

# 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
```

```
import pandas as pd
```

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

Step 2: Load and prepare Data

```
Load your dataset using pd.read_csv() (eg. df =  
pd.read_csv('data.csv'))
```

Separate features (x) and target (y).

Step 3: Initialize parameters:

Set the number of weak classifiers.

$n$  - estimators. Initialize an array weights from for instance weights.

Step 4: Train weak classifiers.

Loop for  $n$  estimators iterations. Predict the target values using the trained weak classifier.

Step 6: make predictions

Step 7: Evaluate the model

Compute the accuracy of the adaboost model on the testing set using accuracy score().

Step 7: output Results

Print or plot the final accuracy and possibly other evaluation metrics.

Program:

```
import numpy as np
```

```
import pandas as pd
```

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

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

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

```
X = df[['x1', 'x2']].values
```

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

```
n_estimators = 5
```

```
n_samples = X_train.shape[0]
```

```
weights = np.ones(n_samples) / n_samples
```

```
classifiers = []
```

```
alpha = []
```



$\text{error} = 0;$

$\text{err} = 1e-10$

$\text{alpha} = 0.6 * \text{np} \cdot \log(1 - \text{err} / \text{err})$

$\text{weights} = \text{weights} * \text{np} \cdot \exp(1 - \text{alpha} * (y_{\text{train}} - y_{\text{pred}}))$

$\text{weights} = 1 / \text{np} \cdot \text{sum}(\text{weights})$

$\text{clf} = \text{clf} \cdot \text{append}(\text{clf})$

$\text{alphas} \cdot \text{append}(\text{alpha})$

$\text{def predict}(x):$

$\text{final\_score} = \text{np} \cdot \text{zeros}(x \cdot \text{shape}[0])$

$\text{for } \text{clf}, \text{alpha} \text{ in zip}(\text{classifiers}, \text{alphas}):$

$\text{pred} = \text{clf} \cdot \text{predict}(x)$

$\text{final\_score} += \text{alpha} * \text{preds}$

$\text{return np} \cdot \text{sign}(\text{final\_score})$

$y_{\text{pred\_test}} = \text{predict}(x_{\text{test}})$

$\text{print}(\text{" Final accuracy on test set: ", acc})$

$\text{print}(\text{" classifier weights alphas: "}, \text{alpha})$

Result:

Thus the python program to implement  
AdaBoosting has been executed successfully.  
and the results have been verified and  
analyzed.

# A python program to implement Gradient Boosting

Aim:

To implement a python program using the gradient boosting model.

Algorithm:

Step 1: Import necessary libraries

```
import numpy as np
```

```
import pandas as pd
```

```
import decision tree regressor from sklearn.  
tree
```

Step 2: Prepare the data

Load your dataset into a dataframe using  
`pd.read_csv("your-dataset.csv")`. Split the  
dataset into features ( $X$ ) and target ( $y$ ).

Step 3: Initialize parameters.

set the number of boosting rounds set  
the weak learners.

Step 4: Initialize the base model.

Compute the initial prediction as the  
mean of the target values. Initialize the  
Predictions to the base model's  
Prediction.



Step 5: Iterate over boosting rounds.

For each boosting round, fit a decision tree to the pseudo-residuals: Append the fitted tree and learning rate to their respective lists.

Step 6: Make predictions on test data

Initialize the test predictions with the base model's predictions.

Step 7: Evaluate the model

Compute the mean squared error on the training data. Compute the mean-squared error on the test data.

Program:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
np.random.seed(42)

df = pd.DataFrame()
df['x'] = x.reshape(100)
df['res1'] = df['y'] - df['pred1']
plt.scatter(df['x'], df['y'])
plt.plot(df['x'], df['pred1'], color='red')
```

```
from sklearn.tree import DecisionTreeRegressor  
from sklearn.metrics import mean_squared_error, r2_score  
X_train, X_test, y_train, y_test = train_test_split(X, y,  
                                                    test_size=0.2,  
                                                    random_state=42)
```

```
tree = DecisionTreeRegressor()
```

```
tree.fit(X_train, y_train)
```

```
np.random.seed(42)
```

```
X = np.random.rand(100, 1) - 0.5
```

```
y = 3 * X[:, 0] * X[:, 0] + 2 + 0.05 * np.random.randn(100)
```

```
g = GradientBoostingRegressor(n_estimators=100)
```

Result:

Thus the python program to implement gradient boosting for the standard uniform distribution has been successfully implemented and the results have been verified and analyzed successfully.