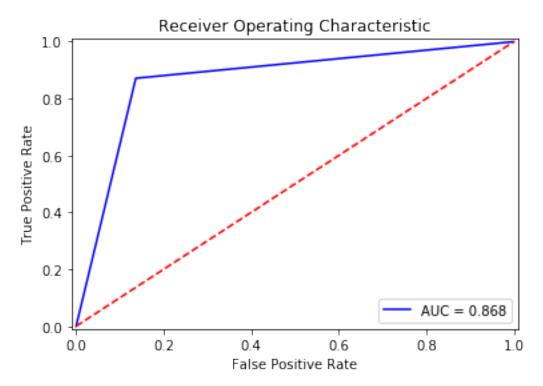
Beast Cancer Data Analysis Part 2

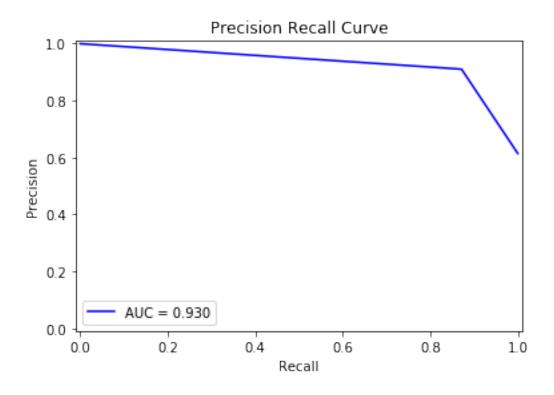
January 15, 2020

1 Decision Tree

```
[14]: import pandas as pd
      import seaborn as sns # for data visualization
      import matplotlib.pyplot as plt # for data visualization
      %matplotlib inline
      import numpy as np
      import os
 [2]: df = pd.read_csv('Breast_cancer_data.csv')
 [4]: df.tail()
      x = df.iloc[:,0:5]
      y = df.iloc[:,5:]
 [9]: from sklearn.tree import DecisionTreeClassifier
      model3 = DecisionTreeClassifier(max_features=3,criterion="entropy")
[10]: from sklearn.model_selection import train_test_split
      x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=0.
       \rightarrow20, random_state=1999)
[11]: model3.fit(x_train,y_train)
[11]: DecisionTreeClassifier(class_weight=None, criterion='entropy', max_depth=None,
                             max_features=3, 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, presort=False,
                             random_state=None, splitter='best')
[12]: y_pred3 = model3.predict(x_test)
[15]: import sklearn.metrics as metrics
      # calculate the fpr and tpr for all thresholds of the classification
      probs = model3.predict_proba(x_test) # probabilities for class 0,1
      preds = probs[:,1] # probabilities for class 1
```

```
fpr, tpr, threshold = metrics.roc_curve(y_test, preds)
roc_auc = metrics.auc(fpr, tpr)
plt.title('Receiver Operating Characteristic')
plt.plot(fpr, tpr, 'b', label = 'AUC = %0.3f' % roc_auc)
plt.legend(loc = 'lower right')
plt.plot([0, 1], [0, 1], 'r--')
plt.xlim([-0.01, 1.01])
plt.ylim([-0.01, 1.01])
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.show()
from sklearn.metrics import precision_recall_curve
from sklearn.metrics import auc
precision, recall, thresholds = precision_recall_curve(y_test, preds)
plt.title('Precision Recall Curve')
plt.plot(recall, precision, 'b', label = 'AUC = %0.3f' % auc(recall, precision))
plt.legend(loc = 'lower left')
plt.xlim([-0.01, 1.01])
plt.ylim([-0.01, 1.01])
plt.ylabel('Precision')
plt.xlabel('Recall')
plt.show()
```





```
[17]: from sklearn.linear_model import LogisticRegression
    from sklearn.metrics import accuracy_score
    acc = accuracy_score(y_test, y_pred3)
    print("Accuracy score using Logistic Regression:", acc)
```

Accuracy score using Logistic Regression: 0.868421052631579

[19]: from sklearn.metrics import classification_report print(classification_report(y_test,y_pred3))

support	f1-score	recall	precision	
44	0.84	0.86	0.81	0
70	0.89	0.87	0.91	1
114	0.87			accuracy
114	0.86	0.87	0.86	macro avg
114	0.87	0.87	0.87	weighted avg

```
[20]: from sklearn.metrics import log_loss
from sklearn.metrics import f1_score
print("log_loss is", '%.03f' %log_loss(y_test, probs))
print("F1 is", '%.03f' %f1_score(y_test, y_pred3, average='weighted'))
```

log_loss is 4.545 F1 is 0.869