Problem statement

A real estate agent want help to predict the house price for regions in USA.He gave us the dataset to work on to use Linear Regression model.Create a model that hepls to determine

Data collection

Importing libraries

```
In [2]:
```

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```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

Importing dataset

```
In [4]:
```

data=pd.read_csv(r"C:\Users\user\Downloads\H.csv")
data

Out[4]:

| Address | Price | Area Population | Avg. Area Number of Bedrooms | Avg. Area Number of Rooms | Avg. Area House Age | Avg. Area Income | |
|--|--------------|--------------------|---------------------------------------|---------------------------------------|------------------------------|---------------------|------|
| 208 Michael Ferry Apt. 674\nLaurabury, NE 3701 | 1.059034e+06 | 23086.80050 | 4.09 | 7.009188 | 5.682861 | 79545.45857 | 0 |
| 188 Johnson Views Suite 079\nLake Kathleen, CA | 1.505891e+06 | 40173.07217 | 3.09 | 6.730821 | 6.002900 | 79248.64245 | 1 |
| 9127 Elizabeth Stravenue\nDanieltown, WI 06482 | 1.058988e+06 | 36882.15940 | 5.13 | 8.512727 | 5.865890 | 61287.06718 | 2 |
| USS Barnett\nFPO AP 44820 | 1.260617e+06 | 34310.24283 | 3.26 | 5.586729 | 7.188236 | 63345.24005 | 3 |
| USNS Raymond\nFPO AE 09386 | 6.309435e+05 | 26354.10947 | 4.23 | 7.839388 | 5.040555 | 59982.19723 | 4 |
| | ••• | ••• | ••• | ••• | ••• | ••• | ••• |
| USNS Williams\nFPO AP 30153-7653 | 1.060194e+06 | 22837.36103 | 3.46 | 6.137356 | 7.830362 | 60567.94414 | 4995 |
| PSC 9258, Box 8489\nAPO AA 42991- 3352 | 1.482618e+06 | 25616.11549 | 4.02 | 6.576763 | 6.999135 | 78491.27543 | 4996 |

| Address | Price | Area Population | Avg. Area Number of Bedrooms | Avg. Area Number of Rooms | Avg. Area House Age | Avg. Area Income | |
|--|--------------|--------------------|---------------------------------------|---------------------------------------|------------------------------|---------------------|------|
| 4215 Tracy Garden Suite 076\nJoshualand, VA 01 | 1.030730e+06 | 33266.14549 | 2.13 | 4.805081 | 7.250591 | 63390.68689 | 4997 |
| USS Wallace\nFPO AE 73316 | 1.198657e+06 | 42625.62016 | 5.44 | 7.130144 | 5.534388 | 68001.33124 | 4998 |
| 37778 George Ridges Apt. 509\nEast Holly, NV 2 | 1.298950e+06 | 46501.28380 | 4.07 | 6.792336 | 5.992305 | 65510.58180 | 4999 |

5000 rows × 7 columns

head

In [11]:

to display first 8 dataset values
data.head(8)

Out[11]:

| t[11]: | | Avg. Area Income | Avg. Area House Age | Avg. Area Number of Rooms | Avg. Area Number of Bedrooms | Area Population | Price | Address |
|--------|---|---------------------|------------------------------|---------------------------------------|------------------------------------|--------------------|--------------|--|
| | 0 | 79545.45857 | 5.682861 | 7.009188 | 4.09 | 23086.80050 | 1.059034e+06 | 208 Michael Ferry Apt. 674\nLaurabury, NE 3701 |
| | 1 | 79248.64245 | 6.002900 | 6.730821 | 3.09 | 40173.07217 | 1.505891e+06 | 188 Johnson Views Suite 079\nLake Kathleen, CA |
| | 2 | 61287.06718 | 5.865890 | 8.512727 | 5.13 | 36882.15940 | 1.058988e+06 | 9127 Elizabeth Stravenue\nDanieltown, WI 06482 |
| | 3 | 63345.24005 | 7.188236 | 5.586729 | 3.26 | 34310.24283 | 1.260617e+06 | USS Barnett\nFPO AP 44820 |
| | 4 | 59982.19723 | 5.040555 | 7.839388 | 4.23 | 26354.10947 | 6.309435e+05 | USNS Raymond\nFPO AE 09386 |
| | 5 | 80175.75416 | 4.988408 | 6.104512 | 4.04 | 26748.42842 | 1.068138e+06 | 06039 Jennifer Islands Apt. 443\nTracyport, KS |
| | 6 | 64698.46343 | 6.025336 | 8.147760 | 3.41 | 60828.24909 | 1.502056e+06 | 4759 Daniel Shoals Suite 442\nNguyenburgh, CO |
| | 7 | 78394.33928 | 6.989780 | 6.620478 | 2.42 | 36516.35897 | 1.573937e+06 | 972 Joyce Viaduct\nLake William, TN 17778-6483 |

info

```
In [10]:
```

```
# to identify missing values
data.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 |

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

describe

In [8]:

to display summary of the dataset
data.describe()

Out[8]:

| | 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% | 61480.562390 | 5.322283 | 6.299250 | 3.140000 | 29403.928700 | 9.975771e+05 |
| 50% | 68804.286405 | 5.970429 | 7.002902 | 4.050000 | 36199.406690 | 1.232669e+06 |
| 75% | 75783.338665 | 6.650808 | 7.665871 | 4.490000 | 42861.290770 | 1.471210e+06 |
| max | 107701.748400 | 9.519088 | 10.759588 | 6.500000 | 69621.713380 | 2.469066e+06 |

columns

```
In [9]:  # to display headings of the dataset
     data.columns
```

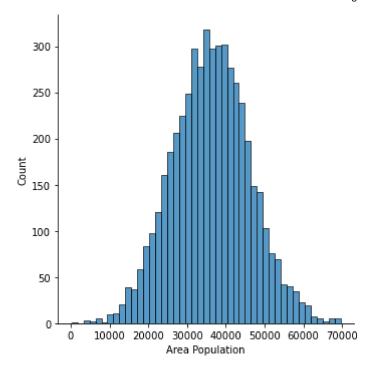
EDA and Visualization

In [12]: sns.pairplot(data) Out[12]: <seaborn.axisgrid.PairGrid at 0x2e63fa16d60> 80000 40000 20000 Avg. Area House Age Avg. 60000 Area Population 00000 2.0 0.5

distribution plot

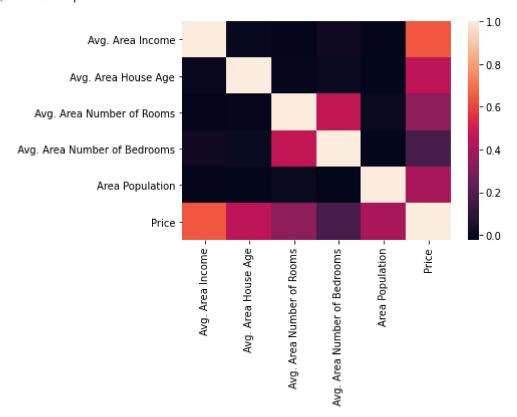
In [34]: sns.displot(data["Area Population"])

Out[34]: <seaborn.axisgrid.FacetGrid at 0x2e642511a00>



correlation

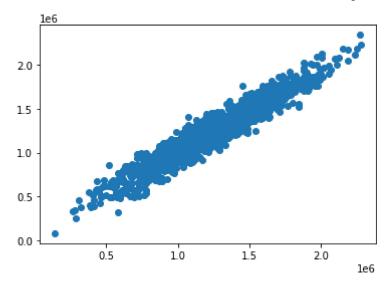
Out[24]: <AxesSubplot:>



To train the model-Model Building

we are going to train Linear Regression model; we need to split out data into two variables x and y where x is independent variable and y is dependent variable on x (output) we could ignore address column as it is not required for our model

```
In [25]:
           x=da[['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of Rooms',
                  'Avg. Area Number of Bedrooms', 'Area Population']]
          y=da['Price']
In [26]:
           # to split my dataset into training and test data
          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 [28]:
           from sklearn.linear model import LinearRegression
           lr= LinearRegression()
          lr.fit(x_train,y_train)
Out[28]: LinearRegression()
In [31]:
           print(lr.intercept_)
          -2641289.1850665007
In [30]:
           coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
           coeff
Out[30]:
                                        Co-efficient
                     Avg. Area Income
                                          21.634539
                   Avg. Area House Age 166212.262248
             Avg. Area Number of Rooms
                                     120882.774731
          Avg. Area Number of Bedrooms
                                        1347.590007
                       Area Population
                                          15.104383
In [32]:
           prediction=lr.predict(x_test)
           plt.scatter(y_test,prediction)
Out[32]: <matplotlib.collections.PathCollection at 0x2e6458cb5e0>
```



In [33]: print(lr.score(x_test,y_test))

0.9132511977358058