In [1]: import numpy as np
 import pandas as pd
 import matplotlib.pyplot as plt
 import seaborn as sns

In [2]: data=pd.read\_csv(r"C:\Users\user\Downloads\10\_USA\_Housing.csv")
 data

#### Out[2]:

Ac	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income	
208 Michael Fer 674\nLaurabu	1.059034e+06	23086.800503	4.09	7.009188	5.682861	79545.458574	0
188 Johnson Suite 079 Kathleer	1.505891e+06	40173.072174	3.09	6.730821	6.002900	79248.642455	1
9127 Eli. Stravenue∖nDani∈ WI 0⊦	1.058988e+06	36882.159400	5.13	8.512727	5.865890	61287.067179	2
USS Barnett\nF	1.260617e+06	34310.242831	3.26	5.586729	7.188236	63345.240046	3
USNS Raymond AE	6.309435e+05	26354.109472	4.23	7.839388	5.040555	59982.197226	4
		•••					
USNS Williams AP 3015	1.060194e+06	22837.361035	3.46	6.137356	7.830362	60567.944140	4995
PSC 925 8489\nAPO AA 4	1.482618e+06	25616.115489	4.02	6.576763	6.999135	78491.275435	4996
4215 Tracy ( Suite 076∖nJosh∟ V	1.030730e+06	33266.145490	2.13	4.805081	7.250591	63390.686886	4997
USS Wallace\nF	1.198657e+06	42625.620156	5.44	7.130144	5.534388	68001.331235	4998
37778 George I Apt. 509\nEas	1.298950e+06	46501.283803	4.07	6.792336	5.992305	65510.581804	4999

5000 rows × 7 columns

In [3]: df=data.head(100)
 df

## Out[3]:

Addı	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income	
208 Michael Ferry 674\nLaurabury 37	1.059034e+06	23086.800503	4.09	7.009188	5.682861	79545.458574	0
188 Johnson Vi Suite 079\nL Kathleen, (	1.505891e+06	40173.072174	3.09	6.730821	6.002900	79248.642455	1
9127 Elizal Stravenue\nDanieltc WI 064	1.058988e+06	36882.159400	5.13	8.512727	5.865890	61287.067179	2
USS Barnett\nFPC 44	1.260617e+06	34310.242831	3.26	5.586729	7.188236	63345.240046	3
USNS Raymond\ni AE 09	6.309435e+05	26354.109472	4.23	7.839388	5.040555	59982.197226	4
69600 Wi Rue\nElizabethl PW 17767-2	1.840236e+06	51815.096419	4.33	6.578352	6.372731	73698.696357	95
001 S Plaza\nJessicastad 25	1.441422e+06	49829.951500	4.41	6.736974	6.374930	66263.910501	96
238 Antr Drive\nAcostaha AS 62753-6	1.498641e+06	22268.075306	4.48	7.834100	6.933352	75394.759582	97
01818 Wi Spurs\nV Andreaton, SD 69! 7	9.547466e+05	21867.476940	2.38	6.316480	6.512270	65984.750645	98
95779 Wa Square\nBakersl PR 49	1.124636e+06	32458.986699	2.37	6.038488	6.982152	66477.262792	99

100 rows × 7 columns

4

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
```

Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	Avg. Area Income	100 non-null	float64
1	Avg. Area House Age	100 non-null	float64
2	Avg. Area Number of Rooms	100 non-null	float64
3	Avg. Area Number of Bedrooms	100 non-null	float64
4	Area Population	100 non-null	float64
5	Price	100 non-null	float64
6	Address	100 non-null	object

dtypes: float64(6), object(1)

memory usage: 5.6+ KB

## In [5]: df.describe()

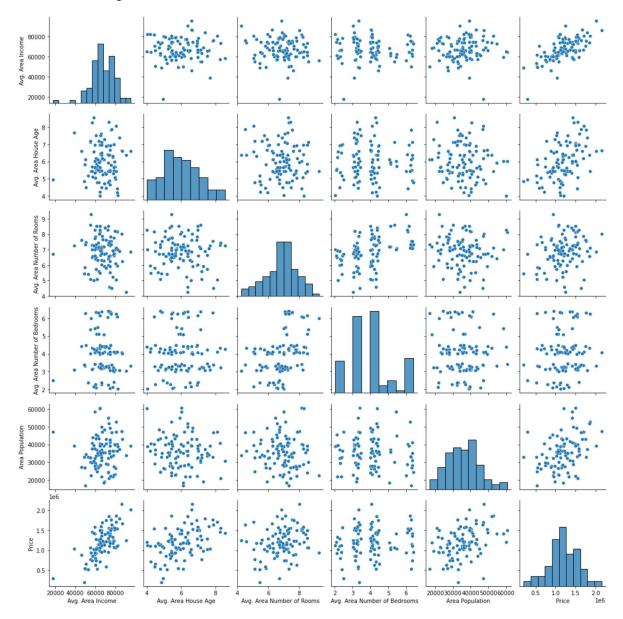
#### Out[5]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
count	100.000000	100.000000	100.000000	100.000000	100.000000	1.000000e+02
mean	67419.354519	6.017796	6.860572	4.004600	35918.684716	1.197750e+06
std	11231.604384	1.032219	0.992326	1.197617	9243.180363	3.622928e+05
min	17796.631190	4.010907	4.242191	2.030000	16810.783311	2.018981e+05
25%	60800.655369	5.343417	6.302292	3.195000	29172.345270	1.009614e+06
50%	66408.895319	5.948502	6.954229	4.040000	35620.473543	1.181989e+06
75%	74544.868943	6.640431	7.447286	4.422500	40517.769280	1.444560e+06
max	95450.293086	8.562611	9.289854	6.410000	60828.249085	2.146925e+06

# In [6]: df.columns

In [7]: sns.pairplot(df)

Out[7]: <seaborn.axisgrid.PairGrid at 0x134bebf8d90>



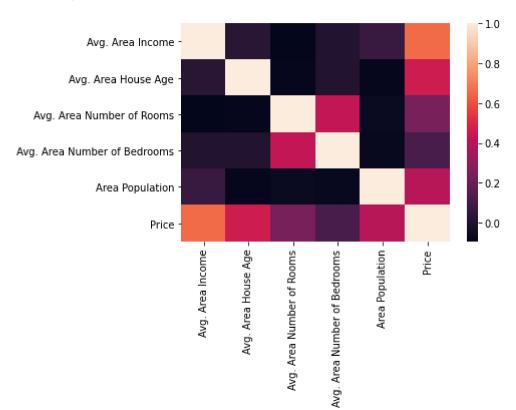
#### Out[8]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
0	79545.458574	5.682861	7.009188	4.09	23086.800503	1.059034e+06
1	79248.642455	6.002900	6.730821	3.09	40173.072174	1.505891e+06
2	61287.067179	5.865890	8.512727	5.13	36882.159400	1.058988e+06
3	63345.240046	7.188236	5.586729	3.26	34310.242831	1.260617e+06
4	59982.197226	5.040555	7.839388	4.23	26354.109472	6.309435e+05
	***				***	
95	73698.696357	6.372731	6.578352	4.33	51815.096419	1.840236e+06
96	66263.910501	6.374930	6.736974	4.41	49829.951500	1.441422e+06
97	75394.759582	6.933352	7.834100	4.48	22268.075306	1.498641e+06
98	65984.750645	6.512270	6.316480	2.38	21867.476940	9.547466e+05
99	66477.262792	6.982152	6.038488	2.37	32458.986699	1.124636e+06

100 rows × 6 columns

In [9]: sns.heatmap(da.corr())

### Out[9]: <AxesSubplot:>



```
In [10]: x=df[['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of Rooms',
                 'Avg. Area Number of Bedrooms', 'Area Population']]
         y=df['Price']
In [11]: from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
In [12]: from sklearn.linear_model import LinearRegression
         lr=LinearRegression()
         lr.fit(x_train,y_train)
Out[12]: LinearRegression()
In [13]:
         print(lr.intercept_)
          -2705255.3129567625
In [14]:
         coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
         coeff
Out[14]:
                                       Co-efficient
                     Avg. Area Income
                                        20.573829
                   Avg. Area House Age 185677.141099
             Avg. Area Number of Rooms 128325.370372
          Avg. Area Number of Bedrooms
                                     -12050 703754
```

15.421778

**Area Population** 

```
Usa (Lasso) - Jupyter Notebook
In [15]:
          prediction=lr.predict(x_test)
          plt.scatter(y_test,prediction)
Out[15]: <matplotlib.collections.PathCollection at 0x134c227a940>
           2.0
           1.8
           1.6
           1.4
           1.2
           1.0
           0.8
           0.6
           0.4
                                 1.2
                                      1.4
                 0.6
                      0.8
                            1.0
                                            1.6
                                                 1.8
                                                      2.0
                                                            2.2
                                                           1e6
In [16]: print(lr.score(x_test,y_test))
          0.9206102508107372
In [17]:
          print(lr.score(x_train,y_train))
          0.9020038095115043
          from sklearn.linear model import Ridge,Lasso
In [18]:
In [19]: rr=Ridge(alpha=10)
          rr.fit(x_train,y_train)
Out[19]: Ridge(alpha=10)
In [20]: |rr.score(x_test,y_test)
Out[20]: 0.9173916645121715
In [21]: la=Lasso(alpha=10)
          la.fit(x_train,y_train)
```

Out[21]: Lasso(alpha=10)

In [22]: la.score(x\_test,y\_test)

Out[22]: 0.9206151912339886

In [ ]: