## Performance Metrics

## April 22, 2022

Compute performance metrics for the given Y and Y\_score without sklearn.

```
[1]: import numpy as np import pandas as pd
```

```
[8]: class PerformanceMetricsABC(object):
         This class is a solution for problem A, B and C.
         def __init__(self, datafile, T):
             self.datafile = datafile
             self.T = T
             self.df = pd.read_csv(filepath_or_buffer=datafile)
             self.df dup = self.df.copy()
             self.df_dup['y_p'] = np.where(self.df_dup['proba'] < self.T, 0.0, 1.0)</pre>
         def compute_cm_values(self, data, c1, c2) -> tuple:
             11 11 11
             This method computes confusion matrix values.
             TN = len(data[(data[c1] == 0.0) & (data[c2] == 0.0)])
             FP = len(data[(data[c1] == 0.0) \& (data[c2] == 1.0)])
             FN = len(data[(data[c1] == 1.0) & (data[c2] == 0.0)])
             TP = len(data[(data[c1] == 1.0) \& (data[c2] == 1.0)])
             return TN, FP, FN, TP
         def compute_cm_rates(self) -> tuple:
             This method computes confusion matrix rates.
             TN, FP, FN, TP = self.compute_cm_values(
                 data=self.df_dup, c1='y', c2='y_p')
             N = TN + FP; P = FN + TP
             TNR = TN / N; FPR = FP / N; FNR = FN / P; TPR = TP / P
             return TNR, FPR, FNR, TPR
         def compute_recall(self) -> float:
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11 11 11
    This method computes recall.
    TN, FP, FN, TP = self.compute_cm_values(
        data=self.df_dup, c1='y', c2='y_p')
    N = TN + FP; P = FN + TP
    return TP / P
def compute_precision(self) -> float:
    This method computes precision.
    TN, FP, FN, TP = self.compute_cm_values(
        data=self.df_dup, c1='y', c2='y_p')
    return TP / (TP + FP)
def compute_f1_score(self) -> float:
    This method computes F1 score.
    r = self.compute_recall()
    p = self.compute_precision()
    return 2 * ((r * p) / (r + p))
def compute_accuracy(self) -> float:
    This method computes accuracy.
    TN, FP, FN, TP = self.compute_cm_values(
        data=self.df_dup, c1='y', c2='y_p')
    return (TN + TP) / (TN + FP + FN + TP)
def compute_auc(self) -> tuple:
    This method computes Area Under ROC Curve score.
    11 11 11
    FPR_arr = list()
    TPR_arr = list()
    A = list()
    sorted_df_dup = self.df_dup.sort_values(by='proba')
    for tau in sorted_df_dup['proba'].tolist():
        y_actual = np.array(sorted_df_dup['y'].to_list())
        y_pred = np.where((sorted_df_dup['proba'] < tau), 0, 1)</pre>
        tau_df = pd.DataFrame()
```

```
tau_df['y'] = y_actual
        tau_df['y_p'] = y_pred
        TN, FP, FN, TP = self.compute_cm_values(
            data=tau_df, c1='y', c2='y_p')
        N = TN + FP; P = FN + TP
        FPR = FP / N; TPR = TP / P
        FPR arr.append(FPR)
        TPR_arr.append(TPR)
        A.append((500 * FN) + (100 * FP))
    FPR_arr = np.array(FPR_arr)
    TPR_arr = np.array(TPR_arr)
    A = np.array(A)
    optimum_t = np.min(a=A)
    auc = np.abs(np.trapz(y=TPR_arr, x=FPR_arr))
    return FPR_arr, TPR_arr, optimum_t, round(auc, 5)
def report(self) -> None:
    11 11 11
    This method prints the performance metrics.
    from matplotlib import pyplot as plt
    from matplotlib import style
    style.use(style='seaborn-whitegrid')
    TN, FP, FN, TP = self.compute_cm_values(
        data=self.df_dup, c1='y', c2='y_p')
    conf_mat = np.matrix(data=[[TN, FP], [FN, TP]])
    f1_score = round(self.compute_f1_score(), 5)
    FPR_arr, TPR_arr, optimum_t, auc = self.compute_auc()
    acc_score = round(self.compute_accuracy(), 5)
    print("Confusion Matrix:\n{}\n".format(conf_mat))
    print("F1 Score: {}\n".format(f1_score))
    plt.figure(figsize=(10, 6))
    plt.plot(FPR_arr, TPR_arr)
    title = 'Receiver Operating Characteristic Curve (AUC = {})'.format(auc)
    plt.title(label=title)
    plt.xlabel(xlabel='FPRs')
    plt.ylabel(ylabel='TPRs')
    plt.show()
```

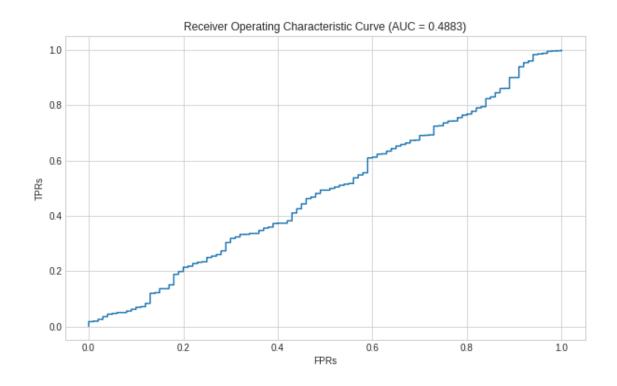
```
print("\nOptimum Threshold: {}\n".format(optimum_t))
print("Accuracy: {}".format(acc_score))
return None
```

A. Compute performance metrics for the given data '5\_a.csv'

```
[3]: prob_A = PerformanceMetricsABC(datafile='5_a.csv', T=0.5) prob_A.report()
```

Confusion Matrix: [[ 0 100] [ 0 10000]]

F1 Score: 0.99502



Optimum Threshold: 10000

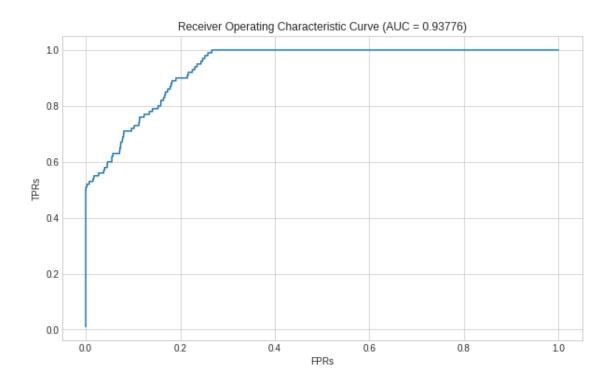
Accuracy: 0.9901

B. Compute performance metrics for the given data '5\_b.csv'

```
[4]: prob_B = PerformanceMetricsABC(datafile='5_b.csv', T=0.5) prob_B.report()
```

Confusion Matrix: [[9761 239] [ 45 55]]

F1 Score: 0.27919



Optimum Threshold: 25000

Accuracy: 0.97188

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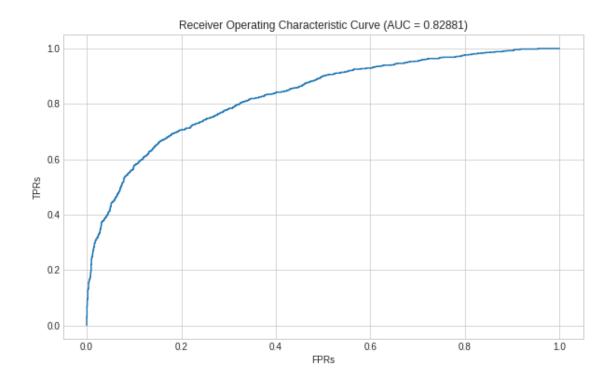
C. Compute the best threshold for the given data '5\_c.csv'

Confusion Matrix:

[[1637 168]

[ 462 585]]

## F1 Score: 0.65



Optimum Threshold: 141000

Accuracy: 0.7791

```
[6]: class PerformanceMetricsD(object):
    """
    This class is a solution for problem D.
    """

def __init__(self, datafile):
    self.datafile = datafile
    self.df = pd.read_csv(filepath_or_buffer=self.datafile)
    self.ai = self.df['y'].to_numpy()
    self.pi = self.df['pred'].to_numpy()
    self.ei = np.abs(self.ai - self.pi)
    self.ai_m = np.mean(a=self.ai)
    self.n = len(self.df)

def compute_mse(self) -> float:
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11 11 11
    This method computes MSE.
    return (1 / self.n) * (self.compute_ssr())
def compute_mape(self) -> float:
    HHHH
    This method computes MAPE.
    return np.sum(self.ei) / np.sum(self.ai)
def compute_sst(self) -> float:
    11 11 11
    This method computes SST.
    return np.sum((self.ai - self.ai_m) ** 2)
def compute_ssr(self) -> float:
    This method computes SSR.
    return np.sum(self.ei ** 2)
def compute_R2(self) -> float:
    HHHH
    This method computes R^2.
    return (1 - (self.compute_ssr() / self.compute_sst()))
def report(self) -> None:
    This method prints the performance metrics.
    mse = round(self.compute_mse(), 5)
    mape = round(self.compute_mape(), 5)
    r2 = round(self.compute_R2(), 5)
    print("MSE: {}\n".format(mse))
    print("MAPE: {}\n".format(mape))
    print("R^2: {}".format(r2))
```

D. Compute performance metrics (for regression) for the given data '5\_d.csv'

```
[7]: prob_D = PerformanceMetricsD(datafile='5_d.csv')
prob_D.report()
```

MSE: 177.1657

MAPE: 0.12912		
R^2: 0.95636		
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