

Bank Authentication Data Set

- from the UCI Repository

NN Project

```
In [22]: import pandas as pd
import seaborn as sns
%matplotlib inline
```

```
In [23]: data = pd.read_csv('bank_note_data.csv')
```

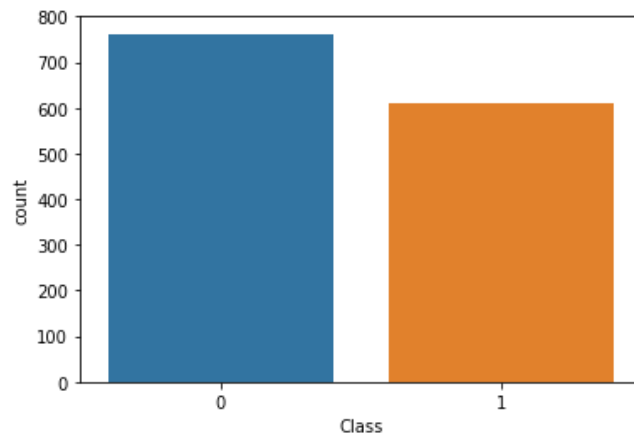
```
In [24]: data.head()
```

```
Out[24]:
```

| | Image.Var | Image.Skew | Image.Curt | Entropy | Class |
|---|-----------|------------|------------|----------|-------|
| 0 | 3.62160 | 8.6661 | -2.8073 | -0.44699 | 0 |
| 1 | 4.54590 | 8.1674 | -2.4586 | -1.46210 | 0 |
| 2 | 3.86600 | -2.6383 | 1.9242 | 0.10645 | 0 |
| 3 | 3.45660 | 9.5228 | -4.0112 | -3.59440 | 0 |
| 4 | 0.32924 | -4.4552 | 4.5718 | -0.98880 | 0 |

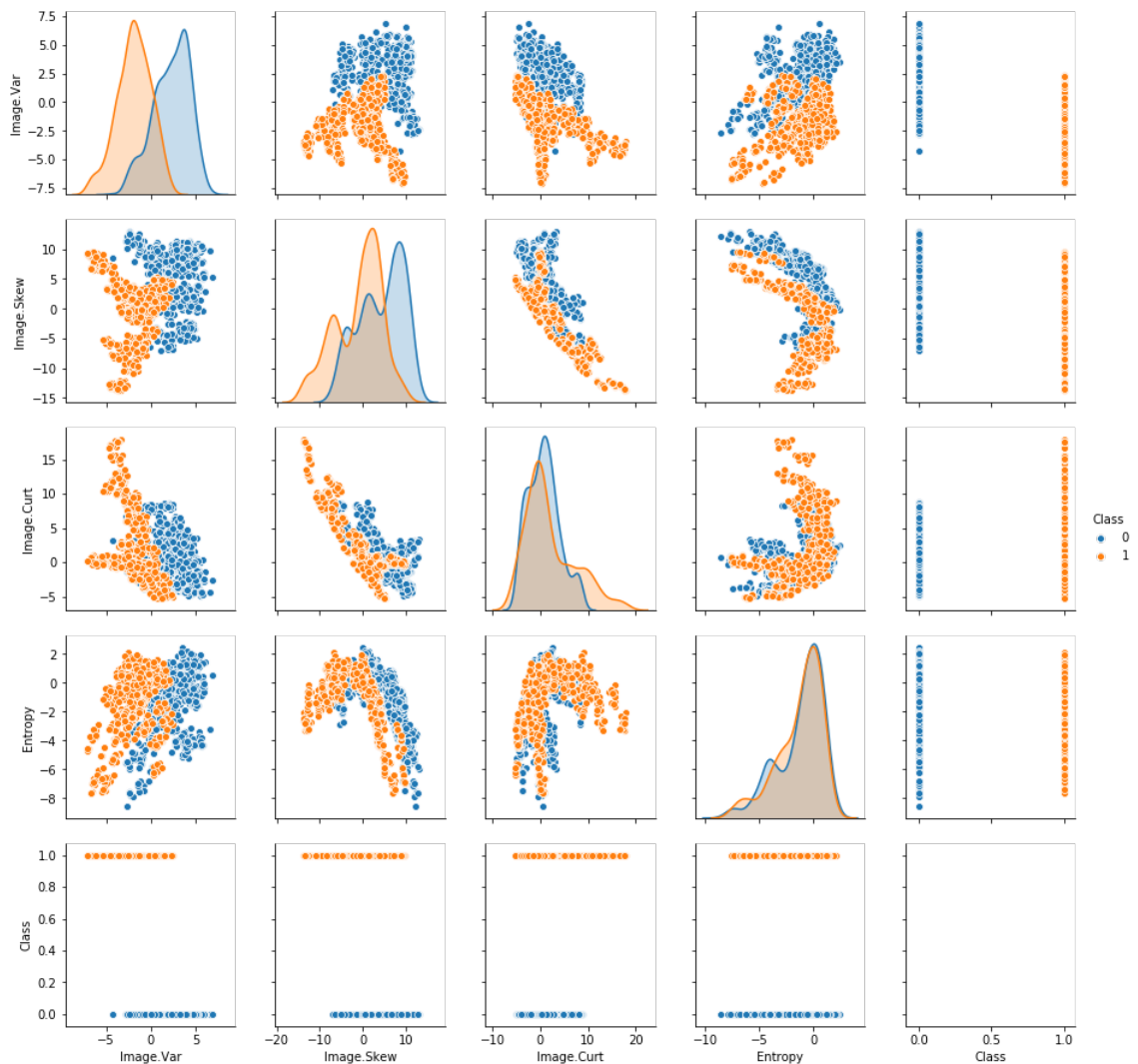
```
In [25]: sns.countplot(x='Class', data=data)
```

```
Out[25]: <matplotlib.axes._subplots.AxesSubplot at 0x1a3fb8cad0>
```



```
In [27]: #Take a look at the data
sns.pairplot(data,hue='Class')
```

```
Out[27]: <seaborn.axisgrid.PairGrid at 0x1a407f5b90>
```



```
In [29]: # Standard Scaling
from sklearn.preprocessing import StandardScaler
```

```
In [30]: # Create an object called scaler
scaler = StandardScaler()
```

```
In [31]: # Fit scaler to features
scaler.fit(data.drop('Class', axis=1))
```

```
Out[31]: StandardScaler(copy=True, with_mean=True, with_std=True)
```

```
In [32]: # Use transform method to transform the features to a scaled version
scaled_feature = scaler.transform(data.drop('Class', axis=1))
```

```
In [33]: # Convert scaled features to a dataframe and check the head to make sure it worke
d
df_feat = pd.DataFrame(scaled_feature, columns = data.columns[:-1])
df_feat.head()
```

```
Out[33]:
```

| | Image.Var | Image.Skew | Image.Curt | Entropy |
|---|-----------|------------|------------|-----------|
| 0 | 1.121806 | 1.149455 | -0.975970 | 0.354561 |
| 1 | 1.447066 | 1.064453 | -0.895036 | -0.128767 |
| 2 | 1.207810 | -0.777352 | 0.122218 | 0.618073 |
| 3 | 1.063742 | 1.295478 | -1.255397 | -1.144029 |
| 4 | -0.036772 | -1.087038 | 0.736730 | 0.096587 |

```
In [34]: # Train Test Split

X = df_feat

y = data['Class']
```

```
In [35]: from sklearn.model_selection import train_test_split
```

```
In [14]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30)
```

```
In [36]: # TensorFlow
import tensorflow as tf
```

```
In [37]: # Create a list of feature Columns
df_feat.columns
```

```
Out[37]: Index(['Image.Var', 'Image.Skew', 'Image.Curt', 'Entropy'], dtype='object')
```

```
In [38]: feat_cols = []

for col in df_feat.columns:
    feat_cols.append(tf.feature_column.numeric_column(col))
```

```
In [39]: feat_cols
```

```
Out[39]: [NumericColumn(key='Image.Var', shape=(1,), default_value=None, dtype=tf.float3
2, normalizer_fn=None),
NumericColumn(key='Image.Skew', shape=(1,), default_value=None, dtype=tf.float3
2, normalizer_fn=None),
NumericColumn(key='Image.Curt', shape=(1,), default_value=None, dtype=tf.float3
2, normalizer_fn=None),
NumericColumn(key='Entropy', shape=(1,), default_value=None, dtype=tf.float32,
normalizer_fn=None)]
```

```
In [40]: # Create an object called classifier. Set 2 classes [10,20,10] hidden unit layer
structure
classifier = tf.estimator.DNNClassifier(hidden_units = [10,20,10],
                                         n_classes = 2,
                                         feature_columns=feat_cols)
```

WARNING: Logging before flag parsing goes to stderr.
W0828 14:19:50.097635 140736573572032 estimator.py:1811] Using temporary folder
as model directory: /var/folders/bg/2b17ybm53nz575_6xtqlzxn0000gn/T/tmp2bw00jez

```
In [41]: # Create a tf.estimator.pandas_input train and batch size
```

```
input_func = tf.estimator.inputs.pandas_input_fn(x = X_train,
                                                  y = y_train,
                                                  batch_size = 20,
                                                  shuffle = True)
```

```
In [43]: # Train classifier to the input function
```

```
classifier.train(input_fn = input_func, steps=500)
```

W0828 14:28:02.788529 140736573572032 deprecation.py:323] From /anaconda3/lib/python3.7/site-packages/tensorflow/python/training/saver.py:1276: checkpoint_exists (from tensorflow.python.training.checkpoint_management) is deprecated and will be removed in a future version.

Instructions for updating:

Use standard file APIs to check for files with this prefix.

W0828 14:28:02.892272 140736573572032 deprecation.py:323] From /anaconda3/lib/python3.7/site-packages/tensorflow/python/training/saver.py:1066: get_checkpoint_mtimes (from tensorflow.python.training.checkpoint_management) is deprecated and will be removed in a future version.

Instructions for updating:

Use standard file utilities to get mtimes.

```
Out[43]: <tensorflow_estimator.python.estimator.canned.dnn.DNNClassifier at 0x1a4167d510>
```

Model Evaluation

```
In [44]: pred_fn = tf.estimator.inputs.pandas_input_fn(x=X_test,
                                                       batch_size = len(X_test),
                                                       shuffle = False)
```

```
In [45]: note_predictions = list(classifier.predict(input_fn = pred_fn))
```

```
In [46]: note_predictions[0]
```

```
Out[46]: {'logits': array([-13.71891], dtype=float32),
          'logistic': array([1.1026859e-06], dtype=float32),
          'probabilities': array([9.999989e-01, 1.101419e-06], dtype=float32),
          'class_ids': array([0]),
          'classes': array([b'0'], dtype=object),
          'all_class_ids': array([0, 1], dtype=int32),
          'all_classes': array([b'0', b'1'], dtype=object)}
```

```
In [47]: final_preds = []
         for pred in note_predictions:
             final_preds.append(pred['class_ids'][0])
```

```
In [48]: #Build the reports
         from sklearn.metrics import classification_report, confusion_matrix
```

```
In [49]: print(confusion_matrix(y_test, final_preds))
```

```
[[231  0]
 [ 0 181]]
```

```
In [50]: print(classification_report(y_test,final_preds))
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 1.00 | 1.00 | 1.00 | 231 |
| 1 | 1.00 | 1.00 | 1.00 | 181 |
| accuracy | | | 1.00 | 412 |
| macro avg | 1.00 | 1.00 | 1.00 | 412 |
| weighted avg | 1.00 | 1.00 | 1.00 | 412 |

Create a Random Forest Classifier and compare the confusion matrix and classification report to the DNN model

```
In [51]: from sklearn.ensemble import RandomForestClassifier
```

```
In [55]: rfc = RandomForestClassifier(n_estimators=200)
```

```
In [56]: rfc.fit(X_train,y_train)
```

```
Out[56]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                                max_depth=None, max_features='auto', max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, n_estimators=200,
                                n_jobs=None, oob_score=False, random_state=None,
                                verbose=0, warm_start=False)
```

```
In [57]: rfc_preds = rfc.predict(X_test)
```

```
In [58]: rfc = RandomForestClassifier(n_estimators=200)
```

```
In [59]: print(classification_report(y_test,rfc_preds))
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 1.00 | 0.99 | 0.99 | 231 |
| 1 | 0.98 | 0.99 | 0.99 | 181 |
| accuracy | | | 0.99 | 412 |
| macro avg | 0.99 | 0.99 | 0.99 | 412 |
| weighted avg | 0.99 | 0.99 | 0.99 | 412 |

```
In [60]: print(confusion_matrix(y_test,rfc_preds))
```

```
[[228  3]
 [ 1 180]]
```

```
In [ ]:
```

```
In [ ]:
```