

In [30]: `import pandas as pd`

In [31]: `pd.__version__`

Out[31]: '2.3.0'

In [32]: `df=pd.read_csv(r"D:\Data Science with AI\Data Science With AI\9th,10th - july-Pa`

In [33]: `df`

Out[33]:

	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 [34]: `id(df)`

Out[34]: 2031496396384

In [35]: `len(df)`

Out[35]: 195

In [36]: `df.columns`

Out[36]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
'IncomeGroup'],
dtype='object')

In [37]: `len(df.columns)`

Out[37]: 5

In [38]: `df.isnull()`

Out[38]:

	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 [39]: `df.isna()`

Out[39]:

	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 [40]: `df.isnull()`

Out[40]:

	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 [41]: `df.isnull()`

Out[41]:

	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 [42]: `df.isnull().sum()`

Out[42]:

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

In [43]: `df.isna().sum()`

```
Out[43]: CountryName      0
CountryCode      0
BirthRate        0
InternetUsers     0
IncomeGroup      0
dtype: int64
```

In [44]: `df.head()`

```
Out[44]:
```

	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 [45]: `df.tail()`

```
Out[45]:
```

	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 [46]: `df.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 [47]: `df[:]`

Out[47]:

	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 [48]: df[1:11]

Out[48]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
1	Afghanistan	AFG	35.253	5.9000	Low income
2	Angola	AGO	45.985	19.1000	Upper middle income
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
10	Azerbaijan	AZE	18.300	58.7000	Upper middle income

In [49]: `df[:: -1]`

Out[49]:

	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 [50]: `df[1:100:10]`

Out[50]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
1	Afghanistan	AFG	35.253	5.9000	Low income
11	Burundi	BDI	44.151	1.3000	Low income
21	Belize	BLZ	23.092	33.6000	Upper middle income
31	Switzerland	CHE	10.200	86.3400	High income
41	Cuba	CUB	10.400	27.9300	Upper middle income
51	Egypt, Arab Rep.	EGY	28.032	29.4000	Lower middle income
61	United Kingdom	GBR	12.200	89.8441	High income
71	Guatemala	GTM	27.465	19.7000	Lower middle income
81	Ireland	IRL	15.000	78.2477	High income
91	Kenya	KEN	35.194	39.0000	Lower middle income

In [51]: `df[10:21]`

Out[51]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
10	Azerbaijan	AZE	18.300	58.70000	Upper middle income
11	Burundi	BDI	44.151	1.30000	Low income
12	Belgium	BEL	11.200	82.17020	High income
13	Benin	BEN	36.440	4.90000	Low income
14	Burkina Faso	BFA	40.551	9.10000	Low income
15	Bangladesh	BGD	20.142	6.63000	Lower middle income
16	Bulgaria	BGR	9.200	53.06150	Upper middle income
17	Bahrain	BHR	15.040	90.00004	High income
18	Bahamas, The	BHS	15.339	72.00000	High income
19	Bosnia and Herzegovina	BIH	9.062	57.79000	Upper middle income
20	Belarus	BLR	12.500	54.17000	Upper middle income

In [52]: df.head(2)

Out[52]:

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

In [53]: df.describe() # by default describe function can write only numerical records

Out[53]:

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

In [54]: df.head(1)

Out[54]:

	CountryName	CountryCode	BirthRate	InternetUsers	IncomeGroup
0	Aruba	ABW	10.244	78.9	High income

In [55]: `df['CountryName']`

Out[55]:

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 [56]: `df['CountryCode']`

Out[56]:

0	ABW
1	AFG
2	AGO
3	ALB
4	ARE
...	...
190	YEM
191	ZAF
192	COD
193	ZMB
194	ZWE

Name: CountryCode, Length: 195, dtype: object

In [57]: `df['CountryName', 'CountryCode']`


```

-----
KeyError                                Traceback (most recent call last)
File ~\AppData\Roaming\Python\Python312\site-packages\pandas\core\indexes\base.p
y:3812, in Index.get_loc(self, key)
    3811 try:
-> 3812     return self._engine.get_loc(casted_key)
    3813 except KeyError as err:

File pandas\_libs\index.pyx:167, in pandas._libs.index.IndexEngine.get_loc()

File pandas\_libs\index.pyx:196, in pandas._libs.index.IndexEngine.get_loc()

File pandas\_libs\hashtable_class_helper.pxi:7088, in pandas._libs.hashtable.PyOb
jectHashTable.get_item()

File pandas\_libs\hashtable_class_helper.pxi:7096, in pandas._libs.hashtable.PyOb
jectHashTable.get_item()

KeyError: ('CountryName', 'CountryCode')

The above exception was the direct cause of the following exception:

KeyError                                Traceback (most recent call last)
Cell In[57], line 1
----> 1 df['CountryName', 'CountryCode']

File ~\AppData\Roaming\Python\Python312\site-packages\pandas\core\frame.py:4107,
in DataFrame.__getitem__(self, key)
    4105 if self.columns.nlevels > 1:
    4106     return self._getitem_multilevel(key)
-> 4107 indexer = self.columns.get_loc(key)
    4108 if is_integer(indexer):
    4109     indexer = [indexer]

File ~\AppData\Roaming\Python\Python312\site-packages\pandas\core\indexes\base.p
y:3819, in Index.get_loc(self, key)
    3814 if isinstance(casted_key, slice) or (
    3815     isinstance(casted_key, abc.Iterable)
    3816     and any(isinstance(x, slice) for x in casted_key)
    3817 ):
    3818     raise InvalidIndexError(key)
-> 3819     raise KeyError(key) from err
    3820 except TypeError:
    3821     # If we have a listlike key, _check_indexing_error will raise
    3822     # InvalidIndexError. Otherwise we fall through and re-raise
    3823     # the TypeError.
    3824     self._check_indexing_error(key)

KeyError: ('CountryName', 'CountryCode')

```

```
In [ ]: df[['CountryName', 'CountryCode']]
```

```
In [ ]: df[['CountryName', 'CountryCode', 'IncomeGroup']]
```

```
In [ ]: df_cat=df[['CountryName', 'CountryCode', 'IncomeGroup']]
df_cat
```

```
In [ ]: print(len(df))
        print(len(df_cat))

In [ ]: print(len(df.columns))

In [ ]: print(len(df_cat.columns))

In [ ]: print((df.columns))

In [ ]: print(df_cat.columns)

In [ ]: df_cat.describe()

In [ ]: df_num=df[['BirthRate','InternetUsers']]

In [ ]: df_num

In [ ]: df.info()

In [ ]: df_cat.info()

In [ ]: df_num.info()

In [ ]: df.describe()

In [ ]: df.describe().transpose() # rows are converted to columns and columns are conver

In [ ]: df.describe().T

In [ ]: df.columns

In [ ]: df.columns=['a','b','c','d','e'] # rename the columns

In [ ]: df.columns

In [ ]: df.head(1)

In [ ]: df.columns=['CountryName','CountryCode','BirthRate','InternetUsers','IncomeGroup

In [ ]: df.head(1)

In [ ]: df[['CountryName','CountryCode','BirthRate','InternetUsers']][4:8] # subset

In [ ]: df[4:8][['CountryName','CountryCode','BirthRate','InternetUsers']]

In [ ]: df.columns

In [ ]: df.BirthRate*df.InternetUsers # multiply two columns

In [ ]: df.head(2)
```

```
In [ ]: df['newcolumn']=df. BirthRate*df.InternetUsers

In [ ]: df.head(5)

In [ ]: len(df.columns)

In [ ]: df=df.drop('newcolumn',axis=1) # axis =1 means it deletes the column and delete

In [ ]: df.head(1)

In [ ]: df=df.drop('newcolumn',axis=1) # column already deleted

In [ ]: df

In [ ]: df.InternetUsers<2 #

In [ ]: df[df.InternetUsers<2]

In [ ]: len(df[df.InternetUsers<2])

In [ ]: df.BirthRate>40

In [ ]: df[df.BirthRate>40]

In [ ]: len(df[df.BirthRate>40])

In [ ]: low_InternetUsers_Country=df.InternetUsers<2
low_InternetUsers_Country

In [ ]: high_birth_rate=df.BirthRate>40
high_birth_rate

In [ ]: low_InternetUsers_Country & high_birth_rate

In [ ]: df[df.InternetUsers<2] & df[df.BirthRate>40]

In [ ]: df.InternetUsers<2 & df.BirthRate>40

In [ ]: Filter=df.InternetUsers < 2
Filter2=df.BirthRate>40

In [ ]: df[Filter & Filter2] # internetusers<2 and birthrate>40

In [ ]: df[low_InternetUsers_Country & high_birth_rate]

In [ ]: df_num

In [ ]: df_cat

In [ ]: df[df.IncomeGroup=='High income']

In [ ]: df.IncomeGroup=='High income'
```

```
In [ ]: df[df.IncomeGroup=='High income']
```

```
In [ ]: df[df.IncomeGroup=='Low income']
```

```
In [ ]: df.IncomeGroup.unique()
```

```
In [59]: df.IncomeGroup.nunique()
```

```
Out[59]: 4
```

we analysis python dataframe or dataset

```
In [61]: import matplotlib.pyplot as plt # visualization
import seaborn as sns # stats visualization ,advanced visualization

%%matplotlib inline #plot the graph in the line
plt.rcParams['figure.figsize'] = (6,2) #width=6,height=2

import warnings
warnings.filterwarnings('ignore') # whenever os will update.ignore the os error
```

```
In [62]: df.head()
```

```
Out[62]:
```

	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 [63]: df.columns
```

```
Out[63]: Index(['CountryName', 'CountryCode', 'BirthRate', 'InternetUsers',
               'IncomeGroup'],
              dtype='object')
```

```
In [64]: df['InternetUsers']
```

```
Out[64]: 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
```

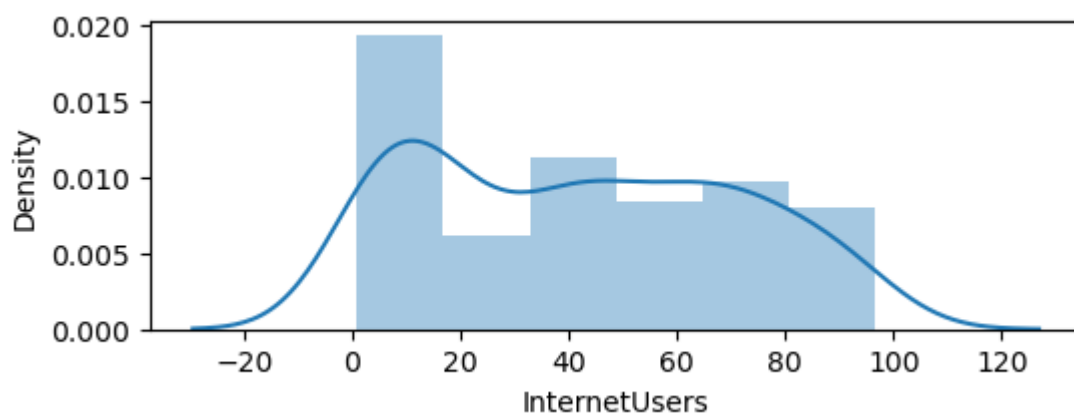
```
In [65]: vis1=plt.distplot(df["InternetUsers"])
```

```
-----
AttributeError                                Traceback (most recent call last)
Cell In[65], line 1
----> 1 vis1=plt.distplot(df["InternetUsers"])

AttributeError: module 'matplotlib.pyplot' has no attribute 'distplot'
```

```
In [ ]: vis1=sns.distplot(df["InternetUsers"])
```

```
In [66]: vis1 = sns.distplot(df["InternetUsers"])
         plt.show(vis1)
```

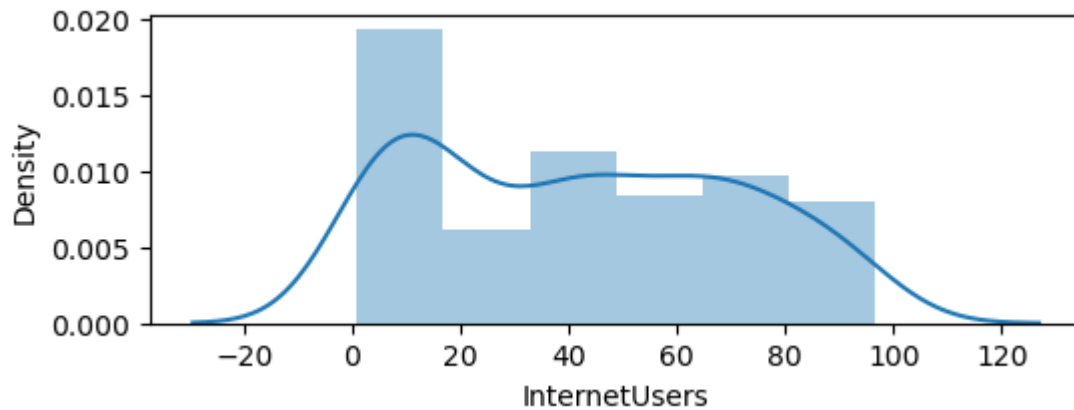


```
In [67]: df['InternetUsers']
```

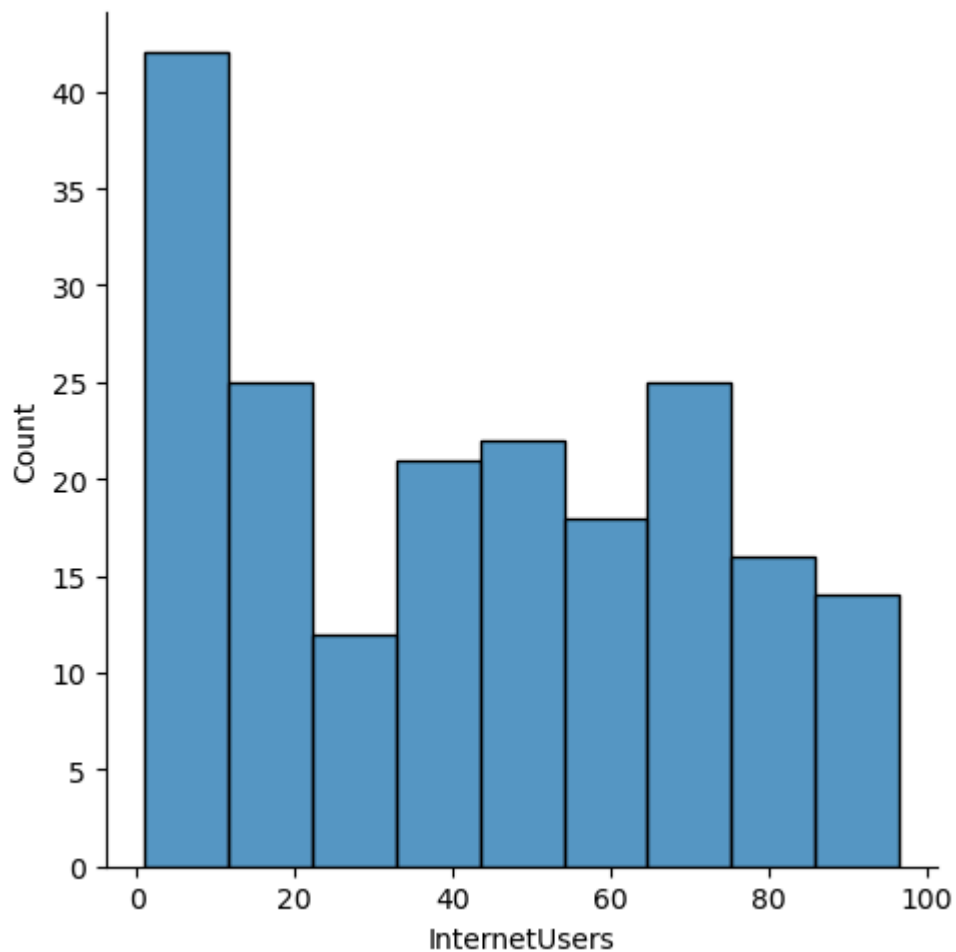
```
Out[67]: 0      78.9
         1       5.9
         2      19.1
         3      57.2
         4      88.0
         ...
        190     20.0
        191     46.5
        192       2.2
        193     15.4
        194     18.5
        Name: InternetUsers, Length: 195, dtype: float64
```

```
In [ ]: df.head()
```

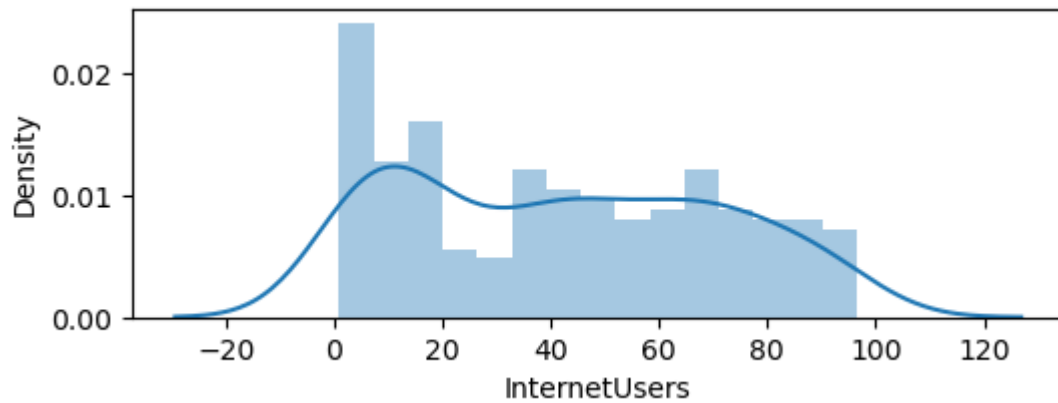
```
In [68]: vis1=sns.distplot(df['InternetUsers']) #distplot means distributed line  
# univariate analysis--plot the graph using 1 variable is called univariate anal
```



```
In [70]: vis2=sns.displot(df['InternetUsers'])
```

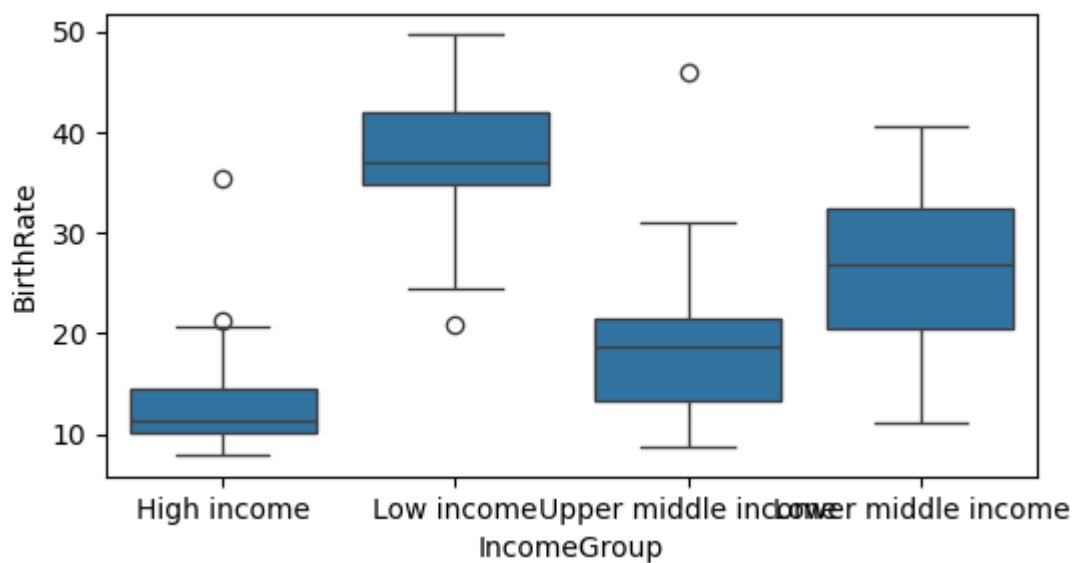


```
In [71]: vis3=sns.distplot(df['InternetUsers'],bins=15)
```



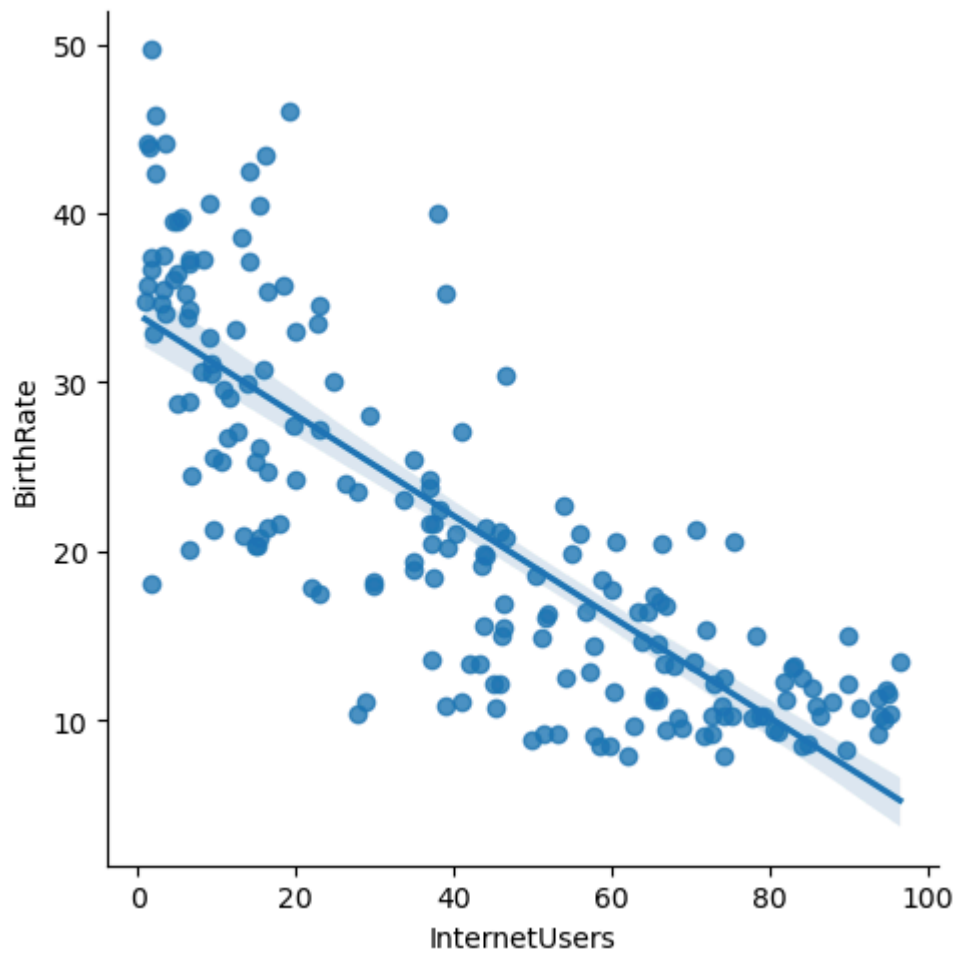
```
In [78]: plt.rcParams['figure.figsize']=6,3
```

```
In [80]: vis4=sns.boxplot(data=df,x='IncomeGroup',y='BirthRate')
# Bivariate means plot the graph using two variables is called Bivariate analysis
```



```
In [81]: # outlier-which doesnot obey the common behaviour
#in statistics outlier is the datapoint which is very far from other objects/obs
# outlier is also called as Anomaly Detection
```

```
In [84]: vis5=sns.lmplot(data=df,x='InternetUsers',y='BirthRate')#lmplot- Linear model plot
```



```
In [88]: vs=sns.lmplot(data=df,x='BirthRate',y='IncomeGroup')
```



```

-----
ValueError                                Traceback (most recent call last)
Cell In[88], line 1
----> 1 vs=sns.lmplot(data=df,x='BirthRate',y='IncomeGroup')

File ~\anaconda3\Lib\site-packages\seaborn\regression.py:640, in lmplot(data, x,
y, hue, col, row, palette, col_wrap, height, aspect, markers, sharex, sharey, hue
_order, col_order, row_order, legend, legend_out, x_estimator, x_bins, x_ci, scat
ter, fit_reg, ci, n_boot, units, seed, order, logistic, lowess, robust, logx, x_p
artial, y_partial, truncate, x_jitter, y_jitter, scatter_kws, line_kws, facet_kw
s)
    637     ax.update_datalim(xys, updatey=False)
    638     ax.autoscale_view(scaley=False)
--> 640 facets.map_dataframe(update_datalim, x=x, y=y)
    642 # Draw the regression plot on each facet
    643 regplot_kws = dict(
    644     x_estimator=x_estimator, x_bins=x_bins, x_ci=x_ci,
    645     scatter=scatter, fit_reg=fit_reg, ci=ci, n_boot=n_boot, units=units,
    (...)
    649     scatter_kws=scatter_kws, line_kws=line_kws,
    650 )

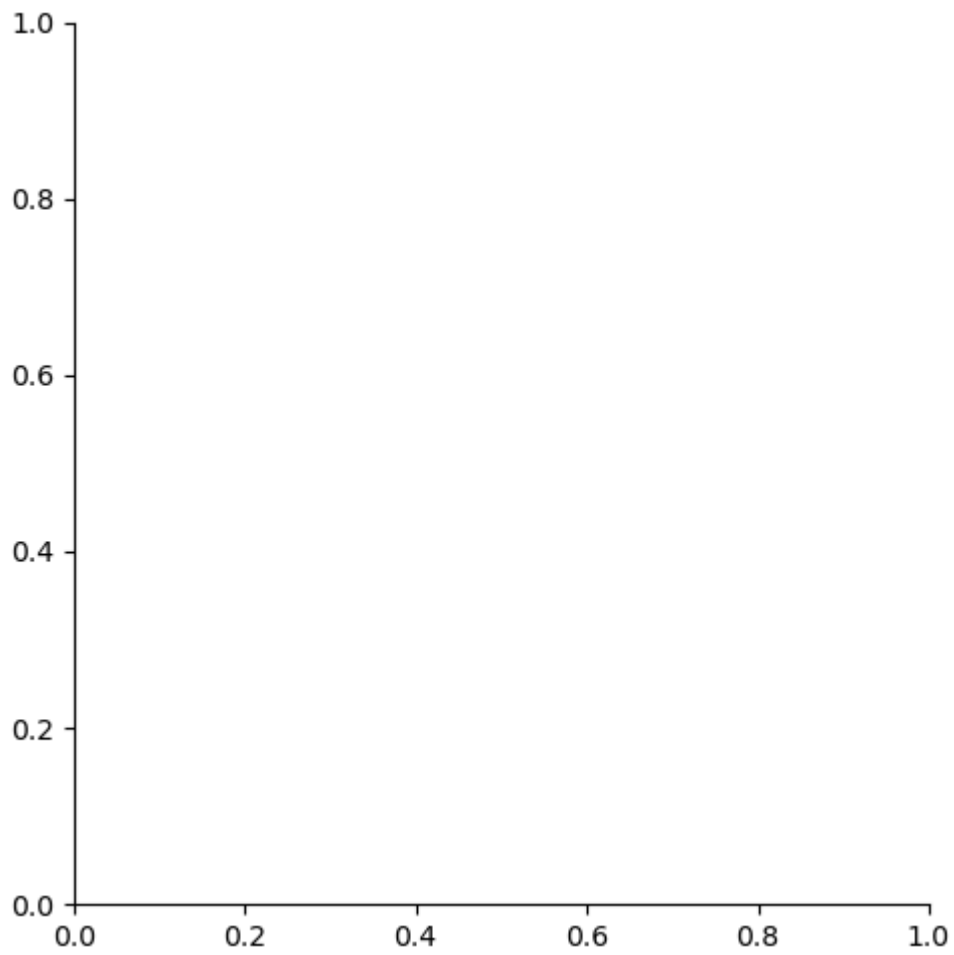
File ~\anaconda3\Lib\site-packages\seaborn\axisgrid.py:825, in FacetGrid.map_data
frame(self, func, *args, **kwargs)
    822     kwargs["data"] = data_ijk
    824     # Draw the plot
--> 825     self._facet_plot(func, ax, args, kwargs)
    827 # For axis labels, prefer to use positional args for backcompat
    828 # but also extract the x/y kwargs and use if no corresponding arg
    829 axis_labels = [kwargs.get("x", None), kwargs.get("y", None)]

File ~\anaconda3\Lib\site-packages\seaborn\axisgrid.py:854, in FacetGrid._facet_p
lot(self, func, ax, plot_args, plot_kws)
    852     plot_args = []
    853     plot_kws["ax"] = ax
--> 854 func(*plot_args, **plot_kws)
    856 # Sort out the supporting information
    857 self._update_legend_data(ax)

File ~\anaconda3\Lib\site-packages\seaborn\regression.py:636, in lmplot.<locals>.
update_datalim(data, x, y, ax, **kws)
    635 def update_datalim(data, x, y, ax, **kws):
--> 636     xys = data[[x, y]].to_numpy().astype(float)
    637     ax.update_datalim(xys, updatey=False)
    638     ax.autoscale_view(scaley=False)

ValueError: could not convert string to float: 'High income'

```



```
In [89]: vs=sns.lmplot(data=df,x='CountryCode',y='IncomeGroup')
```

```

-----
ValueError                                Traceback (most recent call last)
Cell In[89], line 1
----> 1 vs=sns.lmplot(data=df,x='CountryCode',y='IncomeGroup')

File ~\anaconda3\Lib\site-packages\seaborn\regression.py:640, in lmplot(data, x,
y, hue, col, row, palette, col_wrap, height, aspect, markers, sharex, sharey, hue
_order, col_order, row_order, legend, legend_out, x_estimator, x_bins, x_ci, scat
ter, fit_reg, ci, n_boot, units, seed, order, logistic, lowess, robust, logx, x_p
artial, y_partial, truncate, x_jitter, y_jitter, scatter_kws, line_kws, facet_kw
s)
    637     ax.update_datalim(xys, updatey=False)
    638     ax.autoscale_view(scaley=False)
--> 640 facets.map_dataframe(update_datalim, x=x, y=y)
    642 # Draw the regression plot on each facet
    643 regplot_kws = dict(
    644     x_estimator=x_estimator, x_bins=x_bins, x_ci=x_ci,
    645     scatter=scatter, fit_reg=fit_reg, ci=ci, n_boot=n_boot, units=units,
    (...)
    649     scatter_kws=scatter_kws, line_kws=line_kws,
    650 )

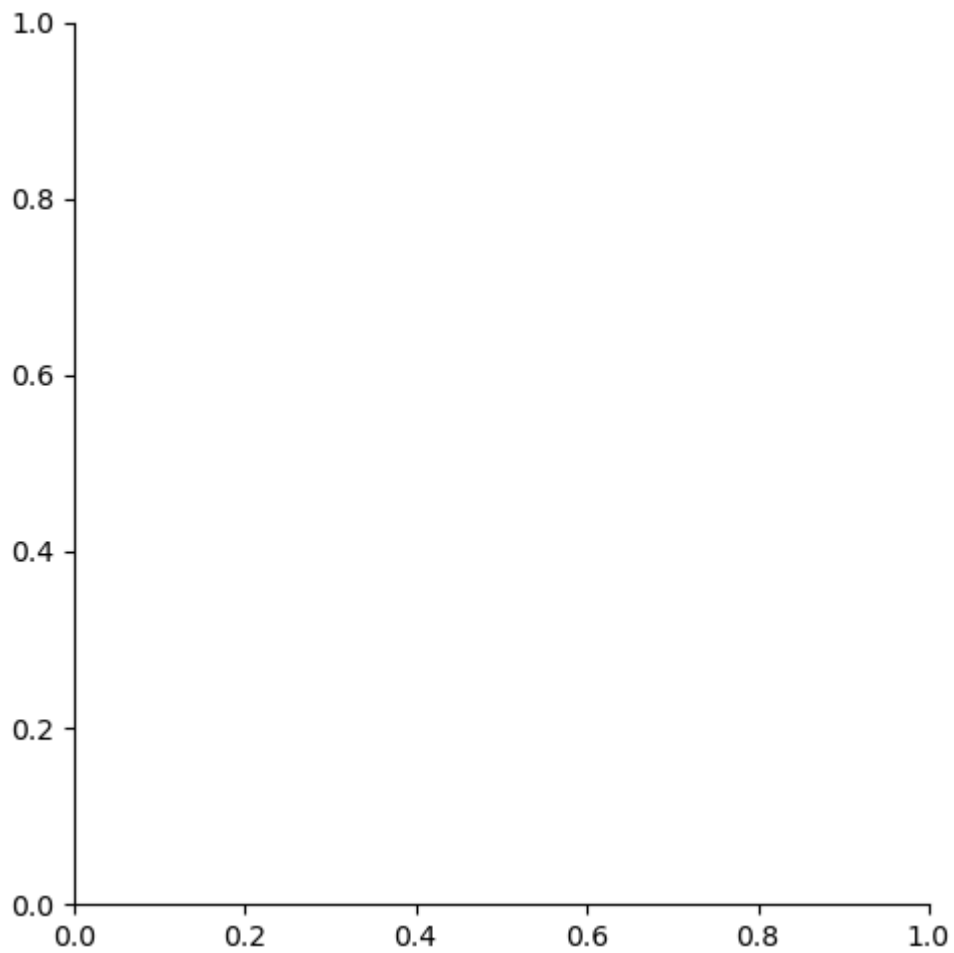
File ~\anaconda3\Lib\site-packages\seaborn\axisgrid.py:825, in FacetGrid.map_data
frame(self, func, *args, **kwargs)
    822     kwargs["data"] = data_ijk
    824     # Draw the plot
--> 825     self._facet_plot(func, ax, args, kwargs)
    827 # For axis labels, prefer to use positional args for backcompat
    828 # but also extract the x/y kwargs and use if no corresponding arg
    829 axis_labels = [kwargs.get("x", None), kwargs.get("y", None)]

File ~\anaconda3\Lib\site-packages\seaborn\axisgrid.py:854, in FacetGrid._facet_p
lot(self, func, ax, plot_args, plot_kws)
    852     plot_args = []
    853     plot_kws["ax"] = ax
--> 854 func(*plot_args, **plot_kws)
    856 # Sort out the supporting information
    857 self._update_legend_data(ax)

File ~\anaconda3\Lib\site-packages\seaborn\regression.py:636, in lmplot.<locals>.
update_datalim(data, x, y, ax, **kws)
    635 def update_datalim(data, x, y, ax, **kws):
--> 636     xys = data[[x, y]].to_numpy().astype(float)
    637     ax.update_datalim(xys, updatey=False)
    638     ax.autoscale_view(scaley=False)

ValueError: could not convert string to float: 'ABW'

```



```
In [90]: vs=sns.lmplot(data=df,x='CountryName',y='IncomeGroup')
```

```

-----
ValueError                                Traceback (most recent call last)
Cell In[90], line 1
----> 1 vs=sns.lmplot(data=df,x='CountryName',y='IncomeGroup')

File ~\anaconda3\Lib\site-packages\seaborn\regression.py:640, in lmplot(data, x,
y, hue, col, row, palette, col_wrap, height, aspect, markers, sharex, sharey, hue
_order, col_order, row_order, legend, legend_out, x_estimator, x_bins, x_ci, scat
ter, fit_reg, ci, n_boot, units, seed, order, logistic, lowess, robust, logx, x_p
artial, y_partial, truncate, x_jitter, y_jitter, scatter_kws, line_kws, facet_kw
s)
    637     ax.update_datalim(xys, updatey=False)
    638     ax.autoscale_view(scaley=False)
--> 640 facets.map_dataframe(update_datalim, x=x, y=y)
    642 # Draw the regression plot on each facet
    643 regplot_kws = dict(
    644     x_estimator=x_estimator, x_bins=x_bins, x_ci=x_ci,
    645     scatter=scatter, fit_reg=fit_reg, ci=ci, n_boot=n_boot, units=units,
    (...)
    649     scatter_kws=scatter_kws, line_kws=line_kws,
    650 )

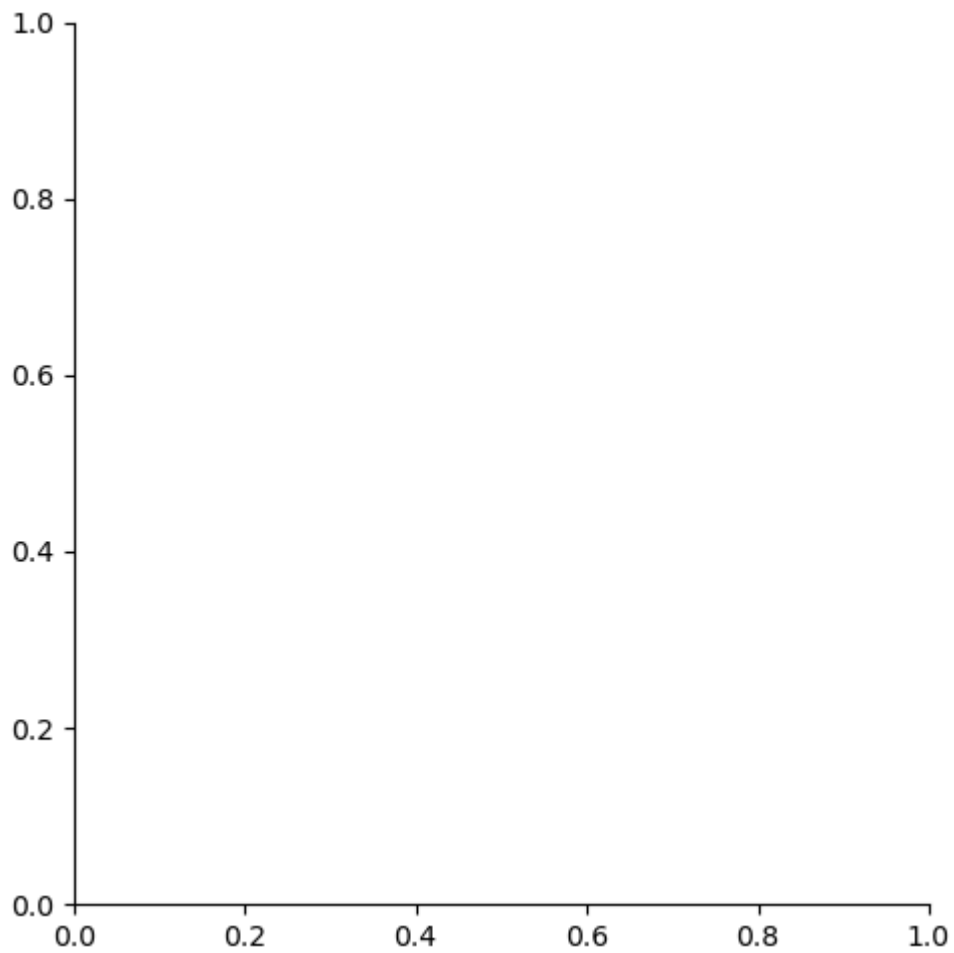
File ~\anaconda3\Lib\site-packages\seaborn\axisgrid.py:825, in FacetGrid.map_data
frame(self, func, *args, **kwargs)
    822     kwargs["data"] = data_ijk
    824     # Draw the plot
--> 825     self._facet_plot(func, ax, args, kwargs)
    827 # For axis labels, prefer to use positional args for backcompat
    828 # but also extract the x/y kwargs and use if no corresponding arg
    829 axis_labels = [kwargs.get("x", None), kwargs.get("y", None)]

File ~\anaconda3\Lib\site-packages\seaborn\axisgrid.py:854, in FacetGrid._facet_p
lot(self, func, ax, plot_args, plot_kws)
    852     plot_args = []
    853     plot_kws["ax"] = ax
--> 854 func(*plot_args, **plot_kws)
    856 # Sort out the supporting information
    857 self._update_legend_data(ax)

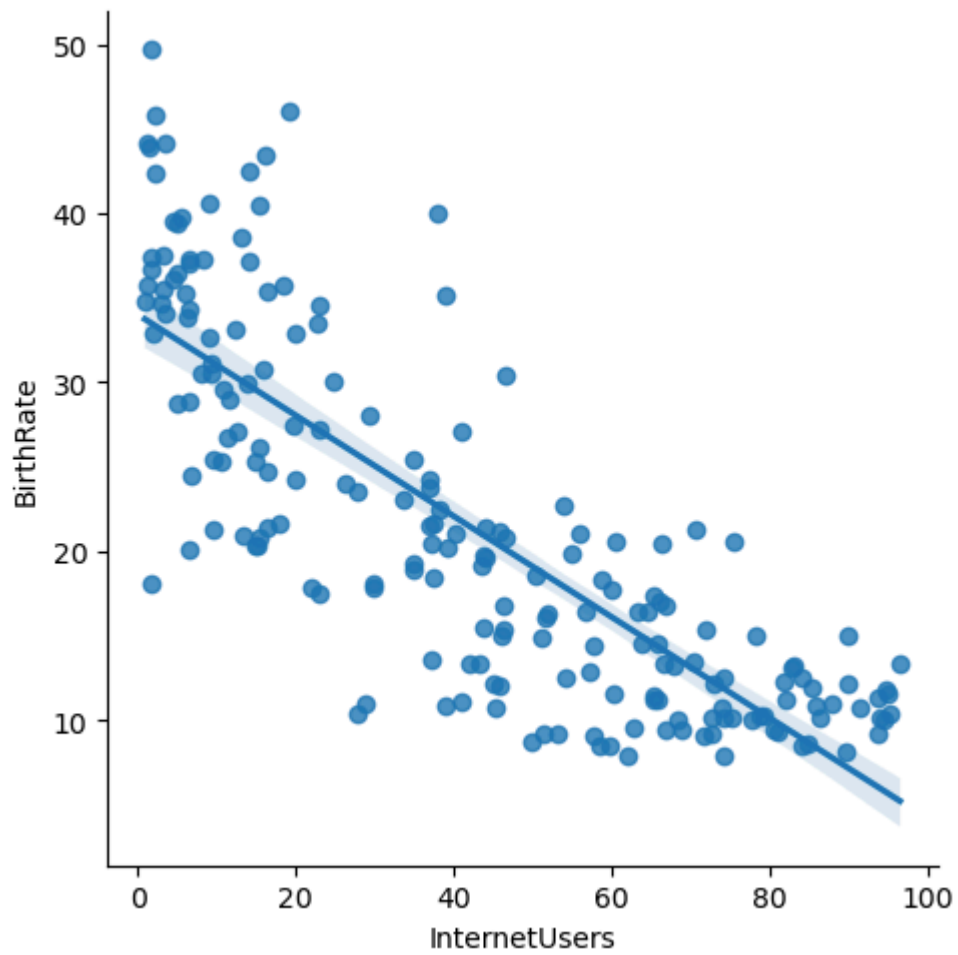
File ~\anaconda3\Lib\site-packages\seaborn\regression.py:636, in lmplot.<locals>.
update_datalim(data, x, y, ax, **kws)
    635 def update_datalim(data, x, y, ax, **kws):
--> 636     xys = data[[x, y]].to_numpy().astype(float)
    637     ax.update_datalim(xys, updatey=False)
    638     ax.autoscale_view(scaley=False)

ValueError: could not convert string to float: 'Aruba'

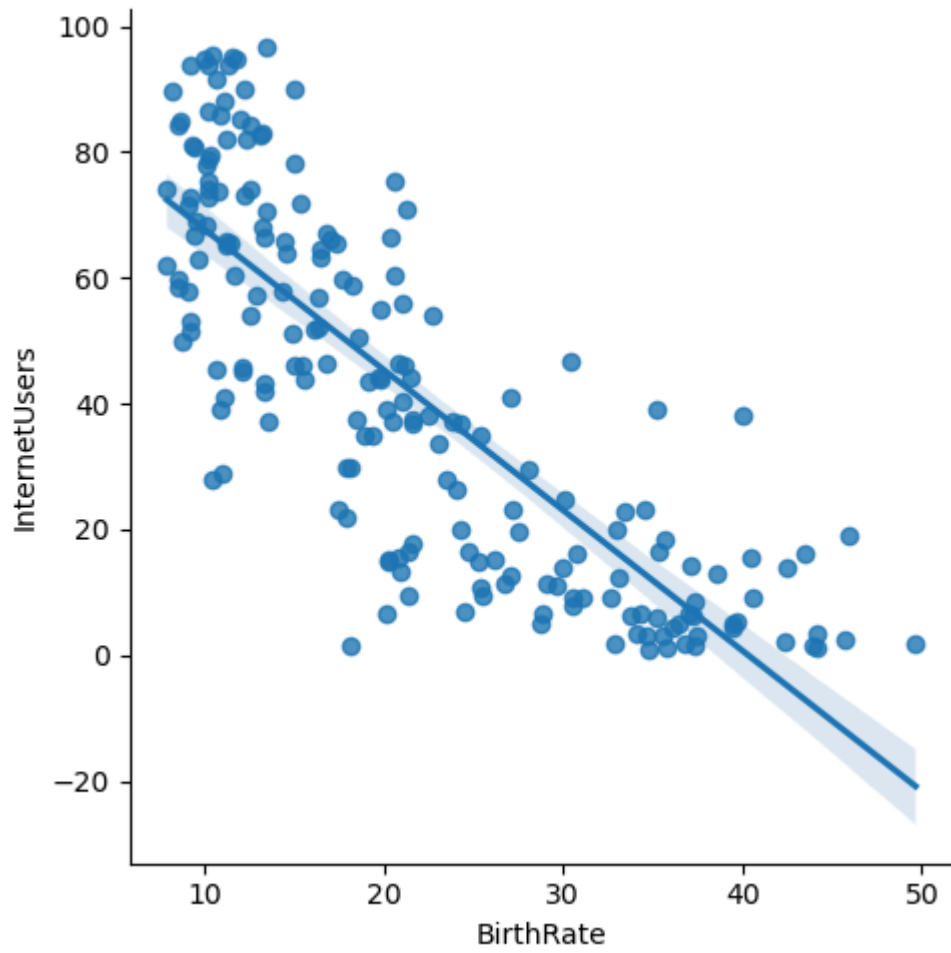
```



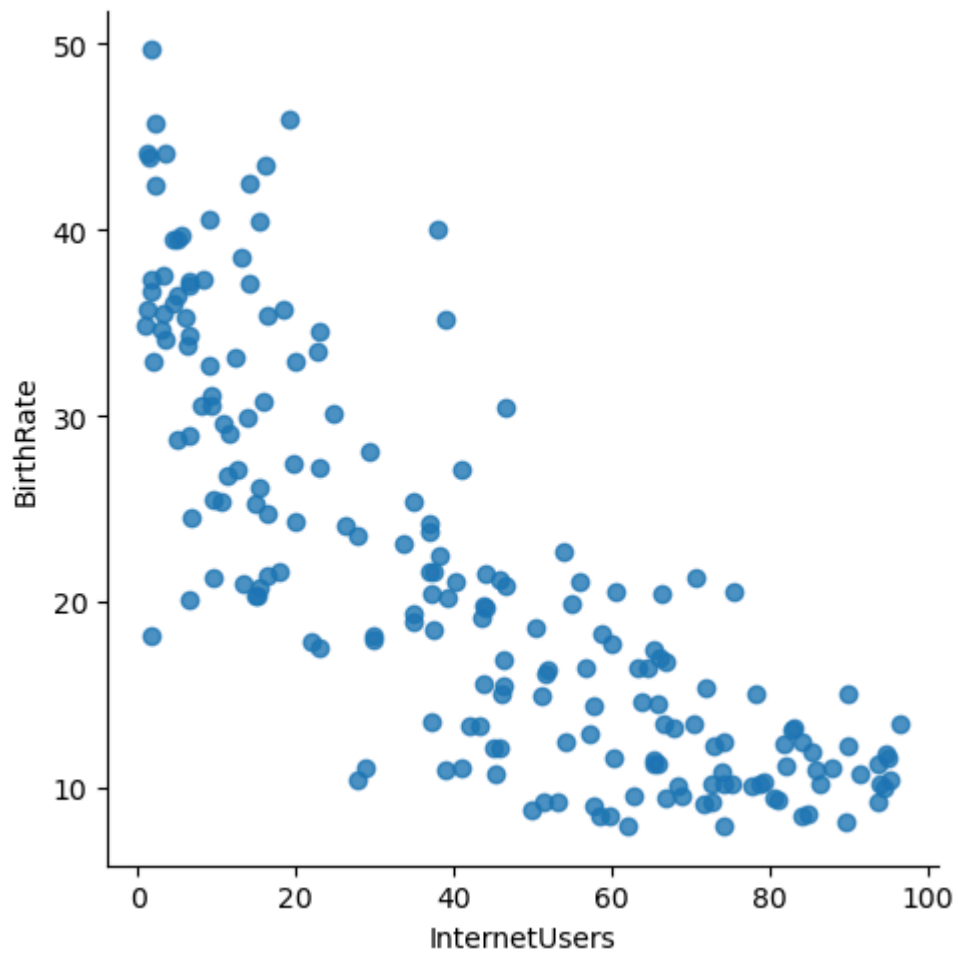
```
In [92]: vis5=sns.lmplot(data=df,x='InternetUsers',y='BirthRate')
```



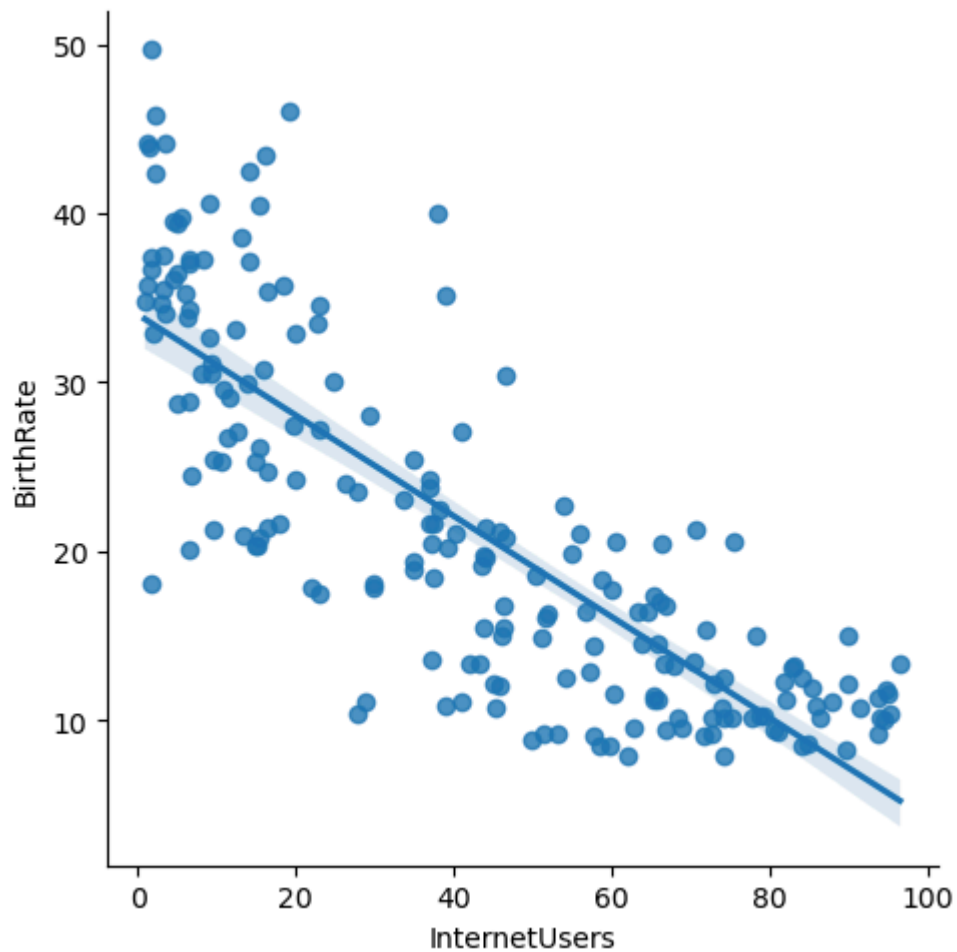
```
In [93]: vis5=sns.lmplot(data=df,x='BirthRate',y='InternetUsers')
```



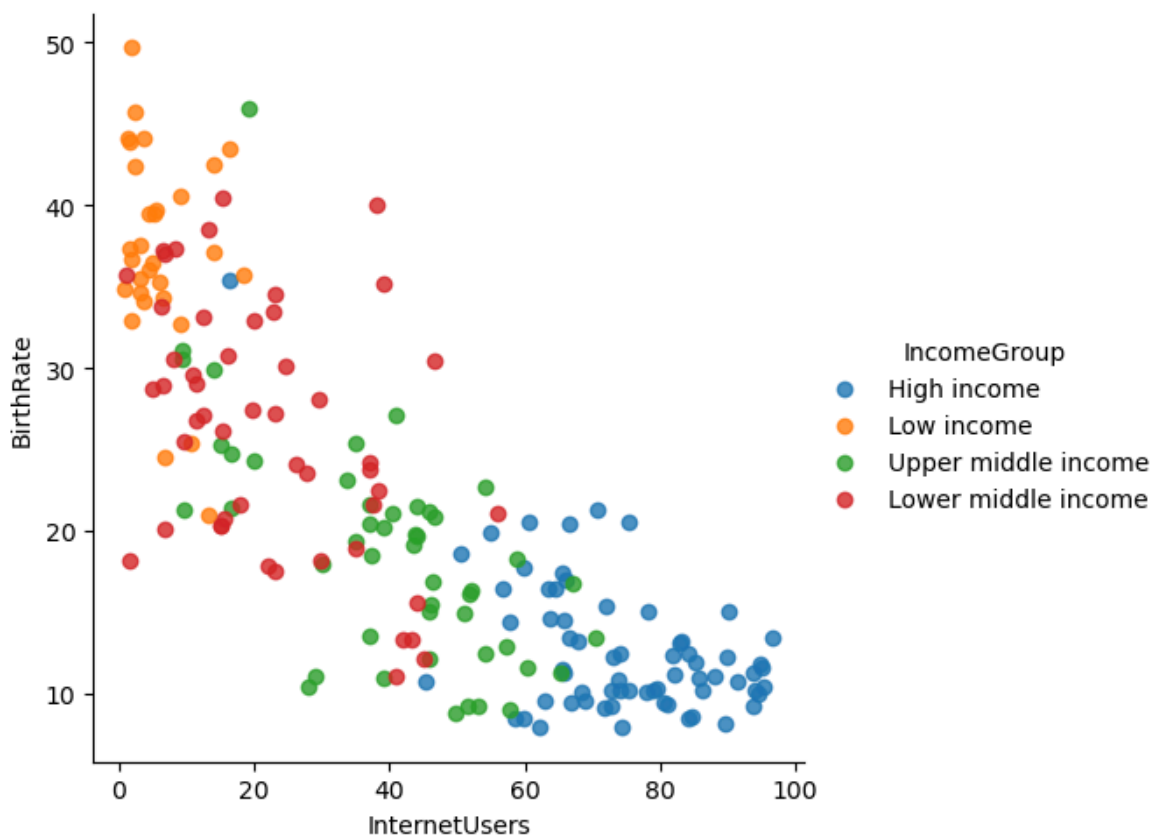
```
In [96]: vis5=sns.lmplot(data=df,x='InternetUsers',y='BirthRate',fit_reg=False) # Regress
```

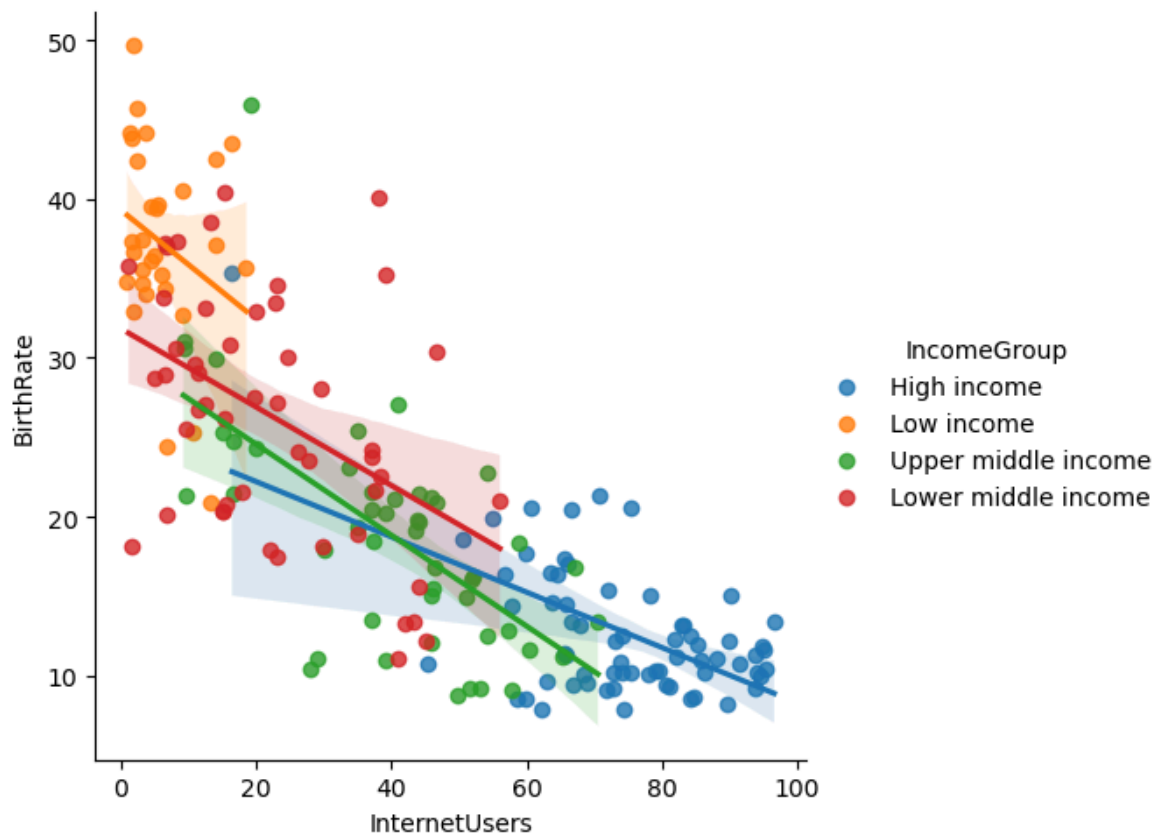
```
In [97]: vis5=sns.lmplot(data=df,x='InternetUsers',y='BirthRate',fit_reg=True)
```



In [98]: `vis8=sns.lmplot(data=df,x='InternetUsers',y='BirthRate',fit_reg=False,hue='Income`



In [100... `vis8=sns.lmplot(data=df,x='InternetUsers',y='BirthRate',fit_reg=True,hue='Income`

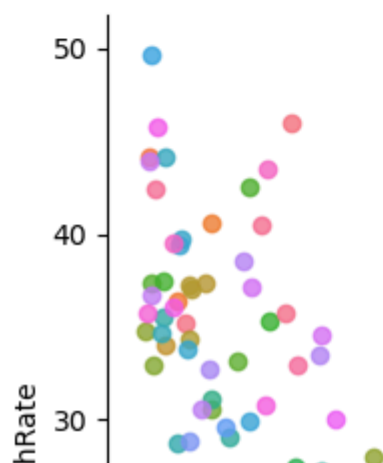


In [106...

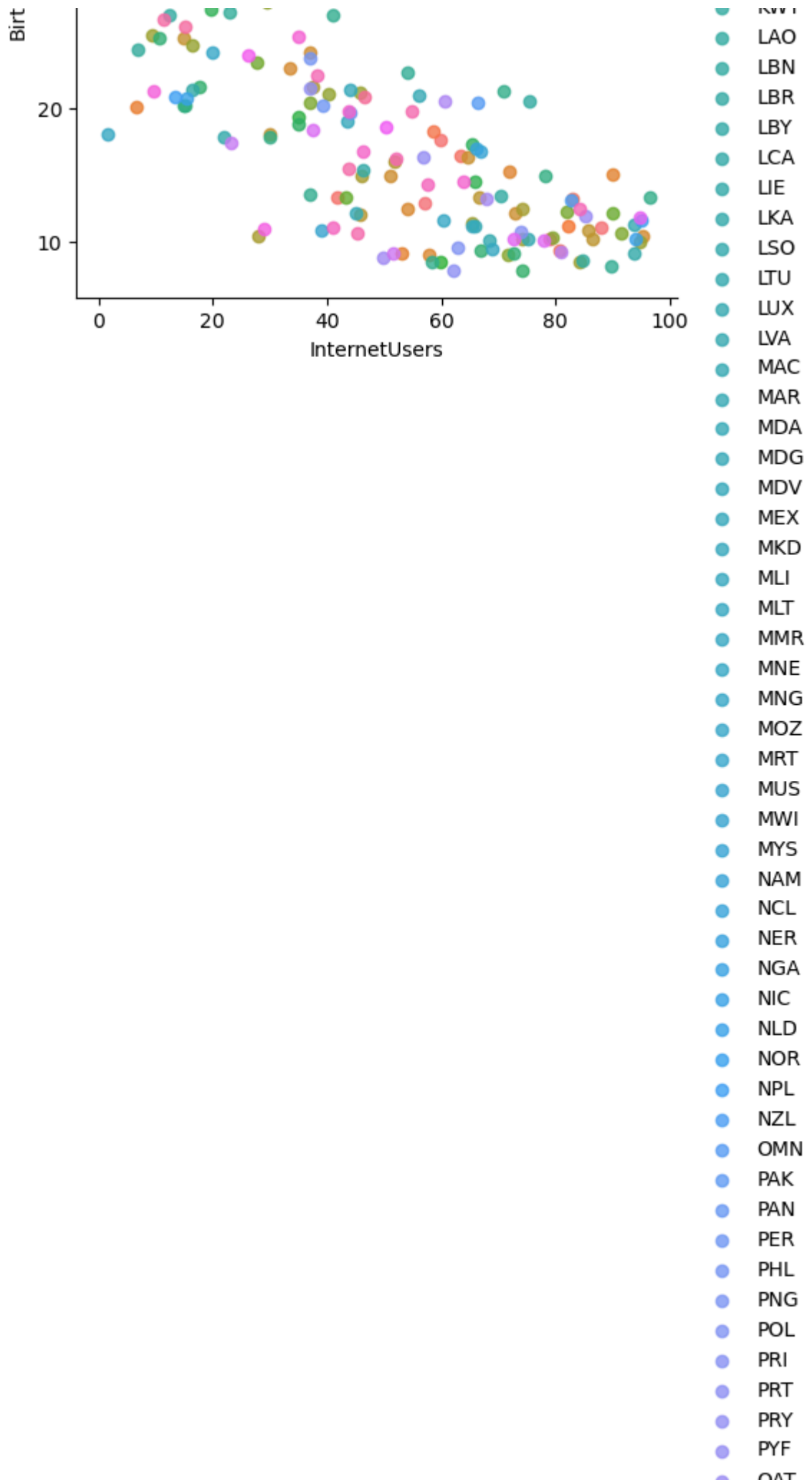
```
vis8=sns.lmplot(data=df,x='InternetUsers',y='BirthRate',fit_reg=False,hue='Count
```

CountryCode

● ABW
● AFG
● AGO
● ALB
● ARE
● ARG
● ARM
● ATG
● AUS
● AUT
● AZE
● BDI
● BEL
● BEN
● BFA
● BGD
● BGR
● BHR
● BHS
● BIH
● BLR
● BLZ
● BMU
● BOL
● BRA
● BRB
● BRN
● BTN
● BWA
● CAF
● CAN
● CHE
● CHL
● CHN
● CIV
● CMR
● COG
● COL
● COM
● CPV
● CRI
● CUB
● CYM
● CYP
● CZE
● DEU
● DJI
● DNK



- DNK
- DOM
- DZA
- ECU
- EGY
- ERI
- ESP
- EST
- ETH
- FIN
- FJI
- FRA
- FSM
- GAB
- GBR
- GEO
- GHA
- GIN
- GMB
- GNB
- GNQ
- GRC
- GRD
- GRL
- GTM
- GUM
- GUY
- HKG
- HND
- HRV
- HTI
- HUN
- IDN
- IND
- IRL
- IRN
- IRQ
- ISL
- ISR
- ITA
- JAM
- JOR
- JPN
- KAZ
- KEN
- KGZ
- KHM
- KIR
- KOR
- KWT



- ROU
- RUS
- RWA
- SAU
- SDN
- SEN
- SGP
- SLB
- SLE
- SLV
- SOM
- SRB
- SSD
- STP
- SUR
- SVK
- SVN
- SWE
- SWZ
- SYC
- SYR
- TCD
- TGO
- THA
- TJK
- TKM
- TLS
- TON
- TTO
- TUN
- TUR
- TZA
- UGA
- UKR
- URY
- USA
- UZB
- VCT
- VEN
- VIR
- VNM
- VUT
- PSE
- WSM
- YEM
- ZAF
- COD
- ZMB
- ZWE

In []: