```
In [1]: from math import sqrt
In [2]: def indicator(TP, FP, FN, TN):
           acc = round(((TP + TN) / (TP + FN + FP + TN)),4)
           recall_positive = round((TP / (TP + FN)),4)
           recall_negative = round((TN / (FP + TN)),4)
precision_positive = round((TP / (TP + FP)),4)
           precision_negative = round((TN / (TN + FN)),4)
           f1_score_positive = round((2*precision_positive*recall_positive / (precision_positive + recall_positive)),4)
           f1_score_negative = round((2*precision_negative*recall_negative / (precision_negative + recall_negative)),4)
           p_0 = (TN + TP) / (TN + FP + FN + TP)
           p_c = ((TN+FN)*(TN+FP) + (FN+TP)*(FP+TP)) / (TN + FP + FN + TP)**2
           kappa = round(((p_0 - p_c) / (1-p_c)),4)
           numerator = (TP * TN) - (FP * FN)
           denominator = sqrt((TP+FP) * (TP+FN) * (TN+FP) * (TN+FN))
           mcc = round((numerator / denominator),4)
           Acc (Accuracy): 不適用於不平衡資料,故不以此為衡量標準
       recall_positive:實際為Positive的樣本中·被正確判斷為Positive的比率
       recall_negative:實際為Negative的樣本中,被正確判斷為Negative的比率
       precision positive:預測為Positive樣本中,實際亦為Positive的比率
       precision negative:預測為Negative的樣本中,實際亦為Negative的比率
       f1 score: precision和recall的調和平均
       kappa: Cohen's kappa係數是用來分析兩個審查者對於類別項目評分的一致性 · (0,1)越高越好
       mcc: Matthews correlation coefficient常用於不平衡類別衡量標準,(-1,1), 1表示完美預測, -1表示完全不一致
        ModelA
In [3]: TP, FP, TN, FN = (853, 341, 7230, 576)
In [4]: indicator(TP, FP, FN, TN)
        1.acc: 0.8981
        2.recall_positive: 0.5969,
        3.recall_negative: 0.955
        4.precision_positive: 0.7144 ,
        5.precision_negative: 0.9262,
        6.f1_score_positive: 0.6504 ,
        7.f1_score_negative: 0.9404,
        8.kappa: 0.5913 ,
        9.mcc: 0.5946
        ModelB
In [5]: TP, FP, TN, FN = (846, 316, 7255, 583)
In [6]: indicator(TP, FP, FN, TN)
        1.acc: 0.9001,
        2.recall_positive: 0.592 ,
        3.recall_negative: 0.9583
        4.precision_positive: 0.7281 ,
        5.precision_negative: 0.9256,
        6.f1_score_positive: 0.653
        7.f1_score_negative: 0.9417,
        8.kappa: 0.5954 ,
        9.mcc: 0.5998
       由以上結果可知:
       1.Acc: ModelA > ModelB, 在不平衡資料當中,此並非為衡量的主要標準
       2.recall_positive: ModelA > ModelB, 表示實際為Positive的樣本中·被正確判斷為Positive的比率ModelA比ModelB高一點
       3.recall_negative: ModelA < ModelB, 表示實際為Negative的樣本中·被正確判斷為Negative的比率ModelA比ModelB低一些
       4.precision_positive: ModelA < ModelB, 表示預測為Positive樣本中·實際亦為Positive的比率ModelA比ModelB低
       5.precision_negative: ModelA > ModelB, 表示預測為Negative的樣本中,實際亦為Negative的比率ModelA比ModelB高
```

6.f1_score_positive: ModelA < ModelB, 表示對於Positive的預測ModelA比ModelB低

 $7.f1_score_negative: ModelA < ModelB$,表示對於Negative的預測ModelA比ModelB低

8.kappa : ModelA < ModelB, 表示ModelB比ModelA好 9.mcc : ModelA < ModelB, 表示ModelB比ModelA好

綜合以上衡量標準, ModelB比ModelA好

In []: