

# MODEL BUILDING

## Model Evaluation

- There are many numbers of model evaluation techniques for the classification type of machine learning models.
- The following are widely used
  - ❖ Accuracy\_score
  - ❖ Confusion matrix
  - ❖ Roc- Auc Curve

### Accuracy

- One of the widely used metrics that computes the performance of classification models is accuracy. The percentage of labels that our model successfully predicted is represented by accuracy. For instance, if our model accurately classified 80 of 100 labels, its accuracy would be 0.80.

### Accuracy score:

- ❖ The accuracy\_score () method of sklearn.metrics, accept the true labels of the sample and the labels predicted by the model as its parameters and computes the accuracy score as a float value, which can likewise be used to obtain the accuracy score in Python. There are several helpful functions to compute typical evaluation metrics in the sklearn.metrics class.

### Recall score:

- Recall\_score computes a precision-recall curve from the ground truth label and a score given by the classifier by varying a decision threshold.

### Roc auc score:

- Roc\_auc\_score compute Area Under the Receiver Operating Characteristic Curve (ROC AUC) from prediction scores.

### Confusion matrix:

- Confusion\_matrix compute confusion matrix to evaluate the accuracy of a classification.

## Step-1:

- After training the data, we need to check accuracy rate by sklearn.metrics library.
- Import **sklearn.metrics** and import **accuracy\_score**, **roc\_auc\_score** and **confusion\_matrix**.
- Then display the accuracy\_score, roc\_auc\_score and confusion\_matrix values as output from Trained models.

The screenshot shows a Jupyter Notebook titled "University Admit Eligibility Predictor.ipynb". The left sidebar displays a file explorer with a list of files and folders, including "Dataset", "Output", "Static", "Templates", "Video", "app.py", "University Admit Eligibility ...", and "university.pkl". The main area of the notebook contains the following code and output:

```
[24]: lr = cls.fit(X_train,Y_train)

[25]: y_pred = lr.predict(X_test)
      y_pred

[25]: array([ True,  True,  True,  True, False,  True,  True,  True,  True,
         True,  True,  True,  True,  True,  True,  True,  True,  True,
         True,  True,  True,  True,  True,  True,  True,  True,  True,
        False,  True,  True,  True,  True,  True,  True,  True,  True,
         True,  True,  True,  True,  True,  True,  True,  True,  True,
         True,  True,  True,  True,  True,  True,  True,  True,  True,
         True,  True,  True,  True,  True,  True,  True,  True,  True,
         True,  True,  True,  True,  True,  True,  True,  True,  True])
```

### 2. Model Evaluation

```
[26]: from sklearn.metrics import accuracy_score,recall_score,roc_auc_score,confusion_matrix

[27]: print("\nAccuracy score:%f" %(accuracy_score(Y_test,y_pred)*100))
      print("Recall score:%f" %(recall_score(Y_test,y_pred)*100))
      print("ROC score : %f\n" %(roc_auc_score(Y_test,y_pred)*100))
      print(confusion_matrix(Y_test,y_pred))

Accuracy score:91.250000
Recall score:97.333333
ROC score : 48.666667

[[ 0  5]
 [ 2 73]]
```

### 3. Save The Model