

## Evaluate the Trained Model

- Calculating the classification report to determine precision, recall F1-score etc..

```
from sklearn.metrics import classification_report  
  
print(classification_report(y_test, predictions))  
  
precision    recall   f1-score   support  
0            0.81      0.88      0.85      2986  
1            0.72      0.60      0.66      1514  
  
accuracy                           0.79      4500  
macro avg       0.77      0.74      0.75      4500  
weighted avg    0.78      0.79      0.78      4500
```

S.N	$A_1$	$A_2$	$E_c D$ $C_1(4,2)$	$E D$ $C_2(3,3)$	$X_4$ $(1,5) C_1$	$Y_4$ $(1,5) C_2$
1	2	5	$X_1(3,6)$	$Y_1(1)$ $\rightarrow C_2$		
2	5	3	$X_2(1,4)$	$Y_2(2,8) \rightarrow C_1$	$X_5$ $(4,4) C_1$	$Y_5$ $(4,4) C_2$
3	3	1	$X_3(1,4)$	$Y_3(4) \rightarrow C_1$		
4	1	5	$X_4(4,2)$	$Y_4(2) \rightarrow C_2$		
5	4	4	$X_5(2)$	$Y_5(1,4) \rightarrow C_2$		

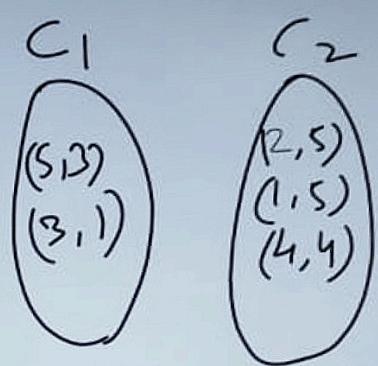
$$\sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

$$= \sqrt{2^2 + 3^2} < 3.6$$

$$(2,5) \begin{pmatrix} X_1 \\ Y_1 \end{pmatrix} (4,2) \begin{pmatrix} X_2 \\ Y_2 \end{pmatrix}$$

$$Y = \sqrt{1^2 + 0^2} = 1$$

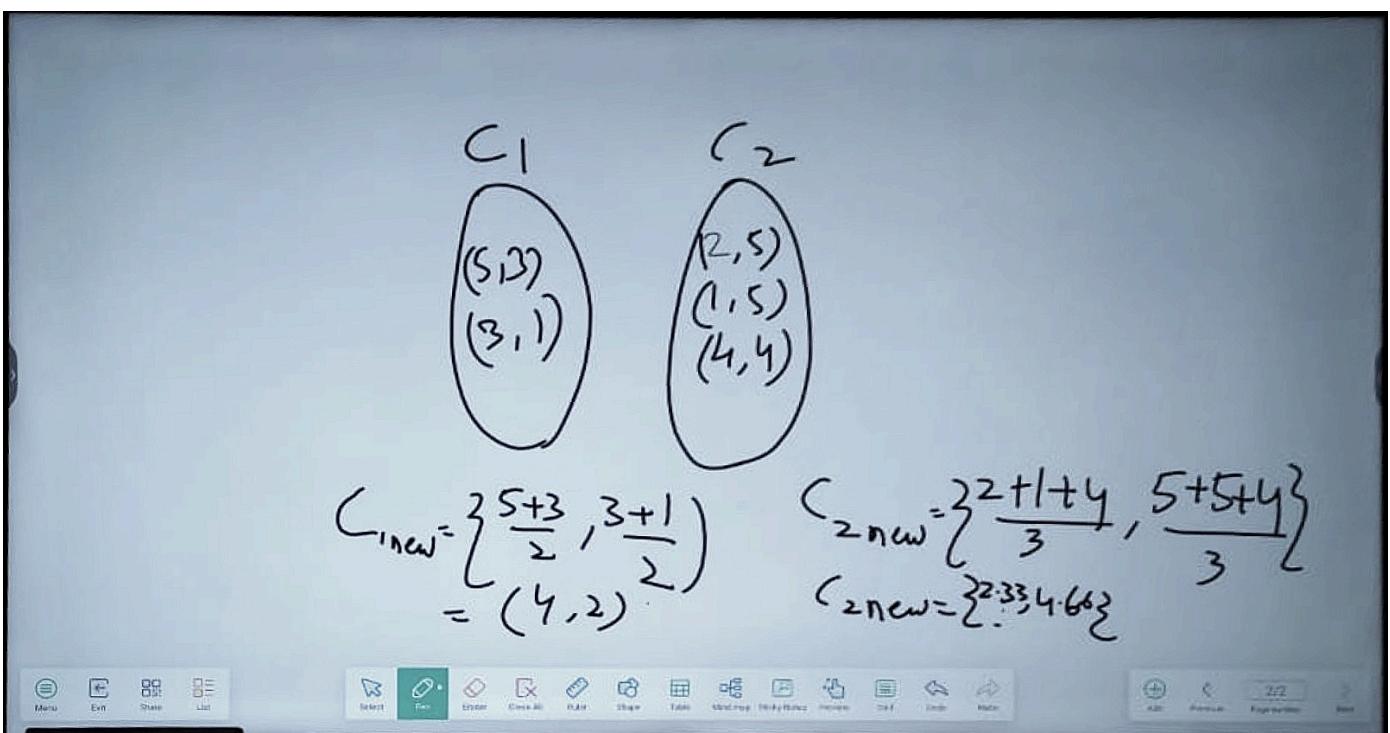
$$(5,3) \begin{pmatrix} X_3 \\ Y_3 \end{pmatrix} (4,2) \begin{pmatrix} X_4 \\ Y_4 \end{pmatrix} (3,1) \begin{pmatrix} X_5 \\ Y_5 \end{pmatrix} (3,5)$$



$$C_{1\text{new}} = \left\{ \frac{5+3}{2}, \frac{3+1}{2} \right\} = (4, 2)$$

$$C_{2\text{new}} = \left\{ \frac{2+1+4}{3}, \frac{5+5+4}{3} \right\}$$

$$C_{2\text{new}} = \{2.33, 4.66\}$$



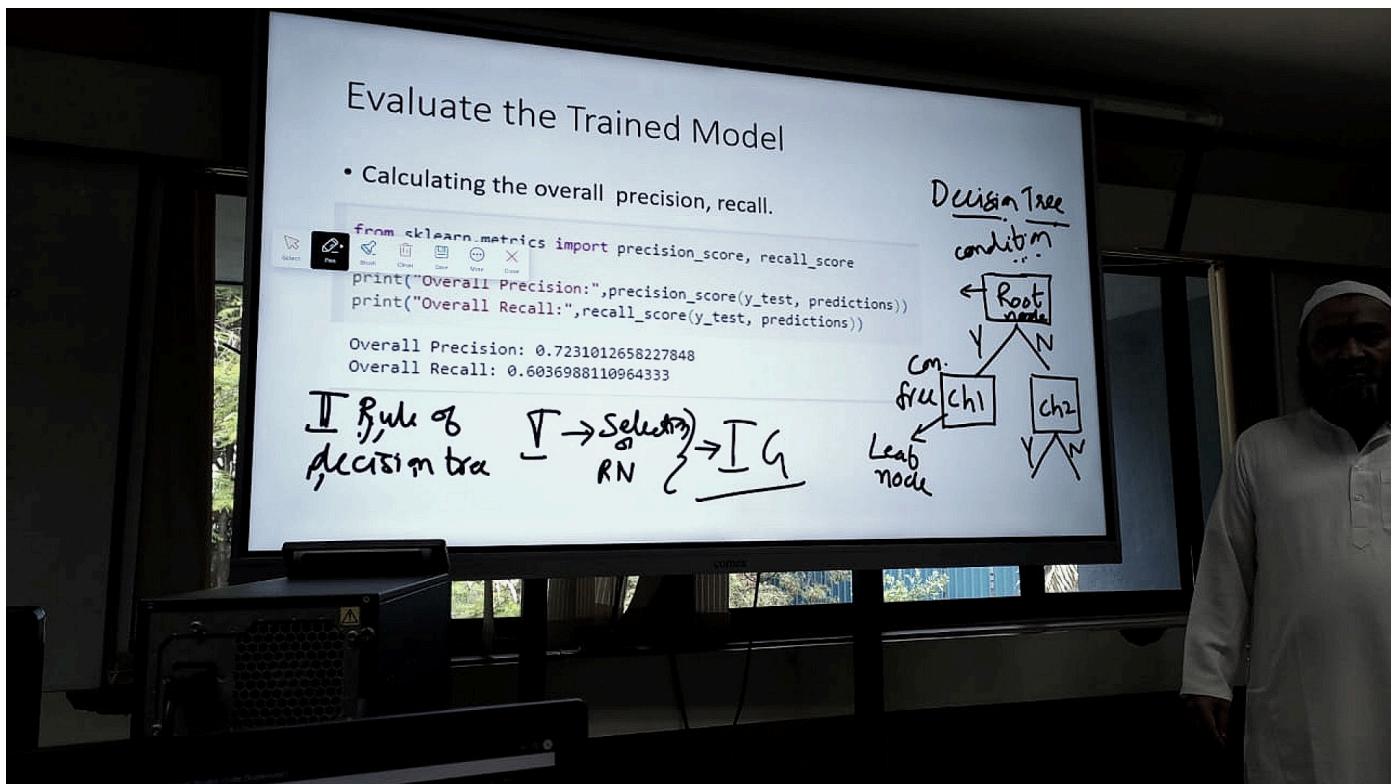
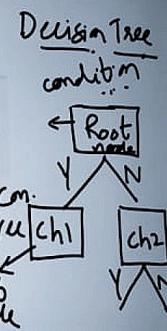
## Evaluate the Trained Model

- Calculating the overall precision, recall.

```
from sklearn.metrics import precision_score, recall_score
print("Overall Precision:",precision_score(y_test, predictions))
print("Overall Recall:",recall_score(y_test, predictions))
```

Overall Precision: 0.7231012658227848  
Overall Recall: 0.6036988110964333

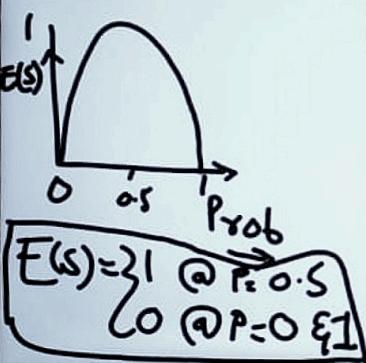
I Rule of decision tree     $\overline{T} \rightarrow \text{Select}(\overline{s}) \rightarrow \overline{T} G$



## Example

- A researcher is trying to identify the root node to design a decision tree classifier to classify the students based on scores greater than or equal to 50% which is a pass and less than 50% which is a fail. The data is given in the below table.

Student ID	No. of Assignments Completed	Number of Hours Studied	Attendance in %	Marks out of 100	Target C
1	Less than 5	Less than 3 hours	Less than 70%	30	F
2	Less than 5	Less than 3 hours	Less than 70%	45	F
3	Less than 5	More than 3 hours	Less than 70%	50	P
4	More than 5	More than 3 hours	Less than 70%	70	P
5	More than 5	More than 3 hours	More than 70%	75	P
6	More than 5	More than 3 hours	More than 70%	85	P



$$E(S) = - \sum_{i=1}^c p_i \log p_i, c = \text{no. of classes}$$

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