6.DECISON TREE:

- Data scientist Interview preparation
- Practical
- June2024

import pandas as pd

boston

print(boston.feature names)

• II.Decison Tree Regressor

Supervised ml :classifcation & regression task

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

##Boston House Pricing Dataset :sklearn # dont run this cell
from sklearn.datasets import load_boston
boston_df=load_boston()
#Note: The Boston housing dataset has been deprecated in scikit-lear
#ethical concerns and is no longer available in the latest version

from sklearn.datasets import fetch_openml

# Load the Boston housing dataset from OpenML
boston = fetch_openml(data_id=531)
```

['CRIM', 'ZN', 'INDUS', 'CHAS', 'NOX', 'RM', 'AGE', 'DIS', 'RAD'

print(boston.data.shape)
print(boston.target.shape)

(506, 13) (506,)

Extract features and target

X = boston.data

y = boston.target

Χ

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	29
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	24
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	24
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	22
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	22
501	0.06263	0.0	11.93	0	0.573	6.593	69.1	2.4786	1	2
502	0.04527	0.0	11.93	0	0.573	6.120	76.7	2.2875	1	2
503	0.06076	0.0	11.93	0	0.573	6.976	91.0	2.1675	1	27
504	0.10959	0.0	11.93	0	0.573	6.794	89.3	2.3889	1	27
505	0.04741	0.0	11.93	0	0.573	6.030	80.8	2.5050	1	2
506 rc	ws × 13 co	olumns								
4										•

У

9 24.0 1 21.6 2 34.7

```
3 33.4

4 36.2

...

501 22.4

502 20.6

503 23.9

504 22.0

505 11.9

Name: MEDV, Length: 506, dtype: float64
```

```
#or you canwrite #independent features
X=pd.DataFrame(boston_df.data,columns=boston_df.feature_names)
#dependent features
y=boston_df.target
```

```
### train test split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.20, random_state=42)
```

II.DECISION TREE REGRESSOR:

- Post pruning
- Pre pruning

```
from sklearn.tree import DecisionTreeRegressor
regressor=DecisionTreeRegressor()
regressor.fit(X_train,y_train)
```

```
→ DecisionTreeRegressor

DecisionTreeRegressor()
```

```
from sklearn.tree import plot_tree
plt.figure(figsize=(15, 13))
plot_tree(regressor, filled=True, feature_names=boston.feature_names)
plt.title("Decision Tree Regressor - Boston Housing Dataset")
plt.show()
```



Decision Tree Regressor - Boston Housing Dataset

```
regressor.fit(X_train,y_train)
     ▼ DecisionTreeRegressor
     DecisionTreeRegressor()
y_pred=regressor.predict(X_test)
from sklearn.metrics import r2 score
score=r2_score(y_pred,y_test)
score
→ 0.7125309163545431
# Hyperparameter Tunning
parameter={
 'criterion':['squared error','friedman mse','absolute error','poiss
  'splitter':['best','random'],
  'max_depth':[1,2,3,4,5,6,7,8,10,11,12],
  'max_features':['auto', 'sqrt', 'log2']
}
regressor=DecisionTreeRegressor()
from sklearn.model selection import GridSearchCV
regressorcv=GridSearchCV(regressor,param_grid=parameter,cv=5,scoring
import warnings
warnings.filterwarnings('ignore')
```

regressorcv.fit(X_train,y_train)



```
► GridSearchCV
► estimator: DecisionTreeRegressor

► DecisionTreeRegressor
```

```
regressorcv.best_params_

{'criterion': 'poisson',
    'max_depth': 5,
    'max_features': 'auto',
    'splitter': 'best'}

y_pred=regressorcv.predict(X_test)

r2_score(y_pred,y_test)

$\infty$ 0.6906700687236822
```

Post pruning

 post pruning :cutting down the specific nodes after building the Decision Tree

```
#post pruning :cutting down the specific nodes after building the De
```

II.Pre pruning:

- It is used before the construction of decision tree wih hyperparameter tuning and crsoss validation
- criterion: Determines how the quality of a split is measured.
- splitter: Determines the strategy used to choose the split at each node.