```
import pandas as pd
 In [6]:
            import numpy as np
 In [8]: data=pd.read_csv("/home/palcement/Desktop/test.txt")
 In [9]: data.head()
 Out[9]:
               srno
                         name
                               pincode
                                         city branch marks
                  1
                          nikhil
                                 534134
                                         attili
                                                  ΙT
                                                         89
                                                         98
                   2 manikanta
                                534133 Attilil
                                                 EEE
             1
In [10]: data=pd.read_csv("/home/palcement/Downloads/fiat500.csv")
In [11]:
           data.describe()
Out[11]:
                            ID engine power
                                              age in days
                                                                     km previous owners
                                                                                                  lat
                                                                                                              lon
                                                                                                                          price
                   1538.000000
                                                                                                                    1538.000000
             count
                                 1538.000000
                                              1538.000000
                                                            1538.000000
                                                                             1538.000000
                                                                                         1538.000000
                                                                                                      1538.000000
             mean
                     769.500000
                                   51.904421
                                              1650.980494
                                                            53396.011704
                                                                                1.123537
                                                                                            43.541361
                                                                                                         11.563428
                                                                                                                    8576.003901
               std
                    444.126671
                                    3.988023
                                              1289.522278
                                                            40046.830723
                                                                                0.416423
                                                                                             2.133518
                                                                                                         2.328190
                                                                                                                    1939.958641
                      1.000000
                                                                                                         7.245400
                                   51.000000
                                               366.000000
                                                            1232.000000
                                                                                1.000000
                                                                                            36.855839
                                                                                                                    2500.000000
              min
                                                                                                         9.505090
              25%
                     385.250000
                                   51.000000
                                               670.000000
                                                            20006.250000
                                                                                1.000000
                                                                                            41.802990
                                                                                                                    7122.500000
              50%
                    769.500000
                                   51.000000
                                              1035.000000
                                                            39031.000000
                                                                                1.000000
                                                                                            44.394096
                                                                                                        11.869260
                                                                                                                    9000.000000
              75%
                   1153.750000
                                   51.000000
                                              2616.000000
                                                            79667.750000
                                                                                1.000000
                                                                                            45.467960
                                                                                                        12.769040
                                                                                                                  10000.000000
              max 1538.000000
                                   77.000000
                                              4658.000000
                                                           235000.000000
                                                                                4.000000
                                                                                            46.795612
                                                                                                        18.365520
                                                                                                                  11100.000000
```

In [13]: data.head(10)

Out[13]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	pop	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	pop	73	3074	106880	1	41.903221	12.495650	5700
5	6	pop	74	3623	70225	1	45.000702	7.682270	7900
6	7	lounge	51	731	11600	1	44.907242	8.611560	10750
7	8	lounge	51	1521	49076	1	41.903221	12.495650	9190
8	9	sport	73	4049	76000	1	45.548000	11.549470	5600
9	10	sport	51	3653	89000	1	45.438301	10.991700	6000

In [14]: data.tail(20)

Out[14]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
1518	1519	lounge	51	397	16053	1	38.122070	13.36112	10500
1519	1520	lounge	51	670	30000	1	45.764648	8.99450	10800
1520	1521	lounge	51	1035	15000	1	41.903221	12.49565	10990
1521	1522	lounge	51	3774	85000	1	44.294300	9.67444	4000
1522	1523	lounge	51	366	14618	1	45.707249	11.47760	10500
1523	1524	pop	51	2251	79800	1	45.512051	10.42701	6450
1524	1525	pop	51	2192	53300	1	40.609531	14.98093	7900
1525	1526	lounge	51	790	41870	1	45.707249	11.47760	9500
1526	1527	lounge	51	1705	23600	1	38.122070	13.36112	9300
1527	1528	pop	51	517	3000	1	40.748241	14.52835	9999
1528	1529	lounge	51	2861	126000	1	43.841980	10.51531	5500
1529	1530	lounge	51	731	22551	1	38.122070	13.36112	9900
1530	1531	lounge	51	670	29000	1	45.764648	8.99450	10800
1531	1532	sport	73	4505	127000	1	45.528511	9.59323	4750
1532	1533	pop	51	1917	52008	1	45.548000	11.54947	9900
1533	1534	sport	51	3712	115280	1	45.069679	7.70492	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.66687	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.41348	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.68227	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.56827	7900

```
In [15]: |data['previous_owners'].unique()
Out[15]: array([1, 2, 3, 4])
In [16]: data["engine power"].unique()
Out[16]: array([51, 74, 73, 62, 63, 66, 77, 58])
In [17]: list(data.columns)
Out[17]: ['ID',
           'model',
           'engine power',
           'age_in_days',
           'km',
           'previous_owners',
           'lat',
           'lon',
           'price']
In [18]: data.groupby(['previous_owners']).count()
Out[18]:
                           ID model engine_power age_in_days
                                                             km
                                                                  lat
                                                                       Ion price
           previous_owners
                       1 1389
                               1389
                                           1389
                                                       1389
                                                           1389
                                                                 1389
                                                                      1389
                                                                           1389
                       2
                          117
                                117
                                            117
                                                        117
                                                            117
                                                                  117
                                                                       117
                                                                            117
                       3
                           23
                                 23
                                             23
                                                        23
                                                              23
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                                                                        23
                                                                             23
                            9
                                  9
                                              9
                                                         9
                                                              9
                                                                   9
                                                                        9
                                                                              9
```

In [19]:	da	ta.g	roupb	y(['model']).count()							
Out[19]:			ID	engine_power	age_in_days	km	previous_owners	la	at lon	price		
	m	odel										
	lo	unge	1094	1094	1094	1094	1094	109	94 1094	1094		
		pop	358	358	358	358	358	35	58 358	358		
	•	sport	86	86	86	86	86	8	86 86	86		
In [22]:	da	ta1=	data.	drop([' <mark>lat</mark> '	,' <mark>ID</mark> '],axi	s=1)						
In [25]:	da	ta.h	iead (5)								
Out[25]:		ID	model	engine_power	age_in_days	km	previous_owner	's	lat	lo	n price)
	0	1	lounge	51	882	25000		1 4	14.907242	8.61156	0 8900)
	1	2	pop	51	1186	32500)	1 4	45.666359	12.24189	0 8800)
	2	3	sport	74	4658	142228		1 4	45.503300	11.41784	0 4200)
	3	4	lounge	51	2739	160000)	1 4	40.633171	17.63460	9 6000)
	4	5	pop	73	3074	106880		1 4	41.903221	12.49565	0 5700)
In [33]:	da ⁻	ta.h	ead(5)								
Out[33]:		ID	model	engine_power	age_in_days	km	previous_owner	's	lat	lo	n price)
	0	1	lounge	51	882	25000		1 4	14.907242	8.61156	0 8900)
	1	2	pop	51	1186	32500	1	1 4	45.666359	12.24189	0 8800)
	2	3	sport	74	4658	142228	}	1 4	45.503300	11.41784	0 4200)
	3	4	lounge	51	2739	160000		1 4	40.633171	17.63460	9 6000)
	4	5	pop	73	3074	106880	•	1 4	41.903221	12.49565	0 5700)

```
In [27]: data['price'].sum()
```

Out[27]: 13189894

In [34]: data2=data.loc[(data.model=='lounge')]

data2

Out[34]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
6	7	lounge	51	731	11600	1	44.907242	8.611560	10750
7	8	lounge	51	1521	49076	1	41.903221	12.495650	9190
11	12	lounge	51	366	17500	1	45.069679	7.704920	10990
1528	1529	lounge	51	2861	126000	1	43.841980	10.515310	5500
1529	1530	lounge	51	731	22551	1	38.122070	13.361120	9900
1530	1531	lounge	51	670	29000	1	45.764648	8.994500	10800
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990

1094 rows × 9 columns

fiat1 - Jupyter Notebook

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	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.61156	8900
1	2	pop	51	1186	32500	1	45.666359	12.24189	8800
6	7	lounge	51	731	11600	1	44.907242	8.61156	10750
7	8	lounge	51	1521	49076	1	41.903221	12.49565	9190
10	11	pop	51	790	43286	1	40.871429	14.43896	8950
1525	1526	lounge	51	790	41870	1	45.707249	11.47760	9500
1526	1527	lounge	51	1705	23600	1	38.122070	13.36112	9300
1527	1528	pop	51	517	3000	1	40.748241	14.52835	9999
1529	1530	lounge	51	731	22551	1	38.122070	13.36112	9900
1530	1531	lounge	51	670	29000	1	45.764648	8.99450	10800

900 rows × 9 columns

20/06/2023 fiat1 - Jupyter Notebook

In [37]: data4=data.loc[(data.model=='lounge') &(data.previous_owners==1)]
 data4

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···	'u		ı	/ 1

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
6	7	lounge	51	731	11600	1	44.907242	8.611560	10750
7	8	lounge	51	1521	49076	1	41.903221	12.495650	9190
11	12	lounge	51	366	17500	1	45.069679	7.704920	10990
1528	1529	lounge	51	2861	126000	1	43.841980	10.515310	5500
1529	1530	lounge	51	731	22551	1	38.122070	13.361120	9900
1530	1531	lounge	51	670	29000	1	45.764648	8.994500	10800
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990

993 rows × 9 columns

20/06/2023 fiat1 - Jupyter Notebook

In [42]: data4=data.loc[(data.model=='pop') |(data.previous_owners==1)]
 data4

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v	u		. –	_	

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	рор	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	рор	73	3074	106880	1	41.903221	12.495650	5700
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	рор	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1420 rows × 9 columns

In [45]: data5=data.loc[(data.model=='lounge') | (data.model=="pop")]
 data5

Out[45]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	рор	51	1186	32500	1	45.666359	12.241890	8800
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	рор	73	3074	106880	1	41.903221	12.495650	5700
5	6	рор	74	3623	70225	1	45.000702	7.682270	7900
1532	1533	pop	51	1917	52008	1	45.548000	11.549470	9900
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1452 rows × 9 columns

In []: