LinearRegression

```
import numpy as np
import pandas as pd
```

data collection

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as pp
import seaborn as sb
```

```
In [3]:
    df = pd.read_csv(r"C:\Users\user\Desktop\10_USA_Housing.csv")
    df
```

Addre:	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income		Out[3]:
208 Michael Ferry Ap 674\nLaurabury, N 3701	1.059034e+06	23086.800503	4.09	7.009188	5.682861	79545.458574	0	
188 Johnson View Suite 079\nLak Kathleen, CA	1.505891e+06	40173.072174	3.09	6.730821	6.002900	79248.642455	1	
9127 Elizabet Stravenue\nDanieltow WI 06482	1.058988e+06	36882.159400	5.13	8.512727	5.865890	61287.067179	2	
USS Barnett\nFPO A 4482	1.260617e+06	34310.242831	3.26	5.586729	7.188236	63345.240046	3	
USNS Raymond\nFP AE 0938	6.309435e+05	26354.109472	4.23	7.839388	5.040555	59982.197226	4	
			•••		•••		•••	
USNS Williams\nFP AP 30153-765	1.060194e+06	22837.361035	3.46	6.137356	7.830362	60567.944140	4995	
PSC 9258, Bc 8489\nAPO AA 4299° 335	1.482618e+06	25616.115489	4.02	6.576763	6.999135	78491.275435	4996	
4215 Tracy Garde Suite 076\nJoshualan VA 01	1.030730e+06	33266.145490	2.13	4.805081	7.250591	63390.686886	4997	
USS Wallace\nFPO A 7331	1.198657e+06	42625.620156	5.44	7.130144	5.534388	68001.331235	4998	
37778 George Ridge Apt. 509\nEast Holl NV 2	1.298950e+06	46501.283803	4.07	6.792336	5.992305	65510.581804	4999	

5000 rows × 7 columns

first 10 rows

In [4]:

df.head(10)

Out[4]:

•	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

In [6]:

df.info()

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

Column Non-Null Count Dtype
--- O Avg. Area Income 5000 non-null float64
1 Avg. Area House Age 5000 non-null float64

Avg. Area Number of Rooms 2 5000 non-null float64 Avg. Area Number of Bedrooms float64 3 5000 non-null float64 4 Area Population 5000 non-null float64 5 Price 5000 non-null 6 Address 5000 non-null object

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

In [7]:

df.describe()

Out[7]:

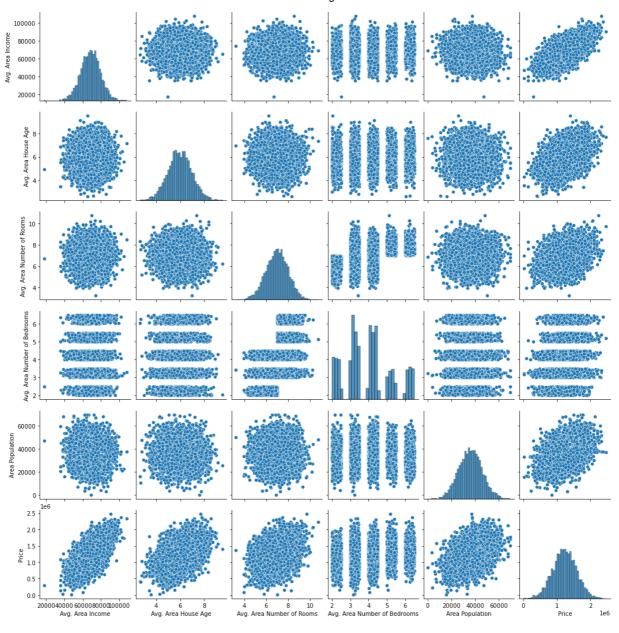
	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
count	5000.000000	5000.000000	5000.000000	5000.000000	5000.000000	5.000000e+03
mean	68583.108984	5.977222	6.987792	3.981330	36163.516039	1.232073e+06
std	10657.991214	0.991456	1.005833	1.234137	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 [9]: df.columns
```

Out[9]: Index(['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of Rooms', 'Avg. Area Number of Bedrooms', 'Area Population', 'Price', 'Address'], dtype='object')

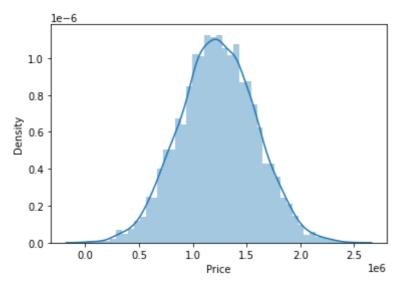
```
In [10]: sb.pairplot(df)
```

Out[10]: <seaborn.axisgrid.PairGrid at 0x2590edcc580>



In [16]: sb.distplot(df["Price"])

Out[16]: <AxesSubplot:xlabel='Price', ylabel='Density'>

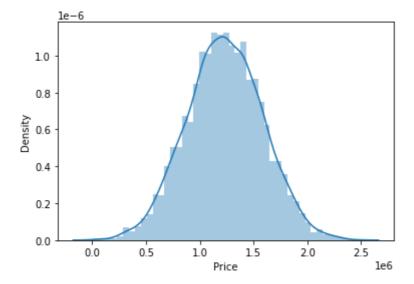


In [17]: sb.distplot(df["Price"])

7/27/23, 3:55 PM linear regression

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

Out[17]: <AxesSubplot:xlabel='Price', ylabel='Density'>



Out[20]:		Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Addres
	0	79545.458574	5.682861	7.009188	4.09	23086.800503	1.059034e+06	208 Michael Ferry Ap 674\nLaurabury, N 3701
	1	79248.642455	6.002900	6.730821	3.09	40173.072174	1.505891e+06	188 Johnson View Suite 079\nLak Kathleen, CA
	2	61287.067179	5.865890	8.512727	5.13	36882.159400	1.058988e+06	9127 Elizabet Stravenue\nDanieltow WI 06482
	3	63345.240046	7.188236	5.586729	3.26	34310.242831	1.260617e+06	USS Barnett\nFPO A 4482
	4	59982.197226	5.040555	7.839388	4.23	26354.109472	6.309435e+05	USNS Raymond\nFP AE 0938
	•••							
	4995	60567.944140	7.830362	6.137356	3.46	22837.361035	1.060194e+06	USNS Williams\nFP AP 30153-765
	4996	78491.275435	6.999135	6.576763	4.02	25616.115489	1.482618e+06	PSC 9258, Bc 8489\nAPO AA 4299° 335
								4215 Tracy Garde

4997 63390.686886 7.250591 4.805081

VA 01

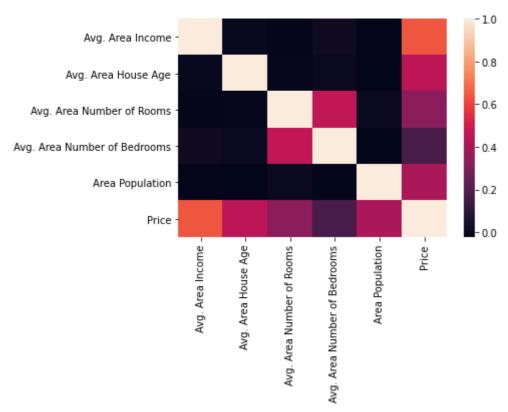
2.13 33266.145490 1.030730e+06 Suite 076\nJoshualan

Addres	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income	
USS Wallace\nFPO A 7331	1.198657e+06	42625.620156	5.44	7.130144	5.534388	68001.331235	4998
37778 George Ridge Apt. 509\nEast Holl NV 2	1.298950e+06	46501.283803	4.07	6.792336	5.992305	65510.581804	4999

5000 rows × 7 columns

```
In [21]: sb.heatmap(df1.corr())
```

Out[21]: <AxesSubplot:>



model building

```
lr = LinearRegression()
           lr.fit(x_train,y_train)
          LinearRegression()
Out[36]:
In [38]:
           print(lr.intercept_)
          -2.3283064365386963e-10
In [42]:
           coef = pd.DataFrame(lr.coef_,x.columns,columns=['Co_efficient'])
           coef
                                         Co_efficient
Out[42]:
                                        3.184052e-15
                      Avg. Area Income
                    Avg. Area House Age -5.061788e-11
             Avg. Area Number of Rooms
                                        7.996921e-11
          Avg. Area Number of Bedrooms
                                       1.311991e-12
                        Area Population
                                        9.261990e-15
                                  Price
                                       1.000000e+00
In [43]:
           print(lr.score(x_test,y_test))
          1.0
In [44]:
           prediction = lr.predict(x_test)
           pp.scatter(y_test,prediction)
Out[44]: <matplotlib.collections.PathCollection at 0x25913950f40>
              le6
          2.0
          1.5
          1.0
          0.5
          0.0
              0.0
                        0.5
                                  1.0
                                            1.5
                                                      2.0
                                                              le6
 In [ ]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as pp
```

In [9]: import seaborn as sb

C:\ProgramData\Anaconda3\lib\site-packages\IPython\core\interactiveshell.py:3165: Dt ypeWarning: Columns (47,73) have mixed types.Specify dtype option on import or set 1 ow memory=False.

has_raised = await self.run_ast_nodes(code_ast.body, cell_name,

Out[10]:

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	STheta	O2Sat	••
0	1	1	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0000A-3	0	10.500	33.4400	NaN	25.64900	NaN	
1	1	2	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0008A-3	8	10.460	33.4400	NaN	25.65600	NaN	
2	1	3	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0010A-7	10	10.460	33.4370	NaN	25.65400	NaN	
3	1	4	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0019A-3	19	10.450	33.4200	NaN	25.64300	NaN	
4	1	5	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0020A-7	20	10.450	33.4210	NaN	25.64300	NaN	
•••											••
864858	34404	864859	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0000A-7	0	18.744	33.4083	5.805	23.87055	108.74	

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	STheta	O2Sat	••
864859	34404	864860	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0002A-3	2	18.744	33.4083	5.805	23.87072	108.74	
864860	34404	864861	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0005A-3	5	18.692	33.4150	5.796	23.88911	108.46	
864861	34404	864862	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0010A-3	10	18.161	33.4062	5.816	24.01426	107.74	
864862	34404	864863	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0015A-3	15	17.533	33.3880	5.774	24.15297	105.66	••

In [11]: df.head(10)

864863 rows × 74 columns

Out[11]:

Cst_Cnt Btl_Cnt Sta_ID Depth_ID Depthm T_degC Salnty O2ml_L STheta O2Sat ... R_PHA

	GS 1_ G 110	200	J.u	- cp	эсри	acgc	Jumry		5111014	02541	•••	
0	1	1	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0000A-3	0	10.50	33.440	NaN	25.649	NaN		Ν
1	1	2	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0008A-3	8	10.46	33.440	NaN	25.656	NaN		Ν
2	1	3	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0010A-7	10	10.46	33.437	NaN	25.654	NaN		N
3	1	4	054.0 056.0	19- 4903CR-	19	10.45	33.420	NaN	25.643	NaN		Ν

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	STheta	O2Sat	•••	R_PHA
				HY-060- 0930- 05400560- 0019A-3								
4	1	5	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0020A-7	20	10.45	33.421	NaN	25.643	NaN		٨
5	1	6	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0030A-7	30	10.45	33.431	NaN	25.651	NaN		٨
6	1	7	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0039A-3	39	10.45	33.440	NaN	25.658	NaN		٨
7	1	8	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0050A-7	50	10.24	33.424	NaN	25.682	NaN		٨
8	1	9	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0058A-3	58	10.06	33.420	NaN	25.710	NaN		٨
9	1	10	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0075A-7	75	9.86	33.494	NaN	25.801	NaN		٨

10 rows × 74 columns

In [12]: | df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 864863 entries, 0 to 864862 Data columns (total 74 columns):

#	Column	Non-Null Count	Dtype
0	Cst_Cnt	864863 non-null	int64
1	Btl_Cnt	864863 non-null	int64
2	Sta_ID	864863 non-null	object
3	Depth_ID	864863 non-null	object

			Untitled21
4	Depthm	864863 non-null	int64
5	T_degC	853900 non-null	
6	Salnty	817509 non-null	
7	O2ml_L	696201 non-null	
8	STheta	812174 non-null	float64
9	02Sat	661274 non-null	float64
10	Oxy_μmol/Kg	661268 non-null	
11	Bt1Num	118667 non-null	float64
12	RecInd	864863 non-null	int64
13	T_prec	853900 non-null	
14	T_qual	23127 non-null	float64
15	S_prec	817509 non-null	float64
16	S_qual	74914 non-null	float64
17	P_qual	673755 non-null	float64
18	0_qual	184676 non-null	float64
19	SThtaq	65823 non-null	float64
20	02Satq	217797 non-null	float64
21	ChlorA	225272 non-null	float64
22	Chlqua	639166 non-null	
23	Phaeop	225271 non-null	
24	Phaqua	639170 non-null	float64
25	PO4uM	413317 non-null	float64
26	PO4q	451786 non-null	
	•		
27	SiO3uM	354091 non-null	
28	SiO3qu	510866 non-null	float64
29	NO2uM	337576 non-null	float64
30	NO2q	529474 non-null	
	•		
31	NO3uM	337403 non-null	float64
32	NO3q	529933 non-null	float64
33	NH3uM	64962 non-null	float64
34	NH3q	808299 non-null	float64
		14432 non-null	
35	C14As1		float64
36	C14A1p	12760 non-null	float64
37	C14A1q	848605 non-null	float64
38	C14As2	14414 non-null	float64
	C14A2p	12742 non-null	float64
39	•		
40	C14A2q	848623 non-null	float64
41	DarkAs	22649 non-null	float64
42	DarkAp	20457 non-null	float64
43	DarkAq	840440 non-null	float64
44	MeanAs	22650 non-null	float64
45	MeanAp	20457 non-null	float64
46	MeanAq	840439 non-null	float64
47	IncTim	14437 non-null	object
48	LightP	18651 non-null	float64
49	R_Depth	864863 non-null	float64
50	R_TEMP	853900 non-null	float64
51	R_POTEMP	818816 non-null	float64
52	_	817509 non-null	float64
	R_SALINITY		
53	R_SIGMA	812007 non-null	float64
54	R_SVA	812092 non-null	float64
55	R DYNHT	818206 non-null	float64
56	R 02	696201 non-null	float64
	_		
57	R_02Sat	666448 non-null	float64
58	R_SI03	354099 non-null	float64
59	R P04	413325 non-null	float64
60	 R_N03	337411 non-null	float64
	_		
61	R_N02	337584 non-null	float64
62	R_NH4	64982 non-null	float64
63	R_CHLA	225276 non-null	float64
64	R PHAEO	225275 non-null	float64
65	R_PRES	864863 non-null	int64
66	R_SAMP	122006 non-null	float64
67	DIC1	1999 non-null	float64
68	DIC2	224 non-null	float64
69	TA1	2084 non-null	float64
			float64
70	TA2	234 non-null	
71	pH2	10 non-null	float64
72	pH1	84 non-null	float64

73 DIC Quality Comment 55 non-null object dtypes: float64(65), int64(5), object(4)

memory usage: 488.3+ MB

In [13]:

df.describe()

Out[13]:		Cst_Cnt	Btl_Cnt	Depthm	T_degC	Salnty	O2ml_L	
	count	864863.000000	864863.000000	864863.000000	853900.000000	817509.000000	696201.000000	8
	mean	17138.790958	432432.000000	226.831951	10.799677	33.840350	3.392468	
	std	10240.949817	249664.587267	316.050259	4.243825	0.461843	2.073256	
	min	1.000000	1.000000	0.000000	1.440000	28.431000	-0.010000	
	25%	8269.000000	216216.500000	46.000000	7.680000	33.488000	1.360000	
	50%	16848.000000	432432.000000	125.000000	10.060000	33.863000	3.440000	
	75%	26557.000000	648647.500000	300.000000	13.880000	34.196900	5.500000	
	max	34404.000000	864863.000000	5351.000000	31.140000	37.034000	11.130000	

8 rows × 70 columns

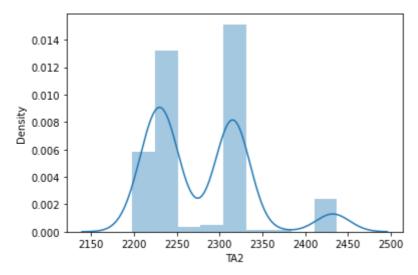
In [14]: df.columns

In [15]:

sb.distplot(df["TA2"])

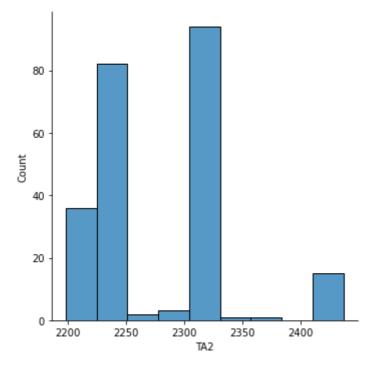
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[15]: <AxesSubplot:xlabel='TA2', ylabel='Density'>



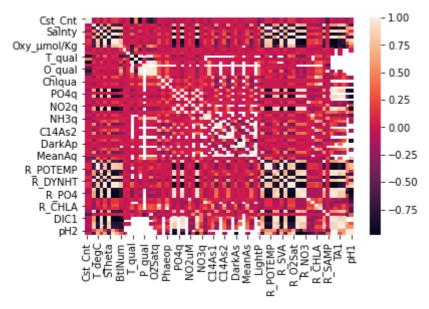
```
In [17]: sb.displot(df["TA2"])
```

Out[17]: <seaborn.axisgrid.FacetGrid at 0x21218412160>



```
In [19]: sb.heatmap(df.corr())
```

Out[19]: <AxesSubplot:>



In []:

```
In [26]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as pp
In [27]:
           import seaborn as sb
In [29]:
           df = pd.read_csv(r"C:\Users\user\Desktop\8_BreastCancerPrediction.csv")
Out[29]:
                      id diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mea
                  842302
                                           17.99
                                                                                     1001.0
             0
                                 Μ
                                                         10.38
                                                                         122.80
                                                                                                      0.1184
                  842517
                                           20.57
                                                          17.77
                                                                         132.90
                                                                                     1326.0
                                                                                                      0.0847
             1
                                 Μ
             2 84300903
                                           19.69
                                                                         130.00
                                                                                     1203.0
                                                                                                      0.1096
                                                         21.25
                                 Μ
               84348301
                                 Μ
                                           11.42
                                                          20.38
                                                                          77.58
                                                                                      386.1
                                                                                                      0.1425
                84358402
                                           20.29
                                                                                                      0.1003
                                                          14.34
                                                                         135.10
                                                                                     1297.0
                                 Μ
                  926424
           564
                                           21.56
                                                          22.39
                                                                         142.00
                                                                                     1479.0
                                                                                                      0.1110
                                 M
           565
                  926682
                                           20.13
                                                          28.25
                                                                         131.20
                                                                                     1261.0
                                                                                                      0.0978
                                 Μ
           566
                  926954
                                 Μ
                                           16.60
                                                         28.08
                                                                         108.30
                                                                                      858.1
                                                                                                      0.0845
           567
                  927241
                                           20.60
                                                          29.33
                                                                         140.10
                                                                                     1265.0
                                                                                                      0.1178
                                 Μ
           568
                                            7.76
                                                                          47.92
                                                                                                      0.0526
                   92751
                                  В
                                                          24.54
                                                                                      181.0
          569 rows × 33 columns
In [30]:
           df.head(10)
Out[30]:
                    id diagnosis radius mean texture_mean perimeter_mean area_mean smoothness_mean
```

		ulugilosis	radias_iiicaii	texture_incur	perimeter_inean	u.cucu.	5ootessear
0	842302	М	17.99	10.38	122.80	1001.0	0.11840
1	842517	М	20.57	17.77	132.90	1326.0	0.08474
2	84300903	М	19.69	21.25	130.00	1203.0	0.1096
3	84348301	М	11.42	20.38	77.58	386.1	0.1425
4	84358402	М	20.29	14.34	135.10	1297.0	0.10030

5 843786 Μ 12.45 15.70 82.57 477.1 0.12780 844359 18.25 19.98 0.09463 6 Μ 119.60 1040.0 84458202 13.71 20.83 90.20 577.9 0.11890 M 8 844981 13.00 21.82 87.50 519.8 0.12730 Μ 84501001

24.04

83.97

475.9

Μ

12.46

0.11860

10 rows × 33 columns

```
In [31]:
```

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 33 columns):

#	Column	Non-Null Count	Dtype
0	id	569 non-null	 int64
1	diagnosis	569 non-null	object
2	radius_mean	569 non-null	float64
3	texture_mean	569 non-null	float64
4	perimeter_mean	569 non-null	float64
5	area_mean	569 non-null	float64
6	smoothness_mean	569 non-null	float64
7	compactness_mean	569 non-null	float64
8	concavity_mean	569 non-null	float64
9	concave points_mean	569 non-null	float64
10	symmetry_mean	569 non-null	float64
11	fractal_dimension_mean	569 non-null	float64
12	radius_se	569 non-null	float64
13	texture_se	569 non-null	float64
14	perimeter_se	569 non-null	float64
15	area_se	569 non-null	float64
16	smoothness_se	569 non-null	float64
17	compactness_se	569 non-null	float64
18	concavity_se	569 non-null	float64
19	concave points_se	569 non-null	float64
20	symmetry_se	569 non-null	float64
21	<pre>fractal_dimension_se</pre>	569 non-null	float64
22	radius_worst	569 non-null	float64
23	texture_worst	569 non-null	float64
24	perimeter_worst	569 non-null	float64
25	area_worst	569 non-null	float64
26	smoothness_worst	569 non-null	float64
27	compactness_worst	569 non-null	float64
28	concavity_worst	569 non-null	float64
29	concave points_worst	569 non-null	float64
30	symmetry_worst	569 non-null	float64
31	<pre>fractal_dimension_worst</pre>	569 non-null	float64
32	Unnamed: 32	0 non-null	float64
	es: float64(31), int64(1)	, object(1)	
memo	ry usage: 146.8+ KB		

In [32]:

df.describe()

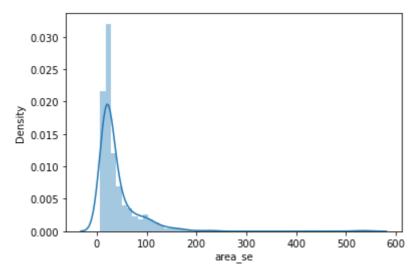
Out[32]:

	id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	(
count	5.690000e+02	569.000000	569.000000	569.000000	569.000000	569.000000	
mean	3.037183e+07	14.127292	19.289649	91.969033	654.889104	0.096360	
std	1.250206e+08	3.524049	4.301036	24.298981	351.914129	0.014064	
min	8.670000e+03	6.981000	9.710000	43.790000	143.500000	0.052630	
25%	8.692180e+05	11.700000	16.170000	75.170000	420.300000	0.086370	
50%	9.060240e+05	13.370000	18.840000	86.240000	551.100000	0.095870	
75%	8.813129e+06	15.780000	21.800000	104.100000	782.700000	0.105300	
max	9.113205e+08	28.110000	39.280000	188.500000	2501.000000	0.163400	

8 rows × 32 columns

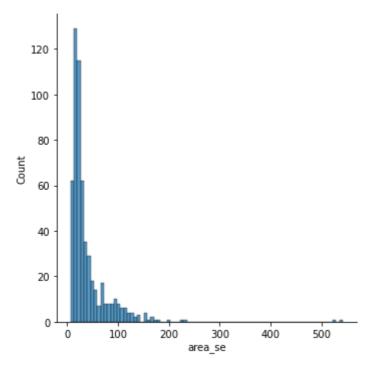
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[34]: <AxesSubplot:xlabel='area_se', ylabel='Density'>



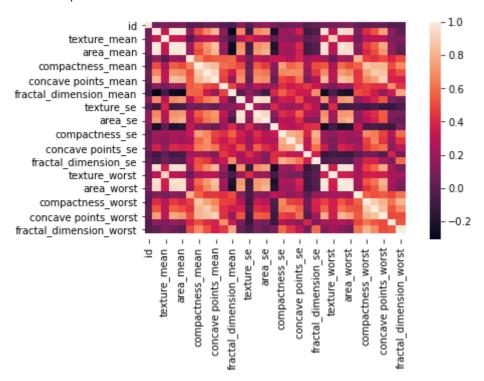
```
In [35]: sb.displot(df["area_se"])
```

Out[35]: <seaborn.axisgrid.FacetGrid at 0x21226b70700>



In [36]: sb.heatmap(df.corr())

Out[36]: <AxesSubplot:>



In []:

import numpy as np
import pandas as pd
import matplotlib.pyplot as pp

In [27]: import seaborn as sb

In [37]: df = pd.read_csv(r"C:\Users\user\Desktop\6_Salesworkload1.csv")
 df

Out[37]:

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLeas
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.
•••						•••			
7653	06.2017	9.0	Sweden	29650.0	Gothenburg	12.0	Checkout	6322.323	0.
7654	06.2017	9.0	Sweden	29650.0	Gothenburg	16.0	Customer Services	4270.479	0.
7655	06.2017	9.0	Sweden	29650.0	Gothenburg	11.0	Delivery	0	0.
7656	06.2017	9.0	Sweden	29650.0	Gothenburg	17.0	others	2224.929	0.
7657	06.2017	9.0	Sweden	29650.0	Gothenburg	18.0	all	39652.2	0.

7658 rows × 14 columns

In [38]:

df.head(10)

Out[38]:

•	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	
	1 0.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	398
	1 10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	82
	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	438
	3 10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	309 [,]

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	165
5	10.2016	1.0	United Kingdom	88253.0	London (I)	6.0	Meat	8270.316	0.0	1713
6	10.2016	1.0	United Kingdom	88253.0	London (I)	13.0	Food	16468.251	0.0	3107
7	10.2016	1.0	United Kingdom	88253.0	London (I)	7.0	Clothing	4698.471	0.0	213
8	10.2016	1.0	United Kingdom	88253.0	London (I)	8.0	Household	1183.272	0.0	54!
9	10.2016	1.0	United Kingdom	88253.0	London (I)	9.0	Hardware	2029.815	0.0	59;

In [39]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7658 entries, 0 to 7657
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype						
0	MonthYear	7658 non-null	object						
1	Time index	7650 non-null	float64						
2	Country	7650 non-null	object						
3	StoreID	7650 non-null	float64						
4	City	7650 non-null	object						
5	Dept_ID	7650 non-null	float64						
6	Dept. Name	7650 non-null	object						
7	HoursOwn	7650 non-null	object						
8	HoursLease	7650 non-null	float64						
9	Sales units	7650 non-null	float64						
10	Turnover	7650 non-null	float64						
11	Customer	0 non-null	float64						
12	Area (m2)	7650 non-null	object						
13	Opening hours	7650 non-null	object						
dtyp	es: float64(7),	object(7)							
memo	memory usage: 837.7+ KB								

In [40]:

df.describe()

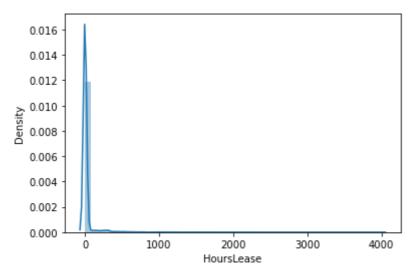
Out[40]:

	Time index	StoreID	Dept_ID	HoursLease	Sales units	Turnover	Customer
count	7650.000000	7650.000000	7650.000000	7650.000000	7.650000e+03	7.650000e+03	0.0
mean	5.000000	61995.220000	9.470588	22.036078	1.076471e+06	3.721393e+06	NaN
std	2.582158	29924.581631	5.337429	133.299513	1.728113e+06	6.003380e+06	NaN
min	1.000000	12227.000000	1.000000	0.000000	0.000000e+00	0.000000e+00	NaN
25%	3.000000	29650.000000	5.000000	0.000000	5.457125e+04	2.726798e+05	NaN
50%	5.000000	75400.500000	9.000000	0.000000	2.932300e+05	9.319575e+05	NaN
75 %	7.000000	87703.000000	14.000000	0.000000	9.175075e+05	3.264432e+06	NaN
max	9.000000	98422.000000	18.000000	3984.000000	1.124296e+07	4.271739e+07	NaN

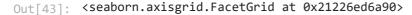
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn ing: `distplot` is a deprecated function and will be removed in a future version. Pl ease adapt your code to use either `displot` (a figure-level function with similar f lexibility) or `histplot` (an axes-level function for histograms).

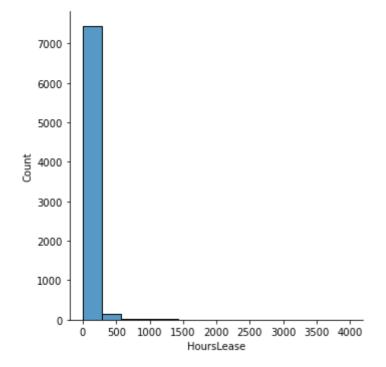
warnings.warn(msg, FutureWarning)

Out[42]: <AxesSubplot:xlabel='HoursLease', ylabel='Density'>



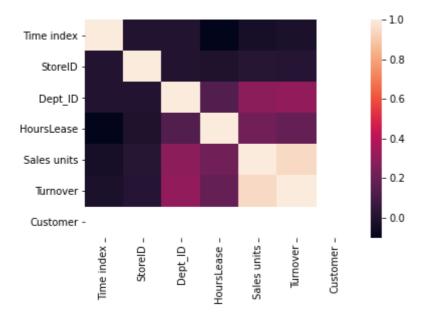
```
In [43]: sb.displot(df["HoursLease"])
```





```
In [44]: sb.heatmap(df.corr())
```





In []:

```
In [26]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as pp
In [27]:
           import seaborn as sb
In [45]:
           df = pd.read_csv(r"C:\Users\user\Desktop\3_Fitness-1.csv")
             Row Labels Sum of Jan Sum of Feb Sum of Mar Sum of Total Sales
Out[45]:
          0
                      Α
                              5.62%
                                         7.73%
                                                      6.16%
                                                                          75
          1
                      В
                              4.21%
                                        17.27%
                                                     19.21%
                                                                          160
          2
                      C
                              9.83%
                                        11.60%
                                                      5.17%
                                                                          101
          3
                      D
                                                      7.88%
                              2.81%
                                        21.91%
                                                                          127
          4
                      Ε
                             25.28%
                                        10.57%
                                                     11.82%
                                                                          179
          5
                      F
                              8.15%
                                        16.24%
                                                     18.47%
                                                                          167
          6
                      G
                             18.54%
                                         8.76%
                                                     17.49%
                                                                          171
          7
                      Н
                             25.56%
                                          5.93%
                                                     13.79%
                                                                          170
             Grand Total
                            100.00%
                                       100.00%
                                                    100.00%
                                                                         1150
In [46]:
           df.head(10)
Out[46]:
             Row Labels Sum of Jan Sum of Feb Sum of Mar Sum of Total Sales
          0
                              5.62%
                                         7.73%
                                                      6.16%
                                                                          75
                      Α
          1
                      В
                              4.21%
                                         17.27%
                                                     19.21%
                                                                          160
          2
                      C
                              9.83%
                                        11.60%
                                                      5.17%
                                                                          101
          3
                      D
                              2.81%
                                        21.91%
                                                      7.88%
                                                                          127
                                                     11.82%
                      Ε
                             25.28%
                                         10.57%
                                                                          179
          5
                      F
                              8.15%
                                         16.24%
                                                     18.47%
                                                                          167
          6
                      G
                             18.54%
                                         8.76%
                                                     17.49%
                                                                          171
          7
                             25.56%
                                          5.93%
                                                     13.79%
                                                                          170
             Grand Total
                            100.00%
                                       100.00%
                                                    100.00%
                                                                         1150
In [47]:
           df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 9 entries, 0 to 8
          Data columns (total 5 columns):
                                      Non-Null Count Dtype
           #
                Column
           0
                Row Labels
                                      9 non-null
                                                        object
                Sum of Jan
                                      9 non-null
                                                        object
           1
                Sum of Feb
                                      9 non-null
                                                        object
```

3 Sum of Mar 9 non-null object 4 Sum of Total Sales 9 non-null int64

dtypes: int64(1), object(4)
memory usage: 488.0+ bytes

```
In [48]: df.describe()
```

[48]:		Sum of Total Sales
	count	9.000000
	mean	255.555556
	std	337.332963
	min	75.000000
	25%	127.000000
	50%	167.000000
	75%	171.000000
	max	1150.000000

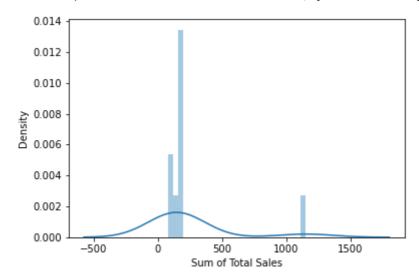
Out

```
In [49]: df.columns
```

```
In [50]: sb.distplot(df["Sum of Total Sales"])
```

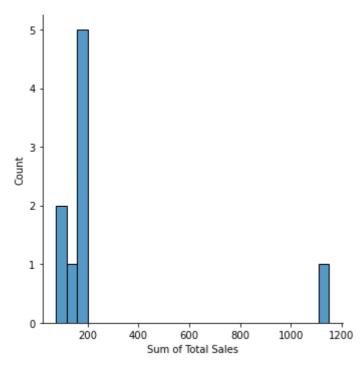
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[50]: <AxesSubplot:xlabel='Sum of Total Sales', ylabel='Density'>



```
In [51]: sb.displot(df["Sum of Total Sales"])
```

Out[51]: <seaborn.axisgrid.FacetGrid at 0x212270c2310>



In [54]: sb.heatmap(df.corr())

Out[54]: <AxesSubplot:>



In []:

```
In [26]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as pp
In [27]:
           import seaborn as sb
In [55]:
           df = pd.read_csv(r"C:\Users\user\Desktop\4_drug200.csv")
Out[55]:
                               BP
                                  Cholesterol Na_to_K
                                                        Drug
               Age
                    Sex
            0
                 23
                       F
                             HIGH
                                        HIGH
                                                25.355
                                                       drugY
            1
                                                13.093 drugC
                 47
                      Μ
                             LOW
                                        HIGH
            2
                 47
                      Μ
                             LOW
                                        HIGH
                                                10.114 drugC
            3
                 28
                       F
                         NORMAL
                                                 7.798
                                                       drugX
                                        HIGH
            4
                 61
                       F
                             LOW
                                        HIGH
                                                18.043
                                                       drugY
          195
                 56
                       F
                             LOW
                                        HIGH
                                                11.567 drugC
          196
                 16
                      Μ
                             LOW
                                        HIGH
                                                12.006 drugC
                         NORMAL
                                                 9.894 drugX
          197
                 52
                      M
                                        HIGH
          198
                         NORMAL
                                     NORMAL
                                                14.020 drugX
                 23
                      M
          199
                 40
                       F
                             LOW
                                     NORMAL
                                                11.349 drugX
         200 rows × 6 columns
In [56]:
           df.head(10)
Out[56]:
                             BP
                                 Cholesterol Na_to_K
             Age
                  Sex
                                                      Drug
          0
               23
                    F
                           HIGH
                                      HIGH
                                              25.355
                                                     drugY
          1
               47
                    Μ
                           LOW
                                      HIGH
                                              13.093 drugC
          2
                           LOW
                                      HIGH
                                              10.114 drugC
               47
                    Μ
                                               7.798
          3
               28
                     F
                       NORMAL
                                      HIGH
                                                     drugX
          4
                    F
                           LOW
                                      HIGH
                                              18.043
                                                     drugY
               61
          5
               22
                       NORMAL
                                      HIGH
                                               8.607
                                                     drugX
          6
               49
                       NORMAL
                                      HIGH
                                              16.275
                                                     drugY
          7
                           LOW
                                      HIGH
                                              11.037 drugC
               41
                    Μ
          8
               60
                    Μ
                       NORMAL
                                      HIGH
                                              15.171
                                                     drugY
          9
               43
                           LOW
                                   NORMAL
                                              19.368 drugY
                    Μ
In [57]:
           df.info()
```

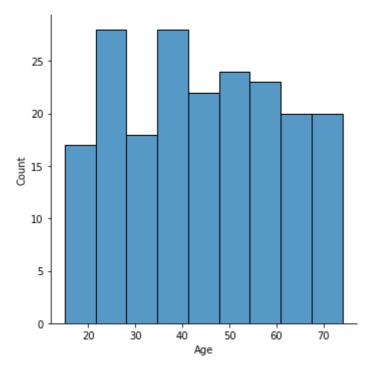
```
<class 'pandas.core.frame.DataFrame'>
          RangeIndex: 200 entries, 0 to 199
          Data columns (total 6 columns):
                             Non-Null Count Dtype
           #
               Column
           0
                             200 non-null
                                              int64
               Age
           1
                             200 non-null
                                              object
               Sex
           2
               BP
                             200 non-null
                                              object
           3
                                              object
               Cholesterol 200 non-null
           4
                                              float64
               Na_to_K
                             200 non-null
           5
                                              object
               Drug
                             200 non-null
          dtypes: float64(1), int64(1), object(4)
          memory usage: 9.5+ KB
In [58]:
           df.describe()
Out[58]:
                              Na_to_K
                      Age
          count 200.000000
                            200.000000
                 44.315000
                             16.084485
          mean
            std
                  16.544315
                             7.223956
                  15.000000
                             6.269000
            min
           25%
                  31.000000
                             10.445500
           50%
                  45.000000
                             13.936500
           75%
                  58.000000
                             19.380000
                 74.000000
                             38.247000
           max
In [59]:
           df.columns
Out[59]: Index(['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K', 'Drug'], dtype='object')
           sb.distplot(df["Age"])
Out[61]: <AxesSubplot:xlabel='Age', ylabel='Density'>
            0.0200
            0.0175
            0.0150
         0.0125
0.0100
            0.0075
            0.0050
            0.0025
            0.0000
                               20
                                                             80
                                         40
                                                   60
                                          Age
```

sb.displot(df["Age"])

In [61]:

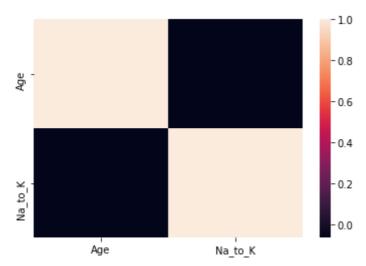
In [62]:

Out[62]: <seaborn.axisgrid.FacetGrid at 0x21218cd3070>



In [63]: sb.heatmap(df.corr())

Out[63]: <AxesSubplot:>



In []:

7/27/23, 4:19 PM

```
Untitled21
In [26]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as pp
In [27]:
           import seaborn as sb
In [64]:
           df = pd.read_csv(r"C:\Users\user\Desktop\10_USA_Housing.csv")
Out[64]:
                                            Avg.
                                  Avg.
                                                  Avg. Area
                                           Area
                    Avg. Area
                                  Area
                                                   Number
                                                                    Area
                                        Number
                                                                                  Price
                                                                                                     Addres
                      Income
                                                         of
                                                               Population
                                 House
                                              of
                                                  Bedrooms
                                   Age
                                          Rooms
                                                                                         208 Michael Ferry Ap
              0 79545.458574 5.682861 7.009188
                                                       4.09
                                                            23086.800503 1.059034e+06
                                                                                           674\nLaurabury, N
                                                                                                       3701
                                                                                            188 Johnson View
              1 79248.642455 6.002900 6.730821
                                                       3.09 40173.072174 1.505891e+06
                                                                                               Suite 079\nLak
                                                                                                Kathleen, CA
                                                                                                9127 Elizabet
              2 61287.067179 5.865890 8.512727
                                                       5.13 36882.159400
                                                                         1.058988e+06
                                                                                       Stravenue\nDanieltow
                                                                                                   WI 06482
                                                                                          USS Barnett\nFPO A
                                                       3.26 34310.242831
                                                                         1.260617e+06
                 63345.240046 7.188236
                                       5.586729
```

4.23

3.46

7.839388

26354.109472 6.309435e+05

22837.361035 1.060194e+06

4.02 25616.115489 1.482618e+06

2.13 33266.145490 1.030730e+06

5.44 42625.620156 1.198657e+06

4.07 46501.283803 1.298950e+06

5000 rows × 7 columns

4995

4998

59982.197226 5.040555

60567.944140 7.830362 6.137356

4996 78491.275435 6.999135 6.576763

4997 63390.686886 7.250591 4.805081

4999 65510.581804 5.992305 6.792336

68001.331235 5.534388 7.130144

In [65]: df.head(10)

4482

AE 0938

USNS Raymond\nFP

USNS Williams\nFP

8489\nAPO AA 42991

4215 Tracy Garde Suite 076\nJoshualan

USS Wallace\nFPO A

37778 George Ridge

Apt. 509\nEast Holl

AP 30153-765

PSC 9258, Bc

335

VA 01

7331

NV 2

Out[65]:

	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

In [66]:

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
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

In [67]:

df.describe()

Out[67]:

	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

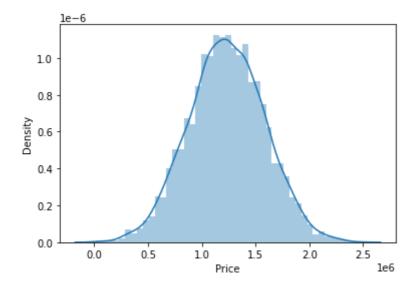
```
In [68]: df.columns
```

Out[68]: Index(['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of Rooms', 'Avg. Area Number of Bedrooms', 'Area Population', 'Price', 'Address'], dtype='object')

In [70]: sb.distplot(df["Price"])

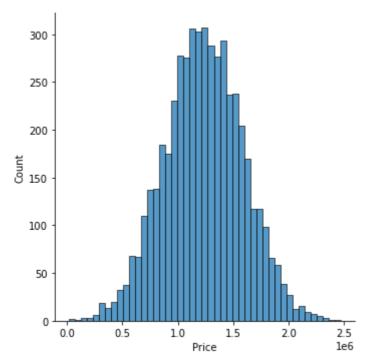
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[70]: <AxesSubplot:xlabel='Price', ylabel='Density'>



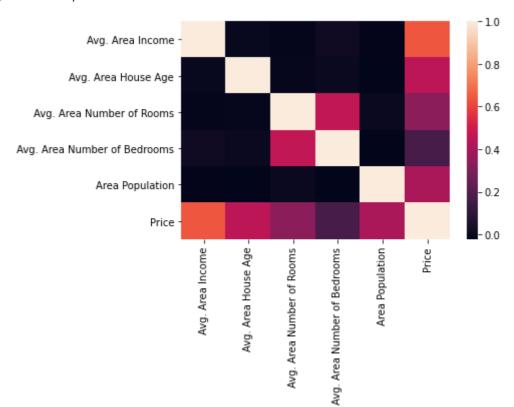
```
In [71]: sb.displot(df["Price"])
```

Out[71]: <seaborn.axisgrid.FacetGrid at 0x21218ed1f40>



In [72]: sb.heatmap(df.corr())

Out[72]: <AxesSubplot:>



In []:

```
In [26]:
            import numpy as np
            import pandas as pd
            import matplotlib.pyplot as pp
In [27]:
            import seaborn as sb
In [73]:
            df = pd.read_csv(r"C:\Users\user\Desktop\7_uber.csv")
                    Unnamed:
Out[73]:
                                              key fare_amount pickup_datetime pickup_longitude pickup_latit
                                       2015-05-07
                                                                       2015-05-07
                 0
                     24238194
                                                             7.5
                                                                                          -73.999817
                                                                                                           40.738
                                  19:52:06.0000003
                                                                      19:52:06 UTC
                                       2009-07-17
                                                                       2009-07-17
                 1
                     27835199
                                                             7.7
                                                                                          -73.994355
                                                                                                           40.728
                                  20:04:56.0000002
                                                                      20:04:56 UTC
                                       2009-08-24
                                                                       2009-08-24
                 2
                     44984355
                                                                                          -74.005043
                                                                                                           40.740
                                                            12.9
                                 21:45:00.00000061
                                                                      21:45:00 UTC
                                       2009-06-26
                                                                       2009-06-26
                     25894730
                                                             5.3
                                                                                          -73.976124
                                                                                                           40.790
                 3
                                  08:22:21.0000001
                                                                      08:22:21 UTC
                                       2014-08-28
                                                                       2014-08-28
                      17610152
                                                            16.0
                                                                                          -73.925023
                                                                                                           40.744
                                17:47:00.000000188
                                                                      17:47:00 UTC
                                       2012-10-28
                                                                       2012-10-28
           199995
                     42598914
                                                                                          -73.987042
                                                                                                           40.739
                                                             3.0
                                 10:49:00.00000053
                                                                      10:49:00 UTC
                                       2014-03-14
                                                                       2014-03-14
           199996
                     16382965
                                                             7.5
                                                                                          -73.984722
                                                                                                           40.736
                                  01:09:00.0000008
                                                                      01:09:00 UTC
                                       2009-06-29
                                                                       2009-06-29
           199997
                     27804658
                                                            30.9
                                                                                          -73.986017
                                                                                                           40.756
                                 00:42:00.00000078
                                                                      00:42:00 UTC
                                                                       2015-05-20
                                       2015-05-20
           199998
                     20259894
                                                            14.5
                                                                                          -73.997124
                                                                                                           40.725
                                  14:56:25.0000004
                                                                      14:56:25 UTC
                                       2010-05-15
                                                                       2010-05-15
           199999
                     11951496
                                                            14.1
                                                                                          -73.984395
                                                                                                           40.720
                                 04:08:00.00000076
                                                                      04:08:00 UTC
          200000 rows × 9 columns
In [74]:
            df.head(10)
Out[74]:
              Unnamed:
                                         key fare_amount pickup_datetime pickup_longitude pickup_latitude (
                                 2015-05-07
                                                                 2015-05-07
               24238194
                                                       7.5
                                                                                    -73.999817
                                                                                                      40.738354
                            19:52:06.0000003
                                                                19:52:06 UTC
                                                                 2009-07-17
                                 2009-07-17
               27835199
                                                                                    -73.994355
                                                                                                      40.728225
                                                       7.7
                            20:04:56.0000002
                                                                20:04:56 UTC
```

44984355

2009-08-24

21:45:00.00000061

2009-08-24

21:45:00 UTC

-74.005043

12.9

40.740770

	Unnamed: 0	key	fare_amount	pickup_datetime	pickup_longitude	pickup_latitude	(
3	25894730	2009-06-26 08:22:21.0000001	5.3	2009-06-26 08:22:21 UTC	-73.976124	40.790844	
4	17610152	2014-08-28 17:47:00.000000188	16.0	2014-08-28 17:47:00 UTC	-73.925023	40.744085	
5	44470845	2011-02-12 02:27:09.0000006	4.9	2011-02-12 02:27:09 UTC	-73.969019	40.755910	
6	48725865	2014-10-12 07:04:00.0000002	24.5	2014-10-12 07:04:00 UTC	-73.961447	40.693965	
7	44195482	2012-12-11 13:52:00.00000029	2.5	2012-12-11 13:52:00 UTC	0.000000	0.000000	
8	15822268	2012-02-17 09:32:00.00000043	9.7	2012-02-17 09:32:00 UTC	-73.975187	40.745767	
9	50611056	2012-03-29 19:06:00.000000273	12.5	2012-03-29 19:06:00 UTC	-74.001065	40.741787	

In [75]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200000 entries, 0 to 199999
Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	Unnamed: 0	200000 non-null	int64
1	key	200000 non-null	object
2	fare_amount	200000 non-null	float64
3	pickup_datetime	200000 non-null	object
4	pickup_longitude	200000 non-null	float64
5	pickup_latitude	200000 non-null	float64
6	dropoff_longitude	199999 non-null	float64
7	dropoff_latitude	199999 non-null	float64
8	passenger_count	200000 non-null	int64
4+,,,,	oc. float(4/5) int	(4/2) object(2)	

dtypes: float64(5), int64(2), object(2)

memory usage: 13.7+ MB

In [76]:

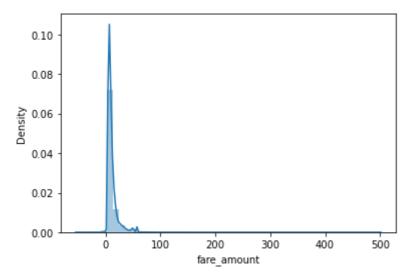
df.describe()

ıt[76]:		Unnamed: 0	fare_amount	pickup_longitude	pickup_latitude	dropoff_longitude	dropoff_la
	count	2.000000e+05	200000.000000	200000.000000	200000.000000	199999.000000	199999.0
	mean	2.771250e+07	11.359955	-72.527638	39.935885	-72.525292	39.9
	std	1.601382e+07	9.901776	11.437787	7.720539	13.117408	6.7
	min	1.000000e+00	-52.000000	-1340.648410	-74.015515	-3356.666300	-881.9
	25%	1.382535e+07	6.000000	-73.992065	40.734796	-73.991407	40.7
	50%	2.774550e+07	8.500000	-73.981823	40.752592	-73.980093	40.7
	75%	4.155530e+07	12.500000	-73.967154	40.767158	-73.963658	40.7
	max	5.542357e+07	499.000000	57.418457	1644.421482	1153.572603	872.6

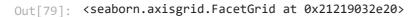
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn ing: `distplot` is a deprecated function and will be removed in a future version. Pl ease adapt your code to use either `displot` (a figure-level function with similar f lexibility) or `histplot` (an axes-level function for histograms).

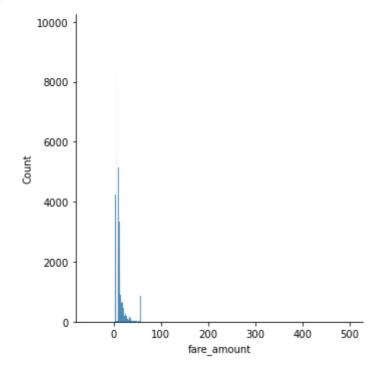
warnings.warn(msg, FutureWarning)

Out[78]: <AxesSubplot:xlabel='fare_amount', ylabel='Density'>



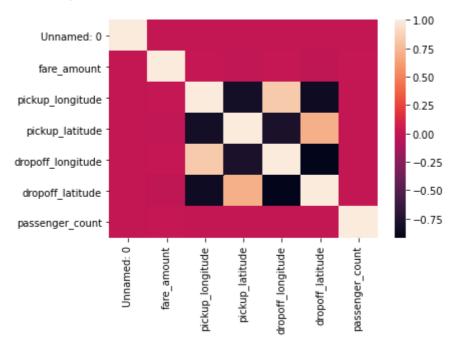
```
In [79]: sb.displot(df["fare_amount"])
```





```
In [80]: sb.heatmap(df.corr())
```





In []:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as pp

In [27]:
import seaborn as sb

In [81]:
    df = pd.read_csv(r"C:\Users\user\Desktop\2015.csv")
    df
```

Out[81]:

:		Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedo
	0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.665!
	1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.6287
	2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.6493
	3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.6697
	4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.6329
	•••									
	153	Rwanda	Sub- Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864	0.592(
	154	Benin	Sub- Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910	0.484
	155	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193	0.1568
	156	Burundi	Sub- Saharan Africa	157	2.905	0.08658	0.01530	0.41587	0.22396	0.1185
	157	Togo	Sub- Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443	0.364

158 rows × 12 columns

In [82]: df.head(10)

Out[82]

]:		Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom
	0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557
	1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877
	2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938
	3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973
	4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297
	5	Finland	Western Europe	6	7.406	0.03140	1.29025	1.31826	0.88911	0.64169
	6	Netherlands	Western Europe	7	7.378	0.02799	1.32944	1.28017	0.89284	0.61576
	7	Sweden	Western Europe	8	7.364	0.03157	1.33171	1.28907	0.91087	0.65980
	8	New Zealand	Australia and New Zealand	9	7.286	0.03371	1.25018	1.31967	0.90837	0.63938
	9	Australia	Australia and New Zealand	10	7.284	0.04083	1.33358	1.30923	0.93156	0.65124

In [83]:

df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 158 entries, 0 to 157 Data columns (total 12 columns):

#	Column	Non-Null Count	Dtypo
#	COTUIIII	NOII-NUII COUIT	Dtype
0	Country	158 non-null	object
1	Region	158 non-null	object
2	Happiness Rank	158 non-null	int64
3	Happiness Score	158 non-null	float64
4	Standard Error	158 non-null	float64
5	Economy (GDP per Capita)	158 non-null	float64
6	Family	158 non-null	float64
7	Health (Life Expectancy)	158 non-null	float64
8	Freedom	158 non-null	float64
9	Trust (Government Corruption)	158 non-null	float64
10	Generosity	158 non-null	float64
11	Dystopia Residual	158 non-null	float64
dtvn	es: $float64(9)$ int64(1) object	+(2)	

dtypes: float64(9), int64(1), object(2)

memory usage: 14.9+ KB

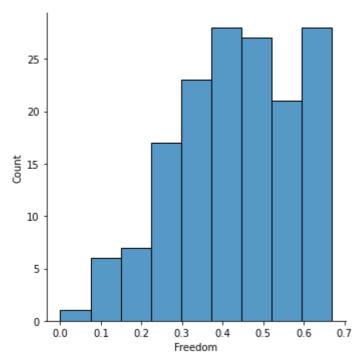
In [84]:

df.describe()

Out[84]: **Economy Standard** Happiness **Happiness** Health (Life (GDP per **Family** Freedom (Govern Rank **Score Error Expectancy**) Capita) Corrug 158.000000 158.000000 158.000000 158.000000 158.000000 158.000000 158.000000 158.00 count 79.493671 5.375734 0.047885 0.846137 0.991046 0.630259 0.428615 mean 0.14 45.754363 0.017146 0.403121 0.272369 1.145010 0.247078 0.150693 0.17 std 1.000000 2.839000 0.018480 0.000000 0.000000 0.000000 0.000000 0.00 min 0.439185 25% 4.526000 0.037268 0.545808 0.856823 40.250000 0.328330 0.0 **50**% 79.500000 5.232500 0.043940 0.910245 1.029510 0.696705 0.435515 0.10 **75%** 118.750000 6.243750 0.052300 1.158448 1.214405 0.811013 0.549092 0.18 158.000000 7.587000 0.136930 1.690420 1.402230 1.025250 0.669730 0.5! max In [85]: df.columns Index(['Country', 'Region', 'Happiness Rank', 'Happiness Score', 'Standard Error', 'Economy (GDP per Capita)', 'Family', 'Health (Life Expectancy)', 'Freedom', 'Trust (Government Corruption)', 'Generosity', 'Dystopia Residual'], dtype='object') In [87]: sb.distplot(df["Freedom"]) <AxesSubplot:xlabel='Freedom', ylabel='Density'> Out[87]: 2.5 2.0 1.5 Density 1.0 0.5 0.0 0.0 0.2 -0.20.4 0.6 0.8 Freedom

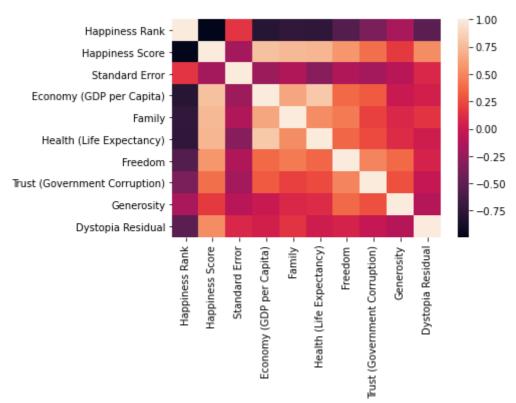
```
In [88]:
          sb.displot(df["Freedom"])
```

Out[88]: <seaborn.axisgrid.FacetGrid at 0x21223af47c0>



In [89]: sb.heatmap(df.corr())

Out[89]: <AxesSubplot:>



In []:

In [26]: import numpy as np import pandas as pd import matplotlib.pyplot as pp In [27]: import seaborn as sb In [90]: df = pd.read_csv(r"C:\Users\user\Desktop\fiat500_VehicleSelection_Dataset.csv") Out[90]: ID model engine_power age_in_days km previous_owners lat lon 0 lounge 51.0 882.0 25000.0 1.0 44.907242 8.611559868 1.0 1 2.0 51.0 1186.0 32500.0 1.0 45.666359 12.24188995 pop 2 3.0 74.0 4658.0 142228.0 1.0 45.503300 sport 11.41784 3 4.0 lounge 51.0 2739.0 160000.0 1.0 40.633171 17.63460922

3074.0

NaN

NaN

NaN

NaN

NaN

106880.0

NaN

NaN

NaN

NaN

NaN

1.0 41.903221

NaN

12.49565029

length

concat

find

search

Null values

73.0

NaN

NaN

NaN

NaN

NaN

1549 rows × 11 columns

5.0

рор

NaN

NaN

NaN

NaN

NaN

4

1544 NaN

1545 NaN

1546 NaN

1547 NaN

1548 NaN

Out[91]:

In [91]: df.head(10)

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	pri
0	1.0	lounge	51.0	882.0	25000.0	1.0	44.907242	8.611559868	89(
1	2.0	рор	51.0	1186.0	32500.0	1.0	45.666359	12.24188995	880
2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.41784	420
3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.63460922	600
4	5.0	pop	73.0	3074.0	106880.0	1.0	41.903221	12.49565029	57(
5	6.0	pop	74.0	3623.0	70225.0	1.0	45.000702	7.68227005	790
6	7.0	lounge	51.0	731.0	11600.0	1.0	44.907242	8.611559868	107!
7	8.0	lounge	51.0	1521.0	49076.0	1.0	41.903221	12.49565029	919
8	9.0	sport	73.0	4049.0	76000.0	1.0	45.548000	11.54946995	560

ID model engine_power age_in_days

```
10.0
                                 51.0
                                          3653.0
                                                  89000.0
                                                                     1.0 45.438301 10.99170017
                                                                                               600
          9
                   sport
In [92]:
          df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 1549 entries, 0 to 1548
          Data columns (total 11 columns):
          #
              Column
                                Non-Null Count
                                                 Dtype
          - - -
          0
              ID
                                1538 non-null
                                                 float64
          1
              model
                                1538 non-null
                                                 object
          2
              engine_power
                                1538 non-null
                                                 float64
          3
              age_in_days
                                1538 non-null
                                                 float64
          4
              km
                                1538 non-null
                                                 float64
          5
              previous_owners 1538 non-null
                                                 float64
          6
                                1538 non-null
                                                 float64
              lat
          7
              lon
                                1549 non-null
                                                 object
          8
              price
                                1549 non-null
                                                 object
          9
              Unnamed: 9
                                0 non-null
                                                 float64
          10 Unnamed: 10
                                1 non-null
                                                 object
          dtypes: float64(7), object(4)
         memory usage: 133.2+ KB
In [93]:
          df.describe()
Out[93]:
                                                                                             Unnar
                        ID engine_power age_in_days
                                                             km previous owners
          count 1538.000000
                             1538.000000
                                        1538.000000
                                                      1538.000000
                                                                      1538.000000 1538.000000
                 769.500000
                               51.904421 1650.980494
                                                     53396.011704
                                                                         1.123537
                                                                                   43.541361
          mean
            std
                 444.126671
                                3.988023
                                        1289.522278
                                                     40046.830723
                                                                         0.416423
                                                                                    2.133518
                   1.000000
                               51.000000
                                          366.000000
                                                                         1.000000
                                                                                   36.855839
                                                      1232.000000
           min
           25%
                 385.250000
                               51.000000
                                          670.000000
                                                     20006.250000
                                                                         1.000000
                                                                                   41.802990
           50%
                 769.500000
                               51.000000
                                         1035.000000
                                                     39031.000000
                                                                         1.000000
                                                                                   44.394096
                1153.750000
                               51.000000
                                         2616.000000
                                                     79667.750000
                                                                         1.000000
                                                                                   45.467960
                               77.000000 4658.000000 235000.000000
                                                                         4.000000
           max 1538.000000
                                                                                   46.795612
In [94]:
          df.columns
         Out[94]:
                dtype='object')
In [95]:
          sb.distplot(df["engine_power"])
```

km previous_owners

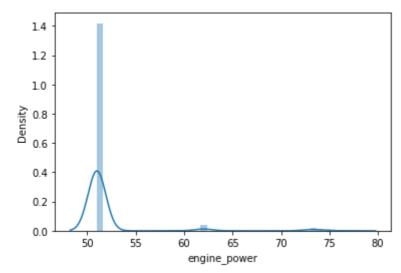
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

pric

lon

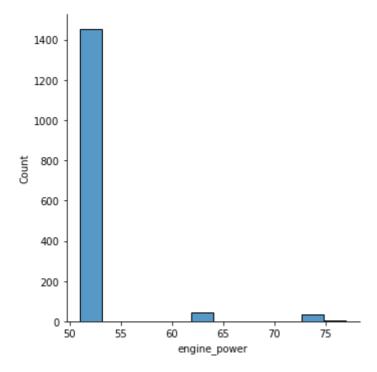
lat

Out[95]: <AxesSubplot:xlabel='engine_power', ylabel='Density'>



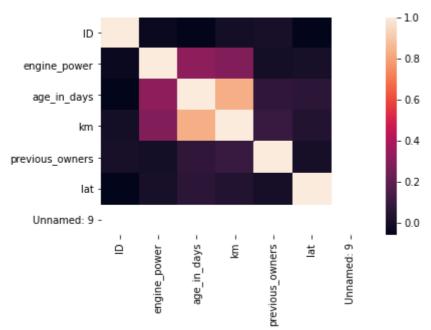
```
In [96]: sb.displot(df["engine_power"])
```

Out[96]: <seaborn.axisgrid.FacetGrid at 0x212241645b0>



```
In [97]: sb.heatmap(df.corr())
```

Out[97]: <AxesSubplot:>



In []: