## DECISION TREE

REGRESSOR

FROM

SCRATCH



```
import numpy as np
class Node:
    def __init__(self,features=None,threshold=None,left=None,right=None,*,value=None):
        self.features=features
        self.threshold=threshold
        self.left=left
        self.right=right
        self.value=value
    def is_leaf_node(self):
        return self.value is not None
```

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```
class DecisionTreeRegressor:
    def __init__(self,max_depth=100,min_sample_split=2,n_features=None) -> None:
        self.root=None
        self.max_depth=max_depth
        self.min_sample_split=min_sample_split
        self.n_features=n_features
        self.n_features=x.shape[1] if not self.n_features else min(self.n_features,X.shape[1])
        self.n_features=X.shape[1] if not self.n_features else min(self.n_features,X.shape[1])
        self.root=self._Tree(X,y)

def _split(self,X_column,threshold):
        left_idxs=np.where(X_column<=threshold)[0]
        right_idxs=np.where(X_column>threshold)[0]
        return left_idxs,right_idxs
```

```
def _Tree(self,X,y,depth=0):
    samples,features=X.shape
    if (depth>=self.max_depth or samples<self.min_sample_split):
        return Node(value=np.mean(y))
    feat_idxs=np.random.choice(self.n_features,features)
    best_feat,best_threshold=self._best_split(X,y,feat_idxs)

if best_threshold!=None:
    left_idxs,right_idxs=self._split(X[:,best_feat],best_threshold)

    left=self._Tree(X[left_idxs],y[left_idxs],depth+1)
    right=self._Tree(X[right_idxs],y[right_idxs],depth+1)
    if depth>self.d:
        self.d=depth
    return Node(best_feat,best_threshold,left,right)
```

```
def _best_split(self,X,y,feat_idxs):
    best_var=-1
    split_idx,split_thres=None,None
    for feature in feat_idxs:
        thresholds=(np.unique(X[:,feature]))

        for threshold in thresholds:
            if threshold!=None:
                var=self._variance_reduction(X[:,feature],y,threshold)

            if var >best_var:
                best_var=var
                split_idx,split_thres=feature,threshold
        return split_idx,split_thres
```

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```
def _variance_reduction(self,X,y,threshold):
    parent_var=np.var(y)
    left_idxs,right_idxs=self._split(X,threshold)
    if len(left_idxs)==0 and len(right_idxs)==0:
        return 0
    n=len(y)
    n_l,n_r=len(y[left_idxs]),len(y[right_idxs])
    left_var,right_var=np.var(y[left_idxs]),np.var(y[right_idxs])
    child_var=((n_l/n)*left_var)+((n_r/n)*right_var)
    return parent_var-child_var
```

```
from sklearn.datasets import load_boston
dataset=load_boston()
X=dataset['data']
Y=dataset['target']
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,Y,test_size=0.1,random_state=1234)
```

from DecisionTreeRegressor import DecisionTreeRegressor
DR=DecisionTreeRegressor(max\_depth=15)
DR.fit(X\_train,y\_train)
y\_pred=DR.predict(X\_test)
from sklearn.metrics import r2\_score
print(r2\_score(y\_test,y\_pred))

#output:
0.8311416



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