## **Decision Tree**

```
In [7]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.ensemble import RandomForestRegressor
         from sklearn.preprocessing import OneHotEncoder
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import mean_squared_error
         from joblib import dump, load
In [8]:
         data = pd.read_csv(r"D:\New folder\dowenload\physics\machine learning\housing.csv
In [9]:
         data
Out[9]:
                 longitude
                           latitude housing_median_age total_rooms total_bedrooms population househ
              0
                   -122.23
                             37.88
                                                  41.0
                                                             0.088
                                                                             129.0
                                                                                        322.0
              1
                   -122.22
                             37.86
                                                  21.0
                                                            7099.0
                                                                            1106.0
                                                                                       2401.0
                                                                                                   11
              2
                   -122.24
                                                  52.0
                             37.85
                                                            1467.0
                                                                             190.0
                                                                                        496.0
              3
                   -122.25
                             37.85
                                                  52.0
                                                            1274.0
                                                                             235.0
                                                                                        558.0
                                                                                                    2
              4
                   -122.25
                             37.85
                                                  52.0
                                                            1627.0
                                                                             280.0
                                                                                        565.0
                                                                                                    2
                                                                                                    3
          20635
                   -121.09
                             39.48
                                                  25.0
                                                            1665.0
                                                                             374.0
                                                                                        845.0
                   -121.21
          20636
                             39.49
                                                  18.0
                                                             697.0
                                                                             150.0
                                                                                        356.0
          20637
                   -121.22
                             39.43
                                                  17.0
                                                            2254.0
                                                                             485.0
                                                                                       1007.0
                   -121.32
                             39.43
                                                            1860.0
                                                                             409.0
          20638
                                                  18.0
                                                                                       741.0
```

270F 0

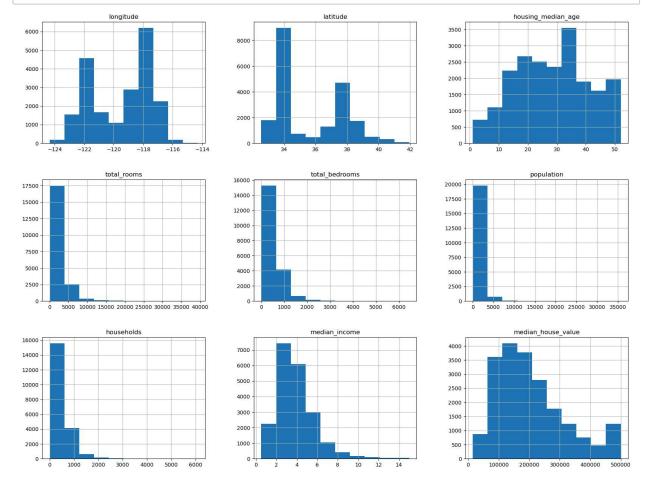
040 0

4007 A

104 04

20 27

## In [10]: data.hist(figsize=(20,15)) plt.show()



```
In [11]: data_cat=data[["ocean_proximity"]]
    encoder = OneHotEncoder()
    encoded_data = encoder.fit_transform(data_cat)
```

## In [ ]:

```
In [14]: encoded_feature_names = encoder.get_feature_names_out(["ocean_proximity"])
        encoded_df = pd.DataFrame(encoded_data.toarray(), columns=encoded_feature_names)
        data_encoded = pd.concat([data, encoded_df], axis=1)
        data_encoded.drop("ocean_proximity", axis=1, inplace=True)
```

```
data_encoded.isnull().sum()
In [15]:
Out[15]: longitude
                                          0
         latitude
                                          0
                                          0
         housing_median_age
         total_rooms
                                          0
         total bedrooms
                                        207
         population
                                          0
         households
                                          0
         median_income
                                          0
         median_house_value
                                          0
         ocean_proximity_<1H OCEAN
                                          0
         ocean_proximity_INLAND
                                          0
         ocean proximity ISLAND
                                          0
         ocean_proximity_NEAR BAY
                                          0
         ocean_proximity_NEAR OCEAN
                                          0
         dtype: int64
         median=data_encoded["total_bedrooms"].median()
In [26]:
         data_encoded["total_bedrooms"].fillna(median,inplace=True)
In [17]: | data_encoded.duplicated().sum()
Out[17]: 0
         X=data encoded.drop('median house value', axis=1)
         Y=data_encoded["median_house_value"]
In [19]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2,random_st
         model = RandomForestRegressor()
In [20]:
In [21]:
         model.fit(X train,Y train)
Out[21]:
          ▼ RandomForestRegressor
          RandomForestRegressor()
In [22]: Y_pred = model.predict(X_test)
In [23]:
         mse = mean_squared_error(Y_test, Y_pred)
         rmse=np.sqrt(mse)
         print("Root Mean Squared Error:",rmse)
```

Root Mean Squared Error: 48692.588455430385

In [28]:	<pre>dump(model, r"C:\Users\abohamam\Desktop\model.h5")</pre>
Out[28]:	['D:\\New folder\\dowenload\\physics\\machine learning\\New Text Document.h5']
In [ ]:	
In [ ]:	
In [ ]:	