In [136]:

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

In [137]:

```
a=pd.read_csv(r"C:\Users\user\Downloads\USA_Housing.csv")
a
```

Out[137]:

·	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income	
208 Michael 674\nLaur	1.059034e+06	23086.800503	4.09	7.009188	5.682861	79545.458574	0
188 Johns Suite C Kathls	1.505891e+06	40173.072174	3.09	6.730821	6.002900	79248.642455	1
9127 Stravenue\nDa W	1.058988e+06	36882.159400	5.13	8.512727	5.865890	61287.067179	2
USS Barnett\	1.260617e+06	34310.242831	3.26	5.586729	7.188236	63345.240046	3
USNS Raymo	6.309435e+05	26354.109472	4.23	7.839388	5.040555	59982.197226	4
USNS Willia AP 30	1.060194e+06	22837.361035	3.46	6.137356	7.830362	60567.944140	4995
PSC (8489\nAPO <i>F</i>	1.482618e+06	25616.115489	4.02	6.576763	6.999135	78491.275435	4996
4215 Trac Suite 076\nJo	1.030730e+06	33266.145490	2.13	4.805081	7.250591	63390.686886	4997
USS Wallace	1.198657e+06	42625.620156	5.44	7.130144	5.534388	68001.331235	4998
37778 Georç Apt. 509\nE	1.298950e+06	46501.283803	4.07	6.792336	5.992305	65510.581804	4999

5000 rows × 7 columns

In [138]:

a=a.head(10)

Out[138]:

Ad	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income	
208 Michael Ferr 674\nLaurabu 3	1.059034e+06	23086.800503	4.09	7.009188	5.682861	79545.458574	0
188 Johnson Suite 079∖ı Kathleen,	1.505891e+06	40173.072174	3.09	6.730821	6.002900	79248.642455	1
9127 Eliz Stravenue\nDanie WI 06	1.058988e+06	36882.159400	5.13	8.512727	5.865890	61287.067179	2
USS Barnett\nFF	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
06039 Jennifer Is Apt. 443\nTrac	1.068138e+06	26748.428425	4.04	6.104512	4.988408	80175.754159	5
4759 Daniel S 442\nNguyenburg	1.502056e+06	60828.249085	3.41	8.147760	6.025336	64698.463428	6
972 Viaduct\nLake W TN 17778	1.573937e+06	36516.358972	2.42	6.620478	6.989780	78394.339278	7
USS Gilbert\nFF	7.988695e+05	29387.396003	2.30	6.393121	5.362126	59927.660813	8
Unit 944 0958\nDPO AE (1.545155e+06	40149.965749	6.10	8.167688	4.423672	81885.927184	9
—							4 6

In [139]:

```
# to find
a.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
Data columns (total 7 columns):
```

#	Column	Non-Null Count	Dtype
0	Avg. Area Income	10 non-null	float64
1	Avg. Area House Age	10 non-null	float64
2	Avg. Area Number of Rooms	10 non-null	float64
3	Avg. Area Number of Bedrooms	10 non-null	float64
4	Area Population	10 non-null	float64
5	Price	10 non-null	float64
6	Address	10 non-null	object

dtypes: float64(6), object(1)
memory usage: 688.0+ bytes

In [140]:

```
# to display summary of statastic
a.describe()
```

Out[140]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
count	10.000000	10.000000	10.000000	10.000000	10.000000	1.000000e+01
mean	70849.075034	5.756976	7.111241	3.807000	35443.678261	1.200363e+06
std	9631.232526	0.866393	0.996334	1.177908	10754.822720	3.315477e+05
min	59927.660813	4.423672	5.586729	2.300000	23086.800503	6.309435e+05
25%	61801.610396	5.120947	6.449960	3.132500	27408.170319	1.058999e+06
50%	71546.401353	5.774376	6.870005	3.725000	35413.300902	1.164377e+06
75%	79471.254544	6.019727	8.070667	4.195000	39333.014162	1.504932e+06
max	81885.927184	7.188236	8.512727	6.100000	60828.249085	1.573937e+06

In [141]:

```
# to display colum heading
a.columns
```

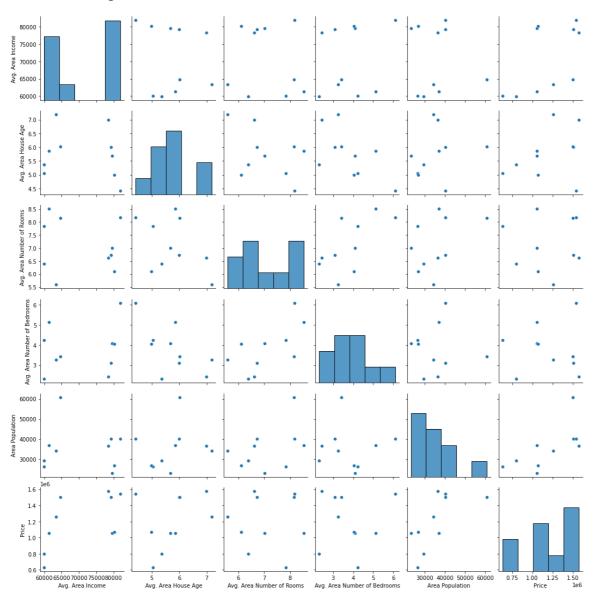
Out[141]:

In [142]:

sns.pairplot(a)

Out[142]:

<seaborn.axisgrid.PairGrid at 0x243f0b3eb20>

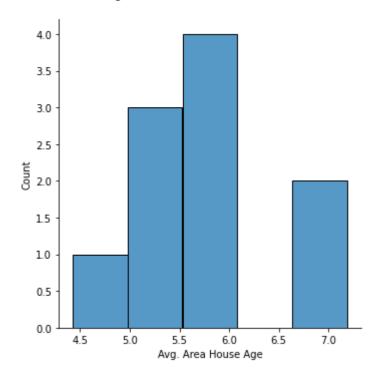


In [143]:

sns.displot(a["Avg. Area House Age"])

Out[143]:

<seaborn.axisgrid.FacetGrid at 0x243f2c21d60>



In [144]:

Out[144]:

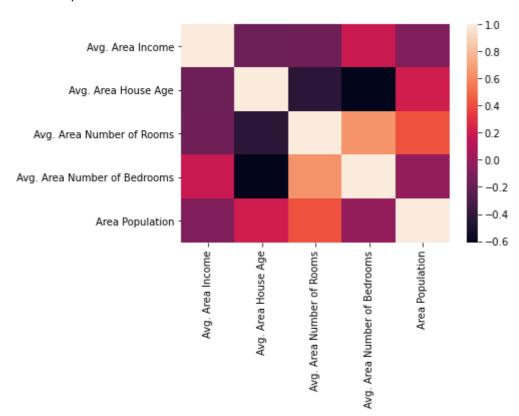
	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population
0	79545.458574	5.682861	7.009188	4.09	23086.800503
1	79248.642455	6.002900	6.730821	3.09	40173.072174
2	61287.067179	5.865890	8.512727	5.13	36882.159400
3	63345.240046	7.188236	5.586729	3.26	34310.242831
4	59982.197226	5.040555	7.839388	4.23	26354.109472
5	80175.754159	4.988408	6.104512	4.04	26748.428425
6	64698.463428	6.025336	8.147760	3.41	60828.249085
7	78394.339278	6.989780	6.620478	2.42	36516.358972
8	59927.660813	5.362126	6.393121	2.30	29387.396003
9	81885.927184	4.423672	8.167688	6.10	40149.965749

In [145]:

```
sns.heatmap(b.corr())
```

Out[145]:

<AxesSubplot:>



In [147]:

In [148]:

```
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 [149]:

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

Out[149]:

LinearRegression()

In [150]:

lr.intercept_

Out[150]:

5.1514348342607263e-14

In [151]:

coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff

Out[151]:

Co-efficient

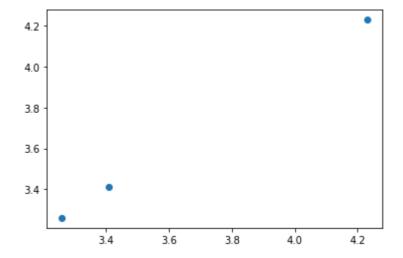
Avg. Area Income 9.344822e-20
Avg. Area House Age 3.142619e-16
Avg. Area Number of Rooms -5.297074e-16
Avg. Area Number of Bedrooms 1.000000e+00
Area Population -1.651153e-18

In [152]:

```
prediction = lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[152]:

<matplotlib.collections.PathCollection at 0x243f31f3610>



In [153]:

```
lr.score(x_test,y_test)
```

Out[153]:

1.0

```
In [154]:
lr.score(x_train,y_train)
Out[154]:
1.0
In [155]:
from sklearn.linear_model import Ridge,Lasso
In [156]:
rr=Ridge(alpha=10)
rr.fit(x_test,y_test)
Out[156]:
Ridge(alpha=10)
In [157]:
rr.score(x_test,y_test)
Out[157]:
0.999999999930226
In [158]:
la=Lasso(alpha=10)
la.fit(x_test,y_test)
Out[158]:
Lasso(alpha=10)
In [159]:
la.score(x_test,y_test)
Out[159]:
0.9993347523195334
In [160]:
from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)
Out[160]:
ElasticNet()
```

```
In [161]:
en.coef_
Out[161]:
array([ 1.33700254e-05, -0.00000000e+00, 0.00000000e+00, 5.12123685e-01,
        2.10633304e-05])
In [162]:
en.intercept_
Out[162]:
0.19862727830626303
In [163]:
prediction=en.predict(x_test)
prediction
Out[163]:
array([3.43776594, 4.09123465, 3.72197928])
In [164]:
en.score(x_test,y_test)
Out[164]:
-0.38238127985761383
EVALUATION METRICS
In [165]:
from sklearn import metrics
In [166]:
print("Mean Absolute Error:", metrics.mean_absolute_error(y_test, prediction))
Mean Absolute Error: 0.45567377105552875
In [167]:
print("Mean Squared Error", metrics.mean_squared_error(y_test, prediction))
Mean Squared Error 0.2512554775101208
In [168]:
print("Root Mean Squared Error",np.sqrt(metrics.mean_squared_error(y_test,prediction)))
```

Root Mean Squared Error 0.5012539052317905

In []:		