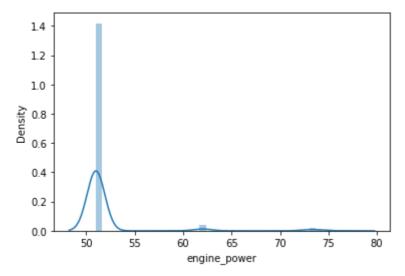
```
In [1]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as pp
In [2]:
           import seaborn as sb
In [3]:
           df = pd.read_csv(r"C:\Users\user\Desktop\fiat500_VehicleSelection_Dataset.csv")
                                                                                            lat
Out[3]:
                   ID
                       model engine_power age_in_days
                                                               km
                                                                   previous_owners
                                                                                                       lon
                                                                                                            pric
             0
                    1
                       lounge
                                          51
                                                      882
                                                            25000
                                                                                   1 44.907242
                                                                                                  8.611560
                                                                                                            890
                    2
             1
                                          51
                                                     1186
                                                            32500
                                                                                     45.666359
                                                                                                12.241890
                                                                                                            880
                         pop
             2
                    3
                                          74
                                                     4658
                                                           142228
                                                                                     45.503300
                                                                                                11.417840
                                                                                                            420
                                                                                   1
                        sport
             3
                    4
                       lounge
                                          51
                                                     2739
                                                           160000
                                                                                     40.633171
                                                                                                 17.634609
                                                                                                            600
             4
                    5
                                          73
                                                     3074
                                                           106880
                                                                                      41.903221
                                                                                                 12.495650
                                                                                                            57C
                          pop
          1533
                1534
                                          51
                                                     3712
                                                           115280
                                                                                     45.069679
                                                                                                  7.704920
                                                                                                            520
                        sport
          1534
                1535
                       lounge
                                          74
                                                     3835
                                                           112000
                                                                                     45.845692
                                                                                                  8.666870
                                                                                                            460
          1535
                1536
                                          51
                                                     2223
                                                            60457
                                                                                     45.481541
                                                                                                  9.413480
                                                                                                            75C
                         pop
                                                     2557
                                                                                     45.000702
                                                                                                  7.682270
          1536
                1537
                       lounge
                                          51
                                                            80750
                                                                                                            599
          1537
                1538
                                          51
                                                     1766
                                                            54276
                                                                                     40.323410 17.568270
                                                                                                            790
                          pop
         1538 rows × 9 columns
In [4]:
           df.head(10)
                                                                                                       price
Out[4]:
                 model
                         engine_power
                                        age_in_days
                                                         km
                                                              previous_owners
                                                                                       lat
                                                                                                 lon
                                                       25000
                                                                                44.907242
          0
              1
                 lounge
                                    51
                                                882
                                                                             1
                                                                                            8.611560
                                                                                                       8900
              2
                                                                                                       8800
          1
                                    51
                                               1186
                                                       32500
                                                                             1
                                                                                45.666359
                                                                                           12.241890
                    pop
          2
              3
                                    74
                                                                                           11.417840
                                                                                                       4200
                   sport
                                               4658
                                                      142228
                                                                                45.503300
                                                      160000
                                                                                                       6000
          3
              4
                 lounge
                                    51
                                               2739
                                                                             1
                                                                                40.633171
                                                                                           17.634609
          4
              5
                    pop
                                    73
                                               3074
                                                      106880
                                                                                41.903221
                                                                                           12.495650
                                                                                                       5700
                                                                                45.000702
                                               3623
                                                                                                       7900
          5
              6
                                    74
                                                       70225
                                                                                            7.682270
                    pop
                 lounge
                                    51
                                                731
                                                       11600
                                                                                44.907242
                                                                                            8.611560
                                                                                                      10750
          6
          7
                                    51
                                               1521
                                                       49076
                                                                                41.903221
                                                                                           12.495650
                                                                                                       9190
              8
                 lounge
          8
              9
                                    73
                                               4049
                                                       76000
                                                                                45.548000
                                                                                           11.549470
                                                                                                       5600
                   sport
          9
             10
                                    51
                                               3653
                                                       89000
                                                                                45.438301
                                                                                           10.991700
                                                                                                       6000
                   sport
In [5]:
           df.info()
```

Out[6]:

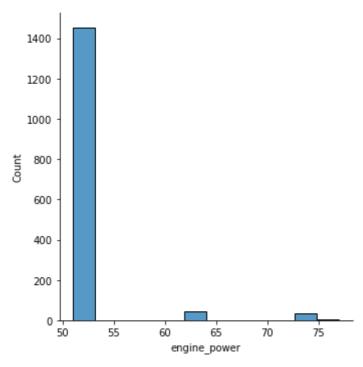
In [7]:

```
<class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1538 entries, 0 to 1537
        Data columns (total 9 columns):
             Column
                               Non-Null Count
                                               Dtype
         0
             ID
                               1538 non-null
                                               int64
         1
             model
                               1538 non-null
                                               object
         2
             engine_power
                               1538 non-null
                                               int64
         3
             age_in_days
                               1538 non-null
                                               int64
         4
                               1538 non-null
                                               int64
         5
             previous_owners 1538 non-null
                                               int64
         6
             lat
                               1538 non-null
                                               float64
         7
             lon
                               1538 non-null
                                               float64
         8
             price
                               1538 non-null
                                               int64
        dtypes: float64(2), int64(6), object(1)
        memory usage: 108.3+ KB
In [6]:
         df.describe()
                       ID engine_power
                                       age_in_days
                                                            km
                                                                previous_owners
                                                                                       lat
        count 1538.000000
                            1538.000000
                                       1538.000000
                                                     1538.000000
                                                                    1538.000000 1538.000000 1538.0
         mean
                769.500000
                              51.904421
                                       1650.980494
                                                    53396.011704
                                                                       1.123537
                                                                                 43.541361
                                                                                             11.5
           std
                444.126671
                               3.988023
                                       1289.522278
                                                    40046.830723
                                                                       0.416423
                                                                                  2.133518
                                                                                              2.3
                  1.000000
                              51.000000
                                        366.000000
                                                     1232.000000
                                                                       1.000000
                                                                                 36.855839
                                                                                              7.2
          min
          25%
                385.250000
                              51.000000
                                        670.000000
                                                    20006.250000
                                                                       1.000000
                                                                                 41.802990
                                                                                              9.5
          50%
                769.500000
                              51.000000
                                       1035.000000
                                                    39031.000000
                                                                       1.000000
                                                                                 44.394096
                                                                                             11.8
              1153.750000
                              51.000000
                                       2616.000000
                                                                       1.000000
          75%
                                                    79667.750000
                                                                                 45.467960
                                                                                             12.7
                                                                       4.000000
              1538.000000
                              77.000000 4658.000000
                                                   235000.000000
                                                                                 46.795612
                                                                                             18.3
          max
         df.columns
        Out[7]:
In [8]:
         sb.distplot(df["engine_power"])
        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[8]: <AxesSubplot:xlabel='engine_power', ylabel='Density'>
```



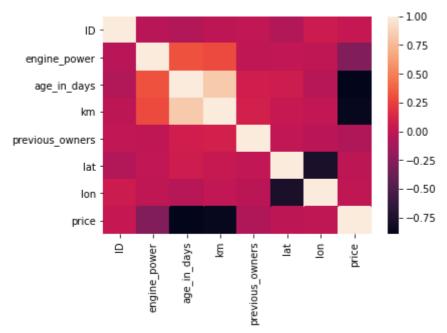
```
In [9]: sb.displot(df["engine_power"])
```

Out[9]: <seaborn.axisgrid.FacetGrid at 0x23e58a56880>



```
In [10]: sb.heatmap(df.corr())
```

Out[10]: <AxesSubplot:>

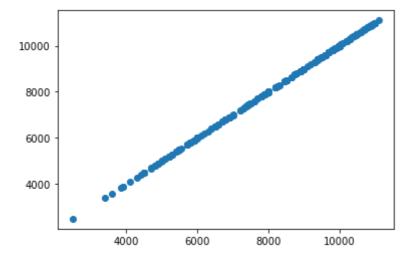


```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

Out[13]: LinearRegression()

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

Out[14]: <matplotlib.collections.PathCollection at 0x23e5ea05850>



```
In [ ]:
```

In []:		
[] .		

Western

Europe

Western

Europe

North

Sub-

Africa Sub-

Saharan

Saharan

Africa

Middle East and

Northern Africa

Sub-

Africa

Sub-

Africa

Saharan

Saharan

America

2

3

4

153

154

155

156

157

Denmark

Norway

Canada

Rwanda

Benin

Syria

Burundi

Togo

```
In [1]:
            import numpy as np
            import pandas as pd
            import matplotlib.pyplot as pp
 In [2]:
            import seaborn as sb
In [15]:
           df = pd.read_csv(r"C:\Users\user\Desktop\2015.csv")
Out[15]:
                                                                      Economy
                                                Happiness Standard
                                     Happiness
                                                                                         Health (Life
                                                                      (GDP per
                  Country
                             Region
                                                                                 Family
                                                                                                     Freedo
                                          Rank
                                                     Score
                                                               Error
                                                                                         Expectancy)
                                                                        Capita)
                            Western
             0 Switzerland
                                              1
                                                     7.587
                                                             0.03411
                                                                        1.39651
                                                                                1.34951
                                                                                             0.94143
                                                                                                       0.6655
                             Europe
                            Western
             1
                                             2
                    Iceland
                                                     7.561
                                                             0.04884
                                                                        1.30232 1.40223
                                                                                             0.94784
                                                                                                       0.6287
                             Europe
```

7.527

7.522

7.427

3.465

3.340

3.006

2.905

2.839

0.03328

0.03880

0.03553

0.03464

0.03656

0.05015

0.08658

0.06727

1.32548

1.45900

1.32629

1.36058

1.33095

1.32261

0.22208 0.77370

0.28665 0.35386

0.66320 0.47489

0.01530 0.41587

0.20868 0.13995

0.87464

0.88521

0.90563

0.42864

0.31910

0.72193

0.22396

0.28443

0.6493

0.6697

0.6329

0.5920

0.4845

0.1568

0.1185

0.3645

3

4

5

...

154

155

156

157

158

158 rows × 12 columns

In [16]:

df.head(10)

Out[16]:

]:		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 [17]:

df.info()

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

Data	columns (cocal 12 columns).		
#	Column	Non-Null Count	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
1.4		. (0)	

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

memory usage: 14.9+ KB

In [18]:

df.describe()

Out[18]:

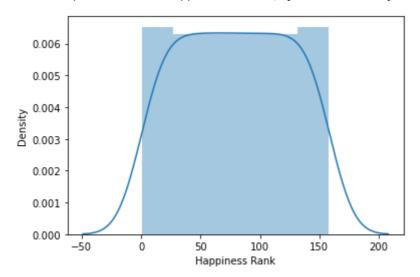
	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	(Govern Corruț
count	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.00
mean	79.493671	5.375734	0.047885	0.846137	0.991046	0.630259	0.428615	0.14
std	45.754363	1.145010	0.017146	0.403121	0.272369	0.247078	0.150693	0.17
min	1.000000	2.839000	0.018480	0.000000	0.000000	0.000000	0.000000	0.00
25%	40.250000	4.526000	0.037268	0.545808	0.856823	0.439185	0.328330	0.00
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
max	158.000000	7.587000	0.136930	1.690420	1.402230	1.025250	0.669730	0.5!

In [19]: df.columns

In [22]: sb.distplot(df["Happiness Rank"])

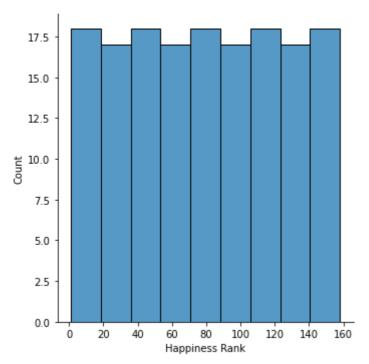
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[22]: <AxesSubplot:xlabel='Happiness Rank', ylabel='Density'>



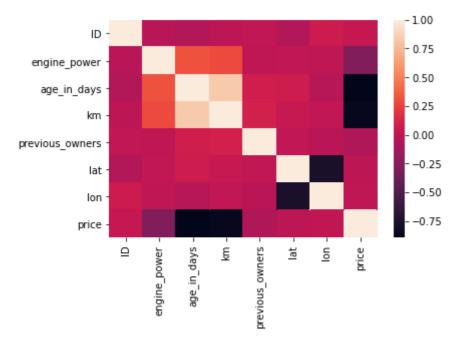
In [23]: sb.displot(df["Happiness Rank"])

Out[23]: <seaborn.axisgrid.FacetGrid at 0x23e5e892100>



```
In [10]: sb.heatmap(df.corr())
```

Out[10]: <AxesSubplot:>



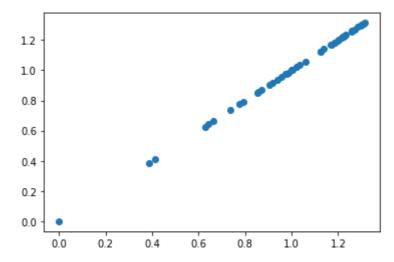
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)

In [29]: from sklearn.linear_model import LinearRegression

```
lr = LinearRegression()
lr.fit(x_train,y_train)
```

Out[29]: LinearRegression()

Out[30]: <matplotlib.collections.PathCollection at 0x23e5eebc9d0>



```
In [ ]:

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
             0
                 842302
                                           17.99
                                                         10.38
                                                                        122.80
                                                                                   1001.0
                                                                                                    0.1184
                                Μ
                  842517
                                Μ
                                           20.57
                                                         17.77
                                                                        132.90
                                                                                   1326.0
                                                                                                    0.0847
             2 84300903
                                           19.69
                                                         21.25
                                                                        130.00
                                                                                   1203.0
                                                                                                    0.1096
                                Μ
             3 84348301
                                М
                                           11.42
                                                         20.38
                                                                         77.58
                                                                                    386.1
                                                                                                    0.1425
                                                                                                    0.1003
             4 84358402
                                           20.29
                                                         14.34
                                                                        135.10
                                                                                   1297.0
                                Μ
          564
                  926424
                                           21.56
                                                         22.39
                                                                        142.00
                                                                                   1479.0
                                                                                                    0.1110
                                Μ
           565
                 926682
                                Μ
                                           20.13
                                                         28.25
                                                                        131.20
                                                                                    1261.0
                                                                                                    0.0978
           566
                 926954
                                Μ
                                           16.60
                                                         28.08
                                                                        108.30
                                                                                    858.1
                                                                                                    0.0845
           567
                 927241
                                Μ
                                           20.60
                                                         29.33
                                                                        140.10
                                                                                    1265.0
                                                                                                    0.1178
          568
                   92751
                                 В
                                           7.76
                                                         24.54
                                                                         47.92
                                                                                    181.0
                                                                                                    0.0526
          569 rows × 33 columns
In [30]:
           df.head(10)
Out[30]:
```

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean
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	0.10960
3	84348301	М	11.42	20.38	77.58	386.1	0.14250
4	84358402	М	20.29	14.34	135.10	1297.0	0.10030
5	843786	М	12.45	15.70	82.57	477.1	0.12780
6	844359	М	18.25	19.98	119.60	1040.0	0.09463
7	84458202	М	13.71	20.83	90.20	577.9	0.11890
8	844981	М	13.00	21.82	87.50	519.8	0.12730
9	84501001	М	12.46	24.04	83.97	475.9	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()

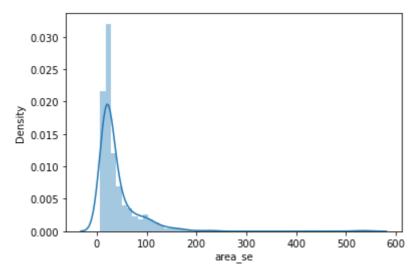
Out[32]:

	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

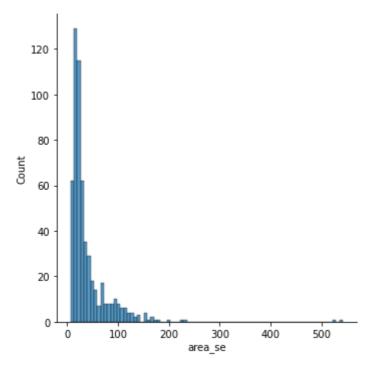
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

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



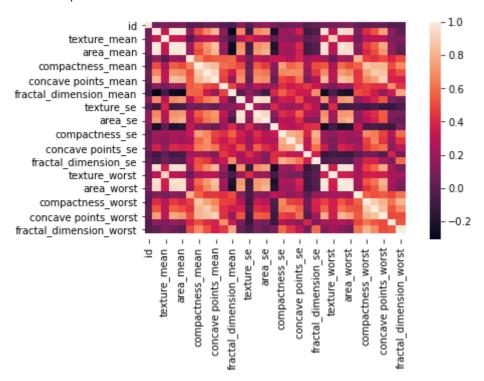
```
In [35]: sb.displot(df["area_se"])
```

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	ry usage: 837.7	+ KB	

In [40]:

df.describe()

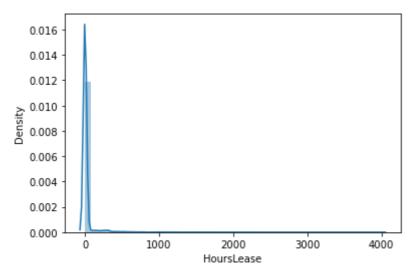
Out[40]:

	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

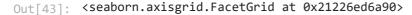
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn ing: `distplot` is a deprecated function and will be removed in a future version. Pl ease adapt your code to use either `displot` (a figure-level function with similar f lexibility) or `histplot` (an axes-level function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

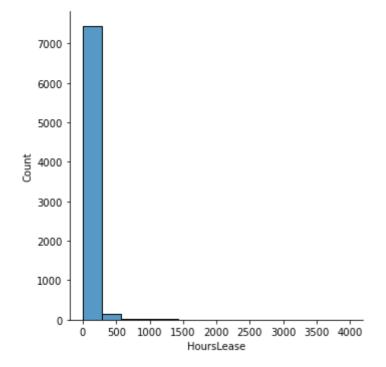
warnings.warn(msg, FutureWarning)

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



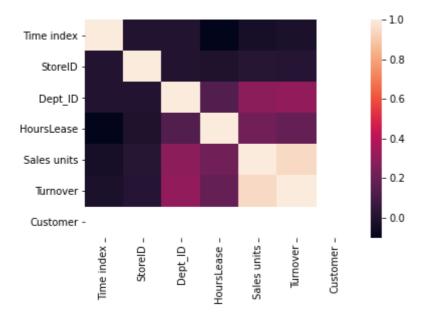
```
In [43]: sb.displot(df["HoursLease"])
```





```
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

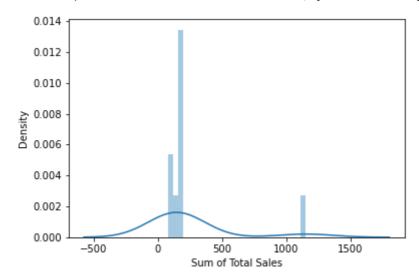
Out

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

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

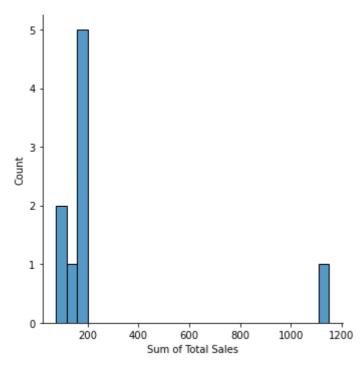
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function 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 [1]: import numpy as np
   import pandas as pd
   import matplotlib.pyplot as pp

In [2]: import seaborn as sb

In [56]: df = nd read csy(r"C:\lsens\user\Desktop\5 Instagram data csy")
```

In [56]: df = pd.read_csv(r"C:\Users\user\Desktop\5_Instagram data.csv")
 df

	uı											
Out[56]:		Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	Follov
	0	3920	2586	1028	619	56	98	9	5	162	35	
	1	5394	2727	1838	1174	78	194	7	14	224	48	
	2	4021	2085	1188	0	533	41	11	1	131	62	
	3	4528	2700	621	932	73	172	10	7	213	23	
	4	2518	1704	255	279	37	96	5	4	123	8	
	•••											
	114	13700	5185	3041	5352	77	573	2	38	373	73	ł
	115	5731	1923	1368	2266	65	135	4	1	148	20	
	116	4139	1133	1538	1367	33	36	0	1	92	34	

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	Follo
117	32695	11815	3147	17414	170	1095	2	75	549	148	2
118	36919	13473	4176	16444	2547	653	5	26	443	611	2.

119 rows × 13 columns

	-	_	L	-	4	

	а	+.nead(10)										
Out[57]:		Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	Follows
	0	3920	2586	1028	619	56	98	9	5	162	35	2
	1	5394	2727	1838	1174	78	194	7	14	224	48	10
	2	4021	2085	1188	0	533	41	11	1	131	62	12
	3	4528	2700	621	932	73	172	10	7	213	23	8
	4	2518	1704	255	279	37	96	5	4	123	8	0
	5	3884	2046	1214	329	43	74	7	10	144	9	2

	Impressions	From Home	From Hashtags		From Other	Saves	Comments	Shares	Likes	Profile Visits	Follows
6	2621	1543	599	333	25	22	5	1	76	26	0
7	3541	2071	628	500	60	135	4	9	124	12	6
8	3749	2384	857	248	49	155	6	8	159	36	4
9	4115	2609	1104	178	46	122	6	3	191	31	6

In [58]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 119 entries, 0 to 118 Data columns (total 13 columns):

Daca	COTAMM13 (COCAT	<u> </u>	
#	Column	Non-Null Count	Dtype
0	Impressions	119 non-null	int64
1	From Home	119 non-null	int64
2	From Hashtags	119 non-null	int64
3	From Explore	119 non-null	int64
4	From Other	119 non-null	int64
5	Saves	119 non-null	int64
6	Comments	119 non-null	int64
7	Shares	119 non-null	int64
8	Likes	119 non-null	int64
9	Profile Visits	119 non-null	int64
10	Follows	119 non-null	int64
11	Caption	119 non-null	object
12	Hashtags	119 non-null	object

dtypes: int64(11), object(2) memory usage: 12.2+ KB

In [59]:

df.describe()

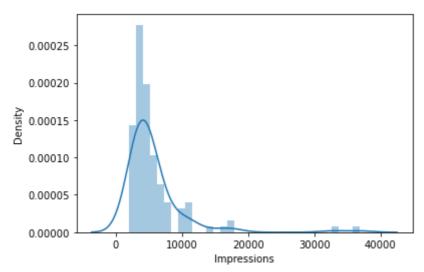
Out[59]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments
count	119.000000	119.000000	119.000000	119.000000	119.000000	119.000000	119.000000
mean	5703.991597	2475.789916	1887.512605	1078.100840	171.092437	153.310924	6.663866
std	4843.780105	1489.386348	1884.361443	2613.026132	289.431031	156.317731	3.544576
min	1941.000000	1133.000000	116.000000	0.000000	9.000000	22.000000	0.000000

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments
25%	3467.000000	1945.000000	726.000000	157.500000	38.000000	65.000000	4.000000
50%	4289.000000	2207.000000	1278.000000	326.000000	74.000000	109.000000	6.000000
75 %	6138.000000	2602.500000	2363.500000	689.500000	196.000000	169.000000	8.000000
max	36919.000000	13473.000000	11817.000000	17414.000000	2547.000000	1095.000000	19.000000

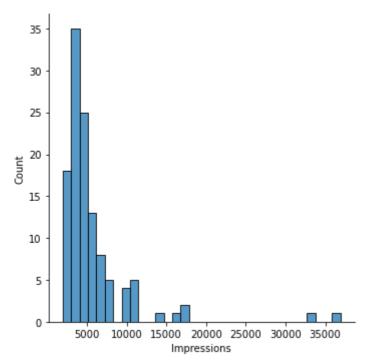
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[61]: <AxesSubplot:xlabel='Impressions', ylabel='Density'>



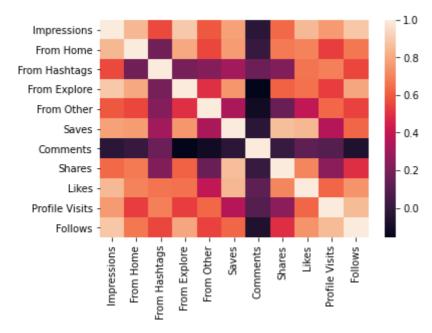
```
In [62]: sb.displot(df["Impressions"])
```

Out[62]: <seaborn.axisgrid.FacetGrid at 0x23e5ea10f10>



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

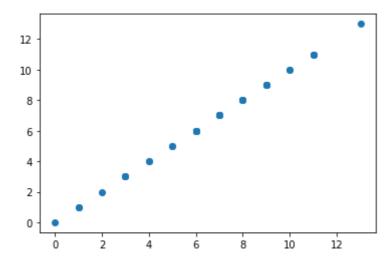
Out[63]: <AxesSubplot:>

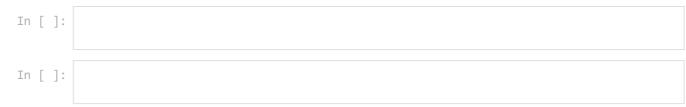


```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

Out[68]: LinearRegression()

Out[69]: <matplotlib.collections.PathCollection at 0x23e603b0e20>





```
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()
```

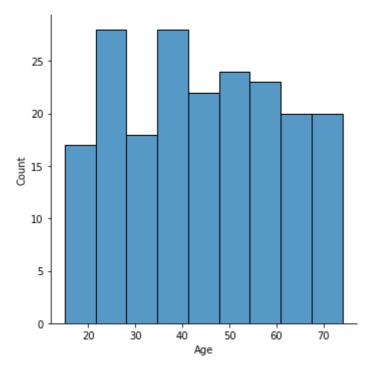
```
<class 'pandas.core.frame.DataFrame'>
          RangeIndex: 200 entries, 0 to 199
          Data columns (total 6 columns):
                             Non-Null Count Dtype
           #
               Column
           0
                             200 non-null
                                              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]:

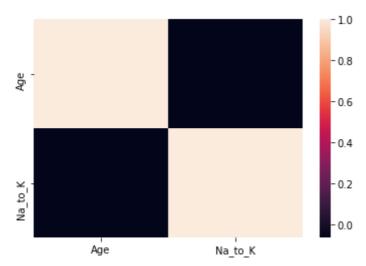
In [62]:

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



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

Out[63]: <AxesSubplot:>



In []:

```
In [1]:
            import numpy as np
            import pandas as pd
            import matplotlib.pyplot as pp
 In [2]:
            import seaborn as sb
In [92]:
            df = pd.read_csv(r"C:\Users\user\Desktop\7_uber.csv")
Out[92]:
                    Unnamed:
                                              key fare_amount pickup_datetime pickup_longitude pickup_latit
                                       2015-05-07
                                                                      2015-05-07
                     24238194
                                                            7.5
                                                                                         -73.999817
                                                                                                          40.738
                                  19:52:06.0000003
                                                                     19:52:06 UTC
                                       2009-07-17
                                                                      2009-07-17
                     27835199
                                                            7.7
                                                                                         -73.994355
                                                                                                          40.728
                                  20:04:56.0000002
                                                                     20:04:56 UTC
                                       2009-08-24
                                                                      2009-08-24
                 2
                     44984355
                                                           12.9
                                                                                         -74.005043
                                                                                                          40.740
                                                                     21:45:00 UTC
                                 21:45:00.00000061
                                       2009-06-26
                                                                      2009-06-26
                 3
                     25894730
                                                            5.3
                                                                                         -73.976124
                                                                                                          40.790
                                  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
                                                            7.5
           199996
                     16382965
                                                                                         -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 [93]:
            df.head(10)
Out[931.
              Unnamed:
```

ut[93]:		Unnamed: 0	key	fare_amount	pickup_datetime	pickup_longitude	pickup_latitude	(
	0	24238194	2015-05-07 19:52:06.0000003	7.5	2015-05-07 19:52:06 UTC	-73.999817	40.738354	
	1	27835199	2009-07-17 20:04:56.0000002	7.7	2009-07-17 20:04:56 UTC	-73.994355	40.728225	
	2	44984355	2009-08-24 21:45:00.00000061	12.9	2009-08-24 21:45:00 UTC	-74.005043	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 [94]:

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
d+\n	oc. float64/E) int	64(2) object(2)	

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

memory usage: 13.7+ MB

In [95]:

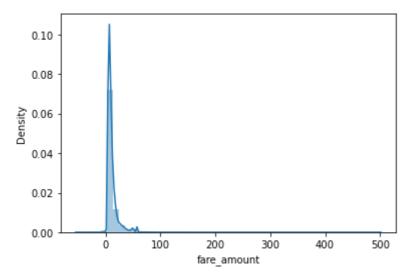
df.describe()

it[95]:	Unnamed: 0		fare_amount	pickup_longitude	pickup_latitude	${\bf dropoff_longitude}$	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
	4						•

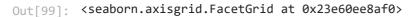
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn ing: `distplot` is a deprecated function and will be removed in a future version. Pl ease adapt your code to use either `displot` (a figure-level function with similar f lexibility) or `histplot` (an axes-level function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

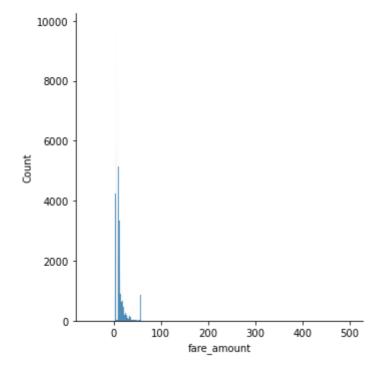
warnings.warn(msg, FutureWarning)

Out[97]: <AxesSubplot:xlabel='fare_amount', ylabel='Density'>



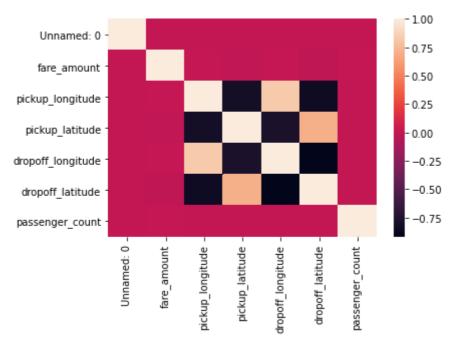
```
In [99]: sb.displot(df["fare_amount"])
```





```
In [100... sb.heatmap(df.corr())
```

Out[100... <AxesSubplot:>



from sklearn.model_selection import train_test_split
 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)

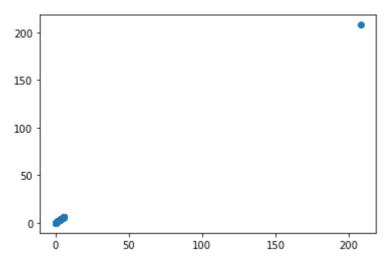
from sklearn.linear_model import LinearRegression

lr = LinearRegression()
 lr.fit(x_train,y_train)

Out[106... LinearRegression()

prediction = lr.predict(x_test)
pp.scatter(y_test,prediction)

Out[107... <matplotlib.collections.PathCollection at 0x23e61cafaf0>



In []:

In []:

In [1]: import numpy as np
 import pandas as pd
 import matplotlib.pyplot as pp

In [2]: import seaborn as sb

In [108... df = pd.read_csv(r"C:\Users\user\Desktop\8_BreastCancerPrediction.csv")
 df

Out[108... id diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mea
0	842302	М	17.99	10.38	122.80	1001.0	0.1184
1	842517	М	20.57	17.77	132.90	1326.0	0.0847
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	0.1003
•••							
564	926424	М	21.56	22.39	142.00	1479.0	0.1110
565	926682	М	20.13	28.25	131.20	1261.0	0.0978
566	926954	М	16.60	28.08	108.30	858.1	0.0845
567	927241	М	20.60	29.33	140.10	1265.0	0.1178
568	92751	В	7.76	24.54	47.92	181.0	0.0526

569 rows × 33 columns

In [109...

df.head(10)

Out[109...

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean
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	0.10960
3	84348301	М	11.42	20.38	77.58	386.1	0.14250
4	84358402	М	20.29	14.34	135.10	1297.0	0.10030
5	843786	М	12.45	15.70	82.57	477.1	0.12780
6	844359	М	18.25	19.98	119.60	1040.0	0.09463
7	84458202	М	13.71	20.83	90.20	577.9	0.11890
8	844981	М	13.00	21.82	87.50	519.8	0.12730
9	84501001	М	12.46	24.04	83.97	475.9	0.11860

10 rows × 33 columns

```
In [110...
```

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	<pre>fractal_dimension_mean</pre>	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
dtype	es: float64(31), int64(1)	, object(1)	
memoi	rv usage: 146.8+ KB		

memory usage: 146.8+ KB

In [111...

df.describe()

Out[111...

	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	

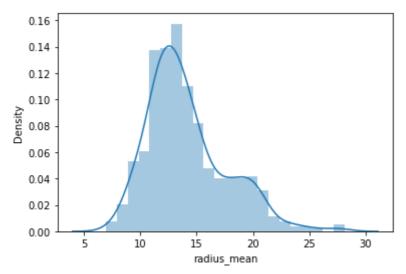
8 rows × 32 columns

```
In [112... df.columns
```

```
In [113... sb.distplot(df["radius_mean"])
```

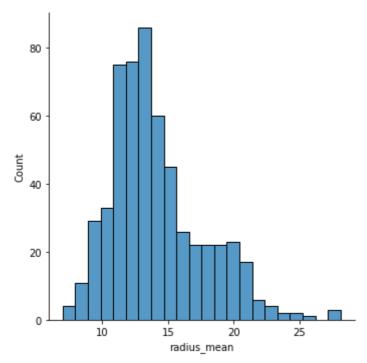
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarn
ing: `distplot` is a deprecated function and will be removed in a future version. Pl
ease adapt your code to use either `displot` (a figure-level function with similar f
lexibility) or `histplot` (an axes-level function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

Out[113... <AxesSubplot:xlabel='radius_mean', ylabel='Density'>



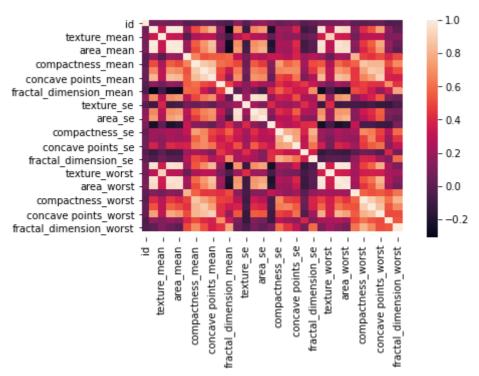
```
In [114... sb.displot(df["radius_mean"])
```

Out[114... <seaborn.axisgrid.FacetGrid at 0x23e607f7ca0>



In [115... sb.heatmap(df.corr())

Out[115... <AxesSubplot:>



```
In [117...
           from sklearn.model_selection import train_test_split
           x train,x test,y train,y test = train test split(x,y,test size=0.3)
          ValueError
                                                     Traceback (most recent call last)
          <ipython-input-117-c38b79e540c3> in <module>
                1 from sklearn.model_selection import train_test_split
          ---> 2 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\model selection\ split.py in trai
          n test split(test size, train size, random state, shuffle, stratify, *arrays)
             2173
             2174
                       n samples = num samples(arrays[0])
          -> 2175
                       n train, n test = validate shuffle split(n samples, test size, train si
             2176
                                                                  default test size=0.25)
             2177
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\model selection\ split.py in val
          idate_shuffle_split(n_samples, test_size, train_size, default_test_size)
                       if n train == 0:
             1856
          -> 1857
                           raise ValueError(
             1858
                               'With n_samples={}, test_size={} and train_size={}, the '
                               'resulting train set will be empty. Adjust any of the '
          ValueError: With n samples=0, test_size=0.3 and train_size=None, the resulting train
          set will be empty. Adjust any of the aforementioned parameters.
In [118...
           from sklearn.linear_model import LinearRegression
           lr = LinearRegression()
           lr.fit(x_train,y_train)
          LinearRegression()
Out[118...
In [119...
           prediction = lr.predict(x_test)
           pp.scatter(y_test,prediction)
          <matplotlib.collections.PathCollection at 0x23e6381c6a0>
Out[119...
           200
          150
          100
            50
                          50
                                    100
                                               150
                                                         200
  In [ ]:
  In [ ]:
```

```
In [1]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as pp
  In [2]:
           import seaborn as sb
In [120...
           df = pd.read_csv(r"C:\Users\user\Desktop\10_USA_Housing.csv")
```

Out[120		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
	4997	63390.686886	7.250591	4.805081	2.13	33266.145490	1.030730e+06	4215 Tracy Garde Suite 076\nJoshualan VA 01
	4998	68001.331235	5.534388	7.130144	5.44	42625.620156	1.198657e+06	USS Wallace\nFPO A 7331
	4999	65510.581804	5.992305	6.792336	4.07	46501.283803	1.298950e+06	37778 George Ridge Apt. 509\nEast Holl NV 2

5000 rows × 7 columns

In [121... df.head(10)

Out[121...

	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 [122...

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 [123...

df.describe()

Out[123...

	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 [124...

df.columns

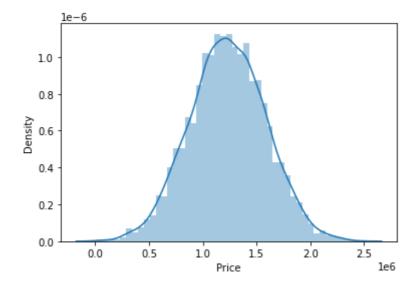
Out[124...

In [125...

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)

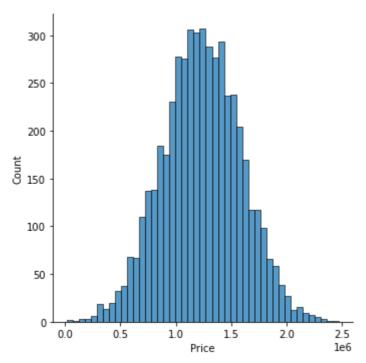
Out[125... <AxesSubplot:xlabel='Price', ylabel='Density'>



In [126...

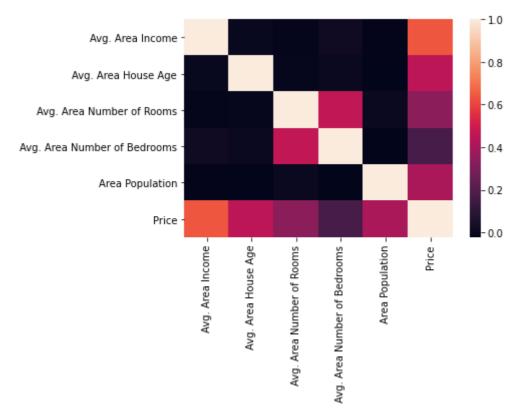
sb.displot(df["Price"])

Out[126... <seaborn.axisgrid.FacetGrid at 0x23e638c6e20>



```
In [127... sb.heatmap(df.corr())
```

Out[127... <AxesSubplot:>



```
from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

```
In [133...
            from sklearn.linear_model import LinearRegression
            lr = LinearRegression()
            lr.fit(x_train,y_train)
           LinearRegression()
Out[133...
In [134...
            prediction = lr.predict(x_test)
            pp.scatter(y_test,prediction)
           <matplotlib.collections.PathCollection at 0x23e618332e0>
Out[134...
           2.5
           2.0
           1.5
           1.0
           0.5
                                           1.5
                       0.5
                                 1.0
                                                     2.0
                                                               2.5
                                                               1e6
  In [ ]:
  In [ ]:
```