

PANDAS

Analysing GDP growth

```
In [53]: import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as sns
```

```
In [54]: stats = pd.read_excel(r'/Users/sasidharbhagavatula/Desktop/data.xlsx')
```

```
In [55]: stats
```

```
Out[55]:
```

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income
1	Afghanistan	AFG	35.253	5.9	Low income
2	Angola	AGO	45.985	19.1	Upper middle income
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income
...
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
191	South Africa	ZAF	20.850	46.5	Upper middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
194	Zimbabwe	ZWE	35.715	18.5	Low income

195 rows × 5 columns

```
In [56]: len(stats)
```

```
Out[56]: 195
```

```
In [57]: stats.columns
```

```
Out[57]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
       'IncomeGroup'],
       dtype='object')
```

```
In [58]: len(stats.columns)
```

```
Out[58]: 5
```

```
In [59]: stats.shape
```

```
Out[59]: (195, 5)
```

```
In [60]: stats.isnull()
```

```
Out[60]:
```

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
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
...
190	False	False	False	False	False
191	False	False	False	False	False
192	False	False	False	False	False
193	False	False	False	False	False
194	False	False	False	False	False

195 rows × 5 columns

```
In [61]: stats.isna()
```

Out[61]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
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
...
190	False	False	False	False	False
191	False	False	False	False	False
192	False	False	False	False	False
193	False	False	False	False	False
194	False	False	False	False	False

195 rows × 5 columns

In [62]: `stats.isnull().sum()`

Out[62]:

CountryName	0
CountryCode	0
BirthRate	0
InternetUsers	0
IncomeGroup	0
dtype:	int64

In [63]: `stats.dtypes`

Out[63]:

CountryName	object
CountryCode	object
BirthRate	float64
InternetUsers	float64
IncomeGroup	object
dtype:	object

In [64]: `stats.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 195 entries, 0 to 194
Data columns (total 5 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   CountryName      195 non-null    object  
 1   CountryCode      195 non-null    object  
 2   BirthRate        195 non-null    float64 
 3   InternetUsers   195 non-null    float64 
 4   IncomeGroup      195 non-null    object  
dtypes: float64(2), object(3)
memory usage: 7.7+ KB
```

```
In [65]: type(stats)
```

```
Out[65]: pandas.core.frame.DataFrame
```

```
In [66]: # top 5 rows
stats.head()
```

```
Out[66]:
```

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income
1	Afghanistan	AFG	35.253	5.9	Low income
2	Angola	AGO	45.985	19.1	Upper middle income
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income

```
In [67]: stats.tail()
```

```
Out[67]:
```

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
191	South Africa	ZAF	20.850	46.5	Upper middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
194	Zimbabwe	ZWE	35.715	18.5	Low income

```
In [68]: stats.tail(1)
```

```
Out[68]:    CountryName  CountryCode  BirthRate  InternetUsers  IncomeGroup
194      Zimbabwe        ZWE     35.715       18.5  Low income
```

```
In [69]: stats.columns
```

```
Out[69]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
               'IncomeGroup'],
               dtype='object')
```

```
In [70]: stats['CountryName']
```

```
Out[70]: 0                  Aruba
1                  Afghanistan
2                  Angola
3                  Albania
4      United Arab Emirates
...
190          Yemen, Rep.
191          South Africa
192          Congo, Dem. Rep.
193          Zambia
194          Zimbabwe
Name: CountryName, Length: 195, dtype: object
```

```
In [71]: stats['BirthRate']
```

```
Out[71]: 0      10.244
1      35.253
2      45.985
3      12.877
4      11.044
...
190     32.947
191     20.850
192     42.394
193     40.471
194     35.715
Name: BirthRate, Length: 195, dtype: float64
```

```
In [72]: stats[['BirthRate', 'CountryName']]
```

```
Out[72]:
```

	BirthRate	CountryName
0	10.244	Aruba
1	35.253	Afghanistan
2	45.985	Angola
3	12.877	Albania
4	11.044	United Arab Emirates
...
190	32.947	Yemen, Rep.
191	20.850	South Africa
192	42.394	Congo, Dem. Rep.
193	40.471	Zambia
194	35.715	Zimbabwe

195 rows × 2 columns

```
In [73]: stats_numeric_data= stats[['BirthRate','InternetUsers']]
```

```
In [74]: stats_numeric_data.head()
```

```
Out[74]:
```

	BirthRate	InternetUsers
0	10.244	78.9
1	35.253	5.9
2	45.985	19.1
3	12.877	57.2
4	11.044	88.0

```
In [75]: stats_catagorical_data= stats[['CountryName','CountryCode','InternetUsers']]
```

```
In [76]: stats_catagorical_data.head()
```

Out[76]:

	CountryName	CountryCode	InternetUsers
0	Aruba	ABW	78.9
1	Afghanistan	AFG	5.9
2	Angola	AGO	19.1
3	Albania	ALB	57.2
4	United Arab Emirates	ARE	88.0

In [77]:

```
#total variables created
print(stats.shape)
print(stats_numeric_data.shape)
print(stats_catagorical_data.shape)
```

```
(195, 5)
(195, 2)
(195, 3)
```

slicing in pandas

In [78]:

```
stats[:]
```

Out[78]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income
1	Afghanistan	AFG	35.253	5.9	Low income
2	Angola	AGO	45.985	19.1	Upper middle income
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income
...
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
191	South Africa	ZAF	20.850	46.5	Upper middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
194	Zimbabwe	ZWE	35.715	18.5	Low income

195 rows × 5 columns

In [79]: #slicing

In [80]: stats[:6]

Out[80]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income
1	Afghanistan	AFG	35.253	5.9	Low income
2	Angola	AGO	45.985	19.1	Upper middle income
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income
5	Argentina	ARG	17.716	59.9	High income

In [81]: stats[3:]

Out [81]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income
5	Argentina	ARG	17.716	59.9	High income
6	Armenia	ARM	13.308	41.9	Lower middle income
7	Antigua and Barbuda	ATG	16.447	63.4	High income
...
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
191	South Africa	ZAF	20.850	46.5	Upper middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
194	Zimbabwe	ZWE	35.715	18.5	Low income

192 rows × 5 columns

In [82]: `stats[3:10]`

Out [82]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
3	Albania	ALB	12.877	57.2000	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0000	High income
5	Argentina	ARG	17.716	59.9000	High income
6	Armenia	ARM	13.308	41.9000	Lower middle income
7	Antigua and Barbuda	ATG	16.447	63.4000	High income
8	Australia	AUS	13.200	83.0000	High income
9	Austria	AUT	9.400	80.6188	High income

In [83]: `stats[3:55:5]`

Out [83]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
3	Albania	ALB	12.877	57.2000	Upper middle income
8	Australia	AUS	13.200	83.0000	High income
13	Benin	BEN	36.440	4.9000	Low income
18	Bahamas, The	BHS	15.339	72.0000	High income
23	Bolivia	BOL	24.236	36.9400	Lower middle income
28	Botswana	BWA	25.267	15.0000	Upper middle income
33	China	CHN	12.100	45.8000	Upper middle income
38	Comoros	COM	34.326	6.5000	Low income
43	Cyprus	CYP	11.436	65.4548	High income
48	Dominican Republic	DOM	21.198	45.9000	Upper middle income
53	Spain	ESP	9.100	71.6350	High income

In [84]: `stats[::-1]`

Out [84]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
194	Zimbabwe	ZWE	35.715	18.5	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
191	South Africa	ZAF	20.850	46.5	Upper middle income
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
...
4	United Arab Emirates	ARE	11.044	88.0	High income
3	Albania	ALB	12.877	57.2	Upper middle income
2	Angola	AGO	45.985	19.1	Upper middle income
1	Afghanistan	AFG	35.253	5.9	Low income
0	Aruba	ABW	10.244	78.9	High income

195 rows × 5 columns

In [85]: `stats[-1::]`

Out [85]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
194	Zimbabwe	ZWE	35.715	18.5	Low income

In [86]: `stats[0:200]`

Out [86]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income
1	Afghanistan	AFG	35.253	5.9	Low income
2	Angola	AGO	45.985	19.1	Upper middle income
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income
...
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
191	South Africa	ZAF	20.850	46.5	Upper middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
194	Zimbabwe	ZWE	35.715	18.5	Low income

195 rows × 5 columns

In [87]: stats

Out[87]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income
1	Afghanistan	AFG	35.253	5.9	Low income
2	Angola	AGO	45.985	19.1	Upper middle income
3	Albania	ALB	12.877	57.2	Upper middle income
4	United Arab Emirates	ARE	11.044	88.0	High income
...
190	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
191	South Africa	ZAF	20.850	46.5	Upper middle income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income
194	Zimbabwe	ZWE	35.715	18.5	Low income

195 rows × 5 columns

In []:

In []:

In []:

In []: