

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

```
In [2]: pd.__version__
```

```
Out[2]: '2.2.3'
```

```
In [3]: stats = pd.read_excel(r'C:\Users\ankus\Desktop\Notes\Country GDP Analysis- proje
```

```
In [4]: stats
```

	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 [5]: len(stats)
```

```
Out[5]: 195
```

```
In [6]: stats.columns
```

```
Out[6]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
               'IncomeGroup'],
               dtype='object')
```

```
In [7]: len(stats.columns)
```

```
Out[7]: 5
```

```
In [8]: stats.shape
```

Out[8]: (195, 5)

In [9]: `stats.isnull()`

	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 [10]: `stats.isna() # both isnull() and isna() gives same output`

	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 [11]: `stats.isnull().sum()`

```
Out[11]: CountryName      0  
CountryCode       0  
BirthRate         0  
InternetUsers    0  
IncomeGroup       0  
dtype: int64
```

```
In [12]: stats.dtypes
```

```
Out[12]: CountryName      object  
CountryCode       object  
BirthRate         float64  
InternetUsers    float64  
IncomeGroup       object  
dtype: object
```

```
In [13]: 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 [14]: type(stats)
```

```
Out[14]: pandas.core.frame.DataFrame
```

```
In [15]: stats.head()           # return top 5 rows by default
```

```
Out[15]:
```

	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 [16]: stats.tail()           #return bottom 5 rows by default
```

Out[16]:

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

In [17]: `stats.head(2)` #top 2 rows

Out[17]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
<b>0</b>	Aruba	ABW	10.244	78.9	High income
<b>1</b>	Afghanistan	AFG	35.253	5.9	Low income

In [18]: `stats.tail(2)` #bottom 2 rows

Out[18]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
<b>193</b>	Zambia	ZMB	40.471	15.4	Lower middle income
<b>194</b>	Zimbabwe	ZWE	35.715	18.5	Low income

In [19]: `stats.columns`

Out[19]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers', 'IncomeGroup'], dtype='object')

In [20]: `stats['CountryName']` # only country name

Out[20]:

```

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 [21]: `stats['BirthRate']`

```
Out[21]: 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 [22]: stats[['BirthRate', 'InternetUsers']]
```

```
Out[22]:   BirthRate  InternetUsers
```

	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
...	...	...
190	32.947	20.0
191	20.850	46.5
192	42.394	2.2
193	40.471	15.4
194	35.715	18.5

195 rows × 2 columns

## #Split the dataset numerical to categorical data

```
In [23]: stats.columns
```

```
Out[23]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
       'IncomeGroup'],
       dtype='object')
```

```
In [24]: stats_numeric_data = stats[['BirthRate', 'InternetUsers']]
stats_numeric_data
```

Out[24]:

	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
...	...	...
190	32.947	20.0
191	20.850	46.5
192	42.394	2.2
193	40.471	15.4
194	35.715	18.5

195 rows × 2 columns

In [25]:

`stats_numeric_data.head()`

Out[25]:

	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 [26]:

`stats_categorical_data = stats[['CountryName', 'CountryCode', 'IncomeGroup']]  
stats_categorical_data`

Out[26]:

	CountryName	CountryCode	IncomeGroup
0	Aruba	ABW	High income
1	Afghanistan	AFG	Low income
2	Angola	AGO	Upper middle income
3	Albania	ALB	Upper middle income
4	United Arab Emirates	ARE	High income
...	...	...	...
190	Yemen, Rep.	YEM	Lower middle income
191	South Africa	ZAF	Upper middle income
192	Congo, Dem. Rep.	COD	Low income
193	Zambia	ZMB	Lower middle income
194	Zimbabwe	ZWE	Low income

195 rows × 3 columns

In [27]:

`stats_categorical_data.head()`

Out[27]:

	CountryName	CountryCode	IncomeGroup
0	Aruba	ABW	High income
1	Afghanistan	AFG	Low income
2	Angola	AGO	Upper middle income
3	Albania	ALB	Upper middle income
4	United Arab Emirates	ARE	High income

In [28]:

`print(stats.shape)  
print(stats_numeric_data.shape)  
print(stats_categorical_data.shape)`(195, 5)  
(195, 2)  
(195, 3)

## Slicing in pandas

In [29]:

`stats[:]`

Out[29]:

	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 [30]: stats[:6]

Out[30]:

	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 [31]: stats[:3]

Out[31]:

	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

In [32]: stats[3:]

Out[32]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
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 [33]: stats[3:10]

Out[33]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
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 [34]: stats[3:50:5]

Out[34]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
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

In [35]: `stats[::-1]`

Out[35]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
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 [36]: `stats[::-2]`

Out[36]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
<b>194</b>	Zimbabwe	ZWE	35.715	18.5	Low income
<b>192</b>	Congo, Dem. Rep.	COD	42.394	2.2	Low income
<b>190</b>	Yemen, Rep.	YEM	32.947	20.0	Lower middle income
<b>188</b>	West Bank and Gaza	PSE	30.394	46.6	Lower middle income
<b>186</b>	Vietnam	VNM	15.537	43.9	Lower middle income
...	...	...	...	...	...
<b>8</b>	Australia	AUS	13.200	83.0	High income
<b>6</b>	Armenia	ARM	13.308	41.9	Lower middle income
<b>4</b>	United Arab Emirates	ARE	11.044	88.0	High income
<b>2</b>	Angola	AGO	45.985	19.1	Upper middle income
<b>0</b>	Aruba	ABW	10.244	78.9	High income

98 rows × 5 columns

In [37]:

stats[50:100]

Out[37]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
<b>50</b>	Ecuador	ECU	21.070	40.353684	Upper middle income
<b>51</b>	Egypt, Arab Rep.	EGY	28.032	29.400000	Lower middle income
<b>52</b>	Eritrea	ERI	34.800	0.900000	Low income
<b>53</b>	Spain	ESP	9.100	71.635000	High income
<b>54</b>	Estonia	EST	10.300	79.400000	High income
<b>55</b>	Ethiopia	ETH	32.925	1.900000	Low income
<b>56</b>	Finland	FIN	10.700	91.514400	High income
<b>57</b>	Fiji	FJI	20.463	37.100000	Upper middle income
<b>58</b>	France	FRA	12.300	81.919800	High income
<b>59</b>	Micronesia, Fed. Sts.	FSM	23.511	27.800000	Lower middle income
<b>60</b>	Gabon	GAB	30.555	9.200000	Upper middle income
<b>61</b>	United Kingdom	GBR	12.200	89.844100	High income
<b>62</b>	Georgia	GEO	13.332	43.300000	Lower middle income
<b>63</b>	Ghana	GHA	33.131	12.300000	Lower middle income
<b>64</b>	Guinea	GIN	37.337	1.600000	Low income
<b>65</b>	Gambia, The	GMB	42.525	14.000000	Low income
<b>66</b>	Guinea-Bissau	GNB	37.503	3.100000	Low income
<b>67</b>	Equatorial Guinea	GNQ	35.362	16.400000	High income
<b>68</b>	Greece	GRC	8.500	59.866300	High income
<b>69</b>	Grenada	GRD	19.334	35.000000	Upper middle income
<b>70</b>	Greenland	GRL	14.500	65.800000	High income
<b>71</b>	Guatemala	GTM	27.465	19.700000	Lower middle income
<b>72</b>	Guam	GUM	17.389	65.400000	High income
<b>73</b>	Guyana	GUY	18.885	35.000000	Lower middle income
<b>74</b>	Hong Kong SAR, China	HKG	7.900	74.200000	High income
<b>75</b>	Honduras	HND	21.593	17.800000	Lower middle income

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
76	Croatia	HRV	9.400	66.747600	High income
77	Haiti	HTI	25.345	10.600000	Low income
78	Hungary	HUN	9.200	72.643900	High income
79	Indonesia	IDN	20.297	14.940000	Lower middle income
80	India	IND	20.291	15.100000	Lower middle income
81	Ireland	IRL	15.000	78.247700	High income
82	Iran, Islamic Rep.	IRN	17.900	29.950000	Upper middle income
83	Iraq	IRQ	31.093	9.200000	Upper middle income
84	Iceland	ISL	13.400	96.546800	High income
85	Israel	ISR	21.300	70.800000	High income
86	Italy	ITA	8.500	58.459300	High income
87	Jamaica	JAM	13.540	37.100000	Upper middle income
88	Jordan	JOR	27.046	41.000000	Upper middle income
89	Japan	JPN	8.200	89.710000	High income
90	Kazakhstan	KAZ	22.730	54.000000	Upper middle income
91	Kenya	KEN	35.194	39.000000	Lower middle income
92	Kyrgyz Republic	KGZ	27.200	23.000000	Lower middle income
93	Cambodia	KHM	24.462	6.800000	Low income
94	Kiribati	KIR	29.044	11.500000	Lower middle income
95	Korea, Rep.	KOR	8.600	84.770000	High income
96	Kuwait	KWT	20.575	75.460000	High income
97	Lao PDR	LAO	27.051	12.500000	Lower middle income
98	Lebanon	LBN	13.426	70.500000	Upper middle income
99	Liberia	LBR	35.521	3.200000	Low income

In [38]: `stats[0:200]`

Out[38]:

	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 [39]: `stats.columns`Out[39]: `Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers', 'IncomeGroup'], dtype='object')`In [40]: `stats.describe() # descriptive statistic only numerical data information`

Out[40]:

	BirthRate	InternetUsers
<b>count</b>	195.000000	195.000000
<b>mean</b>	21.469928	42.076471
<b>std</b>	10.605467	29.030788
<b>min</b>	7.900000	0.900000
<b>25%</b>	12.120500	14.520000
<b>50%</b>	19.680000	41.000000
<b>75%</b>	29.759500	66.225000
<b>max</b>	49.661000	96.546800

In [41]: `stats.describe().transpose()`

		count	mean	std	min	25%	50%	75%	max
	<b>BirthRate</b>	195.0	21.469928	10.605467	7.9	12.1205	19.68	29.7595	49.6610
	<b>InternetUsers</b>	195.0	42.076471	29.030788	0.9	14.5200	41.00	66.2250	96.5468

In [42]: `stats_numeric_data.describe().transpose()`

	count	mean	std	min	25%	50%	75%	max	
	<b>BirthRate</b>	195.0	21.469928	10.605467	7.9	12.1205	19.68	29.7595	49.6610
	<b>InternetUsers</b>	195.0	42.076471	29.030788	0.9	14.5200	41.00	66.2250	96.5468

In [43]: `stats_categorical_data.describe()`

	CountryName	CountryCode	IncomeGroup
count	195	195	195
unique	195	195	4
top	Aruba	ABW	High income
freq	1	1	67

In [44]: `stats.head(2)`

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
<b>0</b>	Aruba	ABW	10.244	78.9	High income
<b>1</b>	Afghanistan	AFG	35.253	5.9	Low income

In [45]: `stats['myCalc'] = stats.BirthRate * stats.InternetUsers`

In [46]: `stats['myCalc']`

```
Out[46]: 0    808.2516
        1    207.9927
        2    878.3135
        3    736.5644
        4    971.8720
        ...
       190   658.9400
      191   969.5250
      192   93.2668
      193   623.2534
      194   660.7275
Name: myCalc, Length: 195, dtype: float64
```

In [47]: `stats.head(2)`

Out[47]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup	myCalc
0	Aruba	ABW	10.244	78.9	High income	808.2516
1	Afghanistan	AFG	35.253	5.9	Low income	207.9927

In [48]: `stats.columns`

Out[48]: `Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers', 'IncomeGroup', 'myCalc'], dtype='object')`

In [49]: `len(stats.columns)`

Out[49]: 6

In [50]: `stats = stats.drop('myCalc', axis=1)`  
`stats`

Out[50]:

	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 [51]: `stats['InternetUsers'] < 2`

```
Out[51]: 0    False
         1    False
         2    False
         3    False
         4    False
        ...
       190   False
       191   False
       192   False
       193   False
       194   False
Name: InternetUsers, Length: 195, dtype: bool
```

```
In [52]: stats[stats['InternetUsers'] < 2]
```

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
11	Burundi	BDI	44.151	1.3	Low income
52	Eritrea	ERI	34.800	0.9	Low income
55	Ethiopia	ETH	32.925	1.9	Low income
64	Guinea	GIN	37.337	1.6	Low income
117	Myanmar	MMR	18.119	1.6	Lower middle income
127	Niger	NER	49.661	1.7	Low income
154	Sierra Leone	SLE	36.729	1.7	Low income
156	Somalia	SOM	43.891	1.5	Low income
172	Timor-Leste	TLS	35.755	1.1	Lower middle income

```
In [53]: len(stats[stats['InternetUsers'] < 2])
```

```
Out[53]: 9
```

```
In [54]: stats[stats['BirthRate'] > 40]
```

Out[54]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
2	Angola	AGO	45.985	19.1	Upper middle income
11	Burundi	BDI	44.151	1.3	Low income
14	Burkina Faso	BFA	40.551	9.1	Low income
65	Gambia, The	GMB	42.525	14.0	Low income
115	Mali	MLI	44.138	3.5	Low income
127	Niger	NER	49.661	1.7	Low income
128	Nigeria	NGA	40.045	38.0	Lower middle income
156	Somalia	SOM	43.891	1.5	Low income
167	Chad	TCD	45.745	2.3	Low income
178	Uganda	UGA	43.474	16.2	Low income
192	Congo, Dem. Rep.	COD	42.394	2.2	Low income
193	Zambia	ZMB	40.471	15.4	Lower middle income

In [55]: `len(stats[stats['BirthRate'] > 40])`

Out[55]: 12

In [56]: `stats[ (stats.BirthRate > 40) & (stats.InternetUsers < 2) ]`

Out[56]:

	<b>CountryName</b>	<b>CountryCode</b>	<b>BirthRate</b>	<b>InternetUsers</b>	<b>IncomeGroup</b>
11	Burundi	BDI	44.151	1.3	Low income
127	Niger	NER	49.661	1.7	Low income
156	Somalia	SOM	43.891	1.5	Low income

In [60]: `stats[stats.IncomeGroup == 'High income']`

Out[60]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.90	High income
4	United Arab Emirates	ARE	11.044	88.00	High income
5	Argentina	ARG	17.716	59.90	High income
7	Antigua and Barbuda	ATG	16.447	63.40	High income
8	Australia	AUS	13.200	83.00	High income
...	...	...	...	...	...
174	Trinidad and Tobago	TTO	14.590	63.80	High income
180	Uruguay	URY	14.374	57.69	High income
181	United States	USA	12.500	84.20	High income
184	Venezuela, RB	VEN	19.842	54.90	High income
185	Virgin Islands (U.S.)	VIR	10.700	45.30	High income

67 rows × 5 columns

In [61]: `stats[stats.IncomeGroup == 'Low income']`

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
1	Afghanistan	AFG	35.253	5.90	Low income
11	Burundi	BDI	44.151	1.30	Low income
13	Benin	BEN	36.440	4.90	Low income
14	Burkina Faso	BFA	40.551	9.10	Low income
29	Central African Republic	CAF	34.076	3.50	Low income
38	Comoros	COM	34.326	6.50	Low income
52	Eritrea	ERI	34.800	0.90	Low income
55	Ethiopia	ETH	32.925	1.90	Low income
64	Guinea	GIN	37.337	1.60	Low income
65	Gambia, The	GMB	42.525	14.00	Low income
66	Guinea-Bissau	GNB	37.503	3.10	Low income
77	Haiti	HTI	25.345	10.60	Low income
93	Cambodia	KHM	24.462	6.80	Low income
99	Liberia	LBR	35.521	3.20	Low income
111	Madagascar	MDG	34.686	3.00	Low income
115	Mali	MLI	44.138	3.50	Low income
120	Mozambique	MOZ	39.705	5.40	Low income
123	Malawi	MWI	39.459	5.05	Low income
127	Niger	NER	49.661	1.70	Low income
132	Nepal	NPL	20.923	13.30	Low income
148	Rwanda	RWA	32.689	9.00	Low income
154	Sierra Leone	SLE	36.729	1.70	Low income
156	Somalia	SOM	43.891	1.50	Low income
158	South Sudan	SSD	37.126	14.10	Low income
167	Chad	TCD	45.745	2.30	Low income
168	Togo	TGO	36.080	4.50	Low income
177	Tanzania	TZA	39.518	4.40	Low income
178	Uganda	UGA	43.474	16.20	Low income
192	Congo, Dem. Rep.	COD	42.394	2.20	Low income
194	Zimbabwe	ZWE	35.715	18.50	Low income

In [62]: `len(stats[stats.IncomeGroup == 'Low income'])`

Out[62]: 30

```
In [63]: stats.IncomeGroup.unique()
```

```
Out[63]: array(['High income', 'Low income', 'Upper middle income',
       'Lower middle income'], dtype=object)
```

```
In [64]: stats.IncomeGroup.nunique()      #no. of unique
```

```
Out[64]: 4
```

```
In [66]: import matplotlib.pyplot as plt
import seaborn as sns

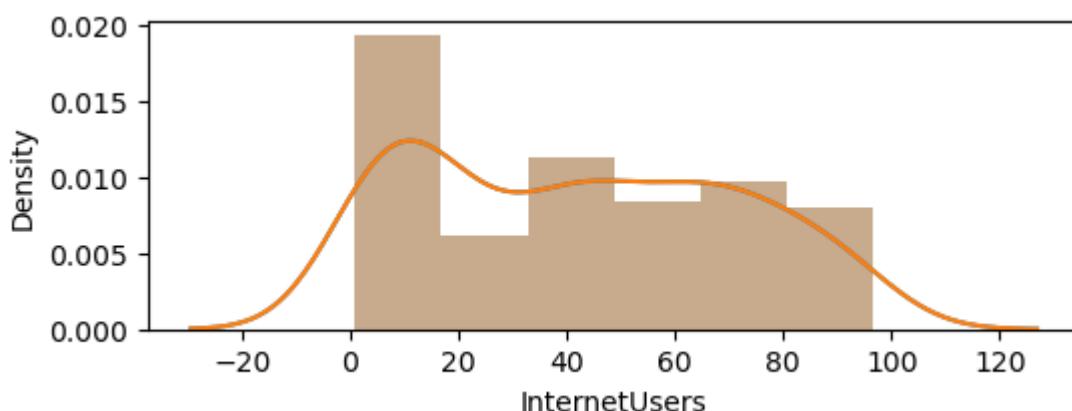
%matplotlib inline
plt.rcParams['figure.figsize'] = 6,2

import warnings
warnings.filterwarnings('ignore')      # ignore as warnings
```

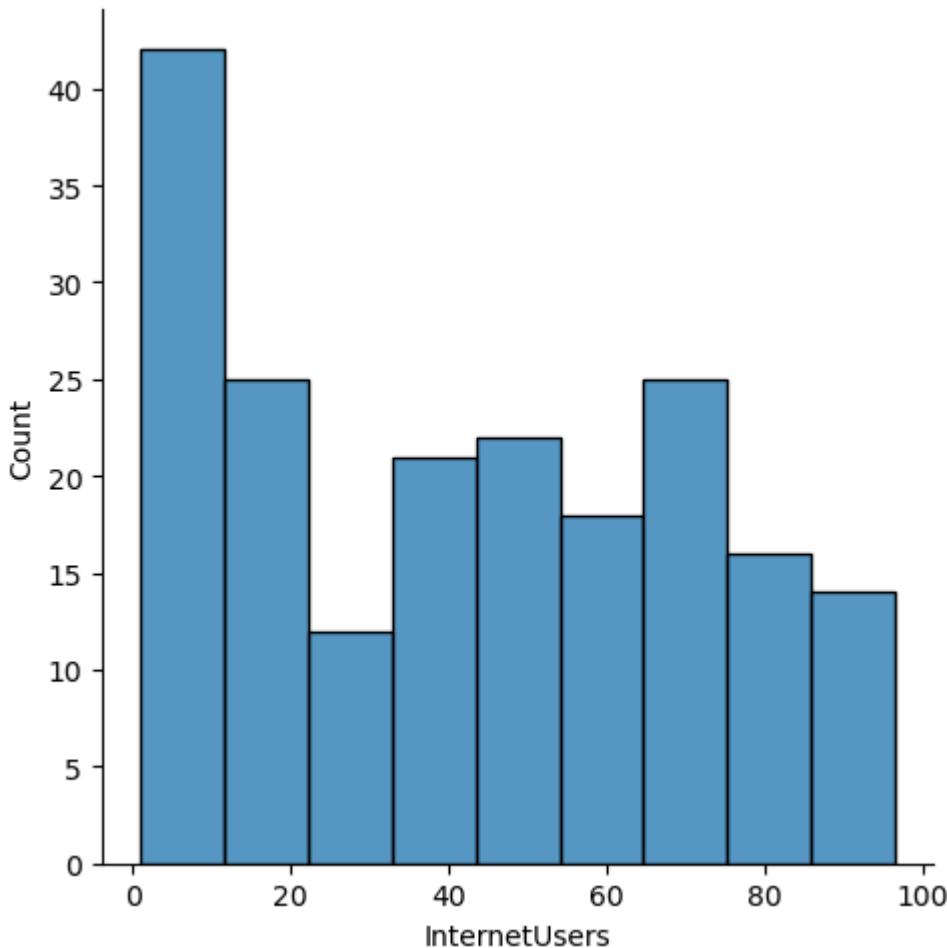
```
In [67]: stats["InternetUsers"]
```

```
Out[67]: 0      78.9
1      5.9
2      19.1
3      57.2
4      88.0
...
190    20.0
191    46.5
192    2.2
193    15.4
194    18.5
Name: InternetUsers, Length: 195, dtype: float64
```

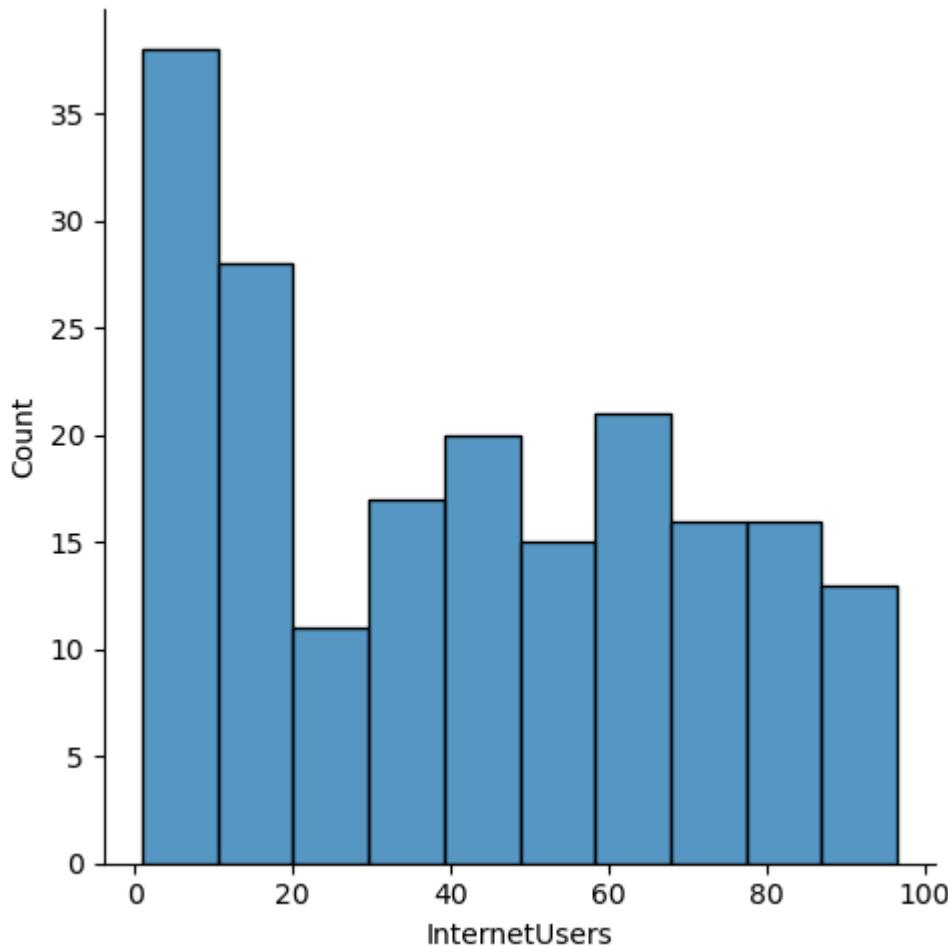
```
In [69]: visl =sns.distplot(stats["InternetUsers"])
visl
plt.show()
```



```
In [70]: visl =sns.distplot(stats["InternetUsers"])
visl
plt.show()
```



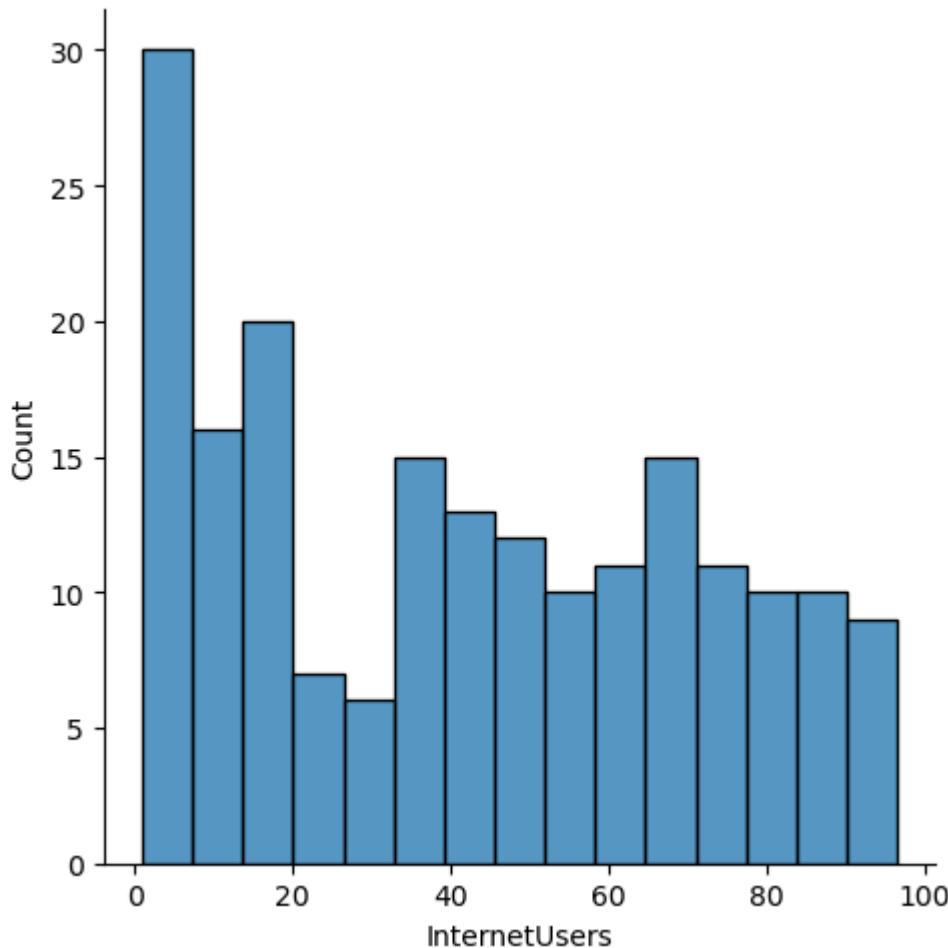
```
In [73]: vis2=sns.displot(stats["InternetUsers"] , bins =10)  
plt.show()
```



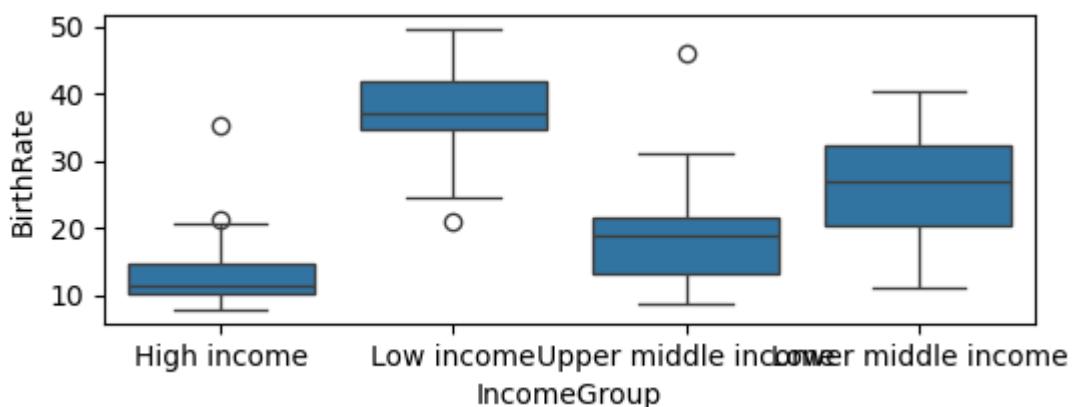
```
In [74]: stats.columns
```

```
Out[74]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
       'IncomeGroup'],
      dtype='object')
```

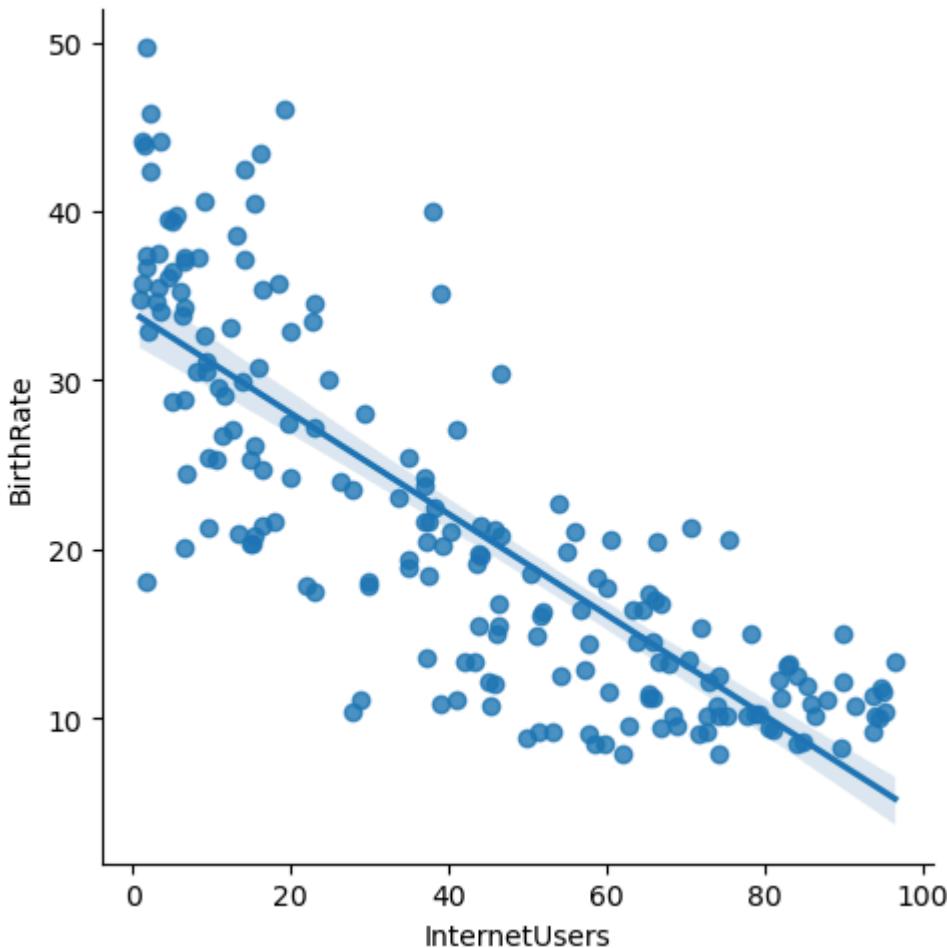
```
In [78]: visl3 =sns.displot(stats["InternetUsers"] , bins =15)
plt.show()
```



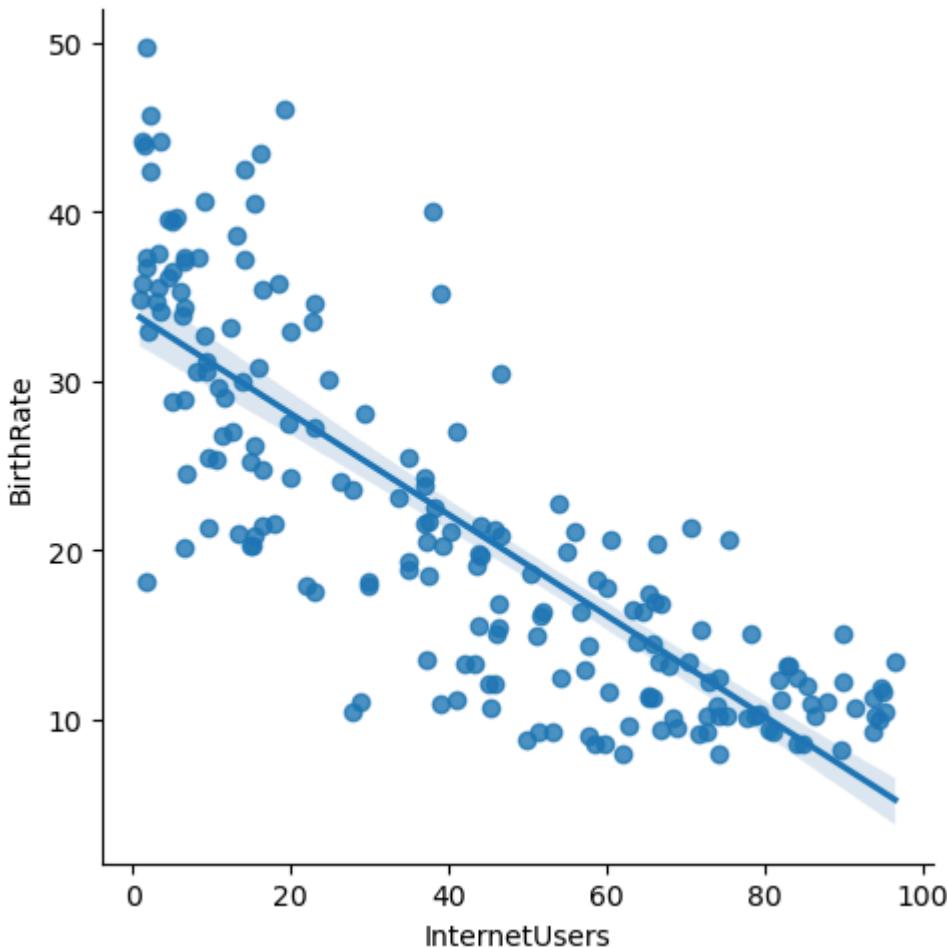
```
In [82]: # BOX PLOTS
vis4 = sns.boxplot(data = stats ,x = "IncomeGroup" ,y = 'BirthRate' ,gap =0)
plt.show()                                     # BI -variate analysis
# o denotes outlier which is not join to any one
```



```
In [88]: vis5 = sns.lmplot(data = stats ,x ="InternetUsers" ,y = 'BirthRate')
plt.show()          # all 195 o shows
```

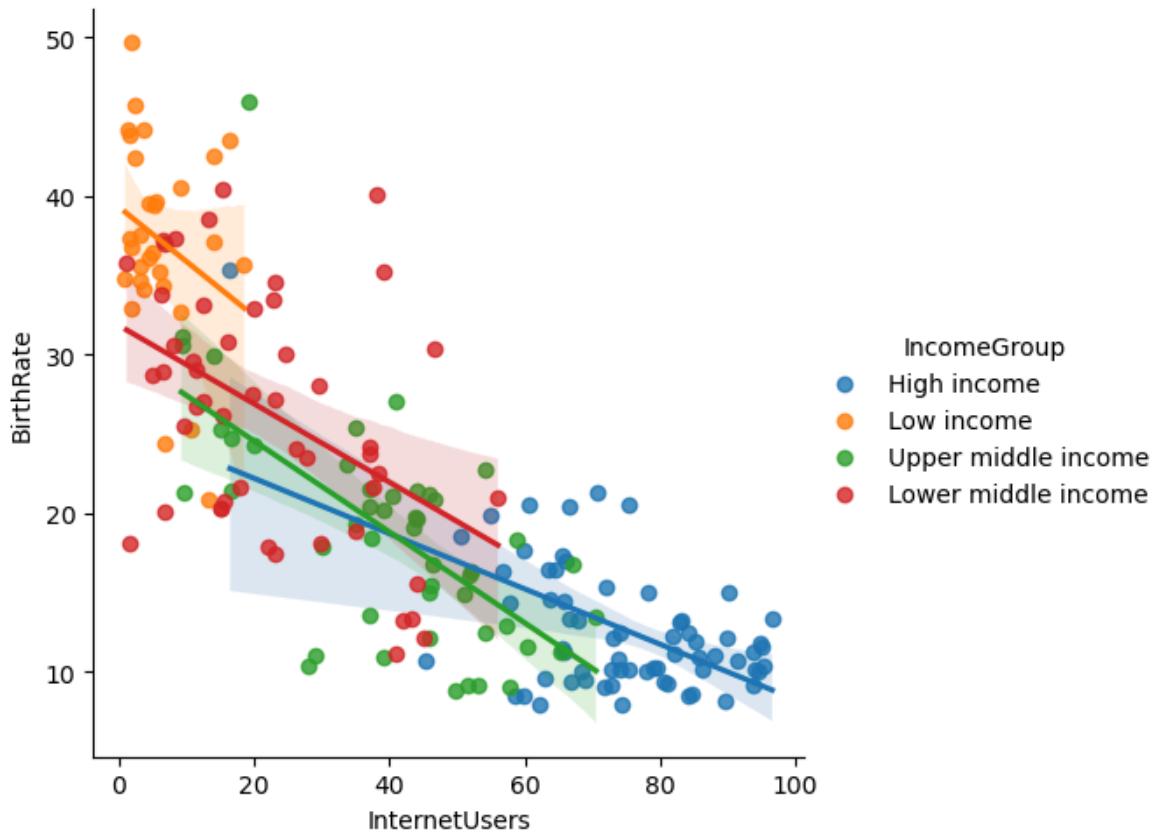


```
In [92]: vis5 = sns.lmplot(data = stats ,x ="InternetUsers" ,y = 'BirthRate' ,fit_reg = True  
plt.show()
```



```
In [ ]: vis5 = sns.lmplot(data = stats ,x ="InternetUsers" ,y = 'BirthRate',fit_reg ='Fa  
plt.show()
```

```
In [95]: vis5 = sns.lmplot(data = stats ,x ="InternetUsers" ,y = 'BirthRate',fit_reg ='Tr  
plt.show()
```



In [ ]: