

## ex 2 set 1

January 12, 2024

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[ ]: EX 2 WORKING WITH DATA USING PANDAS
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[ ]: DATE 18/12/23
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[ ]: URK22AI1048  
HARI HARAN K
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```
[ ]: '''  
#AIM  
    #DATA MANIPULATION WITH PANDAS  
#DESCRIPTION  
    Python Pandas is defined as an open-source library that provides  
    ↪high-performance data  
manipulation in Python. Started by Wes McKinney in 2008 out of a need for a  
    ↪powerful and  
flexible quantitative analysis tool, panda has grown into one of the most  
    ↪popular Python  
libraries. Pandas is built on top of the Numpy package, means Numpy is required  
    ↪for  
operating the Pandas  
  
Create a DataFrame  
data0 = pd.DataFrame()  
  
Read data from .csv  
df = pd.read_csv('filename.csv')  
  
Selection of data from Data Frame  
df['col_name'] #Select one column  
df[['col_name1','col_name2']] # Select more than one column  
  
Filtering data from Data Frame  
df[df['col_name']>value].head()  
  
Filtering missing values  
df[ df[ 'col_name' ].isnull() ]
```

### Manipulating data in Data Frame

The result of all these aggregation functions applied to a row or column is,  
↳ always a

number

Pandas excludes NaN values

The commonly used aggregate functions are min (), max (), count (), sum (),  
↳ mean (),

median (), prod(), std()

# To get aggregate function of all columns

df. aggregate function

# To get aggregate function of specified columns

df['col\_name']. aggregate function

Apply any binary arithmetical operation (+,-,\*,/) to an entire row

s = df['col\_name']/100

### Sorting the data in Data Frame

Sort by column name

df.sort\_values(by='col\_name', ascending = False, inplace = True)

Sort by index value

df.sort\_index(axis = 0, ascending = False, inplace = True)

#inplace=True (overwriting the original DataFrame)

#inplace=False (return a copy of the modified DataFrame)

### Grouping data in Data Frame

Group the data according to some criteria

It is necessary to apply an aggregation function

df.groupby('col\_name').count()

df[['col\_name1', 'col\_name2']].groupby('col\_name2').count()

### Ranking data in Data Frame

The rank is returned on the basis of position after sorting.

Method takes a string input ('average', 'min', 'max', 'first', 'dense') which tells pandas what to do with same values.

Default is average which means assign average of ranks to the similar values.

df.rank()

df['col\_name']. rank(ascending = False, method = 'first')

```
[10]: #URK22AI1048 7Q
import pandas as pd
```

```

df = pd.read_csv('Toyota.csv')

max_km = df['KM'].max()

min_weight = df['Weight'].min()

print("Maximum value in the 'KM' column:", max_km)
print("Minimum value in the 'weight' column:", min_weight)

```

Maximum value in the 'KM' column: 243000  
Minimum value in the 'weight' column: 1015

```

[11]: #URK22AI1048 8Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

mean_cc = df['CC'].mean()

median_cc = df['CC'].median()

print("Mean of the 'CC' column:", mean_cc)
print("Median of the 'CC' column:", median_cc)

```

Mean of the 'CC' column: 1610.3587174348697  
Median of the 'CC' column: 1600.0

```

[13]: #URK22AI1048 9Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

df.loc[df['CC'] == 2000, 'Horse Power'] = df.loc[df['CC'] == 2000, 'HP'] * 2

print(df)

df.to_csv('updated_data.csv', index=False)

```

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	\
0	13500	23	46986	Diesel	90	1	0	2000	3	
1	13750	23	72937	Diesel	90	1	0	2000	3	
2	13950	24	41711	Diesel	90	1	0	2000	3	
3	14950	26	48000	Diesel	90	0	0	2000	3	
4	13750	30	38500	Diesel	90	0	0	2000	3	
..	...	...	...	...	...	...	...	...	...	
494	11950	54	58745	Petrol	110	1	0	1600	4	
495	11250	52	58596	Petrol	110	1	0	1600	3	
496	11750	54	58530	Petrol	110	0	0	1600	5	
497	10950	55	58377	Petrol	110	1	0	1600	3	

498	11250	56	58142	Petrol	110	1	0	1600	5
-----	-------	----	-------	--------	-----	---	---	------	---

	Weight	Horse Power
0	1165	180.0
1	1165	180.0
2	1165	180.0
3	1165	180.0
4	1170	180.0
..	...	...
494	1035	NaN
495	1045	NaN
496	1075	NaN
497	1050	NaN
498	1080	NaN

[499 rows x 11 columns]

```
[28]: #URK22AI1048 10Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

filtered_df = df[df['MetColor'] == 1]

print(filtered_df)
```

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	Weight
0	13500	23	46986	Diesel	90	1	0	2000	3	1165
1	13750	23	72937	Diesel	90	1	0	2000	3	1165
2	13950	24	41711	Diesel	90	1	0	2000	3	1165
6	16900	27	94612	Diesel	90	1	0	2000	3	1245
7	18600	30	75889	Diesel	90	1	0	2000	3	1245
..	...	...	...	...	...	...	...	...	...	...
492	9799	51	59000	Petrol	97	1	0	1400	3	1025
494	11950	54	58745	Petrol	110	1	0	1600	4	1035
495	11250	52	58596	Petrol	110	1	0	1600	3	1045
497	10950	55	58377	Petrol	110	1	0	1600	3	1050
498	11250	56	58142	Petrol	110	1	0	1600	5	1080

[372 rows x 10 columns]

```
[29]: #URK22AI1048 11Q
import pandas as pd
df = pd.read_csv('Toyota.csv')
result_df = df.sort_values(by='Age').head(50)[['Price', 'Age', 'CC']]
print(result_df)
```

	Price	Age	CC
185	18245	1	1600

184	17795	1	1400
183	21500	2	1600
182	21125	2	1400
110	31000	4	2000
111	31275	4	2000
109	32500	4	2000
179	22500	6	1600
177	19950	7	1600
114	22950	7	2000
117	17900	7	1600
181	18700	7	1400
180	18500	7	1600
113	24950	8	2000
115	24990	8	2000
116	21950	8	2000
172	19500	8	1400
112	24950	8	2000
173	18950	8	1400
174	21950	8	1600
175	19950	8	1600
176	18950	8	1600
178	21950	8	1600
171	23750	8	1600
162	19600	9	1600
170	18245	9	1600
169	17795	9	1400
152	18450	10	1400
157	18900	11	1600
98	18750	11	1600
138	23000	11	2000
164	17650	11	1400
104	19450	11	1600
103	18500	11	1600
168	20500	12	1600
153	19500	12	1600
142	19950	13	1600
137	16250	13	1600
133	15950	13	1600
129	15850	13	1600
120	18950	13	1600
102	18500	13	1400
147	24500	13	1600
154	21750	13	1600
122	16350	14	1600
106	18800	14	1600
166	19950	14	1600
149	20950	14	1600
165	19950	14	1600

163 19500 14 1600

```
[39]: #URK22AI1048 12Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

grouped_df = df.groupby(['FuelType', 'MetColor'])

for group_name, group_data in grouped_df:
    print(f"Group: {group_name}")
    print(group_data)
    print("\n")
```

Group: ('CNG', 0)

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	\
189	7750	43	178858	CNG	110	0	0	1600	3	
387	9250	48	142130	CNG	110	0	0	1600	5	

Weight

189	1084
387	1119

Group: ('CNG', 1)

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	\
199	11950	39	98823	CNG	110	1	0	1600	5	
209	11950	37	82743	CNG	110	1	0	1600	5	
223	14950	44	71793	CNG	110	1	0	1600	4	
296	12950	44	41499	CNG	110	1	0	1600	5	
383	8500	55	150000	CNG	110	1	0	1600	3	
436	11500	47	78785	CNG	110	1	0	1600	5	

Weight

199	1119
209	1121
223	1067
296	1103
383	1075
436	1119

Group: ('Diesel', 0)

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	\
3	14950	26	48000	Diesel	90	0	0	2000	3	
4	13750	30	38500	Diesel	90	0	0	2000	3	
5	12950	32	61000	Diesel	90	0	0	2000	3	
9	12950	23	71138	Diesel	69	0	0	1900	3	

43	16950	27	110404	Diesel	90	0	0	2000	5
44	16950	22	100250	Diesel	90	0	0	2000	5
45	19000	23	84000	Diesel	90	0	0	2000	5
109	32500	4	1	Diesel	116	0	0	2000	5
186	6950	43	243000	Diesel	69	0	0	1900	3
187	9500	38	180638	Diesel	90	0	0	2000	4
191	4350	44	158320	Diesel	69	0	0	1800	5
216	13500	33	75699	Diesel	69	0	0	1900	3
378	6500	53	216000	Diesel	69	0	0	1900	3
379	6400	51	198167	Diesel	69	0	0	1900	4
380	7000	53	176000	Diesel	69	0	0	1900	3
386	10250	53	143513	Diesel	69	0	0	1900	5
393	4450	56	129155	Diesel	69	0	0	1800	5
406	10950	51	103018	Diesel	69	0	0	1900	5
416	9950	51	96135	Diesel	72	0	0	2000	3
418	8950	55	94401	Diesel	72	0	0	2000	3
447	10995	49	74656	Diesel	69	0	0	1900	3
487	8950	54	61000	Diesel	69	0	0	2000	5

	Weight
3	1165
4	1170
5	1170
9	1105
43	1255
44	1255
45	1270
109	1480
186	1110
187	1160
191	1110
216	1105
378	1110
379	1095
380	1105
386	1140
393	1110
406	1140
416	1115
418	1115
447	1105
487	1140

Group: ('Diesel', 1)

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	\
0	13500	23	46986	Diesel	90	1	0	2000	3	
1	13750	23	72937	Diesel	90	1	0	2000	3	

2	13950	24	41711	Diesel	90	1	0	2000	3
6	16900	27	94612	Diesel	90	1	0	2000	3
7	18600	30	75889	Diesel	90	1	0	2000	3
46	17950	27	79375	Diesel	90	1	0	2000	5
48	17950	22	72215	Diesel	90	1	0	2000	5
50	17950	22	62636	Diesel	90	1	0	2000	5
68	22250	22	30000	Diesel	110	1	0	2000	5
87	17950	20	66966	Diesel	90	1	0	2000	3
89	21950	19	50005	Diesel	110	1	0	2000	3
91	22250	20	37500	Diesel	90	1	0	2000	3
92	19950	16	34472	Diesel	90	1	0	1995	3
95	19950	17	30351	Diesel	90	1	0	1995	3
110	31000	4	4000	Diesel	116	1	0	2000	5
111	31275	4	1500	Diesel	116	1	0	2000	5
112	24950	8	13253	Diesel	116	1	0	2000	5
113	24950	8	13253	Diesel	116	1	0	2000	5
114	22950	7	10000	Diesel	116	1	0	2000	5
115	24990	8	6000	Diesel	90	1	0	2000	5
116	21950	8	10841	Diesel	90	1	0	2000	5
118	19250	20	63000	Diesel	90	1	0	2000	5
119	22250	17	57313	Diesel	110	1	0	2000	5
121	19950	19	51099	Diesel	90	1	0	2000	5
138	23000	11	25000	Diesel	116	1	0	2000	5
188	11950	40	179860	Diesel	90	1	0	2000	5
190	11950	40	161000	Diesel	69	1	0	1900	3
192	4750	44	131273	Diesel	69	1	0	1800	5
193	11750	40	130062	Diesel	69	1	0	1900	5
194	13250	41	123425	Diesel	69	1	0	1900	5
196	11900	44	110000	Diesel	69	1	0	1900	5
197	14750	39	108847	Diesel	90	1	0	2000	5
203	10450	35	91456	Diesel	69	1	0	1900	3
204	12950	43	89968	Diesel	69	1	0	1900	5
210	13250	41	81106	Diesel	69	1	0	1900	5
211	14750	40	80425	Diesel	90	1	0	2000	5
214	13500	33	78108	Diesel	90	1	0	2000	3
237	13950	35	59500	Diesel	69	1	0	1900	3
243	13500	33	57711	Diesel	90	1	0	2000	3
268	14750	40	48952	Diesel	90	1	0	2000	5
270	13500	33	48928	Diesel	69	1	0	1900	3
272	13500	35	48052	Diesel	69	1	0	1900	3
381	7750	54	174139	Diesel	72	1	0	2000	4
382	8900	45	174000	Diesel	69	1	0	1900	5
384	8950	54	149329	Diesel	72	1	0	2000	5
388	7750	48	140700	Diesel	69	1	0	1900	5
389	9450	54	138394	Diesel	69	1	0	1900	4
390	7750	55	137000	Diesel	72	1	0	2000	5
391	8250	52	135258	Diesel	69	1	0	1900	5
396	9950	53	117913	Diesel	69	1	0	1900	5



397	12450	47	117430	Diesel	90	1	0	2000	3
401	10500	54	115046	Diesel	69	1	0	1900	5
402	5150	56	113997	Diesel	72	1	0	2000	5
412	8950	48	98100	Diesel	69	1	0	1900	5
422	9250	53	90097	Diesel	69	1	0	1900	5
458	8695	50	70440	Diesel	69	1	0	1900	3
463	8750	47	69000	Diesel	69	1	0	1900	5
465	11450	55	68520	Diesel	72	1	0	2000	3
480	11500	48	63000	Diesel	69	1	0	1900	5

	Weight
0	1165
1	1165
2	1165
6	1245
7	1245
46	1255
48	1255
50	1255
68	1275
87	1245
89	1265
91	1260
92	1260
95	1260
110	1480
111	1480
112	1320
113	1320
114	1270
115	1280
116	1270
118	1255
119	1275
121	1255
138	1320
188	1205
190	1105
192	1110
193	1140
194	1140
196	1095
197	1205
203	1110
204	1140
210	1140
211	1205
214	1170

237 1110  
 243 1165  
 268 1205  
 270 1105  
 272 1105  
 381 1100  
 382 1095  
 384 1135  
 388 1110  
 389 1095  
 390 1135  
 391 1140  
 396 1110  
 397 1165  
 401 1140  
 402 1135  
 412 1140  
 422 1140  
 458 1105  
 463 1140  
 465 1115  
 480 1140

Group: ('Petrol', 0)

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	Weight
8	21500	27	19700	Petrol	192	0	0	1800	3	1185
10	20950	25	31461	Petrol	192	0	0	1800	3	1185
11	19950	22	43610	Petrol	192	0	0	1800	3	1185
12	19600	25	32189	Petrol	192	0	0	1800	3	1185
15	22000	28	18739	Petrol	192	0	0	1800	3	1185
..	...	...	...	...	...	...	...	...	...	...
484	9500	54	62519	Petrol	97	0	0	1400	3	1025
486	10750	50	61672	Petrol	110	0	0	1600	5	1075
490	10950	47	60348	Petrol	110	0	0	1600	4	1030
493	11750	51	58761	Petrol	97	0	0	1300	5	1060
496	11750	54	58530	Petrol	110	0	0	1600	5	1075

[103 rows x 10 columns]

Group: ('Petrol', 1)

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	Weight
13	21500	31	23000	Petrol	192	1	0	1800	3	1185
14	22500	32	34131	Petrol	192	1	0	1800	3	1185
16	22750	30	34000	Petrol	192	1	0	1800	3	1185
17	17950	24	21716	Petrol	110	1	0	1600	3	1105
19	16950	30	64359	Petrol	110	1	0	1600	3	1105

..	...	...	...	...	...	...	...	...	...	...	...	...
492	9799	51	59000	Petrol	97		1		0	1400	3	1025
494	11950	54	58745	Petrol	110		1		0	1600	4	1035
495	11250	52	58596	Petrol	110		1		0	1600	3	1045
497	10950	55	58377	Petrol	110		1		0	1600	3	1050
498	11250	56	58142	Petrol	110		1		0	1600	5	1080

[307 rows x 10 columns]

```
[40]: #URK22AI1048 13Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

null_counts = df.isnull().sum()

print("Null counts:")
print(null_counts)

df['CC'].fillna('25', inplace=True)

print("\nDataFrame after filling null values:")
print(df)
```

Null counts:

Price	0
Age	0
KM	0
FuelType	0
HP	0
MetColor	0
Automatic	0
CC	0
Doors	0
Weight	0

dtype: int64

DataFrame after filling null values:

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	Weight
0	13500	23	46986	Diesel	90	1	0	2000	3	1165
1	13750	23	72937	Diesel	90	1	0	2000	3	1165
2	13950	24	41711	Diesel	90	1	0	2000	3	1165
3	14950	26	48000	Diesel	90	0	0	2000	3	1165
4	13750	30	38500	Diesel	90	0	0	2000	3	1170
..	...	...	...	...	...	...	...	...	...	...
494	11950	54	58745	Petrol	110	1	0	1600	4	1035

495	11250	52	58596	Petrol	110	1	0	1600	3	1045
496	11750	54	58530	Petrol	110	0	0	1600	5	1075
497	10950	55	58377	Petrol	110	1	0	1600	3	1050
498	11250	56	58142	Petrol	110	1	0	1600	5	1080

[499 rows x 10 columns]

```
[45]: #URK22AI1048 14Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

new_rows = {'Price': [25000, 30000, 18000],
            'Age': [3, 5, 2],
            'CC': [1800, 2200, 1600],
            'Fuel Type': ['Petrol', 'Diesel', 'Petrol'],
            'Metcolor': [1, 0, 1],
            'HP': [120, 150, 100]}

df = df.append(pd.DataFrame(new_rows), ignore_index=True)

print(df)
```

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	\
0	13500	23	46986.0	Diesel	90	1.0	0.0	2000	3.0	
1	13750	23	72937.0	Diesel	90	1.0	0.0	2000	3.0	
2	13950	24	41711.0	Diesel	90	1.0	0.0	2000	3.0	
3	14950	26	48000.0	Diesel	90	0.0	0.0	2000	3.0	
4	13750	30	38500.0	Diesel	90	0.0	0.0	2000	3.0	
..	...	...	...	...	...	...	...	...	...	
497	10950	55	58377.0	Petrol	110	1.0	0.0	1600	3.0	
498	11250	56	58142.0	Petrol	110	1.0	0.0	1600	5.0	
499	25000	3	NaN	NaN	120	NaN	NaN	1800	NaN	
500	30000	5	NaN	NaN	150	NaN	NaN	2200	NaN	
501	18000	2	NaN	NaN	100	NaN	NaN	1600	NaN	

	Weight	Fuel Type	Metcolor
0	1165.0	NaN	NaN
1	1165.0	NaN	NaN
2	1165.0	NaN	NaN
3	1165.0	NaN	NaN
4	1170.0	NaN	NaN
..	...	...	...
497	1050.0	NaN	NaN
498	1080.0	NaN	NaN
499	NaN	Petrol	1.0
500	NaN	Diesel	0.0

501      NaN      Petrol      1.0

[502 rows x 12 columns]

/tmp/ipykernel\_1340718/2691036192.py:14: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.

```
df = df.append(pd.DataFrame(new_rows), ignore_index=True)
```

```
[1]: #URK22AI1048 15Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

df = df.drop(columns=['HP'])

print(df)
```

	Price	Age	KM	FuelType	MetColor	Automatic	CC	Doors	Weight
0	13500	23	46986	Diesel	1	0	2000	3	1165
1	13750	23	72937	Diesel	1	0	2000	3	1165
2	13950	24	41711	Diesel	1	0	2000	3	1165
3	14950	26	48000	Diesel	0	0	2000	3	1165
4	13750	30	38500	Diesel	0	0	2000	3	1170
..	...	...	...	...	...	...	...	...	...
494	11950	54	58745	Petrol	1	0	1600	4	1035
495	11250	52	58596	Petrol	1	0	1600	3	1045
496	11750	54	58530	Petrol	0	0	1600	5	1075
497	10950	55	58377	Petrol	1	0	1600	3	1050
498	11250	56	58142	Petrol	1	0	1600	5	1080

[499 rows x 9 columns]

```
[2]: #URK22AI1048 16Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

df = df.drop([49, 99])

df.reset_index(drop=True, inplace=True)

print(df)
```

	Price	Age	KM	FuelType	HP	MetColor	Automatic	CC	Doors	Weight
0	13500	23	46986	Diesel	90	1	0	2000	3	1165
1	13750	23	72937	Diesel	90	1	0	2000	3	1165
2	13950	24	41711	Diesel	90	1	0	2000	3	1165
3	14950	26	48000	Diesel	90	0	0	2000	3	1165

4	13750	30	38500	Diesel	90	0	0	2000	3	1170
..	...	...	...	...	...	...	...	...	...	...
492	11950	54	58745	Petrol	110	1	0	1600	4	1035
493	11250	52	58596	Petrol	110	1	0	1600	3	1045
494	11750	54	58530	Petrol	110	0	0	1600	5	1075
495	10950	55	58377	Petrol	110	1	0	1600	3	1050
496	11250	56	58142	Petrol	110	1	0	1600	5	1080

[497 rows x 10 columns]

```
[3]: #URK22AI1048 17Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

filtered_df = df[df['CC'] > 1600]

result_df = filtered_df[['Price', 'KM', 'HP']].head(100)

print(result_df)
```

	Price	KM	HP
0	13500	46986	90
1	13750	72937	90
2	13950	41711	90
3	14950	48000	90
4	13750	38500	90
..	...	...	..
458	8695	70440	69
463	8750	69000	69
465	11450	68520	72
480	11500	63000	69
487	8950	61000	69

[92 rows x 3 columns]

```
[4]: #URK22AI1048 18Q
import pandas as pd

df = pd.read_csv('Toyota.csv')

row_index = 5
col_name = 'Price'

data_loc = df.loc[row_index, col_name]
print(f"Using loc: {data_loc}")
```

```

row_index = 5
col_index = 2

data_iloc = df.iloc[row_index, col_index]
print(f"Using iloc: {data_iloc}")

```

Using loc: 12950  
Using iloc: 61000

```

[6]: #URK22AI1048 19Q
import pandas as pd
df = pd.read_csv('Toyota.csv')
print("Index:")
print(df.index)

print("\nColumn Names:")
print(df.columns)

print("\nDescriptive Statistics:")
print(df.describe())

print("\nSize:")
print(df.size)

print("\nNumber of Dimensions (ndim):")
print(df.ndim)

print("\nShape:")
print(df.shape)

print("\nInfo:")
print(df.info())

```

Index:  
RangeIndex(start=0, stop=499, step=1)

Column Names:  
Index(['Price', 'Age', 'KM', 'FuelType', 'HP', 'MetColor', 'Automatic', 'CC',  
 'Doors', 'Weight'],  
 dtype='object')

Descriptive Statistics:

	Price	Age	KM	HP	MetColor	\
count	499.000000	499.000000	499.000000	499.000000	499.000000	
mean	14072.026052	35.124248	53237.234469	103.951904	0.745491	
std	4059.047783	13.815937	36963.132169	17.882192	0.436022	
min	4350.000000	1.000000	1.000000	69.000000	0.000000	
25%	11450.000000	25.000000	27910.500000	97.000000	0.000000	

50%	12950.000000	39.000000	45725.000000	110.000000	1.000000
75%	16700.000000	44.000000	71759.000000	110.000000	1.000000
max	32500.000000	56.000000	243000.000000	192.000000	1.000000

	Automatic	CC	Doors	Weight
count	499.000000	499.000000	499.000000	499.000000
mean	0.044088	1610.358717	4.236473	1099.643287
std	0.205497	180.135440	0.933638	66.825080
min	0.000000	1300.000000	3.000000	1015.000000
25%	0.000000	1400.000000	3.000000	1060.000000
50%	0.000000	1600.000000	5.000000	1080.000000
75%	0.000000	1600.000000	5.000000	1119.000000
max	1.000000	2000.000000	5.000000	1615.000000

Size:  
4990

Number of Dimensions (ndim):  
2

Shape:  
(499, 10)

Info:  
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 499 entries, 0 to 498  
Data columns (total 10 columns):

#	Column	Non-Null Count	Dtype
0	Price	499 non-null	int64
1	Age	499 non-null	int64
2	KM	499 non-null	int64
3	FuelType	499 non-null	object
4	HP	499 non-null	int64
5	MetColor	499 non-null	int64
6	Automatic	499 non-null	int64
7	CC	499 non-null	int64
8	Doors	499 non-null	int64
9	Weight	499 non-null	int64

dtypes: int64(9), object(1)  
memory usage: 39.1+ KB  
None

```
[ ]: '''
    RESULT
    The above programs were created and executed successfully.
    '''
```