Problem statement

predicting the house price in USA. To create a model to help him estimate of what the house would sell for.

To display top 10 rows

In [3]: 1 df.head(10)

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
5	80175.754159	4.988408	6.104512	4.04	26748.428425	1.068138e+06	06039 Jennifer Islands Apt. 443\nTracyport, KS
6	64698.463428	6.025336	8.147760	3.41	60828.249085	1.502056e+06	4759 Daniel Shoals Suite 442\nNguyenburgh, CO
7	78394.339278	6.989780	6.620478	2.42	36516.358972	1.573937e+06	972 Joyce Viaduct\nLake William, TN 17778-6483
8	59927.660813	5.362126	6.393121	2.30	29387.396003	7.988695e+05	USS Gilbert\nFPO AA 20957
9	81885.927184	4.423672	8.167688	6.10	40149.965749	1.545155e+06	Unit 9446 Box 0958\nDPO AE 97025

Data Cleaning And Pre-Processing

In [4]: 1 df.info()

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

Column	Non-Null Count	Dtype
Avg. Area Income	5000 non-null	float64
Avg. Area House Age	5000 non-null	float64
Avg. Area Number of Rooms	5000 non-null	float64
Avg. Area Number of Bedrooms	5000 non-null	float64
Area Population	5000 non-null	float64
Price	5000 non-null	float64
Address	5000 non-null	object
	Avg. Area Income Avg. Area House Age Avg. Area Number of Rooms Avg. Area Number of Bedrooms Area Population Price	Avg. Area Income 5000 non-null Avg. Area House Age 5000 non-null Avg. Area Number of Rooms 5000 non-null Avg. Area Number of Bedrooms 5000 non-null Area Population 5000 non-null Price 5000 non-null

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

In [5]:

- 1 # Display the statistical summary
- 2 df.describe()

Out[5]:

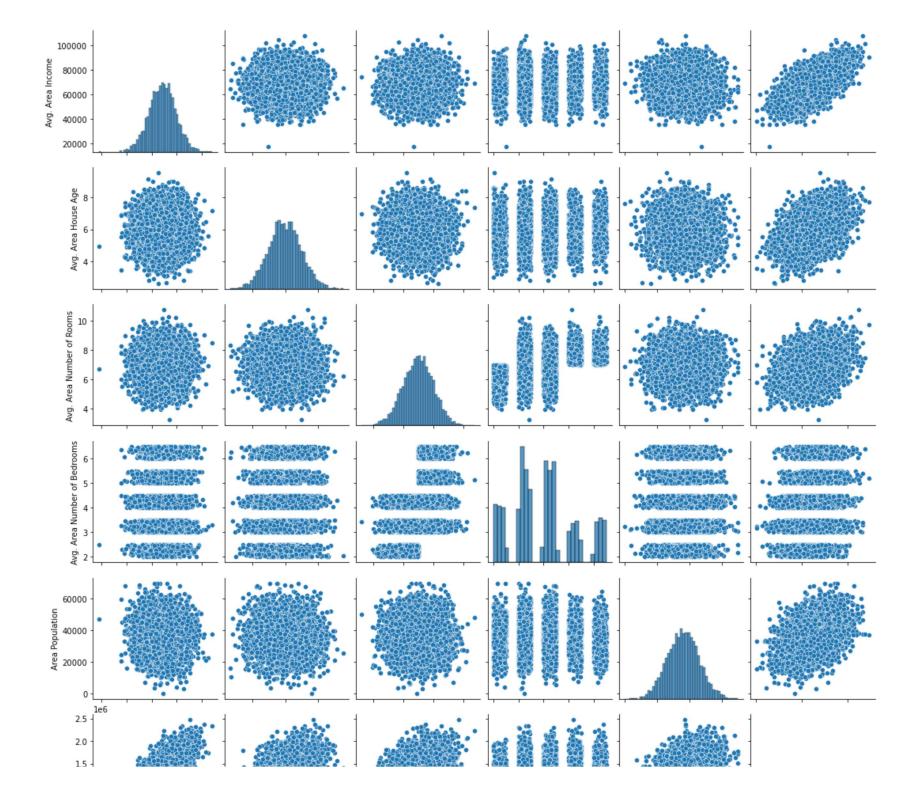
	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
count	5000.000000	5000.000000	5000.000000	5000.000000	5000.000000	5.000000e+03
mean	68583.108984	5.977222	6.987792	3.981330	36163.516039	1.232073e+06
std	10657.991214	0.991456	1.005833	1.234137	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

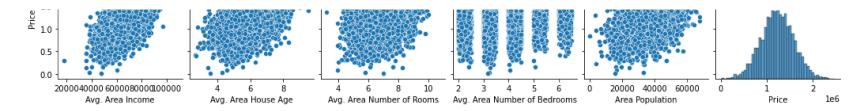
- In [6]: 1 # To display the col headings
 - 2 df.columns

EDA and Visualization

```
In [9]: 1 sns.pairplot(cols)
```

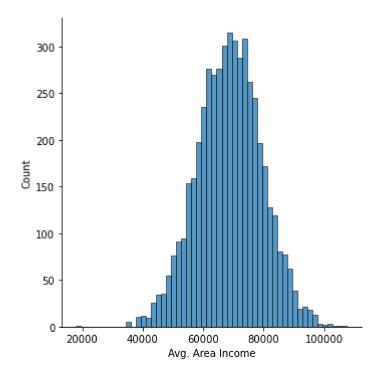
Out[9]: <seaborn.axisgrid.PairGrid at 0x1dddde41d60>





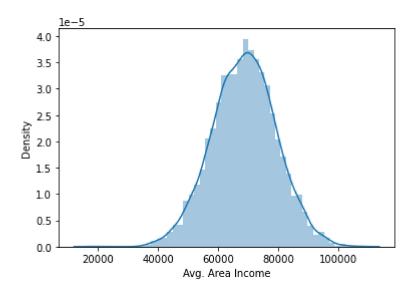
In [11]: 1 sns.displot(df['Avg. Area Income'])

Out[11]: <seaborn.axisgrid.FacetGrid at 0x1dde091efd0>



C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a dep
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

Out[12]: <AxesSubplot:xlabel='Avg. Area Income', ylabel='Density'>



Out[13]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms
0	79545.458574	5.682861	7.009188	4.09
1	79248.642455	6.002900	6.730821	3.09
2	61287.067179	5.865890	8.512727	5.13
3	63345.240046	7.188236	5.586729	3.26
4	59982.197226	5.040555	7.839388	4.23
4995	60567.944140	7.830362	6.137356	3.46
4996	78491.275435	6.999135	6.576763	4.02
4997	63390.686886	7.250591	4.805081	2.13
4998	68001.331235	5.534388	7.130144	5.44
4999	65510.581804	5.992305	6.792336	4.07

5000 rows × 4 columns

```
In [14]: 1 sns.heatmap(df1.corr())
Out[14]: <AxesSubplot:>
```



To train the model - MODEL BUILD

Going to train linear regression model; We split our data into 2 variables x and y where x is independent var(input) and y is dependent on x(output), we could ignore address col as it is not required for our model

To split the dataset into test data

```
In [16]:
           1 # importing lib for splitting test data
           2 from sklearn.model selection import train test split
In [17]: 1 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
In [18]:
           1 from sklearn.linear_model import LinearRegression
           3 lr=LinearRegression()
           4 lr.fit(x_train,y_train)
Out[18]: LinearRegression()
          1 print(lr.intercept_)
In [19]:
         [-1.45519152e-11]
In [20]:
          1 print(lr.score(x test,y test))
         1.0
In [21]:
           1 coeff=pd.DataFrame(lr.coef_)
           2 coeff
Out[21]:
                                     2
          0 1.0 -1.421563e-15 -2.056758e-15 1.100120e-14
```

```
1 pred = lr.predict(x_test)
In [22]:
           plt.scatter(y_test,pred)
Out[22]: <matplotlib.collections.PathCollection at 0x1dde2aafb20>
          100000
           90000
           80000
           70000
           60000
           50000
           40000
                         50000
                                60000
                                      70000
                                            80000
                   40000
                                                  90000 100000
In [23]:
          1 from sklearn.linear model import Ridge, Lasso
           1 rr=Ridge(alpha=10)
In [24]:
           2 rr.fit(x_train,y_train)
Out[24]: Ridge(alpha=10)
In [25]:
          1 rr.score(x_test,y_test)
Out[25]: 1.0
In [26]:
           1 la=Lasso(alpha=10)
           2 la.fit(x_train,y_train)
Out[26]: Lasso(alpha=10)
           1 la.score(x_test,y_test)
In [27]:
Out[27]: 0.9999999999993
```

In []: 1