

```
In [1]: #importing necessary libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: #loading dataset

df = pd.read_csv('data/houseprice.csv')

df.head()
```

```
Out[2]:
```

	bath	balcony	price	total_sqft_int	bhk	price_per_sqft	area_typeSuper built-up Area	area_typeBuilt- up Area	area_typePlot Area	availal
0	3.0	2.0	150.0	1672.0	3	8971.291866	1	0	0	
1	3.0	3.0	149.0	1750.0	3	8514.285714	0	1	0	
2	3.0	2.0	150.0	1750.0	3	8571.428571	1	0	0	
3	2.0	2.0	40.0	1250.0	2	3200.000000	1	0	0	
4	2.0	2.0	83.0	1200.0	2	6916.666667	0	0	1	

5 rows × 108 columns

```
In [3]: df.shape
```

```
Out[3]: (7120, 108)
```

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7120 entries, 0 to 7119
Columns: 108 entries, bath to location_Tumkur Road
dtypes: float64(5), int64(103)
memory usage: 5.9 MB
```

```
In [5]: df.columns
```

```
Out[5]: Index(['bath', 'balcony', 'price', 'total_sqft_int', 'bhk', 'price_per_sqft',
      'area_typeSuper built-up Area', 'area_typeBuilt-up Area',
      'area_typePlot Area', 'availability_Road', 'availability_Road',
      ...,
      'location_Kalena Agrahara', 'location_Horamavu Agara',
      'location_Vidyaranyapura', 'location_BTMT 2nd Stage',
      'location_Hebbal Kempapura', 'location_Hosur Road',
      'location_Horamavu Banaswadi', 'location_Domlur',
      'location_Mahadevpura', 'location_Tumkur Road'],
      dtype='object', length=108)
```

```
In [6]: #seperating features and target
X = df.drop('price', axis=1)
y = df['price']
```

```
In [7]: #splitting dataset into training set and test set
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=1)
```

```
In [8]: #training model
from sklearn.tree import DecisionTreeRegressor

dct = DecisionTreeRegressor()
dct.fit(X_train, y_train)
```

```
Out[8]: ▾ DecisionTreeRegressor
DecisionTreeRegressor()
```

```
In [28]: dct.get_depth()
```

```
Out[28]: 22
```

```
In [9]: #model evaluation
y_pred = dct.predict(X_test)
```

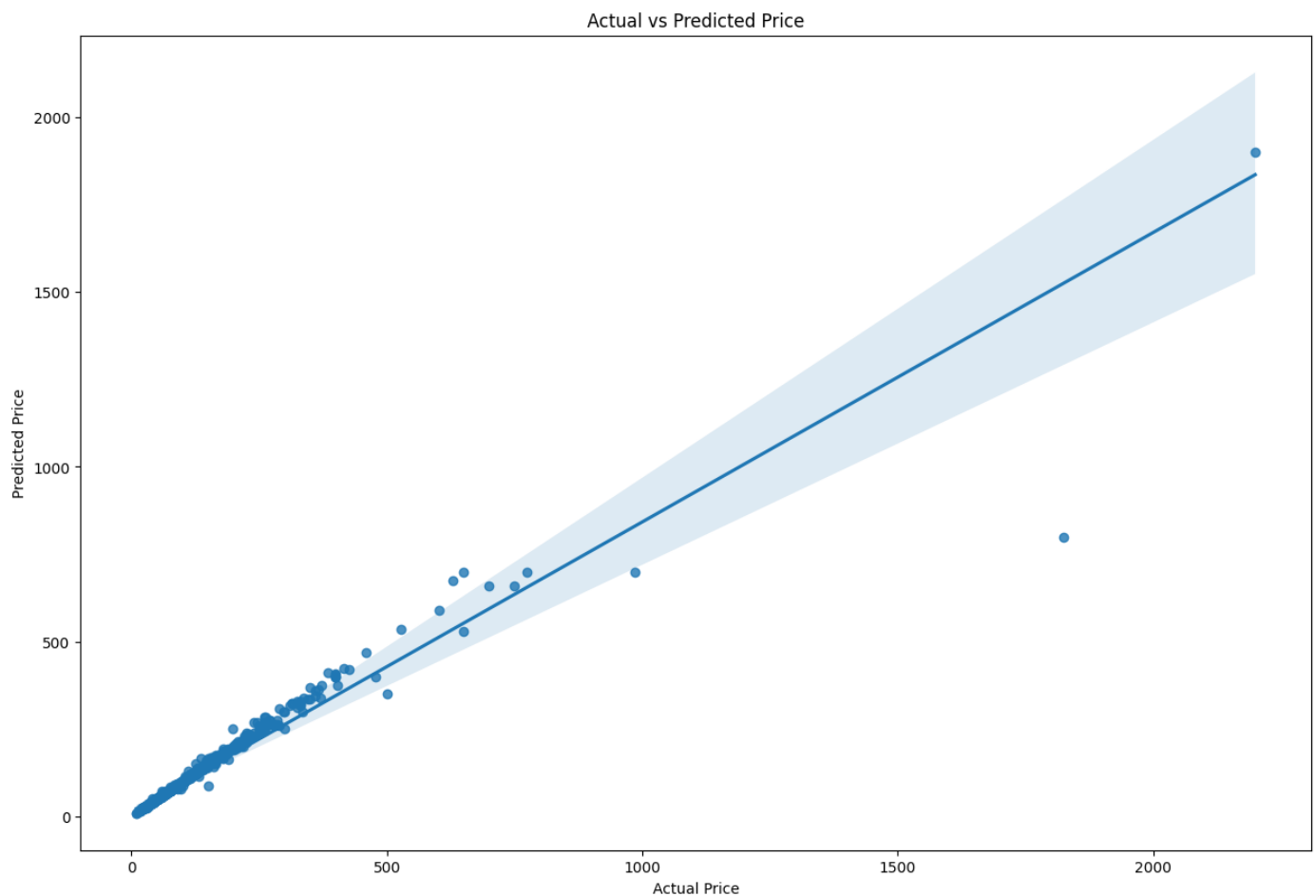
```
In [10]: from sklearn.metrics import r2_score

r2_score(y_test, y_pred)
```

```
Out[10]: 0.9227936812841927
```

```
In [14]: #plotting predicted values vs actual values
```

```
plt.figure(figsize=(15,10))
sns.regplot(x=y_test, y=y_pred)
plt.xlabel("Actual Price")
plt.ylabel('Predicted Price')
plt.title("Actual vs Predicted Price")
plt.show()
```



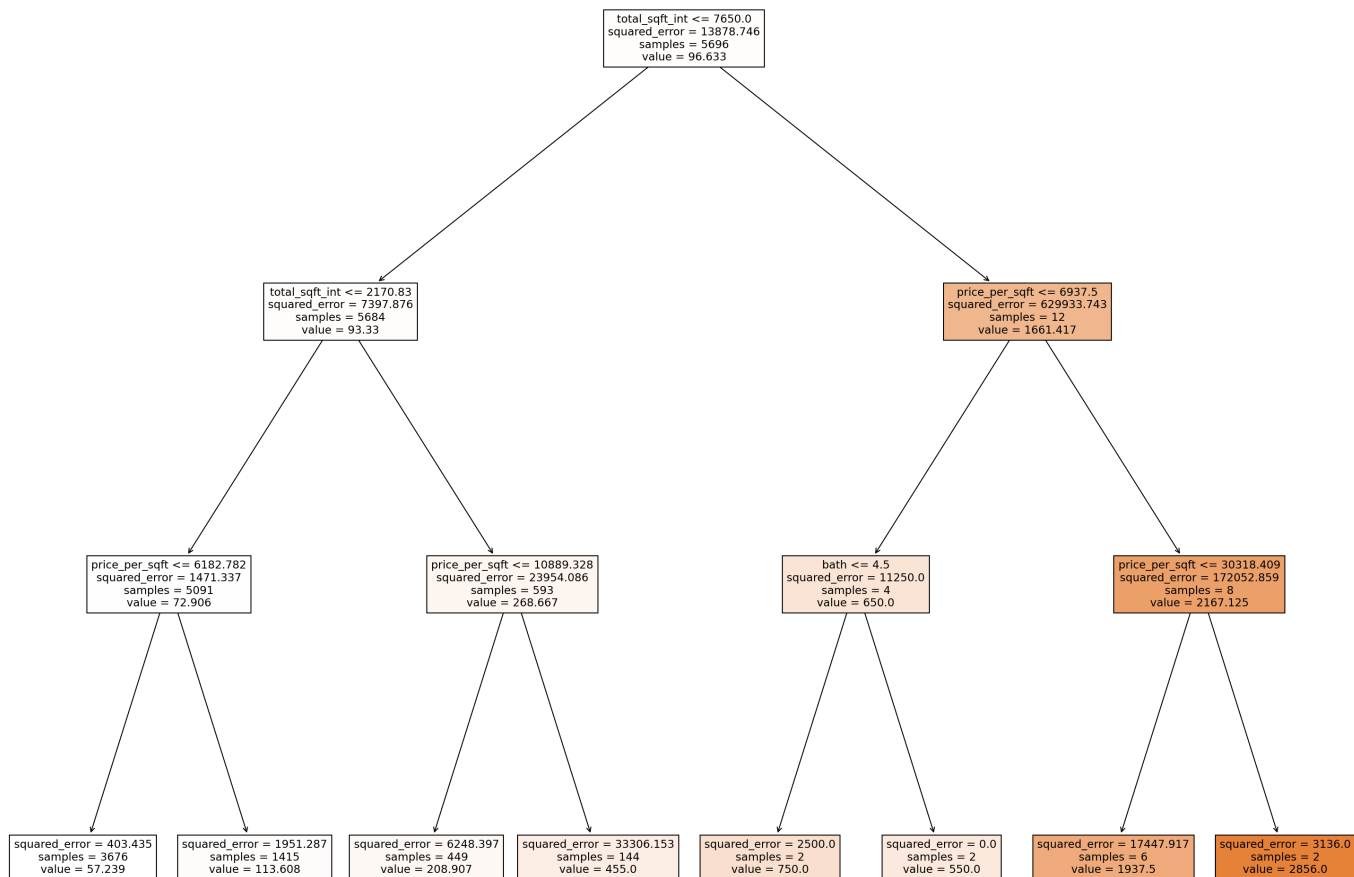
```
In [26]: #visualizing decision tree with max_depth = 3
regressor = DecisionTreeRegressor(max_depth=3, random_state=1)
regressor.fit(X_train, y_train)
y_pred = regressor.predict(X_test)

r2_score(y_test, y_pred)
```

Out[26]: 0.7624851719489383

```
In [27]: from sklearn.tree import plot_tree
plt.figure(figsize=(25,20), dpi=150)
plot_tree(regressor, feature_names=X_train.columns, filled=True)
```

Out[27]: [Text(0.5, 0.875, 'total\_sqft\_int <= 7650.0\nsquared\_error = 13878.746\nsamples = 5696\nvalue = 96.633'),  
Text(0.25, 0.625, 'total\_sqft\_int <= 2170.83\nsquared\_error = 7397.876\nsamples = 5684\nvalue = 93.33'),  
Text(0.125, 0.375, 'price\_per\_sqft <= 6182.782\nsquared\_error = 1471.337\nsamples = 5091\nvalue = 72.906'),  
Text(0.0625, 0.125, 'squared\_error = 403.435\nsamples = 3676\nvalue = 57.239'),  
Text(0.1875, 0.125, 'squared\_error = 1951.287\nsamples = 1415\nvalue = 113.608'),  
Text(0.375, 0.375, 'price\_per\_sqft <= 10889.328\nsquared\_error = 23954.086\nsamples = 593\nvalue = 268.667'),  
Text(0.3125, 0.125, 'squared\_error = 6248.397\nsamples = 449\nvalue = 208.907'),  
Text(0.4375, 0.125, 'squared\_error = 33306.153\nsamples = 144\nvalue = 455.0'),  
Text(0.75, 0.625, 'price\_per\_sqft <= 6937.5\nsquared\_error = 629933.743\nsamples = 12\nvalue = 1661.417'),  
Text(0.625, 0.375, 'bath <= 4.5\nsquared\_error = 11250.0\nsamples = 4\nvalue = 650.0'),  
Text(0.5625, 0.125, 'squared\_error = 2500.0\nsamples = 2\nvalue = 750.0'),  
Text(0.6875, 0.125, 'squared\_error = 0.0\nsamples = 2\nvalue = 550.0'),  
Text(0.875, 0.375, 'price\_per\_sqft <= 30318.409\nsquared\_error = 172052.859\nsamples = 8\nvalue = 2167.125'),  
Text(0.8125, 0.125, 'squared\_error = 17447.917\nsamples = 6\nvalue = 1937.5'),  
Text(0.9375, 0.125, 'squared\_error = 3136.0\nsamples = 2\nvalue = 2856.0')]



In [ ]: !jupyter nbconvert --to webpdf --allow-chromium-download fuel\_efficiency\_prediction.ipyn