

Python pandas and numpy

July 21, 2023

```
[92]: !pip install pandas  
      !pip install numpy
```

```
Requirement already satisfied: pandas in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages  
(2.0.3)  
Requirement already satisfied: python-dateutil>=2.8.2 in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages (from  
pandas) (2.8.2)  
Requirement already satisfied: pytz>=2020.1 in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages (from  
pandas) (2023.3)  
Requirement already satisfied: tzdata>=2022.1 in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages (from  
pandas) (2023.3)  
Requirement already satisfied: numpy>=1.21.0 in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages (from  
pandas) (1.25.1)  
Requirement already satisfied: six>=1.5 in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages (from  
python-dateutil>=2.8.2->pandas) (1.16.0)  
Requirement already satisfied: numpy in  
c:\users\shreyas\appdata\local\programs\python\python311\lib\site-packages  
(1.25.1)
```

```
[93]: import pandas as pd  
      import numpy as np
```

1. Create any Series and print the output

```
[94]: series = pd.Series([1,2,3,4,5,6,7,8,9,10])  
      series
```

```
[94]: 0      1  
      1      2  
      2      3  
      3      4  
      4      5  
      5      6
```

```
6      7
7      8
8      9
9     10
dtype: int64
```

2. Create any dataframe of 10x5 with few nan values and print the output

```
[95]: dataframe = pd.DataFrame({
      "A": [1,2,3,4,5,6,7,8,9,10],
      "B": ["A","B","C","D","E","F","G","H","I","J"],
      "C": [20,20,20,np.nan,29,np.nan,np.nan,np.nan,27,18]
    })
dataframe
```

```
[95]:
```

	A	B	C
0	1	A	20.0
1	2	B	20.0
2	3	C	20.0
3	4	D	NaN
4	5	E	29.0
5	6	F	NaN
6	7	G	NaN
7	8	H	NaN
8	9	I	27.0
9	10	J	18.0

3.Display top 7 and last 6 rows and print the output

```
[96]: dataframe.head(7)
```

```
[96]:
```

	A	B	C
0	1	A	20.0
1	2	B	20.0
2	3	C	20.0
3	4	D	NaN
4	5	E	29.0
5	6	F	NaN
6	7	G	NaN

```
[97]: dataframe.tail(6)
```

```
[97]:
```

	A	B	C
4	5	E	29.0
5	6	F	NaN
6	7	G	NaN
7	8	H	NaN
8	9	I	27.0

```
9  10  J  18.0
```

4. Fill with a constant value and print the output

```
[98]: d1 = dataframe.fillna(value=25)
      d1
```

```
[98]:      A  B      C
0     1  A  20.0
1     2  B  20.0
2     3  C  20.0
3     4  D  25.0
4     5  E  29.0
5     6  F  25.0
6     7  G  25.0
7     8  H  25.0
8     9  I  27.0
9    10  J  18.0
```

5. Drop the column with missing values and print the output

```
[99]: d2 = dataframe.dropna(axis=1)
      d2
```

```
[99]:      A  B
0     1  A
1     2  B
2     3  C
3     4  D
4     5  E
5     6  F
6     7  G
7     8  H
8     9  I
9    10  J
```

6. Drop the row with missing values and print the output

```
[100]: d3 = dataframe.dropna()
      d3
```

```
[100]:      A  B      C
0     1  A  20.0
1     2  B  20.0
2     3  C  20.0
4     5  E  29.0
8     9  I  27.0
9    10  J  18.0
```

7. To check the presence of missing values in your dataframe

```
[101]: dataframe.isnull()
```

```
[101]:
```

	A	B	C
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	True
4	False	False	False
5	False	False	True
6	False	False	True
7	False	False	True
8	False	False	False
9	False	False	False

8. Use operators and check the condition and print the output

```
[102]: d4 = dataframe.loc[dataframe["C"]>18]  
d4
```

```
[102]:
```

	A	B	C
0	1	A	20.0
1	2	B	20.0
2	3	C	20.0
4	5	E	29.0
8	9	I	27.0

9. Display your output using loc and iloc, row and column heading

```
[103]: d5 = dataframe.loc[0:5]  
d5
```

```
[103]:
```

	A	B	C
0	1	A	20.0
1	2	B	20.0
2	3	C	20.0
3	4	D	NaN
4	5	E	29.0
5	6	F	NaN

```
[104]: d6 = dataframe.iloc[0:2]  
d6
```

```
[104]:
```

	A	B	C
0	1	A	20.0
1	2	B	20.0

10. Display the statistical summary of data

```
[105]: dataframe.shape
```

[105]: (10, 3)

```
[106]: dataframe.describe()
```

```
[106]:
```

	A	C
count	10.00000	6.000000
mean	5.50000	22.333333
std	3.02765	4.501851
min	1.00000	18.000000
25%	3.25000	20.000000
50%	5.50000	20.000000
75%	7.75000	25.250000
max	10.00000	29.000000

1 MINI-PROJECT:

```
[107]: import pandas as pd
import numpy as np
```

```
[108]: data = pd.read_csv(r"C:\Users\SHREYAS\Downloads\dataset.csv")
data
```

```
[108]:
```

	Country	Region	Happiness Rank	\
0	Switzerland	Western Europe	1	
1	Iceland	Western Europe	2	
2	Denmark	Western Europe	3	
3	Norway	Western Europe	4	
4	Canada	North America	5	
..	
153	Rwanda	Sub-Saharan Africa	154	
154	Benin	Sub-Saharan Africa	155	
155	Syria	Middle East and Northern Africa	156	
156	Burundi	Sub-Saharan Africa	157	
157	Togo	Sub-Saharan Africa	158	

	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	\
0	7.587	0.03411	1.39651	1.34951	
1	7.561	0.04884	1.30232	1.40223	
2	7.527	0.03328	1.32548	1.36058	
3	7.522	0.03880	1.45900	1.33095	
4	7.427	0.03553	1.32629	1.32261	
..	
153	3.465	0.03464	0.22208	0.77370	
154	3.340	0.03656	0.28665	0.35386	
155	3.006	0.05015	0.66320	0.47489	
156	2.905	0.08658	0.01530	0.41587	
157	2.839	0.06727	0.20868	0.13995	

	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	\
0	0.94143	0.66557		0.41978
1	0.94784	0.62877		0.14145
2	0.87464	0.64938		0.48357
3	0.88521	0.66973		0.36503
4	0.90563	0.63297		0.32957
..
153	0.42864	0.59201		0.55191
154	0.31910	0.48450		0.08010
155	0.72193	0.15684		0.18906
156	0.22396	0.11850		0.10062
157	0.28443	0.36453		0.10731

	Generosity	Dystopia Residual
0	0.29678	2.51738
1	0.43630	2.70201
2	0.34139	2.49204
3	0.34699	2.46531
4	0.45811	2.45176
..
153	0.22628	0.67042
154	0.18260	1.63328
155	0.47179	0.32858
156	0.19727	1.83302
157	0.16681	1.56726

[158 rows x 12 columns]

```
[109]: data.head()
```

```
[109]:
```

	Country	Region	Happiness Rank	Happiness Score	\
0	Switzerland	Western Europe	1	7.587	
1	Iceland	Western Europe	2	7.561	
2	Denmark	Western Europe	3	7.527	
3	Norway	Western Europe	4	7.522	
4	Canada	North America	5	7.427	

	Standard Error	Economy (GDP per Capita)	Family	\
0	0.03411	1.39651	1.34951	
1	0.04884	1.30232	1.40223	
2	0.03328	1.32548	1.36058	
3	0.03880	1.45900	1.33095	
4	0.03553	1.32629	1.32261	

	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	\
0	0.94143	0.66557		0.41978

1	0.94784	0.62877	0.14145
2	0.87464	0.64938	0.48357
3	0.88521	0.66973	0.36503
4	0.90563	0.63297	0.32957

	Generosity	Dystopia	Residual
0	0.29678		2.51738
1	0.43630		2.70201
2	0.34139		2.49204
3	0.34699		2.46531
4	0.45811		2.45176

```
[110]: data.tail()
```

```
[110]:
```

	Country	Region	Happiness Rank	\
153	Rwanda	Sub-Saharan Africa	154	
154	Benin	Sub-Saharan Africa	155	
155	Syria	Middle East and Northern Africa	156	
156	Burundi	Sub-Saharan Africa	157	
157	Togo	Sub-Saharan Africa	158	

	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	\
153	3.465	0.03464	0.22208	0.77370	
154	3.340	0.03656	0.28665	0.35386	
155	3.006	0.05015	0.66320	0.47489	
156	2.905	0.08658	0.01530	0.41587	
157	2.839	0.06727	0.20868	0.13995	

	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	\
153	0.42864	0.59201	0.55191	
154	0.31910	0.48450	0.08010	
155	0.72193	0.15684	0.18906	
156	0.22396	0.11850	0.10062	
157	0.28443	0.36453	0.10731	

	Generosity	Dystopia	Residual
153	0.22628		0.67042
154	0.18260		1.63328
155	0.47179		0.32858
156	0.19727		1.83302
157	0.16681		1.56726

```
[111]: data.describe()
```

```
[111]:
```

	Happiness Rank	Happiness Score	Standard Error	\
count	158.000000	158.000000	158.000000	
mean	79.493671	5.375734	0.047885	

std	45.754363	1.145010	0.017146
min	1.000000	2.839000	0.018480
25%	40.250000	4.526000	0.037268
50%	79.500000	5.232500	0.043940
75%	118.750000	6.243750	0.052300
max	158.000000	7.587000	0.136930

	Economy (GDP per Capita)	Family Health (Life Expectancy)	\
count	158.000000	158.000000	158.000000
mean	0.846137	0.991046	0.630259
std	0.403121	0.272369	0.247078
min	0.000000	0.000000	0.000000
25%	0.545808	0.856823	0.439185
50%	0.910245	1.029510	0.696705
75%	1.158448	1.214405	0.811013
max	1.690420	1.402230	1.025250

	Freedom Trust (Government Corruption)	Generosity	\
count	158.000000	158.000000	158.000000
mean	0.428615	0.143422	0.237296
std	0.150693	0.120034	0.126685
min	0.000000	0.000000	0.000000
25%	0.328330	0.061675	0.150553
50%	0.435515	0.107220	0.216130
75%	0.549092	0.180255	0.309883
max	0.669730	0.551910	0.795880

	Dystopia Residual
count	158.000000
mean	2.098977
std	0.553550
min	0.328580
25%	1.759410
50%	2.095415
75%	2.462415
max	3.602140

```
[112]: data.shape
```

```
[112]: (158, 12)
```

```
[113]: data.size
```

```
[113]: 1896
```

```
[114]: data.isnull()
```



```
[114]:
```

	Country	Region	Happiness Rank	Happiness Score	Standard Error	\
0	False	False	False	False	False	
1	False	False	False	False	False	
2	False	False	False	False	False	
3	False	False	False	False	False	
4	False	False	False	False	False	
..	
153	False	False	False	False	False	
154	False	False	False	False	False	
155	False	False	False	False	False	
156	False	False	False	False	False	
157	False	False	False	False	False	

	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
..	
153	False	False	False	False	
154	False	False	False	False	
155	False	False	False	False	
156	False	False	False	False	
157	False	False	False	False	

	Trust (Government Corruption)	Generosity	Dystopia Residual
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False
..
153	False	False	False
154	False	False	False
155	False	False	False
156	False	False	False
157	False	False	False

[158 rows x 12 columns]

```
[115]: data.fillna(value=0)
```

```
[115]:
```

	Country	Region	Happiness Rank	\
0	Switzerland	Western Europe	1	
1	Iceland	Western Europe	2	
2	Denmark	Western Europe	3	

3	Norway	Western Europe	4
4	Canada	North America	5
..
153	Rwanda	Sub-Saharan Africa	154
154	Benin	Sub-Saharan Africa	155
155	Syria	Middle East and Northern Africa	156
156	Burundi	Sub-Saharan Africa	157
157	Togo	Sub-Saharan Africa	158

	Happiness Score	Standard Error	Economy (GDP per Capita)	Family \
0	7.587	0.03411	1.39651	1.34951
1	7.561	0.04884	1.30232	1.40223
2	7.527	0.03328	1.32548	1.36058
3	7.522	0.03880	1.45900	1.33095
4	7.427	0.03553	1.32629	1.32261
..
153	3.465	0.03464	0.22208	0.77370
154	3.340	0.03656	0.28665	0.35386
155	3.006	0.05015	0.66320	0.47489
156	2.905	0.08658	0.01530	0.41587
157	2.839	0.06727	0.20868	0.13995

	Health (Life Expectancy)	Freedom	Trust (Government Corruption) \
0	0.94143	0.66557	0.41978
1	0.94784	0.62877	0.14145
2	0.87464	0.64938	0.48357
3	0.88521	0.66973	0.36503
4	0.90563	0.63297	0.32957
..
153	0.42864	0.59201	0.55191
154	0.31910	0.48450	0.08010
155	0.72193	0.15684	0.18906
156	0.22396	0.11850	0.10062
157	0.28443	0.36453	0.10731

	Generosity	Dystopia Residual
0	0.29678	2.51738
1	0.43630	2.70201
2	0.34139	2.49204
3	0.34699	2.46531
4	0.45811	2.45176
..
153	0.22628	0.67042
154	0.18260	1.63328
155	0.47179	0.32858
156	0.19727	1.83302
157	0.16681	1.56726

[158 rows x 12 columns]

```
[116]: data.dropna()
```

```
[116]:
```

	Country	Region	Happiness Rank	\
0	Switzerland	Western Europe	1	
1	Iceland	Western Europe	2	
2	Denmark	Western Europe	3	
3	Norway	Western Europe	4	
4	Canada	North America	5	
..	
153	Rwanda	Sub-Saharan Africa	154	
154	Benin	Sub-Saharan Africa	155	
155	Syria	Middle East and Northern Africa	156	
156	Burundi	Sub-Saharan Africa	157	
157	Togo	Sub-Saharan Africa	158	

	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	\
0	7.587	0.03411	1.39651	1.34951	
1	7.561	0.04884	1.30232	1.40223	
2	7.527	0.03328	1.32548	1.36058	
3	7.522	0.03880	1.45900	1.33095	
4	7.427	0.03553	1.32629	1.32261	
..	
153	3.465	0.03464	0.22208	0.77370	
154	3.340	0.03656	0.28665	0.35386	
155	3.006	0.05015	0.66320	0.47489	
156	2.905	0.08658	0.01530	0.41587	
157	2.839	0.06727	0.20868	0.13995	

	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	\
0	0.94143	0.66557	0.41978	
1	0.94784	0.62877	0.14145	
2	0.87464	0.64938	0.48357	
3	0.88521	0.66973	0.36503	
4	0.90563	0.63297	0.32957	
..	
153	0.42864	0.59201	0.55191	
154	0.31910	0.48450	0.08010	
155	0.72193	0.15684	0.18906	
156	0.22396	0.11850	0.10062	
157	0.28443	0.36453	0.10731	

	Generosity	Dystopia	Residual
0	0.29678	2.51738	
1	0.43630	2.70201	

```

2      0.34139      2.49204
3      0.34699      2.46531
4      0.45811      2.45176
..      ...      ...
153    0.22628      0.67042
154    0.18260      1.63328
155    0.47179      0.32858
156    0.19727      1.83302
157    0.16681      1.56726

```

[158 rows x 12 columns]

2 dataset 2

```
[117]: data2 = pd.read_csv(r"C:\Users\SHREYAS\Downloads\dataset2.csv")
data2
```

```
[117]:
```

	ID	model	engine_power	age_in_days	km	previous_owners	\
0	1.0	lounge	51.0	882.0	25000.0	1.0	
1	2.0	pop	51.0	1186.0	32500.0	1.0	
2	3.0	sport	74.0	4658.0	142228.0	1.0	
3	4.0	lounge	51.0	2739.0	160000.0	1.0	
4	5.0	pop	73.0	3074.0	106880.0	1.0	
...	
1554	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1555	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1556	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1557	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1558	NaN	NaN	NaN	NaN	NaN	NaN	NaN

	lat	lon	price	Unnamed: 9	Unnamed: 10	Unnamed: 11	\
0	44.907242	8.611559868	8900	NaN	NaN	NaN	
1	45.666359	12.24188995	8800	NaN	NaN	NaN	
2	45.503300	11.41784	4200	NaN	NaN	NaN	
3	40.633171	17.63460922	6000	NaN	NaN	NaN	
4	41.903221	12.49565029	5700	NaN	NaN	NaN	
...	
1554	NaN	averageif	44028	NaN	NaN	NaN	NaN
1555	NaN	counta	1538	NaN	NaN	NaN	NaN
1556	NaN	left	lou	NaN	NaN	NaN	NaN
1557	NaN	right	ort	NaN	NaN	NaN	NaN
1558	NaN	date	26-11-2002	NaN	NaN	NaN	NaN

	Unnamed: 12
0	NaN
1	NaN

```

2      NaN
3      NaN
4      NaN
...
1554   NaN
1555   NaN
1556   NaN
1557   NaN
1558   NaN

```

[1559 rows x 13 columns]

```
[118]: data2.head()
```

```
[118]:
```

	ID	model	engine_power	age_in_days	km	previous_owners	\
0	1.0	lounge	51.0	882.0	25000.0	1.0	
1	2.0	pop	51.0	1186.0	32500.0	1.0	
2	3.0	sport	74.0	4658.0	142228.0	1.0	
3	4.0	lounge	51.0	2739.0	160000.0	1.0	
4	5.0	pop	73.0	3074.0	106880.0	1.0	

	lat	lon	price	Unnamed: 9	Unnamed: 10	Unnamed: 11	\
0	44.907242	8.611559868	8900	NaN	NaN	NaN	
1	45.666359	12.24188995	8800	NaN	NaN	NaN	
2	45.503300	11.41784	4200	NaN	NaN	NaN	
3	40.633171	17.63460922	6000	NaN	NaN	NaN	
4	41.903221	12.49565029	5700	NaN	NaN	NaN	

```

    Unnamed: 12
0      NaN
1      NaN
2      NaN
3      NaN
4      NaN

```

```
[119]: data2.tail()
```

```
[119]:
```

	ID	model	engine_power	age_in_days	km	previous_owners	lat	\
1554	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1555	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1556	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1557	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1558	NaN	NaN	NaN	NaN	NaN	NaN	NaN	

	lon	price	Unnamed: 9	Unnamed: 10	Unnamed: 11	Unnamed: 12
1554	averageif	44028	NaN	NaN	NaN	NaN
1555	counta	1538	NaN	NaN	NaN	NaN

1556	left	lou	NaN	NaN	NaN	NaN
1557	right	ort	NaN	NaN	NaN	NaN
1558	date	26-11-2002	NaN	NaN	NaN	NaN

```
[120]: data2.describe()
```

```
[120]:
```

	ID	engine_power	age_in_days	km	previous_owners \
count	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000
mean	769.500000	51.904421	1650.980494	53396.011704	1.123537
std	444.126671	3.988023	1289.522278	40046.830723	0.416423
min	1.000000	51.000000	366.000000	1232.000000	1.000000
25%	385.250000	51.000000	670.000000	20006.250000	1.000000
50%	769.500000	51.000000	1035.000000	39031.000000	1.000000
75%	1153.750000	51.000000	2616.000000	79667.750000	1.000000
max	1538.000000	77.000000	4658.000000	235000.000000	4.000000

	lat	Unnamed: 9	Unnamed: 10
count	1538.000000	0.0	0.0
mean	43.541361	NaN	NaN
std	2.133518	NaN	NaN
min	36.855839	NaN	NaN
25%	41.802990	NaN	NaN
50%	44.394096	NaN	NaN
75%	45.467960	NaN	NaN
max	46.795612	NaN	NaN

```
[121]: data2.shape
```

```
[121]: (1559, 13)
```

```
[122]: data2.size
```

```
[122]: 20267
```

```
[123]: data2.isnull()
```

```
[123]:
```

	ID	model	engine_power	age_in_days	km	previous_owners	lat \
0	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False
...
1554	True	True	True	True	True	True	True
1555	True	True	True	True	True	True	True
1556	True	True	True	True	True	True	True
1557	True	True	True	True	True	True	True
1558	True	True	True	True	True	True	True

	lon	price	Unnamed: 9	Unnamed: 10	Unnamed: 11	Unnamed: 12
0	False	False	True	True	True	True
1	False	False	True	True	True	True
2	False	False	True	True	True	True
3	False	False	True	True	True	True
4	False	False	True	True	True	True
...
1554	False	False	True	True	True	True
1555	False	False	True	True	True	True
1556	False	False	True	True	True	True
1557	False	False	True	True	True	True
1558	False	False	True	True	True	True

[1559 rows x 13 columns]

```
[124]: fill = data2.fillna(value = 1)
fill
```

```
[124]:
```

	ID	model	engine_power	age_in_days	km	previous_owners	\
0	1.0	lounge	51.0	882.0	25000.0	1.0	
1	2.0	pop	51.0	1186.0	32500.0	1.0	
2	3.0	sport	74.0	4658.0	142228.0	1.0	
3	4.0	lounge	51.0	2739.0	160000.0	1.0	
4	5.0	pop	73.0	3074.0	106880.0	1.0	
...
1554	1.0	1	1.0	1.0	1.0	1.0	
1555	1.0	1	1.0	1.0	1.0	1.0	
1556	1.0	1	1.0	1.0	1.0	1.0	
1557	1.0	1	1.0	1.0	1.0	1.0	
1558	1.0	1	1.0	1.0	1.0	1.0	

	lat	lon	price	Unnamed: 9	Unnamed: 10	Unnamed: 11	\
0	44.907242	8.611559868	8900	1.0	1.0	1	
1	45.666359	12.24188995	8800	1.0	1.0	1	
2	45.503300	11.41784	4200	1.0	1.0	1	
3	40.633171	17.63460922	6000	1.0	1.0	1	
4	41.903221	12.49565029	5700	1.0	1.0	1	
...
1554	1.000000	averageif	44028	1.0	1.0	1	
1555	1.000000	counta	1538	1.0	1.0	1	
1556	1.000000	left	lou	1.0	1.0	1	
1557	1.000000	right	ort	1.0	1.0	1	
1558	1.000000	date	26-11-2002	1.0	1.0	1	

	Unnamed: 12
0	1

```

1          1
2          1
3          1
4          1
...
1554       1
1555       1
1556       1
1557       1
1558       1

```

[1559 rows x 13 columns]

```
[125]: drop = data2.dropna(axis=1)
drop
```

```
[125]:
           lon      price
0    8.611559868    8900
1   12.24188995    8800
2    11.41784     4200
3   17.63460922    6000
4   12.49565029    5700
...
1554  averageif    44028
1555    counta    1538
1556    left     lou
1557    right     ort
1558    date  26-11-2002

```

[1559 rows x 2 columns]