Practical implementation of Decision Tree Algorithm

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Importing required libraries

In [62]:
```
## Importing Necessary Library
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score
from sklearn.model_selection import GridSearchCV

import warnings
warnings.filterwarnings('ignore')

```

Importing CSV File

In [4]:
```
## Importing CSV File
df = pd.read_csv("https://raw.githubusercontent.com/shrikant-temburwar/Wine-
```

In [5]:
```
## Checking Top 5 Rows
df.head()
```

Out[5]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	pH	sulphates	alcohol
0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4
1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	0.68	9.8
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	9.8
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	9.8
4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4

In [6]:
```
## Checking Bottom 5 Rows
df.tail()
```

Out[6]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	pH	sulphates	alcoh
1594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	10
1595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	11
1596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	11
1597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	0.71	10
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	0.66	11

In [7]:
```
## Checking Shape of the Dataset, here there are total 1599 Rows and 12 Colu
df.shape
```

Out[7]:
```
(1599, 12)
```

In [8]:
```
## Checking Target Column Unique Details.
df['quality'].unique()
```

Out[8]:
```
array([5, 6, 7, 4, 8, 3], dtype=int64)
```

In [10]:
```
## Checking Length of target Columns Unique Details.
len(df['quality'].unique())
```

Out[10]:
```
6
```

In [9]:
```
## Checking Total Count of Target Columns Unique Details
df['quality'].value_counts()

```

Out[9]:
```
5    681
6    638
7    199
4     53
8     18
3     10
Name: quality, dtype: int64
```

In [11]:
```
## Checking Basic Statistics Analysis via Transpose
df.describe().T
```

Out[11]:

	count	mean	std	min	25%	50%	75%	max
fixed acidity	1599.0	8.319637	1.741096	4.60000	7.1000	7.90000	9.200000	15.90000
volatile acidity	1599.0	0.527821	0.179060	0.12000	0.3900	0.52000	0.640000	1.58000
citric acid	1599.0	0.270976	0.194801	0.00000	0.0900	0.26000	0.420000	1.00000
residual sugar	1599.0	2.538806	1.409928	0.90000	1.9000	2.20000	2.600000	15.50000
chlorides	1599.0	0.087467	0.047065	0.01200	0.0700	0.07900	0.090000	0.61100
free sulfur dioxide	1599.0	15.874922	10.460157	1.00000	7.0000	14.00000	21.000000	72.00000
total sulfur dioxide	1599.0	46.467792	32.895324	6.00000	22.0000	38.00000	62.000000	289.00000
density	1599.0	0.996747	0.001887	0.99007	0.9956	0.99675	0.997835	1.00369
pH	1599.0	3.311113	0.154386	2.74000	3.2100	3.31000	3.400000	4.01000
sulphates	1599.0	0.658149	0.169507	0.33000	0.5500	0.62000	0.730000	2.00000
alcohol	1599.0	10.422983	1.065668	8.40000	9.5000	10.20000	11.100000	14.90000
quality	1599.0	5.636023	0.807569	3.00000	5.0000	6.00000	6.000000	8.00000

In [12]:
```
## Checking Duplicate Rows in Dataset and there are total 240 Duplicate rows
df.duplicated().sum()
```

Out[12]: 240

In [13]:
```
## Droping Duplicate Rows from dataset.
df = df.drop_duplicates()
```

In [14]:
```
## Again Checking Duplicate Rows in dataset, Now Dupliacte Row is Zero after
df.duplicated().sum()
```

Out[14]: 0

Creating Independent Feature

In [16]:
```
X = df.drop("quality", axis =1)
```

Creating Dependent Feature

In [17]:
```
y = df['quality']
```

Model Building :-

In [18]:
```
from sklearn.model_selection import train_test_split,GridSearchCV
```

In [36]:
```
## Spliting the Dataset
X_train,X_test,y_train,y_test = train_test_split(X, y, test_size=0.3, random

```

In [37]:
```
## Scaling is not Required in Decision tree Problem

'''
from sklearn.preprocessing import StandardScaler
sclr = StandardScaler()
sclr.fit()
sclr.transform()
'''
```

Out[37]:
```
'\nfrom sklearn.preprocessing import StandardScaler\nsclr = StandardScaler()\ns
clr.fit()\nsclr.transform()\n'
```

Model 1 :- Decision Tree classifier Algorithm

In [38]:
```
from sklearn.tree import DecisionTreeClassifier
model  = DecisionTreeClassifier()
```

In [39]:
```
## Model Fitting with Decision Tree Classifier on Training dataset
model.fit(X_train,y_train)
```

Out[39]:
```
DecisionTreeClassifier

DecisionTreeClassifier()
```

In [40]:
```
model.score(X_train,y_train)
```

Out[40]:
```
1.0
```

In [41]:
```
## Model Prediction on Testing Dataset
y_predict = model.predict(X_test)
```

In [42]:
```
from sklearn.metrics import accuracy_score
```

In [43]:
```
## Accuracy of the Model using Decision Tree Classifier Algorithm
accuracy_score(y_test,y_predict)

```

Out[43]: 0.5073529411764706

Observation

- By using Decision Tree Classifier ,we get Approx 51% Accuracy.

Model 2:- Logistic Regression Algorithm

In [58]:
```
## Importing Logistic Regression Algorithm
from sklearn.linear_model import LogisticRegression

```

In [59]:
```
LR = LogisticRegression()
```

In [60]:
```
LR
```

Out[60]:
```
LogisticRegression
LogisticRegression()
```

In [63]:
```
## Fitting the Training Data in Logistic Regression
LR.fit(X_train,y_train)

```

Out[63]:
```
LogisticRegression
LogisticRegression()
```

In [65]:
```
## Predicting Testing data Value using Logistic regreession
y_predict1 = LR.predict(X_test)
```

In [66]:
```
## Checking Accuracy score using Logistic Regression
accuracy_score(y_test,y_predict1)
```

Out[66]: 0.5784313725490197

Observation

- By using Logistic Regression ,we get Approx 57% Accuracy.

Model 3:- SVC (Support Vector Classifier) Algorithm.

In [67]:
```
## Importing SVC
from sklearn.svm import SVC
```

In [68]:
```
svc = SVC()
```

In [69]:
```
svc
```

Out[69]:
```
SVC()
```

In [70]:
```
## Fitting the Training Data into SVC Model
svc.fit(X_train,y_train)

```

Out[70]:
```
SVC()
```

In [73]:
```
## Predicting Testing data Value using SVC Model
y_predict2 = svc.predict(X_test)

```

In [74]:
```
## Checking Accuracy score using Logistic Regression
accuracy_score(y_test,y_predict2)
```

Out[74]:
```
0.5049019607843137
```

Observation

- By using SVC Model ,we get Approx 50% Accuracy.

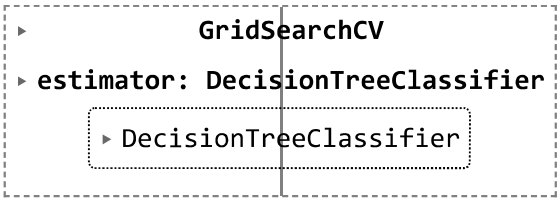
Applying GridSearchCV Hyperparameter tuning

In [46]:
```
## Creating Parameter for GridSearch CV

grid_param = {
    'criterion': ['gini', 'entropy'],
    'max_depth' : range(2,32,1),
    'min_samples_leaf' : range(1,10,1),
    'min_samples_split': range(2,10,1),
    'splitter' : ['best', 'random']

}
```

In [47]:
```
## Importing GridSearchCV Library

from sklearn.model_selection import GridSearchCV
grid_search = GridSearchCV(estimator= model, param_grid = grid_param, cv = 5

```

In [49]:
```
## Fitting our Training Data into GridSearch.
grid_search.fit(X_train,y_train)    ## GridSearchCV take a lot of time to RU

```

Out[49]:



In [50]:
```
## Finding Best param Model using GridSearch
grid_search.best_params_

```

Out[50]:
```
{'criterion': 'gini',
 'max_depth': 4,
 'min_samples_leaf': 6,
 'min_samples_split': 7,
 'splitter': 'random'}
```

Observation

- In GridSearch CV , we found that Best Parameter are as:-
  - a) criterion = 'gini',
  - b) max_depth = 4,
  - c) min_samples_leaf = 6,
  - d) min_samples_split = 7,
  - e) splitter = 'random'

In [52]:
```
## Applying Best Param with our model (Decision Tree)
model_with_best_params = DecisionTreeClassifier( criterion = 'gini',
                                                 max_depth = 4,
                                                 min_samples_leaf = 6,
                                                 min_samples_split = 7,
                                                 splitter = 'random')

```

In [53]:
```
## Fitting Best Param Model to Training Dataset.
model_with_best_params.fit(X_train,y_train)

```

Out[53]:
```
DecisionTreeClassifier(max_depth=4, min_samples_leaf=6, min_samples_split=7,
                       splitter='random')
```

In [54]:
```
## Predictin the Value
y_prediction2 = model_with_best_params.predict(X_test)

```

In [55]:
```
## Checking Accuracy of Model After Doing HyperParameter Tuning using GridSe
accuracy_score(y_test, y_prediction2)

```

Out[55]:
```
0.571078431372549
```

Observation

- After Doing Hyperparameter Tuning with GridSearch CV, we get Approx. 57% Accuracy.

Thank You