```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")
```

In [3]: df = pd.read_csv("Real_estates.csv")
df

Out[3]:

:	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
0	79545.458574	5.682861	7.009188	4.09	23086.800503	1.059034e+06	208 Michael Ferry Apt. 674\nLaurabury, NE 3701
1	79248.642455	6.002900	6.730821	3.09	40173.072174	1.505891e+06	188 Johnson Views Suite 079\nLake Kathleen, CA
2	61287.067179	5.865890	8.512727	5.13	36882.159400	1.058988e+06	9127 Elizabeth Stravenue\nDanieltown, WI 06482
3	63345.240046	7.188236	5.586729	3.26	34310.242831	1.260617e+06	USS Barnett\nFPO AP 44820
4	59982.197226	5.040555	7.839388	4.23	26354.109472	6.309435e+05	USNS Raymond\nFPO AE 09386
4995	60567.944140	7.830362	6.137356	3.46	22837.361035	1.060194e+06	USNS Williams\nFPO AP 30153-7653
4996	78491.275435	6.999135	6.576763	4.02	25616.115489	1.482618e+06	PSC 9258, Box 8489\nAPO AA 42991- 3352
4997	63390.686886	7.250591	4.805081	2.13	33266.145490	1.030730e+06	4215 Tracy Garden Suite 076\nJoshualand, VA 01
4998	68001.331235	5.534388	7.130144	5.44	42625.620156	1.198657e+06	USS Wallace\nFPO AE 73316
4999	65510.581804	5.992305	6.792336	4.07	46501.283803	1.298950e+06	37778 George Ridges Apt. 509\nEast Holly, NV 2

5000 rows × 7 columns

In [4]: df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 5000 entries, 0 to 4999 Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	Avg. Area Income	5000 non-null	float64
1	Avg. Area House Age	5000 non-null	float64
2	Avg. Area Number of Rooms	5000 non-null	float64
3	Avg. Area Number of Bedrooms	5000 non-null	float64
4	Area Population	5000 non-null	float64
5	Price	5000 non-null	float64
6	Address	5000 non-null	object
dtvp	es: float64(6), object(1)		

dtypes: float64(6), object(1)
memory usage: 273.6+ KB

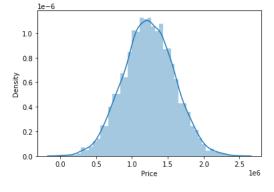
In [6]: df.describe()

Out[6]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
count	5000.000000	5000.000000	5000.000000	5000.000000	5000.000000	5.000000e+03
mean	68583.108984	5.977222	6.987792	3.981330	36163.516039	1.232073e+06
std	10657.991214	0.991456	1.005833	1.234137	9925.650114	3.531176e+05
min	17796.631190	2.644304	3.236194	2.000000	172.610686	1.593866e+04
25%	61480.562388	5.322283	6.299250	3.140000	29403.928702	9.975771e+05
50%	68804.286404	5.970429	7.002902	4.050000	36199.406689	1.232669e+06
75%	75783.338666	6.650808	7.665871	4.490000	42861.290769	1.471210e+06
max	107701.748378	9.519088	10.759588	6.500000	69621.713378	2.469066e+06

In [11]: sns.distplot(df["Price"])

Out[11]: <AxesSubplot: xlabel='Price', ylabel='Density'>



In [12]: df.corr()

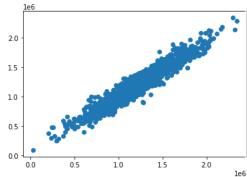
Out[12]:

:		Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
	Avg. Area Income	1.000000	-0.002007	-0.011032	0.019788	-0.016234	0.639734
	Avg. Area House Age	-0.002007	1.000000	-0.009428	0.006149	-0.018743	0.452543
	Avg. Area Number of Rooms	-0.011032	-0.009428	1.000000	0.462695	0.002040	0.335664
	Avg. Area Number of Bedrooms	0.019788	0.006149	0.462695	1.000000	-0.022168	0.171071
	Area Population	-0.016234	-0.018743	0.002040	-0.022168	1.000000	0.408556
	Price	0.639734	0.452543	0.335664	0.171071	0.408556	1.000000

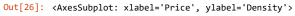
```
In [13]: |df.corr().style.background_gradient()
Out[13]:
                                             Avg. Area Income Avg. Area House Age Avg. Area Number of Rooms Avg. Area Number of Bedrooms Area Population
                                                                                                                                                                       Price
                                                     1.000000
                                                                          -0.002007
                                                                                                                                                         -0.016234 0.639734
                          Avg. Area Income
                                                                                                        -0.011032
                                                                                                                                        0.019788
                       Avg. Area House Age
                                                     -0.002007
                                                                           1.000000
                                                                                                        -0.009428
                                                                                                                                        0.006149
                                                                                                                                                         -0.018743 0.452543
                Avg. Area Number of Rooms
                                                     -0.011032
                                                                           -0.009428
                                                                                                        1.000000
                                                                                                                                        0.462695
                                                                                                                                                          0.002040 0.335664
             Avg. Area Number of Bedrooms
                                                     0.019788
                                                                           0.006149
                                                                                                        0.462695
                                                                                                                                        1.000000
                                                                                                                                                         -0.022168 0.171071
                           Area Population
                                                     -0.016234
                                                                           -0.018743
                                                                                                        0.002040
                                                                                                                                        -0.022168
                                                                                                                                                          1.000000 0.408556
                                                     0.639734
                                                                           0.452543
                                                                                                        0.335664
                                                                                                                                        0.171071
                                                                                                                                                          0.408556 1.000000
                                      Price
In [14]: sns.heatmap(df.corr(),annot=True,cmap="Blues")
Out[14]: <AxesSubplot: >
                         Avg. Area Income -
                                                 -0.002 -0.011 0.02
                                                                      -0.016
                                                                                          - 0.8
                      Avg. Area House Age - -0.002
                                                                               0.45
                                                                                          - 0.6
                Avg. Area Number of Rooms - -0.011 -0.0094
                                                                 0.46
                                                                        0.002
                                                                               0.34
             Avg. Area Number of Bedrooms - 0.02 0.0061 0.46
                                                                        -0.022
                                                                               0.17
                          Area Population - -0.016 -0.019
                                          0.64 0.45
                                                        0.34
                                                                        0.41
                                                                0.17
                                                                                          -00
                                                                 Avg. Area Number of Bedrooms
                                                  Avg. Area House Age
                                                                        Area Population
                                           Avg. Area Income
```

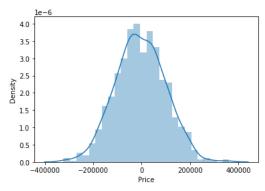
Model Creation

]:		Avg. Area Income	Avg. Area House Age		rea Number of Rooms	Avg. A	rea Number of Bedrooms	Area Population	Price	Address
0	7	79545.458574	5.682861		7.009188		4.09	23086.800503	1.059034e+06	208 Michael Ferry Apt. 674\nLaurabury, NE 3701
1	7	79248.642455	6.002900		6.730821		3.09	40173.072174	1.505891e+06	188 Johnson Views Suite 079\nLake Kathleen, CA
2	6	31287.067179	5.865890		8.512727		5.13	36882.159400	1.058988e+06	9127 Elizabeth Stravenue\nDanieltown, W 06482
3	6	3345.240046	7.188236		5.586729		3.26	34310.242831	1.260617e+06	USS Barnett\nFPO AP 44820
4	5	59982.197226	5.040555		7.839388		4.23	26354.109472	6.309435e+05	USNS Raymond\nFPO AE 0938
		.iloc[:,:-2] .iloc[:,-2]								
: x										
:					vg. Area Numb		Avg. Area Num		Area Population	_
	0 1	79545.45857		5.682861		7.009188		4.09		
	2	79248.64245 61287.06717		5.002900 5.865890		6.730821 8.512727		3.09 5.13		
	3	63345.24004		7.188236		5.586729		3.26		
	4	59982.19722		5.040555		7.839388		4.23		
49	95	60567.94414	0 7	7.830362		6.137356		3.46	22837.36103	5
49	96	78491.27543	s5 6	3.999135		6.576763		4.02	25616.115489	9
49	97	63390.68688	36 7	7.250591		4.805081		2.13	33266.145490)
49	98	68001.33123	5 5	5.534388		7.130144		5.44	42625.620156	3
49	999	65510.58180	14 5	5.992305		6.792336		4.07	46501.283803	3
50	00 ro	ws × 5 column	s							
у										
: 0		1.059034e+								
1 2		1.505891e+ 1.058988e+								
3		1.260617e+								
4		6.309435e+	·05							
499	95	1.060194e+	-06							
499		1.482618e+								
499 499		1.030730e+ 1.198657e+								
499		1.298950e+								
Na	me:	Price, Lengt	h: 5000, dty	pe: float	:64					



```
In [26]: sns.distplot((ytest-ypred))
```





Regression Evaluation Metrics

```
In [27]: from sklearn.metrics import mean_absolute_error,mean_squared_error,r2_score
         mae=mean_absolute_error(ytest,ypred)
mse=mean_squared_error(ytest,ypred)
         rmse=np.sqrt(mse)
         r2=r2_score(ytest,ypred)
         print(f"MAE:- {mae}\n MSE:- {mse}\n RMSE:- {rmse}\n Accuracy:- {r2}")
         MAE:- 82520.32996179513
          MSE:- 10542495109.238657
RMSE:- 102676.65318483388
          Accuracy:- 0.9216017201335713
In [28]: x.columns
In [29]: def predictprice(aai,aaha,aanor,aanob,ap):
             newobs=[[aai,aaha,aanor,aanob,ap]]
             {\tt yp=linreg.predict(newobs)[0]}
             print(f"The price of your Dream House is $ {yp:.2f}")
In [34]: predictprice(70000,15,10,5,10000)
         The price of your Dream House is $ 2721395.26
 In [ ]:
```