

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
from matplotlib import pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error
from sklearn.tree import DecisionTreeRegressor
from sklearn.preprocessing import StandardScaler
```

```
In [2]: data = pd.read_csv("Housing.csv")
```

```
In [3]: data.head()
```

```
Out[3]:
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterhea
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	

```
In [4]: data.shape
```

```
Out[4]: (545, 13)
```

```
In [5]: data.isna().sum()
```

```
Out[5]: price          0
area          0
bedrooms      0
bathrooms     0
stories       0
mainroad      0
guestroom     0
basement      0
hotwaterheating 0
airconditioning 0
parking       0
prefarea      0
furnishingstatus 0
dtype: int64
```

```
In [6]: data.describe()
```

Out[6]:

	price	area	bedrooms	bathrooms	stories	parking
count	5.450000e+02	545.000000	545.000000	545.000000	545.000000	545.000000
mean	4.766729e+06	5150.541284	2.965138	1.286239	1.805505	0.693578
std	1.870440e+06	2170.141023	0.738064	0.502470	0.867492	0.861586
min	1.750000e+06	1650.000000	1.000000	1.000000	1.000000	0.000000
25%	3.430000e+06	3600.000000	2.000000	1.000000	1.000000	0.000000
50%	4.340000e+06	4600.000000	3.000000	1.000000	2.000000	0.000000
75%	5.740000e+06	6360.000000	3.000000	2.000000	2.000000	1.000000
max	1.330000e+07	16200.000000	6.000000	4.000000	4.000000	3.000000

In [7]: data.head()

Out[7]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheat
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	

In [8]: data["mainroad"].value_counts()

Out[8]:

```
mainroad
yes      468
no        77
Name: count, dtype: int64
```

In [9]: Main_road = pd.get_dummies(data["mainroad"], drop_first = True)

In [10]: Main_road = Main_road.astype(int)

In [11]: data = pd.concat([Main_road, data], axis = 1)

In [12]: data = data.rename(columns = {"yes" : "main_road"})

```
In [13]: guest_room = pd.get_dummies(data["guestroom"], drop_first = True)
guest_room = guest_room.astype(int)
data = pd.concat([guest_room, data], axis = 1)
```

In [14]: data = data.rename(columns = {"yes" : "guest_room"})

```
In [15]: base_ment = pd.get_dummies(data["basement"], drop_first = True)
base_ment = base_ment.astype(int)
data = pd.concat([base_ment, data], axis = 1)
data = data.rename(columns = {"yes" : "base_ment"})
```

```
In [16]: hot_water = pd.get_dummies(data["hotwaterheating"], drop_first = True)
hot_water = hot_water.astype(int)
```

```
data = pd.concat([hot_water , data] , axis = 1)
data = data.rename(columns = {"yes" : "hot_water"})
```

```
In [17]: air_condition = pd.get_dummies(data["airconditioning"] , drop_first = True)
air_condition = air_condition.astype(int)
data = pd.concat([air_condition , data] , axis = 1)
data = data.rename(columns = {"yes" : "air_condition"})
```

```
In [18]: df = data.drop(["mainroad" , "guestroom" ,"basement" , "hotwaterheating" , "aircondi
```

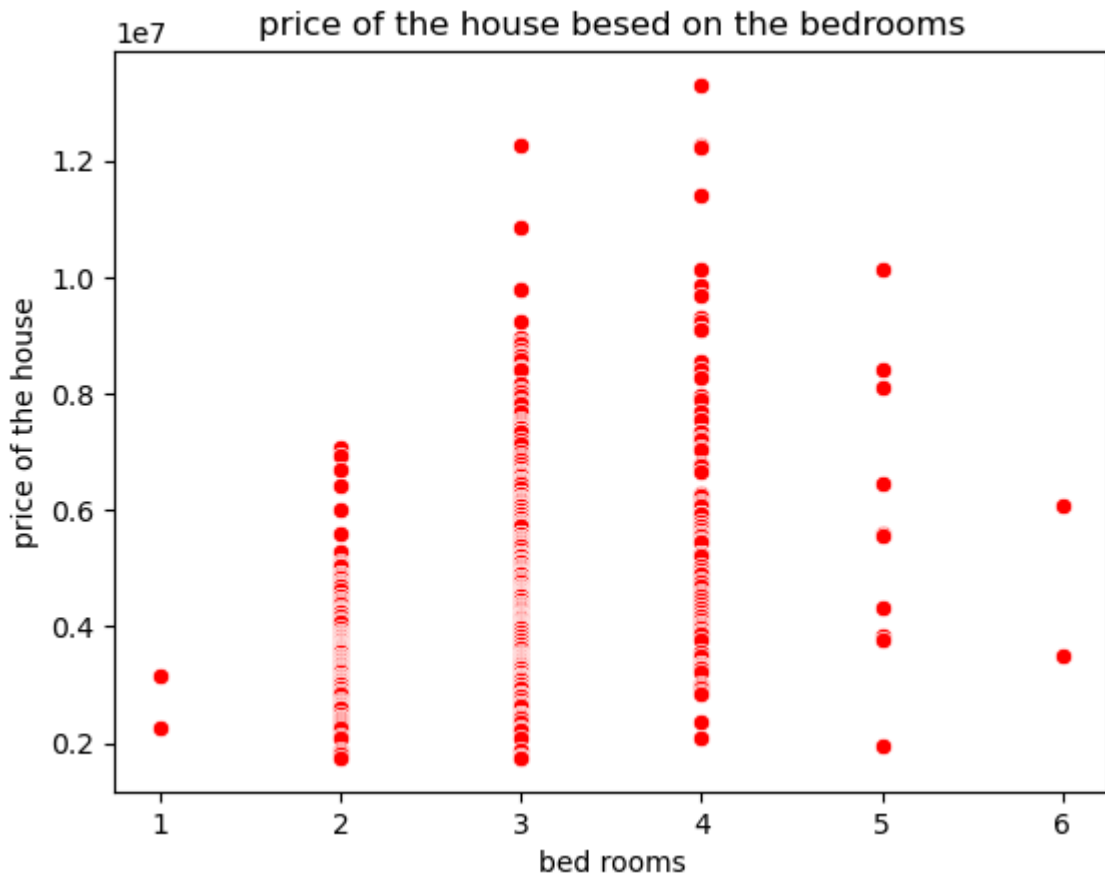
```
In [19]: df.head()
```

```
Out[19]:
```

	air_condition	hot_water	base_ment	guest_room	main_road	price	area	bedrooms	bath
0	1	0	0	0	1	13300000	7420	4	
1	1	0	0	0	1	12250000	8960	4	
2	0	0	1	0	1	12250000	9960	3	
3	1	0	1	0	1	12215000	7500	4	
4	1	0	1	1	1	11410000	7420	4	

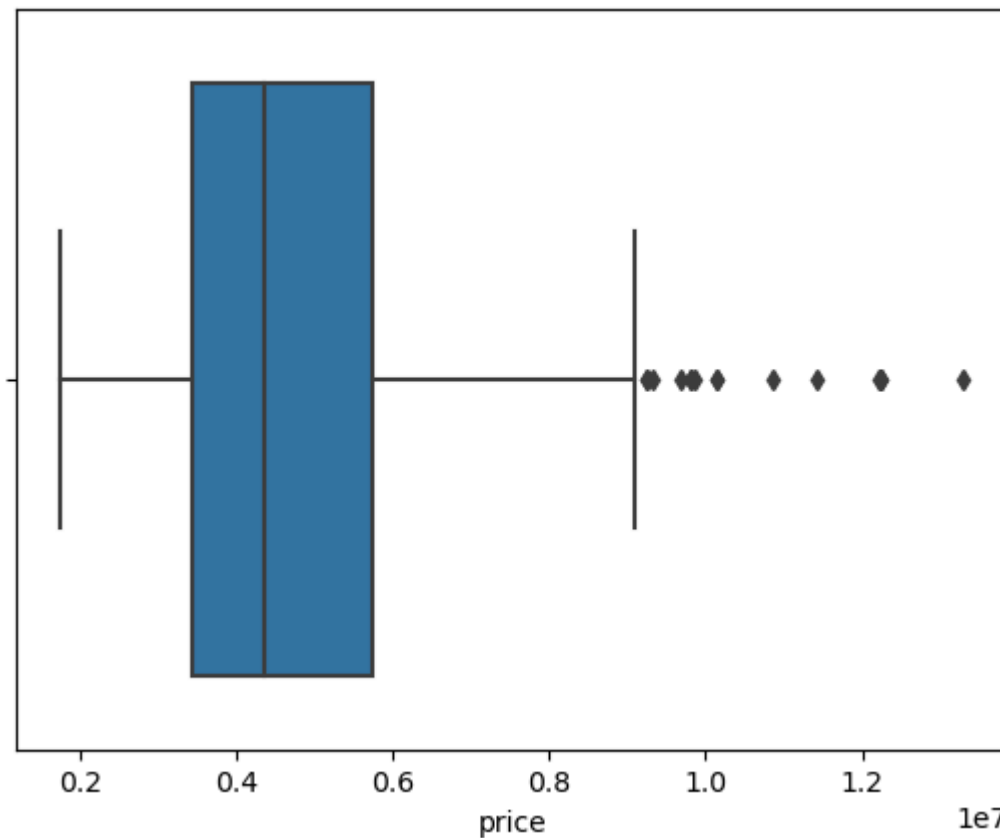
```
In [20]: sns.scatterplot(data = df , x = "bedrooms" , y = "price" , color = "red")
plt.xlabel("bed rooms")
plt.ylabel("price of the house")
plt.title("price of the house based on the bedrooms")
```

```
Out[20]: Text(0.5, 1.0, 'price of the house based on the bedrooms')
```



```
In [21]: sns.boxplot(x = df["price"])
```

Out[21]: <Axes: xlabel='price'>



In [22]: `sns.distplot(df["price"])`

C:\Users\godde\AppData\Local\Temp\ipykernel_20732\50337492.py:1: UserWarning:

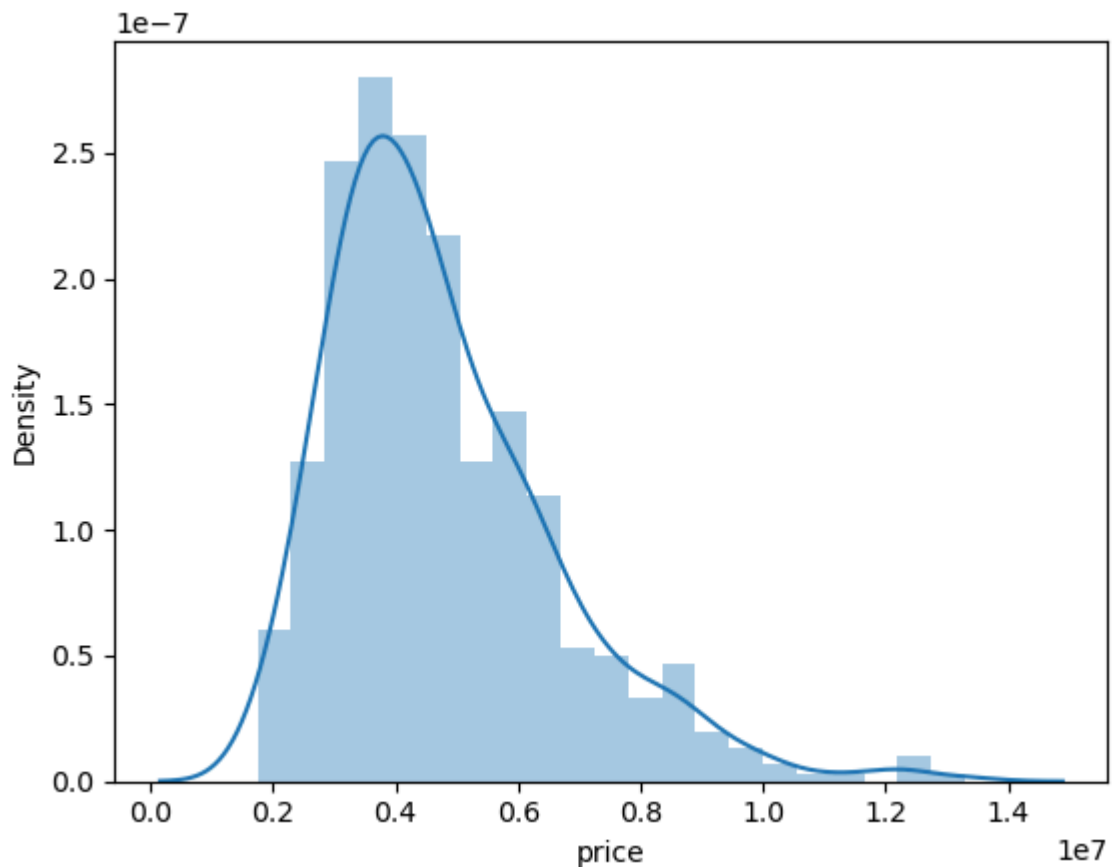
``distplot` is a deprecated function and will be removed in seaborn v0.14.0.`

Please adapt your code to use either ``displot`` (a figure-level function with similar flexibility) or ``histplot`` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["price"])
```

Out[22]: <Axes: xlabel='price', ylabel='Density'>



```
In [23]: Q1 = df["price"].quantile(0.25)
Q2 = df["price"].quantile(0.50)
Q3 = df["price"].quantile(0.75)
print("Q1 = ", Q1)
print("Q2 = ", Q2)
print("Q3 = " , Q3)
```

```
Q1 = 3430000.0
Q2 = 4340000.0
Q3 = 5740000.0
```

```
In [24]: IQR = Q3-Q1
```

```
In [25]: IQR
```

```
Out[25]: 2310000.0
```

```
In [26]: upper_level = Q3+1.5*IQR
lower_level = Q1-1.5*IQR
print("upper_level = " , upper_level)
print("lower_level = " , lower_level)
```

```
upper_level = 9205000.0
lower_level = -35000.0
```

```
In [27]: df["price"] = np.where(df["price"] > upper_level , upper_level,
                               np.where(df["price"] < lower_level , lower_level , df["price"])
```

```
In [28]: sns.distplot(df["price"])
```

```
C:\Users\godde\AppData\Local\Temp\ipykernel_20732\50337492.py:1: UserWarning:
```

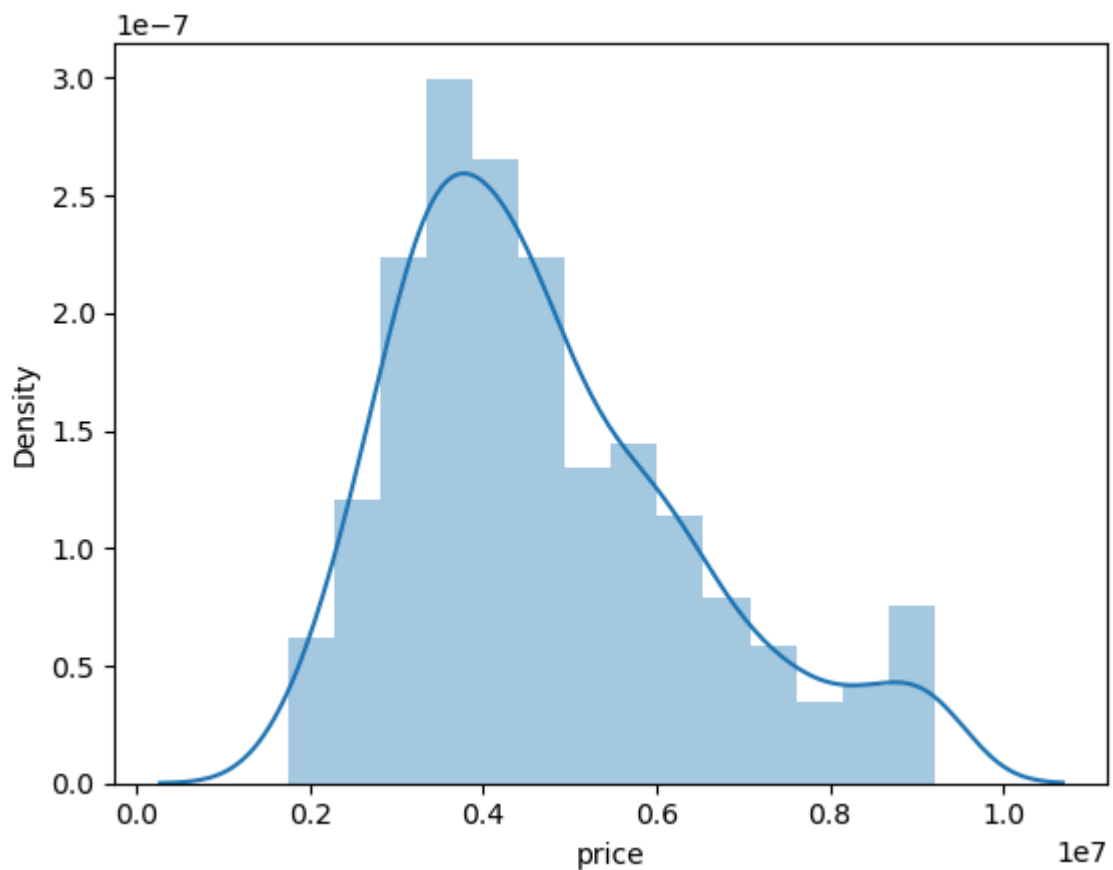
```
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.
```

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

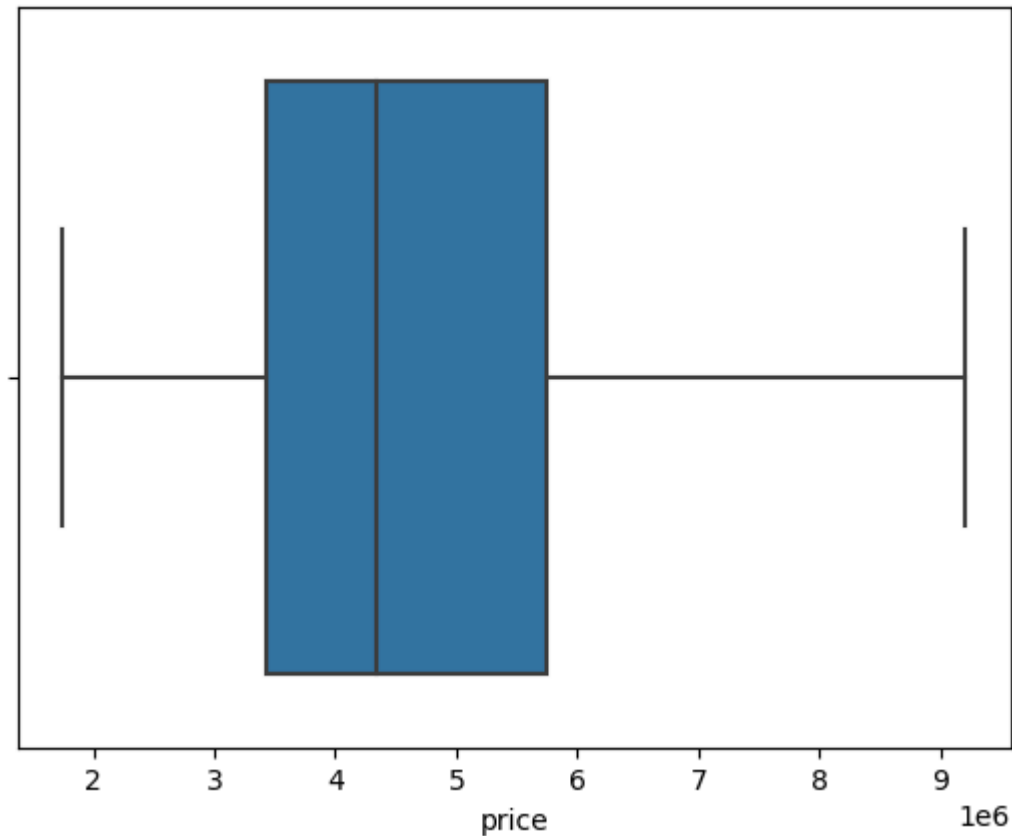
```
sns.distplot(df["price"])
```

```
Out[28]: <Axes: xlabel='price', ylabel='Density'>
```



```
In [29]: sns.boxplot(x = df["price"])
```

```
Out[29]: <Axes: xlabel='price'>
```



In [30]: `df.head()`

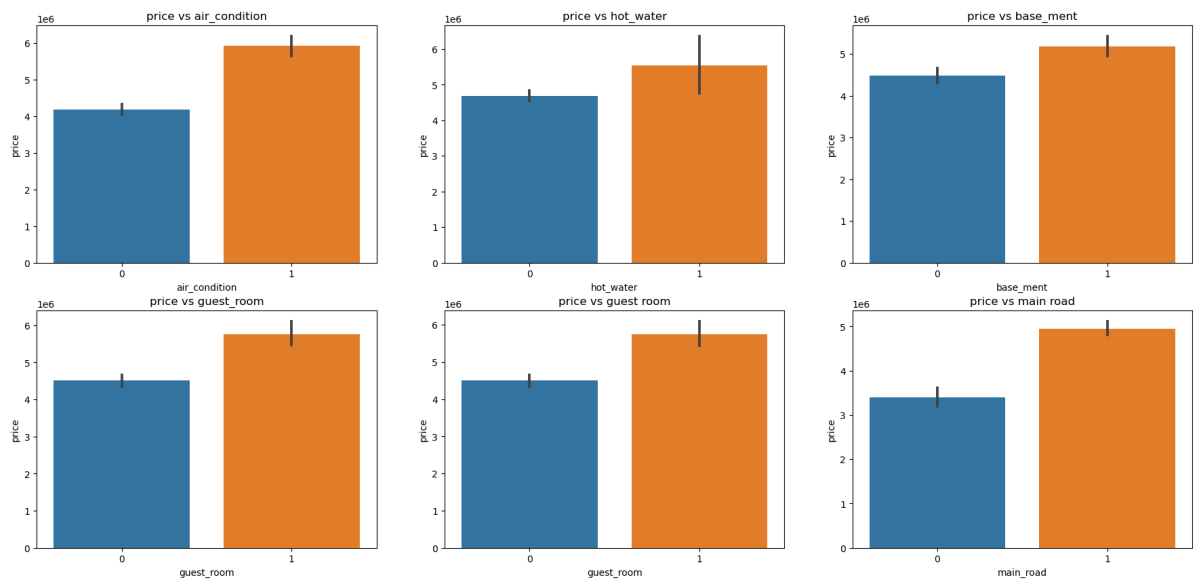
Out[30]:

	air_condition	hot_water	base_ment	guest_room	main_road	price	area	bedrooms	bath
0	1	0	0	0	1	9205000.0	7420	4	
1	1	0	0	0	1	9205000.0	8960	4	
2	0	0	1	0	1	9205000.0	9960	3	
3	1	0	1	0	1	9205000.0	7500	4	
4	1	0	1	1	1	9205000.0	7420	4	

In [31]:

```
plt.figure(figsize = (22,10))
plt.subplot(2,3,1)
sns.barplot(data = df , x = "air_condition" , y = "price")
plt.title("price vs air_condition")
plt.subplot(2,3,2)
sns.barplot(data = df , x = "hot_water" , y = "price")
plt.title("price vs hot_water")
plt.subplot(2,3,3)
sns.barplot(data = df , x = "base_ment" , y = "price")
plt.title("price vs base_ment")
plt.subplot(2,3,4)
sns.barplot(data = df , x = "guest_room" , y = "price")
plt.title("price vs guest_room")
plt.subplot(2,3,5)
sns.barplot(data = df , x = "guest_room" , y = "price")
plt.title("price vs guest room")
plt.subplot(2,3,6)
sns.barplot(data = df , x = "main_road" , y = "price")
plt.title("price vs main road")
```

Out[31]: Text(0.5, 1.0, 'price vs main road')



In [32]: `df.head()`

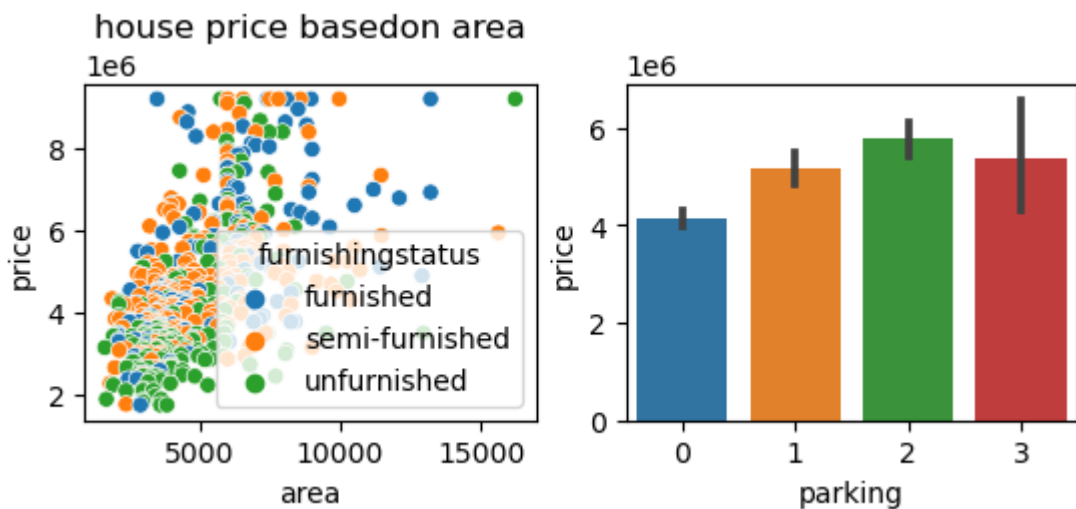
Out[32]:

	air_condition	hot_water	base_ment	guest_room	main_road	price	area	bedrooms	bath
0	1	0	0	0	1	9205000.0	7420	4	
1	1	0	0	0	1	9205000.0	8960	4	
2	0	0	1	0	1	9205000.0	9960	3	
3	1	0	1	0	1	9205000.0	7500	4	
4	1	0	1	1	1	9205000.0	7420	4	

In [33]:

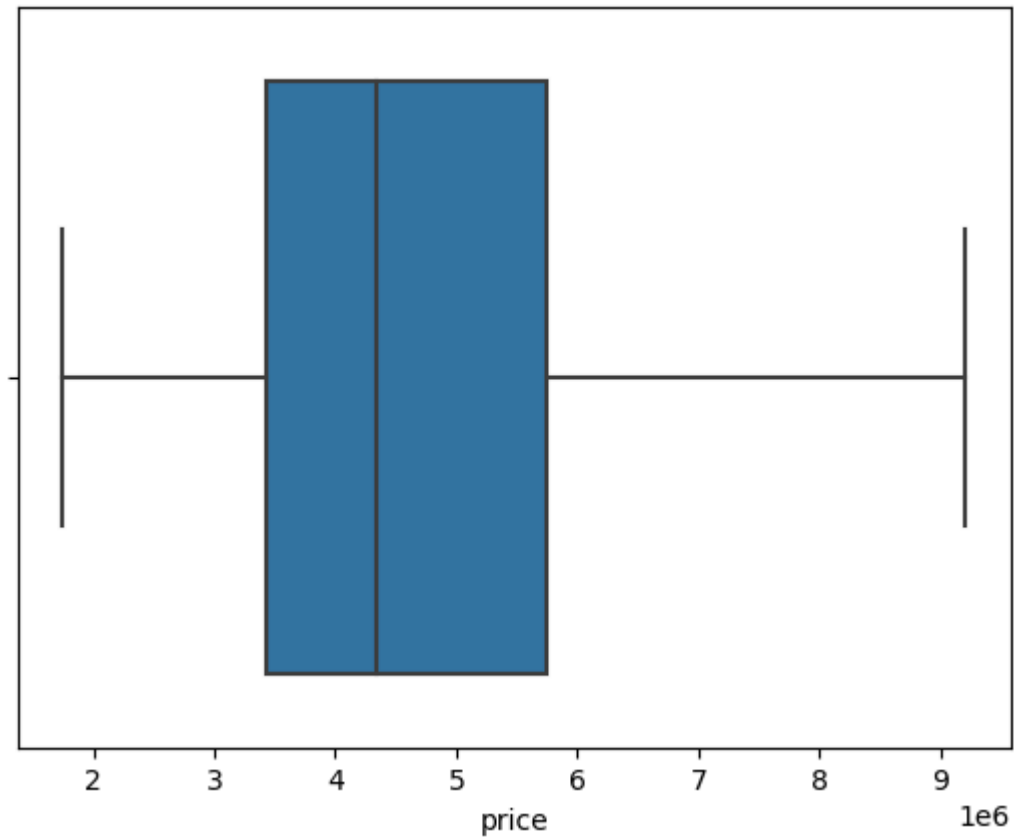
```
plt.subplot(2,2,1)
sns.scatterplot(data = df , x = "area" , y = "price" , hue = "furnishingstatus")
plt.title("house price basedon area")
plt.subplot(2,2,2)
sns.barplot(data = df , x = "parking" , y = "price" )
```

Out[33]: <Axes: xlabel='parking', ylabel='price'>



In [34]: `sns.boxplot(x = df["price"])`

Out[34]: <Axes: xlabel='price'>



```
In [35]: data.head()
```

```
Out[35]:
```

	air_condition	hot_water	base_ment	guest_room	main_road	price	area	bedrooms	bath
0	1	0	0	0	1	13300000	7420	4	
1	1	0	0	0	1	12250000	8960	4	
2	0	0	1	0	1	12250000	9960	3	
3	1	0	1	0	1	12215000	7500	4	
4	1	0	1	1	1	11410000	7420	4	

```
In [36]: df = df.drop("furnishingstatus",axis = 1)
```

```
In [37]: df
```

Out[37]:

	air_condition	hot_water	base_ment	guest_room	main_road	price	area	bedrooms	ba
0	1	0	0	0	1	9205000.0	7420	4	
1	1	0	0	0	1	9205000.0	8960	4	
2	0	0	1	0	1	9205000.0	9960	3	
3	1	0	1	0	1	9205000.0	7500	4	
4	1	0	1	1	1	9205000.0	7420	4	
...
540	0	0	1	0	1	1820000.0	3000	2	
541	0	0	0	0	0	1767150.0	2400	3	
542	0	0	0	0	1	1750000.0	3620	2	
543	0	0	0	0	0	1750000.0	2910	3	
544	0	0	0	0	1	1750000.0	3850	3	

545 rows × 11 columns

In [38]: `x = df.drop("price" , axis = 1)`

In [39]: `y = df[["price"]]`

In [40]: `x`

Out[40]:

	air_condition	hot_water	base_ment	guest_room	main_road	area	bedrooms	bathrooms	s
0	1	0	0	0	1	7420	4	2	
1	1	0	0	0	1	8960	4	4	
2	0	0	1	0	1	9960	3	2	
3	1	0	1	0	1	7500	4	2	
4	1	0	1	1	1	7420	4	1	
...
540	0	0	1	0	1	3000	2	1	
541	0	0	0	0	0	2400	3	1	
542	0	0	0	0	1	3620	2	1	
543	0	0	0	0	0	2910	3	1	
544	0	0	0	0	1	3850	3	1	

545 rows × 10 columns

In [41]: `y`

Out[41]:

	price
0	9205000.0
1	9205000.0
2	9205000.0
3	9205000.0
4	9205000.0
...	...
540	1820000.0
541	1767150.0
542	1750000.0
543	1750000.0
544	1750000.0

545 rows × 1 columns

In [42]: `scalar = StandardScaler()`

In [43]: `x = scalar.fit_transform(x)`

In [44]: `x_train , x_test , y_train , y_test = train_test_split(x , y , test_size = 0.2 , ra`

In [45]: `x_train`

Out[45]:

```
array([[ 1.4726183 , -0.2192645 , -0.73453933, ..., -0.57018671,
         0.22441013, -0.80574124],
       [-0.67906259, -0.2192645 ,  1.3613975 , ...,  1.42181174,
         0.22441013, -0.80574124],
       [-0.67906259, -0.2192645 , -0.73453933, ..., -0.57018671,
        -0.92939666, -0.80574124],
       ...,
       [-0.67906259,  4.5607017 ,  1.3613975 , ..., -0.57018671,
         0.22441013,  0.35597563],
       [ 1.4726183 , -0.2192645 , -0.73453933, ..., -0.57018671,
        -0.92939666,  1.51769249],
       [-0.67906259, -0.2192645 ,  1.3613975 , ..., -0.57018671,
         0.22441013, -0.80574124]])
```

In [46]: `x_test`

Out[46]:

```
array([[ 1.4726183 , -0.2192645 , -0.73453933, ...,  1.42181174,
         1.37821692,  1.51769249],
       [ 1.4726183 , -0.2192645 ,  1.3613975 , ..., -0.57018671,
        -0.92939666, -0.80574124],
       [ 1.4726183 , -0.2192645 ,  1.3613975 , ..., -0.57018671,
         0.22441013, -0.80574124],
       ...,
       [ 1.4726183 , -0.2192645 ,  1.3613975 , ..., -0.57018671,
         1.37821692,  1.51769249],
       [-0.67906259, -0.2192645 , -0.73453933, ..., -0.57018671,
        -0.92939666,  0.35597563],
       [ 1.4726183 , -0.2192645 ,  1.3613975 , ...,  1.42181174,
        -0.92939666,  1.51769249]])
```

```
In [47]: y_train
```

```
Out[47]:
```

	price
--	-------

505	2653000.0
-----	-----------

238	4613000.0
-----	-----------

246	4550000.0
-----	-----------

2	9205000.0
---	-----------

307	4165000.0
-----	-----------

...	...
-----	-----

153	5530000.0
-----	-----------

528	2275000.0
-----	-----------

74	6650000.0
----	-----------

176	5250000.0
-----	-----------

338	3885000.0
-----	-----------

436 rows × 1 columns

```
In [48]: y_test
```

```
Out[48]:
```

	price
--	-------

0	9205000.0
---	-----------

118	5950000.0
-----	-----------

200	4900000.0
-----	-----------

323	4025000.0
-----	-----------

71	6755000.0
----	-----------

...	...
-----	-----

179	5215000.0
-----	-----------

125	5943000.0
-----	-----------

319	4060000.0
-----	-----------

183	5145000.0
-----	-----------

180	5215000.0
-----	-----------

109 rows × 1 columns

```
In [49]: model = DecisionTreeRegressor()
```

```
In [50]: model
```

```
Out[50]:
```

▼ DecisionTreeRegressor

DecisionTreeRegressor()

```
In [51]: model.fit(x_train , y_train)
```

```
Out[51]: ▼ DecisionTreeRegressor  
DecisionTreeRegressor()
```

```
In [52]: y_prediction = model.predict(x_test)
```

```
In [53]: y_prediction[71]
```

```
Out[53]: 7210000.0
```

```
In [54]: y_test[:10]
```

```
Out[54]:
```

	price
0	9205000.0
118	5950000.0
200	4900000.0
323	4025000.0
71	6755000.0
198	4935000.0
446	3150000.0
533	2100000.0
494	2730000.0
536	1960000.0

```
In [55]: error = mean_squared_error(y_test , y_prediction)
```

```
In [56]: error
```

```
Out[56]: 2516478792002.956
```

```
In [57]: error_2 = mean_squared_error(y_prediction , y_test)
```

```
In [58]: error_2
```

```
Out[58]: 2516478792002.956
```

random forest

```
In [59]: from sklearn.ensemble import RandomForestRegressor
```

```
In [60]: rfr = RandomForestRegressor(1000)
```

```
In [61]: rfr
```

```
Out[61]: ▼ RandomForestRegressor  
RandomForestRegressor(n_estimators=1000)
```

```
In [62]: rfr.fit(x_train , y_train)
```

C:\Users\godde\anaconda3\Lib\site-packages\sklearn\base.py:1151: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

```
return fit_method(estimator, *args, **kwargs)
```

```
Out[62]: ▼ RandomForestRegressor
RandomForestRegressor(n_estimators=1000)
```

```
In [63]: prediction = rfr.predict(x_test)
```

```
In [64]: prediction[:10]
```

```
Out[64]: array([ 8395531.66666667,  5887441.          ,  3906136.5          ,
        4185863.5          ,  6707848.          ,  5155118.5          ,
        4935154.          ,  3722481.81666667,  4178216.          ,
        3413746.          ])
```

```
In [65]: y_test
```

```
Out[65]:
```

	price
0	9205000.0
118	5950000.0
200	4900000.0
323	4025000.0
71	6755000.0
...	...
179	5215000.0
125	5943000.0
319	4060000.0
183	5145000.0
180	5215000.0

109 rows × 1 columns

```
In [66]: error_3 = mean_squared_error(y_test , prediction)
```

```
In [67]: error_3
```

```
Out[67]: 1111613325196.62
```

```
In [ ]:
```

support vector machine

```
In [73]: from sklearn.svm import SVR
```

```
In [91]: model = SVR(kernel = "poly")
```

```
In [92]: model
```

```
Out[92]: SVR
SVR(kernel='poly')
```

```
In [93]: model.fit(x_train , y_train)
```

C:\Users\godde\anaconda3\Lib\site-packages\sklearn\utils\validation.py:1184: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
y = column_or_1d(y, warn=True)

```
Out[93]: SVR
SVR(kernel='poly')
```

```
In [94]: y_prediction_2 = model.predict(x_test)
```

```
In [95]: y_prediction[:10]
```

```
Out[95]: array([9205000., 4900000., 4007500., 3290000., 6580000., 5075000.,
        4620000., 4550000., 4900000., 3234000.])
```

```
In [96]: y_test
```

```
Out[96]:
```

	price
0	9205000.0
118	5950000.0
200	4900000.0
323	4025000.0
71	6755000.0
...	...
179	5215000.0
125	5943000.0
319	4060000.0
183	5145000.0
180	5215000.0

109 rows × 1 columns

```
In [97]: error_4 = mean_squared_error(y_test , y_prediction)
```

```
In [98]: error_4
```

```
Out[98]: 2516478792002.956
```

```
In [ ]:
```

In []:

In []:

In []: