Basic operations using NumPy and Pandas

Importing libraries

```
In [1]: import numpy as np import pandas as pd
```

importing the dataset

```
In [3]: data=pd.read_csv(r"C:\Users\user\Desktop\Ash\fiat500_VehicleSelection_Dataset.csv")
```

Selecting first 10 rows using head()

In [6]: data.head(10)

Out[6]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price	Unnamed: 9	Unnamed: 10	Unnam
0	1.0	lounge	51.0	882.0	25000.0	1.0	44.907242	8.611560	8900.0	NaN	NaN	CONCATE
1	2.0	рор	51.0	1186.0	32500.0	1.0	45.666359	12.241890	8800.0	NaN	NaN	Length
2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.417840	4200.0	NaN	NaN	No. of w days btw 23 to 17
3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.634609	6000.0	NaN	NaN	
4	5.0	pop	73.0	3074.0	106880.0	1.0	41.903221	12.495650	5700.0	NaN	NaN	Ave
5	6.0	pop	74.0	3623.0	70225.0	1.0	45.000702	7.682270	7900.0	NaN	NaN	С
6	7.0	lounge	51.0	731.0	11600.0	1.0	44.907242	8.611560	10750.0	NaN	NaN	
7	8.0	lounge	51.0	1521.0	49076.0	1.0	41.903221	12.495650	9190.0	NaN	NaN	
8	9.0	sport	73.0	4049.0	76000.0	1.0	45.548000	11.549470	5600.0	NaN	NaN	
9	10.0	sport	51.0	3653.0	89000.0	1.0	45.438301	10.991700	6000.0	NaN	NaN	
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Selecting last 10 rows using tail()

In [7]: data.tail(10)
Out[7]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price	Unnamed: 9	l
1531	1532.0	sport	73.0	4505.0	127000.0000	1.0	45.528511	9.59323	4.750000e+03	NaN	_
1532	1533.0	рор	51.0	1917.0	52008.0000	1.0	45.548000	11.54947	9.900000e+03	NaN	
1533	1534.0	sport	51.0	3712.0	115280.0000	1.0	45.069679	7.70492	5.200000e+03	NaN	
1534	1535.0	lounge	74.0	3835.0	112000.0000	1.0	45.845692	8.66687	4.600000e+03	NaN	
1535	1536.0	рор	51.0	2223.0	60457.0000	1.0	45.481541	9.41348	7.500000e+03	NaN	
1536	1537.0	lounge	51.0	2557.0	80750.0000	1.0	45.000702	7.68227	5.990000e+03	NaN	
1537	1538.0	рор	51.0	1766.0	54276.0000	1.0	40.323410	17.56827	7.900000e+03	NaN	
1538	NaN	NaN	NaN	2080506.0	53396.0117	NaN	66966.613720	17784.55279	1.318989e+07	NaN	
1539	NaN	NaN	1538.0	NaN	NaN	NaN	NaN	NaN	1.714086e+04	NaN	
1540	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1.318989e+07	NaN	
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To get the statistical data of the table

	ID	er	ngine_power	age_in_days		km	previous_owners	lat	lon	price	Unnamed: 9
38.	.000000		1539.000000	1.539000e+03	1539.000	0000	1538.000000	1539.000000	1539.000000	1.541000e+03	0.0
69.	.500000		52.870045	3.001763e+03	53396.011	704	1.123537	87.026139	23.111829	2.568905e+04	NaN
14.	.126671		38.090731	5.300702e+04	40033.809	481	0.416423	1705.913084	453.050800	4.747172e+05	NaN
1.	.000000		51.000000	3.660000e+02	1232.000	0000	1.000000	36.855839	7.245400	2.500000e+03	NaN
35.	.250000		51.000000	6.700000e+02	20012.500	0000	1.000000	41.802990	9.505090	7.190000e+03	NaN
39.	.500000		51.000000	1.035000e+03	39038.000	0000	1.000000	44.399971	11.869260	9.000000e+03	NaN
53.	.750000		51.000000	2.616000e+03	79535.500	0000	1.000000	45.467960	12.774520	1.000000e+04	NaN
38.	.000000		1538.000000	2.080506e+06	235000.000	0000	4.000000	66966.613720	17784.552790	1.318989e+07	NaN
											•

To find the row and column of the table

```
In [10]: data.shape
Out[10]: (1541, 15)
```

To find the size of the table

```
In [11]: data.size
Out[11]: 23115
```

Find missing values

In [8]: data.describe()

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price	Unnamed: 9	Unnamed: 10	Unnamed: 11	Unname
0	False	False	False	False	False	False	False	False	False	True	True	False	Fal
1	False	False	False	False	False	False	False	False	False	True	True	False	Fal
2	False	False	False	False	False	False	False	False	False	True	True	False	Fal
3	False	False	False	False	False	False	False	False	False	True	True	False	Fal
4	False	False	False	False	False	False	False	False	False	True	True	False	Fal
1536	False	False	False	False	False	False	False	False	False	True	True	True	Tr
1537	False	False	False	False	False	False	False	False	False	True	True	True	Tr
1538	True	True	True	False	False	True	False	False	False	True	True	True	Tr
1539	True	True	False	True	True	True	True	True	False	True	True	True	Tr
1540	True	True	True	True	True	True	True	True	False	True	True	True	Tr

filling the missing values as 4

In [12]: data.isna()

Out[12]:

In [13]: data.fillna(value=4)

Out[13]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price	Unnamed: 9
0	1.0	lounge	51.0	882.0	25000.0000	1.0	44.907242	8.611560	8.900000e+03	4.0
1	2.0	рор	51.0	1186.0	32500.0000	1.0	45.666359	12.241890	8.800000e+03	4.0
2	3.0	sport	74.0	4658.0	142228.0000	1.0	45.503300	11.417840	4.200000e+03	4.0
3	4.0	lounge	51.0	2739.0	160000.0000	1.0	40.633171	17.634609	6.000000e+03	4.0
4	5.0	pop	73.0	3074.0	106880.0000	1.0	41.903221	12.495650	5.700000e+03	4.0
1536	1537.0	lounge	51.0	2557.0	80750.0000	1.0	45.000702	7.682270	5.990000e+03	4.0
1537	1538.0	рор	51.0	1766.0	54276.0000	1.0	40.323410	17.568270	7.900000e+03	4.0
1538	4.0	4	4.0	2080506.0	53396.0117	4.0	66966.613720	17784.552790	1.318989e+07	4.0
1539	4.0	4	1538.0	4.0	4.0000	4.0	4.000000	4.000000	1.714086e+04	4.0
1540	4.0	4	4.0	4.0	4.0000	4.0	4.000000	4.000000	1.318989e+07	4.0
1541 ı	ows × 1	5 colum	nns							

Data Set 2 [2015]

Importing the dataset ¶

In [14]: data1=pd.read_csv(r"C:\Users\user\Desktop\Ash\2015.csv")

Select first 10 rows

In [17]: data1.head(10)

Out[17]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity	Dystopia Residua
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678	2.51738
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630	2.70201
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139	2.49204
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699	2.46531
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811	2.45176
5	Finland	Western Europe	6	7.406	0.03140	1.29025	1.31826	0.88911	0.64169	0.41372	0.23351	2.6195
6	Netherlands	Western Europe	7	7.378	0.02799	1.32944	1.28017	0.89284	0.61576	0.31814	0.47610	2.4657(
7	Sweden	Western Europe	8	7.364	0.03157	1.33171	1.28907	0.91087	0.65980	0.43844	0.36262	2.37119
8	New Zealand	Australia and New Zealand	9	7.286	0.03371	1.25018	1.31967	0.90837	0.63938	0.42922	0.47501	2.2642{
9	Australia	Australia and New Zealand	10	7.284	0.04083	1.33358	1.30923	0.93156	0.65124	0.35637	0.43562	2.2664€
4 6												

Select the last 10 rows

In [18]: data1.tail(10)

Out[18]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity	Dysto Resid
148	Chad	Sub- Saharan Africa	149	3.667	0.03830	0.34193	0.76062	0.15010	0.23501	0.05269	0.18386	1.942
149	Guinea	Sub- Saharan Africa	150	3.656	0.03590	0.17417	0.46475	0.24009	0.37725	0.12139	0.28657	1.99 ⁻
150	Ivory Coast	Sub- Saharan Africa	151	3.655	0.05141	0.46534	0.77115	0.15185	0.46866	0.17922	0.20165	1.41
151	Burkina Faso	Sub- Saharan Africa	152	3.587	0.04324	0.25812	0.85188	0.27125	0.39493	0.12832	0.21747	1.464
152	Afghanistan	Southern Asia	153	3.575	0.03084	0.31982	0.30285	0.30335	0.23414	0.09719	0.36510	1.952
153	Rwanda	Sub- Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864	0.59201	0.55191	0.22628	0.67(
154	Benin	Sub- Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910	0.48450	0.08010	0.18260	1.63(
155	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193	0.15684	0.18906	0.47179	0.328
156	Burundi	Sub- Saharan Africa	157	2.905	0.08658	0.01530	0.41587	0.22396	0.11850	0.10062	0.19727	1.83(
157	Togo	Sub- Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443	0.36453	0.10731	0.16681	1.567
4												•

To get the statistical data of the table

```
In [20]: data1.describe()
```

Out[20]:

Out[24]: 1896

	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity	Dystopia Residual
count	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000
mean	79.493671	5.375734	0.047885	0.846137	0.991046	0.630259	0.428615	0.143422	0.237296	2.098977
std	45.754363	1.145010	0.017146	0.403121	0.272369	0.247078	0.150693	0.120034	0.126685	0.553550
min	1.000000	2.839000	0.018480	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.328580
25%	40.250000	4.526000	0.037268	0.545808	0.856823	0.439185	0.328330	0.061675	0.150553	1.759410
50%	79.500000	5.232500	0.043940	0.910245	1.029510	0.696705	0.435515	0.107220	0.216130	2.095415
75%	118.750000	6.243750	0.052300	1.158448	1.214405	0.811013	0.549092	0.180255	0.309883	2.462415
max	158.000000	7.587000	0.136930	1.690420	1.402230	1.025250	0.669730	0.551910	0.795880	3.602140

To find the row and columns of the table

```
In [23]: data1.shape
Out[23]: (158, 12)
```

Find the size of the table

```
In [24]: data1.size
```

Find the missing values of the table

In [25]: data1.isna()

Out[25]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity	Dystopia Residual
0	False	False	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False	False
153	False	False	False	False	False	False	False	False	False	False	False	False
154	False	False	False	False	False	False	False	False	False	False	False	False
155	False	False	False	False	False	False	False	False	False	False	False	False
156	False	False	False	False	False	False	False	False	False	False	False	False
157	False	False	False	False	False	False	False	False	False	False	False	False

158 rows × 12 columns

To delete to rows containing empty cells

In [26]: data1.dropna()

Out[26]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity	Dystor Residu
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678	2.517
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630	2.702
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139	2.492
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699	2.465
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811	2.451
153	Rwanda	Sub- Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864	0.59201	0.55191	0.22628	0.670
154	Benin	Sub- Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910	0.48450	0.08010	0.18260	1.633
155	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193	0.15684	0.18906	0.47179	0.328
156	Burundi	Sub- Saharan Africa	157	2.905	0.08658	0.01530	0.41587	0.22396	0.11850	0.10062	0.19727	1.833
157	Togo	Sub- Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443	0.36453	0.10731	0.16681	1.567
158 r	rows × 12 co	lumns										

158 rows × 12 columns

In [] •	
TH [] •	