages/sklearn/neighbors/base.py in kneighbors(self, X,
n neighbors, return distance)
    452
                        delayed query(
    453
                           self. tree, X[s], n neighbors, return distance)
--> 454
                        for s in gen even slices(X.shape[0], n jobs)
    455
                    )
    456
                else:
/usr/local/lib/python3.6/dist-packages/joblib/parallel.py in call (self, iterable)
                    # remaining jobs.
   1001
   1002
                    self._iterating = False
-> 1003
                    if self.dispatch_one_batch(iterator):
   1004
                        self. iterating = self. original iterator is not None
   1005
/usr/local/lib/python3.6/dist-packages/joblib/parallel.py in dispatch one batch(self, iterator)
                        return False
    832
    833
                    else:
--> 834
                        self. dispatch (tasks)
   835
                        return True
    836
/usr/local/lib/python3.6/dist-packages/joblib/parallel.py in dispatch(self, batch)
                with self._lock:
    752
                    job idx = len(self. jobs)
--> 753
                    job = self._backend.apply_async(batch, callback=cb)
   754
                    # A job can complete so quickly than its callback is
                    # called before we get here, causing self._jobs to
    755
/usr/local/lib/python3.6/dist-packages/joblib/ parallel backends.py in apply async(self, func,
callback)
   199
            def apply async(self, func, callback=None):
   200
                """Schedule a func to be run"""
--> 201
                result = ImmediateResult(func)
   202
                if callback:
   203
                    callback(result)
/usr/local/lib/python3.6/dist-packages/joblib/ parallel backends.py in init (self, batch)
                # Don't delay the application, to avoid keeping the input
    580
    581
                # arguments in memory
--> 582
                self.results = batch()
   583
    584
           def get(self):
/usr/local/lib/python3.6/dist-packages/joblib/parallel.py in __call__(self)
                with parallel backend(self. backend, n jobs=self. n jobs):
                    return [func(*args, **kwargs)
    255
--> 256
                            for func, args, kwargs in self.items]
   257
    258
            def len (self):
/usr/local/lib/python3.6/dist-packages/joblib/parallel.py in tcomp>(.0)
                with parallel backend(self._backend, n_jobs=self._n_jobs):
   254
    255
                    return [func(*args, **kwargs)
--> 256
                            for func, args, kwargs in self.items]
    257
   258
            def __len__(self):
/usr/local/lib/python3.6/dist-packages/sklearn/neighbors/base.py in tree query parallel helper(tr
ee, data, n_neighbors, return_distance)
    289
            under PyPy.
    290
--> 291
            return tree.query(data, n neighbors, return distance)
   292
   293
KeyboardInterrupt:
In [34]:
from sklearn.metrics import classification_report, confusion_matrix
print(confusion_matrix(y_test, y_pred))
nrint (classification report (v test v nred))
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[[ 9146 19
             1
                    Ω
                          0.1
   6 13390
              41 20
ſ
        39 2303 0 0]
     0
         2
[
 Γ
                1.00 1.00
0.99 1.00
          0
                                  1.00
                                            9166
                                         13457
          1
                                  0.99
                         0.98
                                  0.98
                 0.98
                                           2342
          2
                                            215
          3
                 0.90
                          0.85
                                0.00
                                   0.87
                         0.00
          4
                 0.00
                                              15
                                  0.99 25195
   accuracy
macro avg 0.77
weighted avg 0.99
                      0.76
0.99
                                0.77
0.99
                                         25195
25195
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1437:
UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with
no predicted samples.
 'precision', 'predicted', average, warn_for)
In [35]:
from sklearn import metrics
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
Accuracy: 0.9930938678309188
In [0]:
from imblearn.metrics import sensitivity score, specificity score, geometric mean score
In [28]:
sensitivity score(y test, y pred, average='macro')
Out[28]:
0.8637591298763134
In [37]:
specificity_score(y_test, y_pred, average='macro')
Out[37]:
0.9976004637070919
In [38]:
geometric_mean_score(y_test, y_pred, average='macro')
Out[38]:
0.8733299138641619
In [0]:
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