In [61]: import pandas as pd
In [62]: data=pd.read\_csv("/home/placement/Downloads/fiat500.csv")
In [63]: data.describe()

## Out[63]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price
count	1538.000000	1538.000000	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
min	1.000000	51.000000	366.000000	1232.000000	1.000000	36.855839	7.245400	2500.000000
25%	385.250000	51.000000	670.000000	20006.250000	1.000000	41.802990	9.505090	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 [64]: data.head()

## Out[64]:

	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	рор	73	3074	106880	1	41.903221	12.495650	5700

```
In [65]: data['previous owners'].unique()
Out[65]: array([1, 2, 3, 4])
In [66]: list(data.columns)
Out[66]: ['ID',
           'model',
           'engine power',
           'age in days',
           'km',
           'previous owners',
           'lat',
           'lon',
           'price']
In [67]: data.groupby(['previous owners']).count()
Out[67]:
                           ID model engine_power age_in_days
                                                            km
                                                                  lat lon price
           previous_owners
                      1 1389
                               1389
                                           1389
                                                      1389
                                                           1389 1389 1389
                                                                           1389
                       2
                          117
                                117
                                            117
                                                       117
                                                            117
                                                                 117
                                                                       117
                                                                            117
                                             23
                                                        23
                                                             23
                                                                  23
                                                                             23
                       3
                           23
                                 23
                                                                       23
                                              9
                            9
                                  9
                                                         9
                                                              9
                                                                   9
                                                                        9
                                                                             9
                       4
```

In [68]: data.groupby(['model']).count()

Out[68]:

	ID	engine_power	age_in_days	km	previous_owners	lat	Ion	price
model								
lounge	1094	1094	1094	1094	1094	1094	1094	1094
pop	358	358	358	358	358	358	358	358
sport	86	86	86	86	86	86	86	86

In [69]: data1=data.drop(['lat','ID'],axis=1)

In [70]: data.head(5)

Out[70]:

	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	рор	73	3074	106880	1	41.903221	12.495650	5700

```
In [71]: data1.head(5)
```

### Out[71]:

	model	engine_power	age_in_days	km	previous_owners	lon	price
0	lounge	51	882	25000	1	8.611560	8900
1	pop	51	1186	32500	1	12.241890	8800
2	sport	74	4658	142228	1	11.417840	4200
3	lounge	51	2739	160000	1	17.634609	6000
4	pop	73	3074	106880	1	12.495650	5700

```
In [72]: data['price'].sum
```

```
Out[72]: <bound method NDFrame._add_numeric_operations.<locals>.sum of 0 8900
```

1 8800 2 4200 3 6000 4 5700

1533 5200 1534 4600 1535 7500

1536 5990 1537 7900

Name: price, Length: 1538, dtype: int64>

In [73]: data2=data.loc[(data.model=='lounge')]
 data2

Out[73]:

	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

Out[74]:

	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
6	7	lounge	51	731	11600	1	44.907242	8.61156	10750
11	12	lounge	51	366	17500	1	45.069679	7.70492	10990
12	13	lounge	51	456	18450	1	45.426571	11.78813	9700
19	20	lounge	51	425	20030	1	45.354389	11.86926	10500
1520	1521	lounge	51	1035	15000	1	41.903221	12.49565	10990
1522	1523	lounge	51	366	14618	1	45.707249	11.47760	10500
1526	1527	lounge	51	1705	23600	1	38.122070	13.36112	9300
1527	1528	рор	51	517	3000	1	40.748241	14.52835	9999
1529	1530	lounge	51	731	22551	1	38.122070	13.36112	9900

492 rows × 9 columns

Out[75]:

	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

907 rows × 9 columns

Out[76]:

		ID	engine_power	age_in_days	km	previous_owners	lat	lon	price
_	model								
-	lounge	734	734	734	734	734	734	734	734
	рор	162	162	162	162	162	162	162	162
	sport	11	11	11	11	11	11	11	11

#### Out[78]:

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

907 rows × 9 columns

In [79]: data1=data.drop(['model'],axis=1)
 data1

# Out[79]:

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

1538 rows × 8 columns

In [80]: corr=datal.corr()
cor

Out[80]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price
ID	1.000000	-0.034059	-0.060753	-0.006537	0.007803	-0.058207	0.058941	0.028516
engine_power	-0.034059	1.000000	0.319190	0.285495	-0.005030	0.005721	-0.005032	-0.277235
age_in_days	-0.060753	0.319190	1.000000	0.833890	0.075775	0.062982	-0.042667	-0.893328
km	-0.006537	0.285495	0.833890	1.000000	0.097539	0.035519	0.004839	-0.859373
previous_owners	0.007803	-0.005030	0.075775	0.097539	1.000000	0.001697	-0.026836	-0.076274
lat	-0.058207	0.005721	0.062982	0.035519	0.001697	1.000000	-0.766646	-0.011733
lon	0.058941	-0.005032	-0.042667	0.004839	-0.026836	-0.766646	1.000000	-0.003541
price	0.028516	-0.277235	-0.893328	-0.859373	-0.076274	-0.011733	-0.003541	1.000000

In [81]: import pandas as pd

In [82]: data=pd.read\_csv("/home/placement/Downloads/fiat500.csv")

In [83]: data.head()

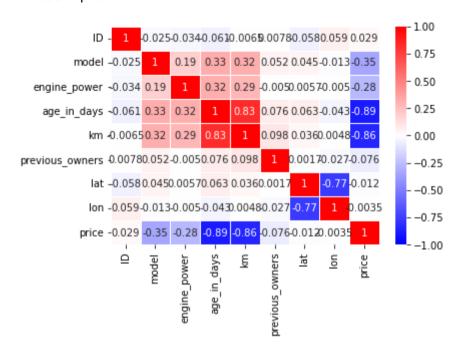
Out[83]:

	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

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
ID	1.000000	-0.024740	-0.034059	-0.060753	-0.006537	0.007803	-0.058207	0.058941	0.028516
model	-0.024740	1.000000	0.189906	0.326508	0.319580	0.052480	0.044901	-0.013200	-0.349885
engine_power	-0.034059	0.189906	1.000000	0.319190	0.285495	-0.005030	0.005721	-0.005032	-0.277235
age_in_days	-0.060753	0.326508	0.319190	1.000000	0.833890	0.075775	0.062982	-0.042667	-0.893328
km	-0.006537	0.319580	0.285495	0.833890	1.000000	0.097539	0.035519	0.004839	-0.859373
previous_owners	0.007803	0.052480	-0.005030	0.075775	0.097539	1.000000	0.001697	-0.026836	-0.076274
lat	-0.058207	0.044901	0.005721	0.062982	0.035519	0.001697	1.000000	-0.766646	-0.011733
lon	0.058941	-0.013200	-0.005032	-0.042667	0.004839	-0.026836	-0.766646	1.000000	-0.003541
price	0.028516	-0.349885	-0.277235	-0.893328	-0.859373	-0.076274	-0.011733	-0.003541	1.000000

```
In [87]: import seaborn as sns
sns.heatmap(cor, vmax=1, vmin=-1, annot=True, linewidths=.5, cmap='bwr')
```

Out[87]: <AxesSubplot:>



```
In [ ]:
```