EDA

importing libraries/modules

```
In [5]: !pip install xgboost
         Requirement already satisfied: xgboost in c:\users\weite\anaconda3\lib\site-packages
         (2.1.1)
         Requirement already satisfied: numpy in c:\users\weite\anaconda3\lib\site-packages (fi
         om xgboost) (1.26.4)
         Requirement already satisfied: scipy in c:\users\weite\anaconda3\lib\site-packages (fr
         om xgboost) (1.11.4)
In [52]: import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         import sklearn
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression
         from sklearn.tree import DecisionTreeRegressor
         from sklearn.ensemble import GradientBoostingRegressor
         from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
         import xgboost
         from xgboost import XGBRegressor
         import warnings
         warnings.filterwarnings('ignore')
```



loading dataset

In [53]: df = pd.read_csv('USA_Housing.csv')
df

Out[53]:

	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, N' 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 [54]: df.shape

Out[54]: (5000, 7)



Out[55]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
4990	52723.876555	5.452237	8.124571	6.39	14802.088438	4.795006e+05	86727 Kelly Plaza\nLake Veronica, IL 04474
4991	74102.191890	5.657841	7.683993	3.13	24041.270592	1.263721e+06	2871 John Lodge\nAmychester, GU 61734-5597
4992	87499.125743	6.403473	4.836091	4.02	40815.199679	1.568701e+06	Unit 2096 Bc 9559\nDPO AE 8098; 879
4993	69639.140896	5.007510	7.778375	6.05	54056.128430	1.381831e+06	5259 David Causeway Apt. 975\nSouth Alexstad,
4994	73060.846226	5.293682	6.312253	4.16	22695.695480	9.053549e+05	5224 Lamb Passage∖nNancystad, GA 16579
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

In [56]: df.head()

Out[56]:

	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

In [57]: 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 Avg. Area House Age 5000 non-null float64 1 Avg. Area Number of Rooms 5000 non-null float64 Avg. Area Number of Bedrooms 5000 non-null float64 Area Population 5000 non-null float64 4 5 Price 5000 non-null float64 Address 5000 non-null 6 object

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



In [58]: df.columns

In [59]: df.describe()

Out[59]:

	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 [60]: | df.nunique()

Out[60]: Avg. Area Income 5000

Avg. Area House Age 5000
Avg. Area Number of Rooms 5000
Avg. Area Number of Bedrooms 255
Area Population 5000
Price 5000
Address 5000

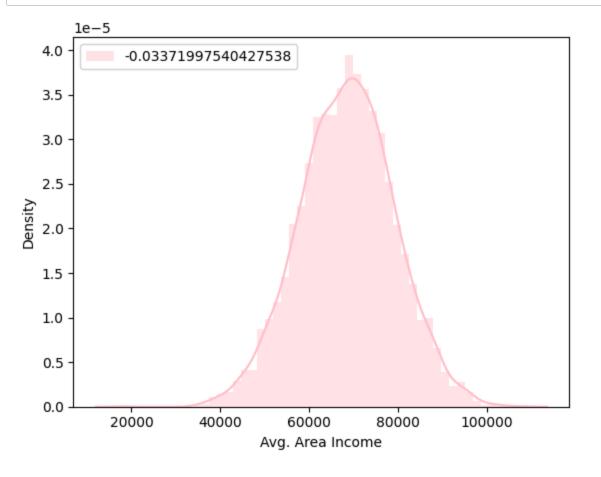
dtype: int64

```
In [61]: df['Avg. Area Income']
Out[61]: 0
                 79545.458574
         1
                 79248.642455
         2
                 61287.067179
         3
                 63345.240046
         4
                 59982.197226
         4995
                 60567.944140
         4996
                 78491.275435
         4997
                 63390.686886
         4998
                 68001.331235
                 65510.581804
         4999
         Name: Avg. Area Income, Length: 5000, dtype: float64
```

handling missig data

```
In [62]: |df.isnull().sum()
Out[62]: Avg. Area Income
                                          0
         Avg. Area House Age
                                          0
         Avg. Area Number of Rooms
                                          0
         Avg. Area Number of Bedrooms
                                          0
         Area Population
                                          0
         Price
                                          0
         Address
                                          0
         dtype: int64
In [63]: df['Avg. Area Income'].skew()
```

Out[63]: -0.03371997540427538



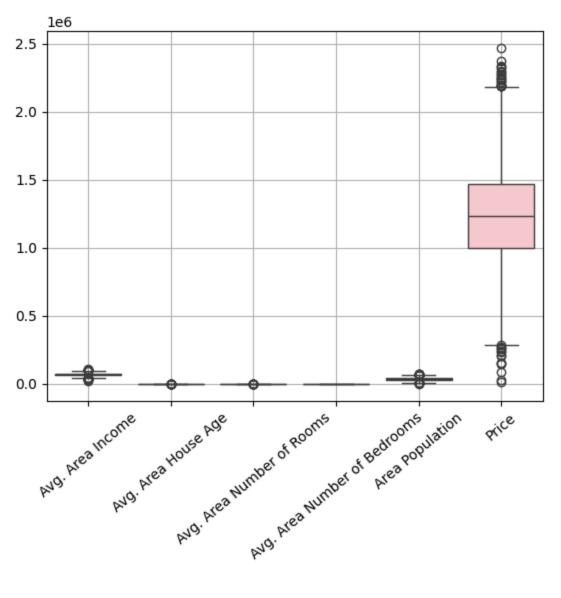
```
In [65]: print(df['Avg. Area Income'].mean())
         print(df['Avg. Area Income'].median())
         df['Avg. Area Income'].fillna(df['Avg. Area Income'].mean(), inplace=True)
         68583.10898395974
         68804.28640371616
In [66]: | df.isnull().sum()
Out[66]: Avg. Area Income
                                          0
         Avg. Area House Age
                                          0
         Avg. Area Number of Rooms
                                          0
         Avg. Area Number of Bedrooms
                                          0
         Area Population
         Price
                                          0
         Address
                                          0
         dtype: int64
```

duplicate data

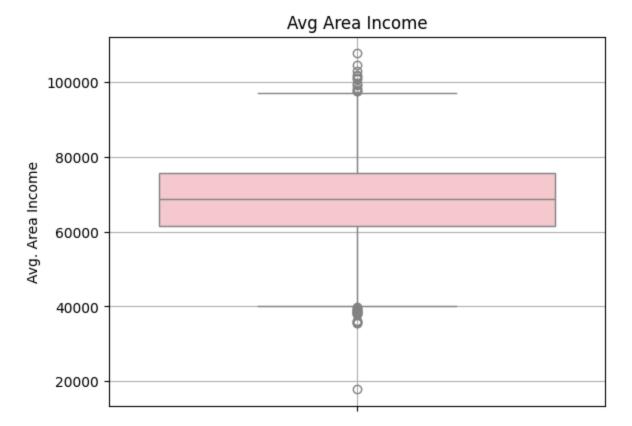
```
In [67]: print(df.duplicated().sum())
          df[df.duplicated()]
          0
Out[67]:
                Avg. Area
                              Avg. Area
                                          Avg. Area Number
                                                             Avg. Area Number of
                                                                                        Area
                                                                                             Price Address
                  Income
                             House Age
                                                 of Rooms
                                                                     Bedrooms
                                                                                   Population
In [68]: | df.drop_duplicates(keep='first', inplace=True)
```

outliers

```
In [69]: sns.boxplot(df,color='pink')
  plt.xticks(rotation=40)
  plt.grid()
  plt.show()
```



```
In [70]: sns.boxplot(df['Avg. Area Income'],color='pink')
    plt.title('Avg Area Income')
    plt.grid()
    plt.show()
```





Out[71]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
428	97112.361252	5.914725	6.133646	3.10	51470.067638	1.917584e+06	35841 Christine Shore\nWallmouth, NC 98956-5357
558	99629.013581	5.431863	7.351398	5.45	36950.739057	1.883481e+06	PSC 0420, Box 6835\nAPO AE 0835
693	107701.748378	7.143522	8.518608	3.29	37619.439929	2.332111e+06	41017 Eri Village\nJonathanport, CO 37205
962	101928.858060	4.829586	9.039382	4.08	22804.991935	1.938866e+06	856 Harris Centers Suite 940\nNicholasport, IL
1096	97548.310413	5.460973	6.609396	2.50	39089.415712	2.026303e+06	349 Sean Forges\nWhitakerville, WY 85138-4153
1734	104702.724257	5.575523	6.932106	3.22	22560.527135	1.742432e+06	14230 Douglas River Suite 570\nConniechester,
1891	101144.323930	6.350845	7.231771	3.09	35772.524007	2.007556e+06	233 Wilson Ranch Suite 086\nWest Christine, IN
2300	98468.253641	7.035383	6.629233	3.05	50676.312404	2.275455e+06	USNV Hoffman\nFPO AA 45311-5701
2719	101599.670580	7.798746	7.480512	6.39	37523.864670	2.370231e+06	52280 Steven Street\nRobertchester, IA 40405-0504
3483	97881.587279	5.034395	7.575905	5.46	37152.799341	1.859161e+06	01230 Peter Loop Suite 135\nEdwardstad, SC 67538
3541	102881.120902	6.471249	5.693536	3.12	21051.531294	1.754938e+06	784 Arnold Prairie Apt. 787\nJamesside, NM 04270
3798	97669.064491	6.888763	6.739379	4.14	43203.271060	2.102244e+06	0646 Martin Pass∖nKimberlyberg, GU 73453
3947	97076.164503	5.960820	6.606164	2.46	39915.656778	1.969194e+06	20831 Heather Common Suite 998\nLake Elizabeth
4087	100741.298585	5.870726	6.644853	4.33	26041.487616	1.644923e+06	5155 Virginia Station\nFischerberg, PA 75078-8129
4400	99317.823145	5.495861	7.182721	6.03	50350.352292	2.219724e+06	87549 Garcia Path\nSolomonside, ID 72003

In [72]: len(df[df['Avg. Area Income'] > 97000])

In [73]: df[df['Avg. Area Income'] < 40000]



Out[73]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
							209 Natasha Stream
12	39033.809237	7.671755	7.250029	3.10	39220.361467	1.042814e+06	Suite 961\nHuffmanland, NE
39	17796.631190	4.949557	6.713905	2.50	47162.183643	3.023558e+05	9932 Eric Circles\nLake Martha, WY 34611-6127
411	36100.444227	5.778489	5.497450	2.29	44901.857338	5.995040e+05	842 Duane Brook Apt. 380\nMonicaview, AP 01639.
844	39411.652788	4.385845	7.047435	4.44	45851.398296	5.394834e+05	1085 Michele Glens Apt. 517\nSergiotown, VT 20367
1271	37971.207566	4.291224	5.807510	3.24	33267.767728	3.114052e+04	98398 Terrance Pines\nSouth Joshua, MT 00544-8919
1459	35963.330809	3.438547	8.264122	3.28	24435.777302	1.430274e+05	166 Terry Grove\nSouth Michaelhaven, PR 18054
1597	39294.036523	5.928585	5.960676	4.36	43183.516104	7.811375e+05	829 Bonnie Rue Suite 653\nYorkville, MP 35797
2025	38139.919045	5.577267	6.348068	2.13	45899.738402	7.237501e+05	2899 Katherine Junction\nNorth Richardview, SC
2092	35608.986237	6.935839	7.827589	6.35	20833.007623	4.493316e+05	652 Stanton Island\nAdamsview, VA 56957-9960
2242	38868.250311	6.965104	8.966906	4.22	25432.076773	7.590447e+05	86840 Jonathon Field\nStevenport, KY 45694-2395
2597	38734.005216	5.641762	6.297908	2.31	38890.892760	4.011486e+05	4317 Heather Port\nHicksland, SC 77008-9209
3069	35454.714659	6.855708	6.018647	4.50	59636.402553	1.077806e+06	Unit 4700 Box 1880\nDPO AP 18074
3144	38571.963670	7.425292	5.723009	3.47	47386.793614	9.684116e+05	6613 Cynthia Loop Suite 555\nLake Heatherstad,
3183	38122.524488	6.336109	7.762551	5.12	38067.552184	8.996093e+05	783 Stacey Glen\nWest Katherineberg, NV 67007
4449	39777.606906	5.804627	7.147719	3.16	38725.424303	6.960145e+05	5231 David Shoals\nEast Chaseland, MO 54930-5094
4716	38530.124478	4.265906	8.026969	4.47	67727.229051	1.267987e+06	14763 Jeffrey Islands\nAmytown, WV 47279-7543

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
4744	39653.770031	5.205089	6.951617	2.32	40275.599326	3.959013e+05	58979 Jimmy Place Apt. 907∖nEast Bryanbury, MN
4844	37908.675863	6.233813	7.252916	6.07	39632.079786	8.804028e+05	901 Peter Fort Apt. 157\nTaylorfort, ND 06249
4855	35797.323122	5.544221	7.795138	5.00	24844.200190	2.998630e+05	645 Mary Radial\nEast Roberto, CA 23652- 5430

In [74]: len(df[df['Avg. Area Income'] < 40000])</pre>

Out[74]: 19



In [75]: df[(df['Avg. Area Income'] > 97000) | (df['Avg. Area Income'] < 40000)]

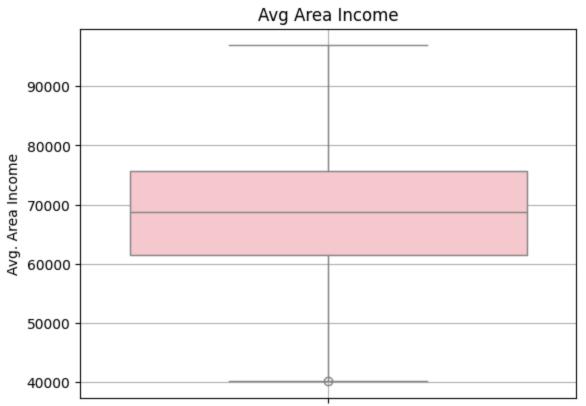


Out[75]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
12	39033.809237	7.671755	7.250029	3.10	39220.361467	1.042814e+06	209 Natasha Stream Suite 961\nHuffmanland, NE
39	17796.631190	4.949557	6.713905	2.50	47162.183643	3.023558e+05	9932 Eric Circles\nLake Martha, WY 34611-6127
411	36100.444227	5.778489	5.497450	2.29	44901.857338	5.995040e+05	842 Duane Brook Apt. 380\nMonicaview, AP 01639.
428	97112.361252	5.914725	6.133646	3.10	51470.067638	1.917584e+06	35841 Christine Shore\nWallmouth, NC 98956-5357
558	99629.013581	5.431863	7.351398	5.45	36950.739057	1.883481e+06	PSC 0420, Box 6835\nAPO AE 08359
693	107701.748378	7.143522	8.518608	3.29	37619.439929	2.332111e+06	41017 Eric Village\nJonathanport, CO 37205
844	39411.652788	4.385845	7.047435	4.44	45851.398296	5.394834e+05	1085 Michele Glens Apt. 517\nSergiotown, VT 20367
962	101928.858060	4.829586	9.039382	4.08	22804.991935	1.938866e+06	856 Harris Centers Suite 940\nNicholasport, IL
1096	97548.310413	5.460973	6.609396	2.50	39089.415712	2.026303e+06	349 Sean Forges\nWhitakerville, WY 85138-4153
1271	37971.207566	4.291224	5.807510	3.24	33267.767728	3.114052e+04	98398 Terrance Pines\nSouth Joshua, MT 00544-8919
1459	35963.330809	3.438547	8.264122	3.28	24435.777302	1.430274e+05	166 Terry Grove\nSouth Michaelhaven, PR 18054
1597	39294.036523	5.928585	5.960676	4.36	43183.516104	7.811375e+05	829 Bonnie Rue Suite 653\nYorkville, MP 35797
1734	104702.724257	5.575523	6.932106	3.22	22560.527135	1.742432e+06	14230 Douglas River Suite 570\nConniechester,
1891	101144.323930	6.350845	7.231771	3.09	35772.524007	2.007556e+06	233 Wilson Ranch Suite 086\nWest Christine, IN
2025	38139.919045	5.577267	6.348068	2.13	45899.738402	7.237501e+05	2899 Katherine Junction\nNorth Richardview, SC
2092	35608.986237	6.935839	7.827589	6.35	20833.007623	4.493316e+05	652 Stanton Island\nAdamsview, VA 56957-9960
2242	38868.250311	6.965104	8.966906	4.22	25432.076773	7.590447e+05	86840 Jonathon Field\nStevenport, KY 45694-2395
2300	98468.253641	7.035383	6.629233	3.05	50676.312404	2.275455e+06	USNV Hoffman\nFPO AA 45311-5701

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
2597	38734.005216	5.641762	6.297908	2.31	38890.892760	4.011486e+05	4317 Heather Port\nHicksland, SC 77008-9209
2719	101599.670580	7.798746	7.480512	6.39	37523.864670	2.370231e+06	52280 Steven Street\nRobertchester, IA 40405-0504
3069	35454.714659	6.855708	6.018647	4.50	59636.402553	1.077806e+06	Unit 4700 Box 1880\nDPO AP 18074
3144	38571.963670	7.425292	5.723009	3.47	47386.793614	9.684116e+05	6613 Cynthia Loo Suite 555\nLak Heatherstad,
3183	38122.524488	6.336109	7.762551	5.12	38067.552184	8.996093e+05	783 Stacey Glen\nWest Katherineberg, NV 67007
3483	97881.587279	5.034395	7.575905	5.46	37152.799341	1.859161e+06	01230 Peter Loop Suite 135\nEdwardstad, SC 67538
3541	102881.120902	6.471249	5.693536	3.12	21051.531294	1.754938e+06	784 Arnold Prairie Apt. 787\nJamesside, NM 04270
3798	97669.064491	6.888763	6.739379	4.14	43203.271060	2.102244e+06	0646 Martin Pass\nKimberlyberg, GU 73453
3947	97076.164503	5.960820	6.606164	2.46	39915.656778	1.969194e+06	20831 Heather Common Suite 998\nLake Elizabeth
4087	100741.298585	5.870726	6.644853	4.33	26041.487616	1.644923e+06	5155 Virginia Station∖nFischerberg, PA 75078-8129
4400	99317.823145	5.495861	7.182721	6.03	50350.352292	2.219724e+06	87549 Garcia Path\nSolomonside, ID 72003
4449	39777.606906	5.804627	7.147719	3.16	38725.424303	6.960145e+05	5231 David Shoals\nEast Chaseland, MO 54930- 5094
4716	38530.124478	4.265906	8.026969	4.47	67727.229051	1.267987e+06	14763 Jeffrey Islands\nAmytown, WV 47279-7543
4744	39653.770031	5.205089	6.951617	2.32	40275.599326	3.959013e+05	58979 Jimmy Place Apt. 907\nEast Bryanbury, MN
4844	37908.675863	6.233813	7.252916	6.07	39632.079786	8.804028e+05	901 Peter Fort Apt. 157\nTaylorfort, ND 06249
4855	35797.323122	5.544221	7.795138	5.00	24844.200190	2.998630e+05	645 Mary Radial\nEast Roberto, CA 23652- 5430

```
In [76]: len(df[(df['Avg. Area Income'] > 97000) | (df['Avg. Area Income'] < 40000)])</pre>
Out[76]: 34
In [77]: |df[(df['Avg. Area Income'] > 97000) | (df['Avg. Area Income'] < 40000)].index</pre>
Out[77]: Index([ 12,
                        39, 411, 428, 558, 693, 844, 962, 1096, 1271, 1459, 1597,
                1734, 1891, 2025, 2092, 2242, 2300, 2597, 2719, 3069, 3144, 3183, 3483,
                3541, 3798, 3947, 4087, 4400, 4449, 4716, 4744, 4844, 4855],
               dtype='int64')
In [78]: df.shape
Out[78]: (5000, 7)
In [79]: df.drop(index=[12, 39, 411, 428, 558, 693, 844, 962, 1096, 1271, 1459,
                     1597, 1734, 1891, 2025, 2092, 2242, 2300, 2597, 2719, 3069, 3144,
                     3183, 3483, 3541, 3798, 3947, 4087, 4400, 4449, 4716, 4744, 4844,
                     4855], axis=0, inplace=True)
In [80]: df.shape
Out[80]: (4966, 7)
In [81]: | sns.boxplot(df['Avg. Area Income'], color='pink')
         plt.title('Avg Area Income')
         plt.grid()
         plt.show()
```



```
In [82]: df[(df['Avg. Area Income'] > 97000) | (df['Avg. Area Income'] < 40000)]
```

Out[82]:

Avg. Area Income

Avg. Area House Age Avg. Area Number of Rooms

Avg. Area Number of Bedrooms

Area Propulation

Price Address

feature encoding or droping

df['Address'].value_counts() Out[83]: Address 208 Michael Ferry Apt. 674\nLaurabury, NE 37010-5101 1 89265 Robert Haven Apt. 492\nSmithburgh, FM 58565-1105 1 PSC 7034, Box 6131\nAPO AA 05662-4293 1 21042 Wilson Islands Suite 238\nFischerchester, MP 42425-4129 1 Unit 8831 Box 5748\nDPO AE 73012-7314 1 054 Carter Crescent Suite 674\nGlennport, WA 11140 1 8460 Kathleen Mission Apt. 482\nPort Amytown, KY 72016 1 3737 Hartman Rue\nReneestad, ID 69250-7718 1 3465 Latoya Well\nNelsonmouth, MI 55741-4287 1 37778 George Ridges Apt. 509\nEast Holly, NV 29290-3595 1 Name: count, Length: 4966, dtype: int64 In [84]: df.shape Out[84]: (4966, 7)df.drop(columns=['Address'],axis=1,inplace=True) In [85]:

In [86]: d

Out[86]:

Avg. Area Avg. Area Avg. Area Number Avg. Area Number of Area **Price House Age** Income of Rooms **Bedrooms Population** 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 4995 60567.944140 7.830362 6.137356 3.46 22837.361035 1.060194e+06 4996 4.02 78491.275435 6.999135 6.576763 25616.115489 1.482618e+06 4997 63390.686886 7.250591 4.805081 2.13 33266.145490 1.030730e+06 4998 68001.331235 5.534388 7.130144 5.44 42625.620156 1.198657e+06 4999 65510.581804 5.992305 6.792336 4.07 46501.283803 1.298950e+06

correlation

In [87]: df.corr()

Out[87]:

	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	-0.006274	-0.012836	0.018610	-0.011947	0.629465
Avg. Area House Age	-0.006274	1.000000	-0.008912	0.005030	-0.018811	0.455607
Avg. Area Number of Rooms	-0.012836	-0.008912	1.000000	0.463133	0.004193	0.340249
Avg. Area Number of Bedrooms	0.018610	0.005030	0.463133	1.000000	-0.021517	0.171979
Area Population	-0.011947	-0.018811	0.004193	-0.021517	1.000000	0.414990
Price	0.629465	0.455607	0.340249	0.171979	0.414990	1.000000

In [88]: df.corr()['Price']

Out[88]: Avg. Area Income 0.629465
Avg. Area House Age 0.455607
Avg. Area Number of Rooms 0.340249
Avg. Area Number of Bedrooms 0.171979
Area Population 0.414990
Price 1.000000

Name: Price, dtype: float64

model building

```
In [89]: X=df.iloc[:,:-1]
y=df['Price']
```

```
Χ
In [90]:
Out[90]:
                                                                               Avg. Area Number of
                        Avg. Area
                                   Avg. Area House
                                                      Avg. Area Number of
                                                                                                             Area
                         Income
                                                                  Rooms
                                                                                        Bedrooms
                                                                                                       Population
                                              Age
               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
                                                                                                     36882.159400
                                                                                              5.13
               3
                    63345.240046
                                          7.188236
                                                                 5.586729
                                                                                              3.26
                                                                                                     34310.242831
               4
                    59982.197226
                                          5.040555
                                                                 7.839388
                                                                                              4.23
                                                                                                     26354.109472
            4995
                    60567.944140
                                          7.830362
                                                                 6.137356
                                                                                              3.46
                                                                                                     22837.361035
            4996
                    78491.275435
                                          6.999135
                                                                 6.576763
                                                                                              4.02
                                                                                                     25616.115489
            4997
                    63390.686886
                                          7.250591
                                                                 4.805081
                                                                                              2.13
                                                                                                     33266.145490
            4998
                    68001.331235
                                                                 7.130144
                                                                                                     42625.620156
                                          5.534388
                                                                                              5.44
           4999
                    65510.581804
                                          5.992305
                                                                 6.792336
                                                                                              4.07
                                                                                                     46501.283803
           4966 rows × 5 columns
In [91]:
          У
Out[91]:
           0
                    1.059034e+06
           1
                    1.505891e+06
           2
                    1.058988e+06
           3
                    1.260617e+06
           4
                    6.309435e+05
           4995
                    1.060194e+06
           4996
                    1.482618e+06
           4997
                    1.030730e+06
           4998
                    1.198657e+06
           4999
                    1.298950e+06
           Name: Price, Length: 4966, dtype: float64
In [96]: |X_train,X_test,y_train,y_test=train_test_split(X,y, test_size=0.2,random_state=42)
In [97]:
          X_train.shape
Out[97]:
           (3972, 5)
In [98]:
          X_test.shape
Out[98]: (994, 5)
In [99]:
          y_test.shape
```

Out[99]: (994,)

```
In [100]: |y_train.shape
Out[100]: (3972,)
  In [ ]: # linear regression
In [101]: | lr=LinearRegression()
In [102]: |lr.fit(X_train,y_train)
Out[102]: LinearRegression()
           In a Jupyter environment, please rerun this cell to show the HTML representation or trust the
           notebook.
           On GitHub, the HTML representation is unable to render, please try loading this page with
           nbviewer.org.
In [104]: |print(f'training accuracy : { round(lr.score(X train,y train)*100)}%')
           print(f'test accuracy : { round(lr.score(X_test,y_test)*100)}%')
           training accuracy : 92%
           test accuracy : 92%
In [105]: # decision tree
In [106]: | dt=DecisionTreeRegressor(max_depth=9)
           dt.fit(X_train,y_train)
Out[106]: DecisionTreeRegressor(max_depth=9)
           In a Jupyter environment, please rerun this cell to show the HTML representation or trust the
           notebook.
           On GitHub, the HTML representation is unable to render, please try loading this page with
           nbviewer.org.
In [107]: |print(f'training accuracy : { round(dt.score(X_train,y_train)*100)}%')
           print(f'test accuracy : { round(dt.score(X_test,y_test)*100)}%')
           training accuracy : 92%
           test accuracy: 78%
In [108]: # gradient boosting regressor
In [111]: | gb=GradientBoostingRegressor()
           gb.fit(X_train,y_train)
Out[111]: GradientBoostingRegressor()
           In a Jupyter environment, please rerun this cell to show the HTML representation or trust the
           notebook.
```

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [113]: print(f'training accuracy : { round(gb.score(X_train,y_train)*100)}%')
print(f'test accuracy : { round(gb.score(X_test,y_test)*100)}%')
```

training accuracy : 93%
test accuracy : 91%

model training

In [114]: y_train_pred=lr.predict(X_train)
y_test_pred=lr.predict(X_test)

In [115]: X_train[:3]

Out[115]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population
2502	55306.141974	6.125857	6.410318	4.06	32142.947348
2770	84802.787605	5.101614	7.844812	6.04	35748.331474
2521	61909.041438	6.228343	6.593138	4.29	28953.925377

In [116]: y_train[:3]

Out[116]: 2502 6.479827e+05 2770 1.437984e+06 2521 1.063964e+06

Name: Price, dtype: float64

In [117]: y_train_pred[:3]

Out[117]: array([838521.68881121, 1537885.93430884, 972616.39243308])

slopes /coefficients

Out[119]: array([2.16406066e+01, 1.66104091e+05, 1.20673492e+05, 1.75557397e+03, 1.51408551e+01])

```
In [124]: pd.DataFrame(lr.coef_, index=X.columns ,columns=['Coefficients'])

Out[124]: Coefficients

Avg. Area Income 21.640607

Avg. Area House Age 166104.090506

Avg. Area Number of Rooms 120673.492264

Avg. Area Number of Bedrooms 1755.573974

Area Population 15.140855
```



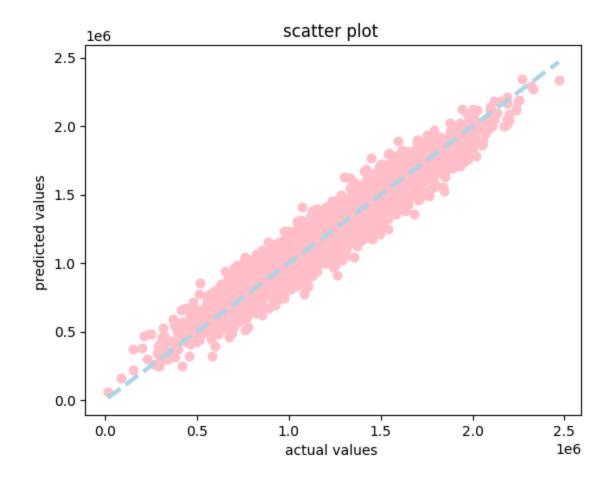
model evaluation

```
In [126]: #training data
In [127]: mean_squared_error(y_train,y_train_pred)
Out[127]: 10268623193.67436
In [128]: np.sqrt(mean_squared_error(y_train,y_train_pred))
Out[128]: 101334.21531582686
In [129]: mean_absolute_error(y_train,y_train_pred)
Out[129]: 81605.20894176647
In [131]: r2_score(y_train,y_train_pred)
Out[131]: 0.9161466729297922
```

```
In [138]: def evaluation(actual, predicted):
              mse=mean_squared_error(actual,predicted)
              rmse=np.sqrt(mean_squared_error(actual,predicted))
              mae=mean_absolute_error(actual,predicted)
              r2=r2_score(actual,predicted)
              print(f'MSE : {mse}')
              print(f'RMSE : {rmse}')
              print(f'MAE : {mae}')
              print(f'ACCURACY : {r2}')
              plt.scatter(actual, predicted , color='pink' , marker='o')
              plt.plot([actual.min(),actual.max()],[actual.min(),actual.max()],
                        'lightblue',lw=3,ls='--')
              plt.title('scatter plot ')
              plt.xlabel('actual values ')
              plt.ylabel('predicted values ')
              plt.show()
```

In [139]: evaluation(y_train,y_train_pred)

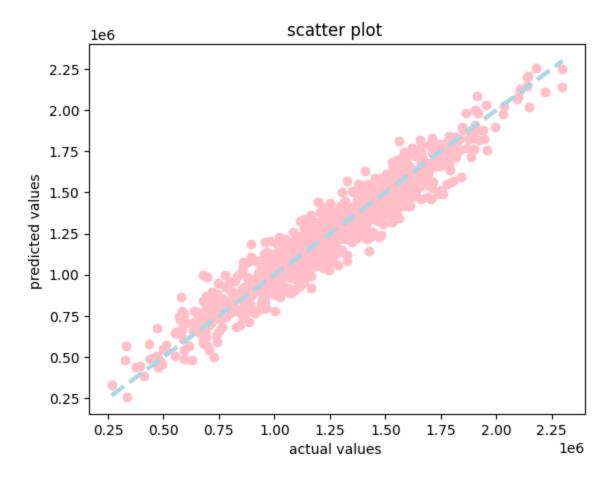
MSE : 10268623193.67436 RMSE : 101334.21531582686 MAE : 81605.20894176647 ACCURACY : 0.9161466729297922



In [140]: evaluation(y_test,y_test_pred)

MSE: 9769360588.245775 RMSE: 98840.07582072049 MAE: 79388.31116422128

ACCURACY: 0.9180015018430217



In []:

4