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## A python Program to Implement Ada Boosting

Aim:-

To implement a python program for  
Ada Boosting.

Algorithm :-

Step 1:- Import Necessary Libraries

Import numpy as np

Step 2:- Load and prepare Data

Step 3:- Initialize parameters

Step 4:- Train weak classifiers

Step 5:- Make predictions

Step 6:- Evaluate the model

Program:-

```
import pandas as pd
```

```
import numpy as np
```

```
from sklearn.tree import DecisionTreeClassifier
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.metrics import accuracy_score
```

# Step 2: Load and Prepare Data

```
df = pd.DataFrame({
```

```
'X1': [2, 3, 4, 5, 6, 6, 7, 9, 9],
```

```
'X2': [5, 3, 6, 8, 1, 9, 5, 8, 9, 2]
```

```
'label': [1, 1, 0, 1, 0, 1, 0, 0, 0]
```

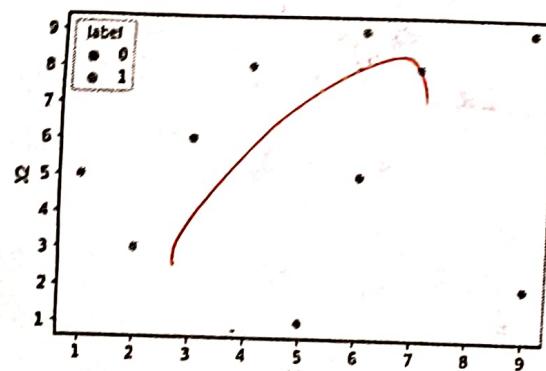
y)

```
X = df[['X1', 'X2']].values
```

```
y = df['label'].values
```

	$x_1$	$x_2$	label	weights
cover	6	6	0	0.1
cover	6	6	0	0.1
cover	0	1	1	0.1
cover	6	6	0	0.1
cover	7	7	1	0.1
cover	5	6	1	0.1
cover	1	2	1	0.1
cover	8	9	0	0.1
cover	4	5	0	0.1
soft	6	6	0	0.1

feature (1) has oppositely sign weights  
feature (2) has same sign weights



	$x_1$	$x_2$	label	weights	
cover	1	2	3	1	0.1
cover	6	6	5	0	0.1
cover	5	6	9	1	0.1
cover	1	2	3	1	0.1
cover	5	6	9	1	0.1
cover	8	9	9	0	0.1
cover	8	9	9	0	0.1
cover	8	9	9	0	0.1
cover	5	6	9	1	0.1
cover	8	9	9	0	0.1

230 300  
2 class, 0.1 step

last experiment of logistic regression in  
gradient descent

	x1	x2	label	weights	y_pred
6	6	5	0	0.1	0
6	6	5	0	0.1	0
0	1	5	1	0.1	1
6	6	5	0	0.1	0
7	7	8	1	0.1	0
5	6	9	1	0.1	0
1	2	3	1	0.1	1
8	9	9	0	0.1	0
4	5	1	0	0.1	0
6	6	5	0	0.1	0

	x1	x2	label	weights	y_pred	updated_weights
6	6	5	0	0.1	0	0.033622
6	6	5	0	0.1	0	0.033622
0	1	5	1	0.1	1	0.033622
6	6	5	0	0.1	0	0.033622
7	7	8	1	0.1	0	0.297427
5	6	9	1	0.1	0	0.297427
1	2	3	1	0.1	1	0.033622
8	9	9	0	0.1	0	0.033622
4	5	1	0	0.1	0	0.033622
6	6	5	0	0.1	0	0.033622

x1	x2	label	weights	beta_0	postulated_weights	current_weights	updated_weights
6	6	5	0.1	0	0.033622	0.033622	0.033622
6	6	5	0.1	0	0.033622	0.033622	0.033622
0	1	5	0.1	0	0.033622	0.033622	0.033622
6	6	5	0.1	0	0.033622	0.033622	0.033622
7	7	8	0.1	0	0.297427	0.297427	0.297427
5	6	9	0.1	0	0.297427	0.297427	0.297427
1	2	3	0.1	0	0.033622	0.033622	0.033622
8	9	9	0.1	0	0.033622	0.033622	0.033622
4	5	1	0.1	0	0.033622	0.033622	0.033622
6	6	5	0.1	0	0.033622	0.033622	0.033622

Logistic regression model

gradient descent

gradient descent

# Convert y into [-1, +1] form for AdaBoost-M1  
y\_transformed = np.where(y == 1, 1, -1)  
x\_transformed test\_size=0.3, random\_state=42)

# Step 3: Initialise parameters

n\_estimators = 5

n\_samples = x\_train.shape[0]

weights = np.ones(n\_samples) / n\_samples

classifiers = []

alphas = []

# Step 4: Train weak classifiers

for i in range(n\_estimators):

clf = DecisionTreeClassifier(max\_depth=1)

clf.fit(x\_train, y\_train, sample\_weight=weights)

y\_pred = clf.predict(x\_train)

# Calculate weighted error

err = np.sum(weights \* (y\_pred != y\_train)) /  
np.sum(weights)

if err == 0:

err = 1e-10

# Compute alpha

alpha = 0.5 \* np.log((1-err)/err)

# Update weights

weights = weights \* np.exp(-alpha \* y\_train \* y\_pred)

weights /= np.sum(weights) # normalise

classifiers.append(clf)

alphas.append(alpha)

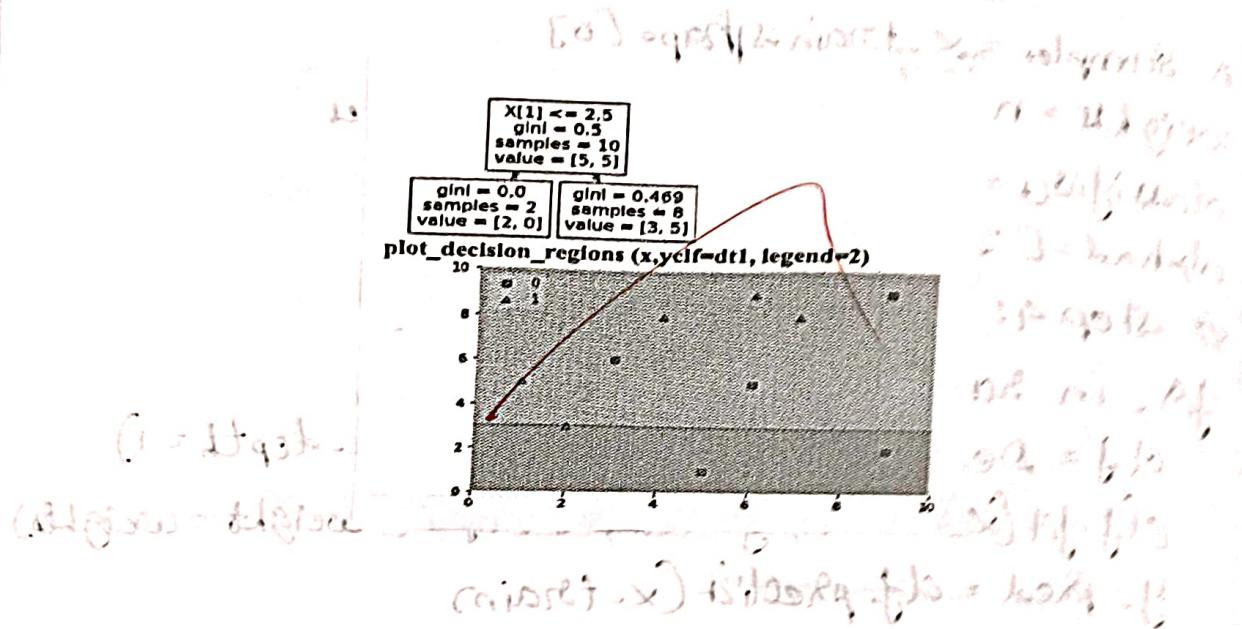
# Step 5: Make prediction

def predict(x):

final\_shape = np.zeros(x.shape[0])

for clf, alpha in zip(classifiers, alphas):

1.5.2.  $\text{dt} = \text{DecisionTreeClassifier}(criterion='gini', max_depth=2)$   
 $\text{dt}.fit(\text{x}, \text{y})$   
 Text(0.5 \* 73, 0.5 \* 73, 'x[0] == 3.5) & gini < 0.469 & samples < 10, value=[5, 5],  
 Text(0.25 \* 73, 0.25 \* 73, 'gini > 0.469 & samples < 10, value=[5, 5],  
 Text(0.75 \* 73, 0.25, 'gini < 0.375 & samples < 10, value=[10, 21, 5])



$(1 - \text{dt}_1.\text{value}[0][0]) * \text{dt}_1.\text{value}[0][1] = 0.25$   
 $(\text{dt}_1.\text{value}[1][0]) * \text{dt}_1.\text{value}[1][1] = 0.75$

$(0.25)(0.25 - 0.75) + (0.75)(0.75 - 0.25) = 0.25$   
 $(0.25 - 0.75)^2 + (0.75 - 0.25)^2 = 0.25$   
 $\sqrt{0.25} = 0.5$

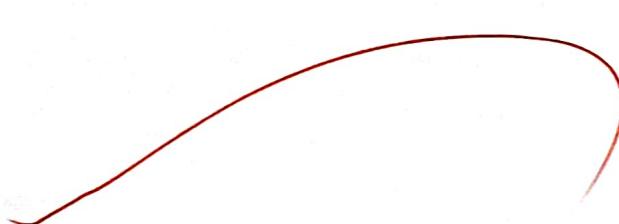
$(0.25)(0.25 - 0.75) + (0.75)(0.75 - 0.25) = 0.25$   
 $(0.25 - 0.75)^2 + (0.75 - 0.25)^2 = 0.25$   
 $\sqrt{0.25} = 0.5$

$(0.25)(0.25 - 0.75) + (0.75)(0.75 - 0.25) = 0.25$   
 $(0.25 - 0.75)^2 + (0.75 - 0.25)^2 = 0.25$   
 $\sqrt{0.25} = 0.5$

```
preds = clf.predict(x)
final_score = alpha * preds
detecn = np.sign(final_score)
y_pred_test = predict(x-test)

# Step 6: Evaluate Model
acc = accuracy_score(y-test, y_pred-test)

# Step 7: Output Results
print("Final Accuracy on Test Set:", acc)
print("Classifier weights (alpha):", alpha)
```

  
Result:-  
Thus the python program to implement  
~~Adaboosting~~ has been successfully  
executed.