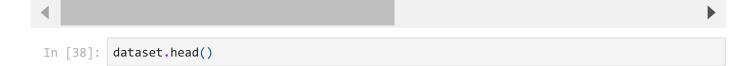
```
In [35]:
         import pandas as pd
         import numpy as nm
         import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn.preprocessing import OneHotEncoder
         from sklearn.metrics import mean_absolute_error
         from sklearn.model_selection import train_test_split
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.metrics import accuracy_score
         from sklearn.metrics import mean_absolute_percentage_error
         from sklearn.linear_model import LinearRegression
         from sklearn import svm
         from sklearn.svm import SVC
In [36]:
         dataset=pd.read_csv('Houseprice.csv')
In [37]:
         dataset
```

Out[37]:		date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	c
	0	2014- 05-02 00:00:00	3.130000e+05	3.0	1.50	1340	7912	1.5	0	0	
	1	2014- 05-02 00:00:00	2.384000e+06	5.0	2.50	3650	9050	2.0	0	4	
	2	2014- 05-02 00:00:00	3.420000e+05	3.0	2.00	1930	11947	1.0	0	0	
	3	2014- 05-02 00:00:00	4.200000e+05	3.0	2.25	2000	8030	1.0	0	0	
	4	2014- 05-02 00:00:00	5.500000e+05	4.0	2.50	1940	10500	1.0	0	0	
	•••			•••	•••	•••	•••		•••		
	4595	2014- 07-09 00:00:00	3.081667e+05	3.0	1.75	1510	6360	1.0	0	0	
	4596	2014- 07-09 00:00:00	5.343333e+05	3.0	2.50	1460	7573	2.0	0	0	
	4597	2014- 07-09 00:00:00	4.169042e+05	3.0	2.50	3010	7014	2.0	0	0	
	4598	2014- 07-10 00:00:00	2.034000e+05	4.0	2.00	2090	6630	1.0	0	0	
	4599	2014- 07-10 00:00:00	2.206000e+05	3.0	2.50	1490	8102	2.0	0	0	

4600 rows × 18 columns



Out[38]:		date	price	bedroom	s bathro	oms sq	qft_living	sqft_lot	floors	waterfro	nt view	conditio
	0	2014- 05-02 00:00:00	313000.0	3.	0	1.50	1340	7912	1.5		0 0	
	1	2014- 05-02 00:00:00	2384000.0	5.	0	2.50	3650	9050	2.0		0 4	
	2	2014- 05-02 00:00:00	342000.0	3.	.0	2.00	1930	11947	1.0		0 0	
	3	2014- 05-02 00:00:00	420000.0	3	0	2.25	2000	8030	1.0		0 0	
	4	2014- 05-02 00:00:00	550000.0	4.	0	2.50	1940	10500	1.0		0 0	
4												•
In [39]:	da	taset.ta	nil()									
[]			()									
Out[39]:		da		price l	pedrooms	bathro	oms sqf	t_living	sqft_lot	floors	waterfron	t view
		<b>da</b>	a <b>te</b> 14- 09 308166		pedrooms 3.0		ooms sqf 1.75	<b>t_living</b> 1510	<b>sqft_lot</b> 6360	floors		t view
	45	95 07- 00:00	14- .09 308166 :00 14- .09 534333	.666667							(	
	45	95 07- 00:00 96 07- 00:00	14- .09 308166 .00 14- .09 534333 .00 14- .09 416904	.333333	3.0		1.75	1510	6360	1.0		0 0
	45 45	95 07- 00:00 96 07- 00:00 20° 97 07-	14- .09 308166 :00 14- .09 534333 :00 14- .09 416904 :00 14- .10 203400	.333333	3.0		2.50	1510	6360 7573	2.0		0 0
	45 45 45	95 07- 00:00 96 07- 00:00 97 07- 00:00 20' 97 07- 00:00 20' 98 07-	14- .09 308166 .00 14- .09 534333 .00 14- .09 416904 .00 14- .10 203400 .00 14- .10 220600	.333333	3.0 3.0 3.0		1.75 2.50 2.50	1510 1460 3010	6360 7573 7014	2.0		0 0
	45 45 45	95 07- 00:00 96 07- 00:00 97 07- 00:00 98 07- 00:00 99 07-	14- .09 308166 .00 14- .09 534333 .00 14- .09 416904 .00 14- .10 203400 .00 14- .10 220600	.666667 .333333 .166667	3.0 3.0 4.0		1.75 2.50 2.50 2.00	1510 1460 3010 2090	6360 7573 7014 6630	1.0 2.0 2.0		0 0 0 0 0 0 0

3, 9:51 PM					Нои	Housepriceprediction						
Out[40]:		date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condi	
	2676	2014- 06-17 00:00:00	506000.0	3.0	1.75	2180	7700	1.0	0	0		
	2212	2014- 06-10 00:00:00	165000.0	3.0	1.00	970	6600	1.0	0	0		
4											•	
In [41]:	data	set.info	()									
	<pre><class 'pandas.core.frame.dataframe="" (total="" 0="" 18="" 4599="" 4600="" columns="" columns):<="" data="" entries,="" pre="" rangeindex:="" to=""></class></pre>											
	#	Column	N	on-Null Co	unt Dtype	<u>:</u>						
			-	600	11	. <b>.</b>						
	0 1	date price		600 non-nu 600 non-nu	9							
	2	bedrooms		600 non-nu								
	3	bathroom		600 non-nu								
	4	sqft_liv		600 non-nu								
	5	sqft_lot	: 4	600 non-nu	ıll int64							
	6	floors		600 non-nu								
	7	waterfro		600 non-nu								
	8	view	4	600 non-nu	ıll int64							

int64

int64

int64

int64

int64

object object

object

object

memory usage: 647.0+ KB

condition

sqft\_basement

10 sqft\_above

13 yr\_renovated

statezip 17 country

dataset.describe()

street

city

12 yr built

14

15

16

In [42]:

4600 non-null

4600 non-null

4600 non-null

4600 non-null

4600 non-null

4600 non-null

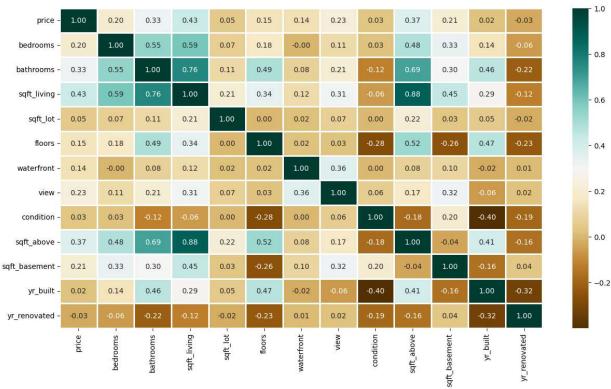
4600 non-null 4600 non-null

4600 non-null

dtypes: float64(4), int64(9), object(5)

Out[42]:			price	e bedroo	ms k	oathro	oms	sqf	t_living	sq	ft_lot		floors	waterfror	nt
	count	4.600	0000e+0	3 4600.000	000 46	600.00	0000	4600.	.000000	4.600000	e+03	4600.0	000000	4600.00000	00
	mean	5.519	0630e+0	5 3.400	870	2.16	0815	2139.	.346957	1.485252	e+04	1.5	12065	0.00717	<b>'</b> 4
	std	5.638	347e+0	5 0.908	848	0.78	3781	963.	206916	3.588444	e+04	0.5	38288	0.08440	)4
	min	0.000	0000e+0	0.000	000	0.00	0000	370.	.000000	6.380000	e+02	1.0	000000	0.00000	00
	25%	3.228	3750e+0	5 3.000	000	1.75	0000	1460	.000000	5.000750	e+03	1.0	000000	0.00000	00
	50%	4.609	9435e+0	5 3.000	000	2.25	0000	1980	.000000	7.683000	e+03	1.5	00000	0.00000	00
	75%	6.549	625e+0	5 4.000	000	2.50	0000	2620	.000000	1.100125	e+04	2.0	000000	0.00000	00
	max	2.659	000e+0	7 9.000	000	8.00	0000	13540	.000000	1.074218	e+06	3.5	00000	1.00000	00
														•	
43]:	datas		ape												
3]:	(4600	, 18)													
4]:	datas	et.is	null()												
1]:		date	price	bedrooms	bathro	ooms	sqft_	living	sqft_lot	floors	wate	rfront	view	condition	s
	0	False	False	False		False		False	False	False		False	False	False	
	1	False	False	False		False		False	False	False		False	False	False	
	2	False	False	False		False		False	False	False		False	False	False	
	3	False	False	False		False		False	False	False		False	False	False	
	4	False	False	False		False		False	False	False		False	False	False	
	•••		•••	•••					•••						
	4595	False	False	False		False		False	False	False		False	False	False	
	4596	False	False	False		False		False	False	False		False	False	False	
	4597	False	False	False		False		False	False	False		False	False	False	
	4598	False	False	False		False		False	False	False		False	False	False	
	4599	False	False	False		False		False	False	False		False	False	False	
	4600 r	ows ×	18 colu	umns											
[45]:	objec	t_col	s = li	types == st(obj[ob al variab	j].ind	lex)	objec	t_col:	s))						
	num_c	ols =	list(	dtypes == int_[int_ ariables:	].inde	ex)	cols)	)							

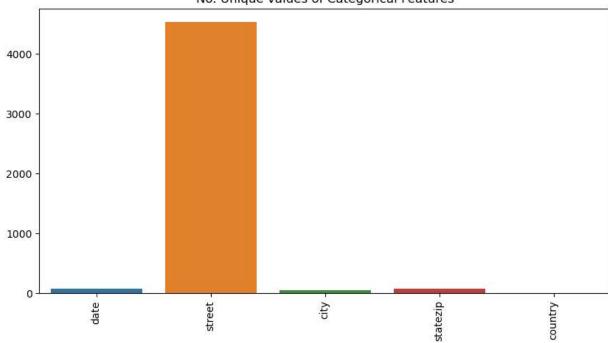
```
fl = (dataset.dtypes == 'float')
          fl_cols = list(fl[fl].index)
          print("Float variables:",len(fl_cols))
         Categorical variables: 5
         Integer variables: 0
         Float variables: 4
In [46]: plt.figure(figsize=(15, 8)
          sns.heatmap(dataset.corr(),
                      cmap = 'BrBG',
                      fmt = '.2f',
                      linewidths = 2,
                      annot = True)
         <AxesSubplot:>
Out[46]:
```



```
unique values = []
In [48]:
         for col in object cols:
            unique_values.append(dataset[col].unique().size)
          plt.figure(figsize=(10,5))
          plt.title('No. Unique values of Categorical Features')
          plt.xticks(rotation=90)
          sns.barplot(x=object_cols,y=unique_values)
```

<AxesSubplot:title={'center':'No. Unique values of Categorical Features'}> Out[48]:

## No. Unique values of Categorical Features



Out[51]:		date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	c
	0	2014- 05-02 00:00:00	3.130000e+05	3.0	1.50	1340	7912	1.5	0	0	
	1	2014- 05-02 00:00:00	2.384000e+06	5.0	2.50	3650	9050	2.0	0	4	
	2	2014- 05-02 00:00:00	3.420000e+05	3.0	2.00	1930	11947	1.0	0	0	
	3	2014- 05-02 00:00:00	4.200000e+05	3.0	2.25	2000	8030	1.0	0	0	
	4	2014- 05-02 00:00:00	5.500000e+05	4.0	2.50	1940	10500	1.0	0	0	
	•••	•••					***	•••			
4!	595	2014- 07-09 00:00:00	3.081667e+05	3.0	1.75	1510	6360	1.0	0	0	
4!	596	2014- 07-09 00:00:00	5.343333e+05	3.0	2.50	1460	7573	2.0	0	0	
4!	597	2014- 07-09 00:00:00	4.169042e+05	3.0	2.50	3010	7014	2.0	0	0	
4!	598	2014- 07-10 00:00:00	2.034000e+05	4.0	2.00	2090	6630	1.0	0	0	
4:	599	2014- 07-10 00:00:00	2.206000e+05	3.0	2.50	1490	8102	2.0	0	0	

4600 rows × 18 columns

**→** 

In [52]: new\_dataset.isnull().sum()

```
date
                          0
Out[52]:
                          0
         price
         bedrooms
                          0
         bathrooms
                          0
         sqft living
                          0
         sqft_lot
                          0
         floors
                           0
         waterfront
                          0
         view
                           0
                          0
         condition
         sqft above
                          0
         sqft basement
                          0
         yr built
                          0
         yr renovated
                          0
         street
                          0
                          0
         city
         statezip
                          0
                           0
         country
         dtype: int64
In [53]: from sklearn.preprocessing import OneHotEncoder
         s = (new_dataset.dtypes == 'object')
         object_cols = list(s[s].index)
          print("Categorical variables:")
          print(object cols)
          print('No. of. categorical features: ',
                len(object_cols))
         Categorical variables:
         ['date', 'street', 'city', 'statezip', 'country']
         No. of. categorical features: 5
         OH encoder = OneHotEncoder(sparse=False)
In [55]:
         OH cols = pd.DataFrame(OH encoder.fit transform(new dataset[object cols]))
         OH cols.index = new dataset.index
         OH cols.columns = OH_encoder.get_feature_names()
          df final = new dataset.drop(object cols, axis=1)
          df final = pd.concat([df final, OH cols], axis=1)
         C:\Users\reddy\anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWar
         ning: Function get feature names is deprecated; get feature names is deprecated in 1.
         0 and will be removed in 1.2. Please use get_feature_names_out instead.
           warnings.warn(msg, category=FutureWarning)
In [57]: X = df_final.drop(['price'], axis=1)
         Y = df final['price']
In [58]:
```

Out[58]:

	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	sqft_above	sc
0	3.0	1.50	1340	7912	1.5	0	0	3	1340	
1	5.0	2.50	3650	9050	2.0	0	4	5	3370	
2	3.0	2.00	1930	11947	1.0	0	0	4	1930	
3	3.0	2.25	2000	8030	1.0	0	0	4	1000	
4	4.0	2.50	1940	10500	1.0	0	0	4	1140	
•••				•••	•••					
4595	3.0	1.75	1510	6360	1.0	0	0	4	1510	
4596	3.0	2.50	1460	7573	2.0	0	0	3	1460	
4597	3.0	2.50	3010	7014	2.0	0	0	3	3010	
4598	4.0	2.00	2090	6630	1.0	0	0	3	1070	
4599	3.0	2.50	1490	8102	2.0	0	0	4	1490	

4600 rows × 4729 columns

```
In [59]:
                 3.130000e+05
Out[59]:
                 2.384000e+06
                 3.420000e+05
         3
                 4.200000e+05
                 5.500000e+05
         4595
                 3.081667e+05
         4596
                 5.343333e+05
         4597
                 4.169042e+05
         4598
                 2.034000e+05
         4599
                 2.206000e+05
         Name: price, Length: 4600, dtype: float64
         Features1 = ['bedrooms', 'bathrooms', 'sqft living', 'sqft lot', 'floors']
In [62]:
         target = 'price'
         X1 = dataset[Features1]
         y1 = dataset[target]
         X_train, X_test, y_train, y_test = train_test_split(X1, y1, test_size=0.5, random_stat
In [63]:
         X1
```

Out[63]:		bedrooms	bathrooms	sqft_living	sqft_lot	floors
	0	3.0	1.50	1340	7912	1.5
	1	5.0	2.50	3650	9050	2.0
	2	3.0	2.00	1930	11947	1.0
	3	3.0	2.25	2000	8030	1.0
	4	4.0	2.50	1940	10500	1.0
	•••					•••
	4595	3.0	1.75	1510	6360	1.0
	4596	3.0	2.50	1460	7573	2.0
	4597	3.0	2.50	3010	7014	2.0
	4598	4.0	2.00	2090	6630	1.0
	4599	3.0	2.50	1490	8102	2.0

4600 rows × 5 columns

```
In [64]:
         у1
                  3.130000e+05
Out[64]:
         1
                  2.384000e+06
          2
                  3.420000e+05
         3
                  4.200000e+05
                  5.500000e+05
         4595
                  3.081667e+05
         4596
                  5.343333e+05
         4597
                  4.169042e+05
         4598
                  2.034000e+05
         4599
                  2.206000e+05
         Name: price, Length: 4600, dtype: float64
In [65]:
          lr = LinearRegression()
          lr.fit(X_train, y_train)
In [66]:
         LinearRegression()
Out[66]:
In [67]:
         y_pred = lr.predict(X_test)
In [68]:
         y_pred
         array([363715.81566704, 399446.44586553, 834230.23979669, ...,
Out[68]:
                 748123.84167885, 569569.52453238, 658083.41823349])
          score = lr.score(X_test, y_test)
In [69]:
          print("lr R^2 Score:", score)
         lr R^2 Score: 0.10933671026238057
          new_house = pd.DataFrame({'bedrooms': [2], 'bathrooms': [2.5], 'sqft_living': [600],
In [70]:
          predicted_price = lr.predict(new_house)
```

```
print("Predicted Price:", predicted_price[0])
Predicted Price: 146977.00302329706
```

In [ ]:

9/6/23, 9:51 PM