

Internet Users Worldwide

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1 Project: Internet Users Worldwide

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1.1 Table of Contents

Introduction

Data Wrangling

Exploratory Data Analysis

Conclusions

1.2 Introduction

The main idea behind studying the usage of internet around the world is simply to better understand distribution of access and possibly make some projections. Basically we have only three dimensions on this dataset: the Country name, the Year, and the total number of persons with internet access. Based on this data it is possible to calculate an extra field of growth rate, which can be expressed as a percentage. Having such variations on our hands, we can answer some questions like what was the biggest internet growth from one year to another in one country in total number of users, or even if there has ever been a decrease in the number of users in any country (maybe due to some war, the arab spring, crisis in Syria?) at any given point.

Data was taken from <https://www.gapminder.org/data/>

```
In [77]: # Use this cell to set up import statements for all of the packages that you
#        plan to use.
```

```
# Remember to include a 'magic word' so that your visualizations are plotted
# inline with the notebook. See this page for more:
# http://ipython.readthedocs.io/en/stable/interactive/magics.html
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
% matplotlib inline
```

Data Wrangling

We will be using Pandas Dataframes for this analysis.

1.2.1 General Properties

```
In [78]: # Load your data and print out a few lines. Perform operations to inspect data
# types and look for instances of missing or possibly errant data.
internet_users = pd.read_excel('Internet user total.xlsx')
internet_users.head()
```

```
Out[78]:
```

	Internet users, total number	1990	1991	1992	1993	1994	\
0	Afghanistan	0.0	NaN	NaN	NaN	NaN	
1	Albania	0.0	NaN	NaN	NaN	NaN	
2	Algeria	0.0	NaN	NaN	NaN	100.090832	
3	American Samoa	0.0	NaN	NaN	NaN	NaN	
4	Andorra	0.0	NaN	NaN	NaN	NaN	

	1995	1996	1997	1998	...	2002	\
0	NaN	NaN	NaN	NaN	...	1123.920517	
1	350.820117	1002.157509	1502.476346	2002.216783	...	12052.645369	
2	500.465235	500.469039	3002.798156	6005.526744	...	500441.425787	
3	NaN	NaN	NaN	NaN	...	NaN	
4	NaN	995.756049	1979.716334	4424.113974	...	7774.565416	

	2003	2004	2005	2006	2007	\
0	22569.277533	2.824412e+04	3.380450e+05	5.988651e+05	5.537710e+05	
1	30194.039918	7.563375e+04	1.898870e+05	3.033497e+05	4.765945e+05	
2	700615.293472	1.501387e+06	1.921982e+06	2.462986e+06	3.204578e+06	
3	NaN	NaN	NaN	NaN	NaN	
4	9780.916489	2.020683e+04	2.929038e+04	3.908782e+04	5.768109e+04	

	2008	2009	2010	2011
0	5.490559e+05	1.085510e+06	1256469.72	NaN
1	7.590813e+05	1.315402e+06	1441927.80	NaN
2	3.504773e+06	3.924904e+06	4433526.00	NaN
3	NaN	NaN	NaN	NaN
4	5.783693e+04	6.571155e+04	68739.84	NaN

[5 rows x 23 columns]

We notice that some lines have only NaN values, which makes no sense. Those should be removed. Also, we notice that some nations do not have values for old dates, which makes sense since the internet started rising globally according to Wikipedia (not as a research tool between Universities) just since early 90s. For this reason, I will choose the year 2001 as a reference, and if the country has a NaN value the entire line will be removed. Afterward, all NaN values will be replaced by zeroes. We will also rename the index column to "Country".

```
In [79]: internet_users = internet_users[np.isfinite(internet_users['2001'])]
internet_users.fillna(0,inplace=True)
internet_users = internet_users.rename(columns={'Internet users, total number':'Country'})
internet_users.head()
```

```

Out [79]:
      Country 1990 1991 1992 1993      1994      1995 \
0    Afghanistan 0.0 0.0 0.0 0.0 0.000000 0.000000
1      Albania 0.0 0.0 0.0 0.0 0.000000 350.820117
2      Algeria 0.0 0.0 0.0 0.0 100.090832 500.465235
5      Angola 0.0 0.0 0.0 0.0 0.000000 0.000000
6 Antigua and Barbuda 0.0 0.0 0.0 0.0 0.000000 1510.651721

      1996      1997      1998 ...      2002      2003 \
0 0.000000 0.000000 0.000000 ... 1123.920517 22569.277533
1 1002.157509 1502.476346 2002.216783 ... 12052.645369 30194.039918
2 500.469039 3002.798156 6005.526744 ... 500441.425787 700615.293472
5 96.624245 725.799034 2424.499171 ... 40260.379005 57158.963169
6 2013.777727 2517.193852 3020.806471 ... 10063.375000 14078.562638

      2004      2005      2006      2007      2008 \
0 2.824412e+04 3.380450e+05 5.988651e+05 5.537710e+05 5.490559e+05
1 7.563375e+04 1.898870e+05 3.033497e+05 4.765945e+05 7.590813e+05
2 1.501387e+06 1.921982e+06 2.462986e+06 3.204578e+06 3.504773e+06
5 7.417261e+04 1.885300e+05 3.244978e+05 5.608117e+05 8.297463e+05
6 2.010192e+04 2.913262e+04 5.321297e+04 6.020606e+04 6.518681e+04

      2009      2010      2011
0 1.085510e+06 1256469.72 0.0
1 1.315402e+06 1441927.80 0.0
2 3.924904e+06 4433526.00 0.0
5 1.113307e+06 1908191.20 0.0
6 6.514908e+04 70968.00 0.0

```

[5 rows x 23 columns]

Now, we can have a more general view of the whole dataset with the describe function:

```
In [80]: internet_users.describe()
```

```

Out [80]:
      1990      1991      1992      1993      1994 \
count 1.960000e+02 1.960000e+02 1.960000e+02 1.960000e+02 1.960000e+02
mean 1.340288e+04 2.185336e+04 3.495840e+04 5.044544e+04 1.030947e+05
std 1.424906e+05 2.137989e+05 3.205830e+05 4.275411e+05 9.213485e+05
min 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
25% 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
50% 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
75% 0.000000e+00 0.000000e+00 0.000000e+00 9.366564e+02 4.467333e+03
max 1.988024e+06 2.975535e+06 4.453200e+06 5.924598e+06 1.281192e+07

      1995      1996      1997      1998      1999 \
count 1.960000e+02 1.960000e+02 1.960000e+02 1.960000e+02 1.960000e+02
mean 1.984319e+05 3.694079e+05 5.921310e+05 9.231412e+05 1.388812e+06
std 1.768884e+06 3.191477e+06 4.324240e+06 6.129312e+06 7.656223e+06

```

min	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00
25%	0.000000e+00	1.532582e+02	7.825498e+02	1.988207e+03	4.968147e+03
50%	3.504101e+02	2.000100e+03	4.992679e+03	1.014368e+04	2.813967e+04
75%	1.952791e+04	4.411411e+04	8.699426e+04	1.501114e+05	2.450025e+05
max	2.460056e+07	4.423269e+07	5.893568e+07	8.305303e+07	1.001255e+08

	...	2002	2003	2004	2005 \
count	...	1.960000e+02	1.960000e+02	1.960000e+02	1.960000e+02
mean	...	3.306814e+06	3.872760e+06	4.503447e+06	5.135800e+06
std	...	1.421755e+07	1.562491e+07	1.728795e+07	1.872051e+07
min	...	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00
25%	...	2.441795e+04	3.006752e+04	3.670026e+04	4.923302e+04
50%	...	1.235731e+05	1.529147e+05	2.126868e+05	3.392482e+05
75%	...	1.040248e+06	1.406971e+06	1.866206e+06	2.622681e+06
max	...	1.695767e+08	1.797180e+08	1.904301e+08	2.017430e+08

		2006	2007	2008	2009	2010 \
count	1.960000e+02	1.960000e+02	1.960000e+02	1.960000e+02	1.960000e+02	
mean	5.786826e+06	6.900786e+06	7.974482e+06	9.010653e+06	1.029808e+07	
std	2.016719e+07	2.453651e+07	2.902002e+07	3.386430e+07	3.906940e+07	
min	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	
25%	6.745723e+04	8.333297e+04	1.026825e+05	1.123229e+05	1.481696e+05	
50%	4.836016e+05	5.793525e+05	7.880628e+05	9.252343e+05	1.293429e+06	
75%	3.076977e+06	3.391346e+06	4.237466e+06	4.594224e+06	5.170963e+06	
max	2.064934e+08	2.267134e+08	3.001903e+08	3.857886e+08	4.600780e+08	

	2011
count	196.0
mean	0.0
std	0.0
min	0.0
25%	0.0
50%	0.0
75%	0.0
max	0.0

[8 rows x 22 columns]

1.2.2 Data Cleaning (Replace this with more specific notes!)

There is something odd about the year 2011: all values are zeroes. It seems this whole column should be removed.

```
In [81]: internet_users = internet_users.drop(columns=['2011'])
internet_users.head()
```

```
Out[81]:
```

	Country	1990	1991	1992	1993	1994	1995 \
0	Afghanistan	0.0	0.0	0.0	0.0	0.000000	0.000000

1		Albania	0.0	0.0	0.0	0.0	0.000000	350.820117
2		Algeria	0.0	0.0	0.0	0.0	100.090832	500.465235
5		Angola	0.0	0.0	0.0	0.0	0.000000	0.000000
6		Antigua and Barbuda	0.0	0.0	0.0	0.0	0.000000	1510.651721

	1996	1997	1998	...	2001 \
0	0.000000	0.000000	0.000000	...	1118.180665
1	1002.157509	1502.476346	2002.216783	...	10026.047580
2	500.469039	3002.798156	6005.526744	...	200180.427339
5	96.624245	725.799034	2424.499171	...	19570.007120
6	2013.777727	2517.193852	3020.806471	...	7045.119378

	2002	2003	2004	2005	2006 \
0	1123.920517	22569.277533	2.824412e+04	3.380450e+05	5.988651e+05
1	12052.645369	30194.039918	7.563375e+04	1.898870e+05	3.033497e+05
2	500441.425787	700615.293472	1.501387e+06	1.921982e+06	2.462986e+06
5	40260.379005	57158.963169	7.417261e+04	1.885300e+05	3.244978e+05
6	10063.375000	14078.562638	2.010192e+04	2.913262e+04	5.321297e+04

	2007	2008	2009	2010
0	5.537710e+05	5.490559e+05	1.085510e+06	1256469.72
1	4.765945e+05	7.590813e+05	1.315402e+06	1441927.80
2	3.204578e+06	3.504773e+06	3.924904e+06	4433526.00
5	5.608117e+05	8.297463e+05	1.113307e+06	1908191.20
6	6.020606e+04	6.518681e+04	6.514908e+04	70968.00

[5 rows x 22 columns]

It seems odd that the description of the dataset from the source says "Internet users, total number", but numbers do not have integer values. How is it possible to have a fraction of a human being? After looking at the original source (apparently, gapminder is a hub of data), it seems the total number of internet users per country is in fact the product of the percentage of internet users by the population of the country. So, in order to make more sense, it seems another function will have to be applied to round the number up.

```
In [82]: columns = []
         for year in range(1990,2011):
             columns.append(str(year))
         internet = internet_users[columns].applymap(np.int64)
         internet.insert(loc=0,column='Country',value=internet_users['Country'])
         internet.set_index('Country', inplace=True)
         internet_users = internet
         internet_users.head()
```

```
Out[82]:
```

	1990	1991	1992	1993	1994	1995	1996	1997	1998 \
Country									
Afghanistan	0	0	0	0	0	0	0	0	0
Albania	0	0	0	0	0	350	1002	1502	2002

Algeria	0	0	0	0	100	500	500	3002	6005
Angola	0	0	0	0	0	0	96	725	2424
Antigua and Barbuda	0	0	0	0	0	1510	2013	2517	3020

	1999	...	2001	2002	2003	2004	2005	\
Country		...						
Afghanistan	0	...	1118	1123	22569	28244	338045	
Albania	2502	...	10026	12052	30194	75633	189886	
Algeria	60054	...	200180	500441	700615	1501386	1921981	
Angola	9723	...	19570	40260	57158	74172	188529	
Antigua and Barbuda	4027	...	7045	10063	14078	20101	29132	

	2006	2007	2008	2009	2010
Country					
Afghanistan	598865	553770	549055	1085510	1256469
Albania	303349	476594	759081	1315401	1441927
Algeria	2462985	3204577	3504773	3924903	4433526
Angola	324497	560811	829746	1113306	1908191
Antigua and Barbuda	53212	60206	65186	65149	70968

[5 rows x 21 columns]

So now it seems we have nice data and we are ready to start our investigation!

Exploratory Data Analysis

So, first thing we could do to analyze our data is to get a nice overview of maximum and minimums. Let's start by getting a list of which country had the max and min values in each year.

In [83]: `internet_users.apply(lambda x: x.idxmax(), axis=0)`

Out [83]:

1990	United States
1991	United States
1992	United States
1993	United States
1994	United States
1995	United States
1996	United States
1997	United States
1998	United States
1999	United States
2000	United States
2001	United States
2002	United States
2003	United States
2004	United States
2005	United States
2006	United States
2007	United States
2008	China

```

2009          China
2010          China
dtype: object

```

So far we have our first interesting observation about the data: American leadership was taken out by China in 2008 in terms of absolute numbers, which is understandable considering the size of chinese poupluation.

```
In [84]: internet_users.apply(lambda x: x.idxmin(), axis=0)
```

```

Out [84]: 1990          Afghanistan
          1991          Afghanistan
          1992          Afghanistan
          1993          Afghanistan
          1994          Afghanistan
          1995          Afghanistan
          1996          Afghanistan
          1997          Afghanistan
          1998          Afghanistan
          1999          Afghanistan
          2000          Afghanistan
          2001  Korea, Dem. Rep.
          2002          Australia
          2003          Australia
          2004          Australia
          2005          Belarus
          2006          Guyana
          2007          Guyana
          2008  Korea, Dem. Rep.
          2009          Eritrea
          2010          Guam
dtype: object

```

```
In [85]: internet_users.min(axis=0)
```

```

Out [85]: 1990      0
          1991      0
          1992      0
          1993      0
          1994      0
          1995      0
          1996      0
          1997      0
          1998      0
          1999      0
          2000      0
          2001      0
          2002      0
          2003      0

```

```

2004    0
2005    0
2006    0
2007    0
2008    0
2009    0
2010    0
dtype: int64

```

It is weird to see Australia in the list with lowest internet access, but what seems to be happening is that the country apparently is missing data for those years. The same must be happening to all countries that had internet internet access in any previous year and had a zero value in any subsequent time.

```
In [86]: internet_users.loc[list_countries_min_values_to_check]
```

```

Out[86]:
          1990  1991  1992  1993  1994  1995  1996  1997  1998  1999  ...  \
Country
Kiribati      0     0     0     0     0     0     0     0    499   999  ...
Tonga         0     0     0     0     0    118   157   493   741   991  ...
Samoa          0     0     0     0     0     0     0    299   399   499  ...
Gibraltar     0     0     0     0     0     0     0    725  1137  1615  ...
Vanuatu        0     0     0     0     0     0     97   244   487   975  ...

          2001  2002  2003  2004  2005  2006  2007  2008  2009  2010
Country
Kiribati   1999  2178  2663  3163  3679  4209  5702  6757  7842  8959
Tonga      2782  2879  2975  3963  4953  5946  7341  8346  10351 12486
Samoa      3002  4006  5015  5526  6042  8078  8610  9147  10944 12815
Gibraltar  5846  6625  8047  9468  11359 13243 15119 16986 19037 19008
Vanuatu    5365  6828  7805  9757  10732 12681 15121 16576 17534 19172

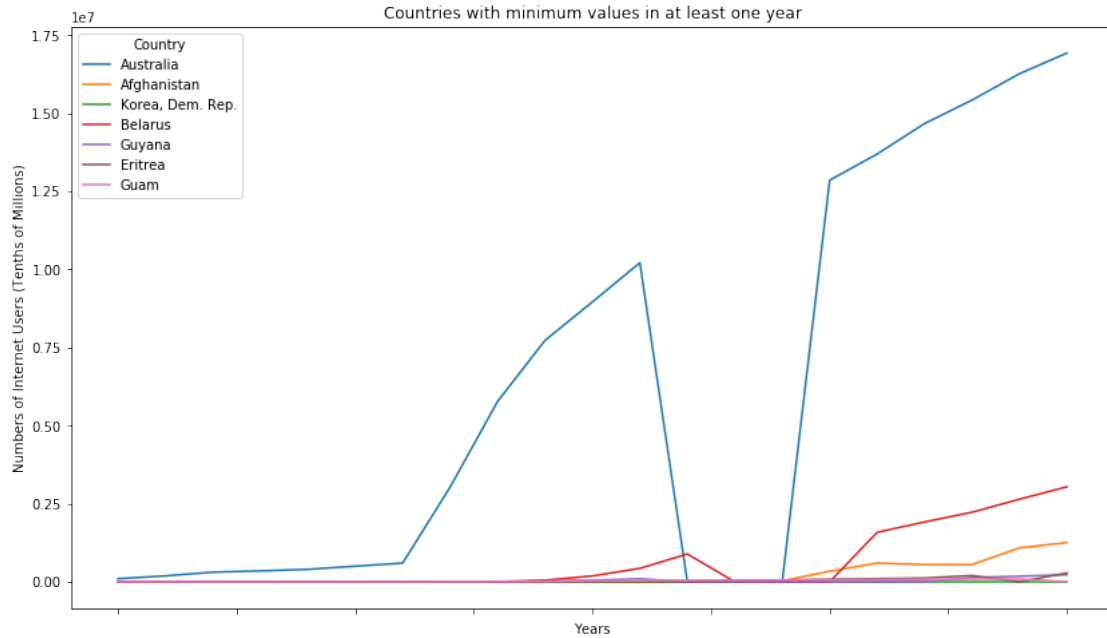
```

```
[5 rows x 21 columns]
```

```

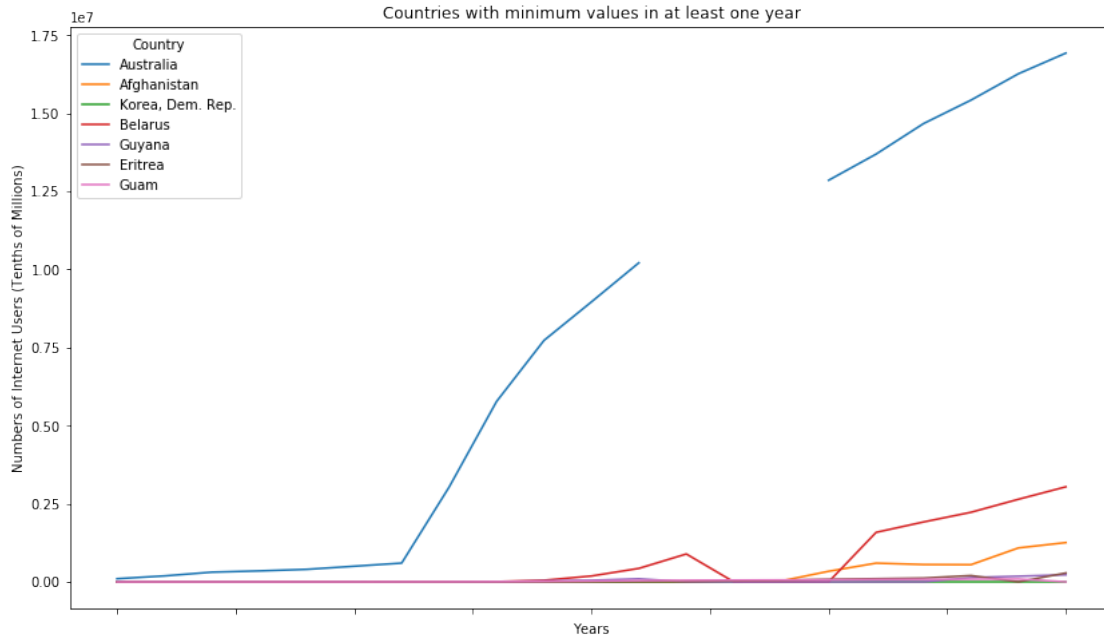
In [87]: list_countries_min_values_to_check = ['Australia', 'Afghanistan', 'Korea, Dem. Rep.', 'B
transposed_internet_users = internet_users.loc[list_countries_min_values_to_check].tr
transposed_internet_users.set(xlabel='Years', ylabel='Numbers of Internet Users (Tent
plt.show()

```

So it seems we might have two possible alternatives. First is to fill those missing gaps with some sort of approximation of the value expected. For this we probably would have to make a function to calculate those values. Second would be to try to remove the 0 values for those specific points and try to force matplotlib to smooth the graph.

```
In [88]: internet_users.at['Australia','2002']=None
internet_users.at['Australia','2003']=None
internet_users.at['Australia','2004']=None
transposed_internet_users = internet_users.loc[list_countries_min_values_to_check].transpose()
transposed_internet_users.set(xlabel='Years', ylabel='Numbers of Internet Users (Tenths of Millions)')
plt.show()
```



It seems it does not perform such task automatically, so we have to force it somehow. After some Google research, found out that in fact my mistake was to manually set all those values to zero in the beginning. Instead, I should have removed the NaN values with the `dropna()` function. So let's start it over.

```
In [89]: internet_users = pd.read_excel('Internet user total.xlsx')
internet_users = internet_users[np.isfinite(internet_users['2001'])]
internet_users = internet_users.rename(columns={'Internet users, total number': 'Country'})
internet_users = internet_users.drop(columns=['2011'])
internet = internet_users.fillna(value=0)
internet = internet.round(0)
internet_users = internet[internet_users.notna()]
#internet_users = internet_users.fillna(value='')
internet_users.set_index('Country', drop=True, inplace=True)
internet_users
```

```
Out [89]:
```

	1990	1991	1992	1993	\
Country					
Afghanistan	0.0	NaN	NaN	NaN	
Albania	0.0	NaN	NaN	NaN	
Algeria	0.0	NaN	NaN	NaN	
Angola	0.0	NaN	NaN	NaN	
Antigua and Barbuda	0.0	NaN	NaN	NaN	
Argentina	0.0	NaN	1004.0	10034.0	
Armenia	0.0	NaN	NaN	NaN	
Aruba	0.0	NaN	NaN	NaN	
Australia	100029.0	190043.0	310055.0	350047.0	

Austria	10000.0	20000.0	50000.0	60000.0
Azerbaijan	0.0	NaN	NaN	NaN
Bahamas	0.0	NaN	NaN	NaN
Bahrain	0.0	NaN	NaN	NaN
Bangladesh	0.0	NaN	NaN	NaN
Barbados	0.0	NaN	NaN	NaN
Belarus	0.0	NaN	NaN	NaN
Belgium	100.0	2003.0	10009.0	20009.0
Benin	0.0	NaN	NaN	NaN
Bermuda	0.0	NaN	NaN	NaN
Bhutan	0.0	NaN	NaN	NaN
Bolivia	0.0	NaN	NaN	NaN
Bosnia and Herzegovina	0.0	NaN	NaN	NaN
Botswana	0.0	0.0	NaN	NaN
Brazil	0.0	5003.0	20013.0	40029.0
Brunei	0.0	NaN	NaN	NaN
Bulgaria	0.0	NaN	NaN	200.0
Burkina Faso	0.0	NaN	NaN	NaN
Burundi	0.0	0.0	0.0	0.0
Cambodia	0.0	NaN	NaN	NaN
Cameroon	0.0	NaN	NaN	NaN
...
Sudan	0.0	NaN	NaN	NaN
Suriname	0.0	NaN	NaN	NaN
Swaziland	0.0	NaN	NaN	NaN
Sweden	50000.0	99999.0	129995.0	149991.0
Switzerland	39756.0	79530.0	119374.0	149352.0
Syria	0.0	0.0	0.0	0.0
Tajikistan	0.0	NaN	NaN	NaN
Tanzania	0.0	NaN	NaN	NaN
Thailand	0.0	30.0	200.0	7968.0
Togo	0.0	0.0	0.0	0.0
Tonga	0.0	NaN	NaN	NaN
Trinidad and Tobago	0.0	NaN	NaN	NaN
Tunisia	0.0	NaN	NaN	NaN
Turkey	0.0	NaN	NaN	4818.0
Turkmenistan	0.0	NaN	NaN	NaN
Uganda	0.0	NaN	NaN	NaN
Ukraine	0.0	NaN	NaN	401.0
United Arab Emirates	0.0	NaN	NaN	NaN
United Kingdom	49980.0	99956.0	149920.0	299807.0
United States	1988024.0	2975535.0	4453200.0	5924598.0
Uruguay	0.0	NaN	NaN	NaN
Uzbekistan	0.0	NaN	NaN	NaN
Vanuatu	0.0	NaN	NaN	NaN
Venezuela	0.0	NaN	2493.0	8776.0
Vietnam	0.0	NaN	NaN	NaN
Virgin Islands (U.S.)	0.0	NaN	NaN	NaN

West Bank and Gaza	0.0	NaN	NaN	NaN
Yemen, Rep.	0.0	NaN	NaN	NaN
Zambia	0.0	NaN	NaN	NaN
Zimbabwe	0.0	NaN	NaN	NaN
	1994	1995	1996	1997 \
Country				
Afghanistan	NaN	NaN	NaN	NaN
Albania	NaN	351.0	1002.0	1502.0
Algeria	100.0	500.0	500.0	3003.0
Angola	NaN	NaN	97.0	726.0
Antigua and Barbuda	NaN	1511.0	2014.0	2517.0
Argentina	15044.0	30072.0	50092.0	100123.0
Armenia	300.0	1700.0	3000.0	3500.0
Aruba	NaN	NaN	2300.0	NaN
Australia	400033.0	500006.0	600014.0	3032449.0
Austria	110000.0	150000.0	550001.0	760006.0
Azerbaijan	110.0	160.0	499.0	1993.0
Bahamas	NaN	2686.0	4956.0	3915.0
Bahrain	NaN	1936.0	4848.0	9762.0
Bangladesh	NaN	NaN	NaN	919.0
Barbados	NaN	20.0	1029.0	2077.0
Belarus	50.0	300.0	3001.0	5002.0
Belgium	70000.0	99958.0	299756.0	499407.0
Benin	NaN	NaN	99.0	1476.0
Bermuda	NaN	4199.0	9997.0	14995.0
Bhutan	NaN	NaN	NaN	NaN
Bolivia	NaN	4991.0	14976.0	34951.0
Bosnia and Herzegovina	NaN	NaN	500.0	2000.0
Botswana	NaN	1023.0	2557.0	5110.0
Brazil	60050.0	170164.0	740838.0	1311724.0
Brunei	NaN	2946.0	9817.0	14721.0
Bulgaria	1650.0	10000.0	60000.0	100000.0
Burkina Faso	NaN	NaN	106.0	2112.0
Burundi	0.0	0.0	49.0	493.0
Cambodia	NaN	NaN	NaN	686.0
Cameroon	NaN	NaN	NaN	990.0
...
Sudan	0.0	0.0	0.0	391.0
Suriname	NaN	500.0	1000.0	4493.0
Swaziland	NaN	10.0	496.0	892.0
Sweden	299985.0	450000.0	800125.0	2100857.0
Switzerland	189346.0	249305.0	321250.0	1070663.0
Syria	0.0	0.0	0.0	4844.0
Tajikistan	NaN	NaN	NaN	NaN
Tanzania	NaN	NaN	499.0	2496.0
Thailand	22831.0	44633.0	69581.0	219680.0
Togo	0.0	0.0	460.0	9183.0

Tonga	NaN	118.0	158.0	493.0
Trinidad and Tobago	NaN	1995.0	4988.0	14963.0
Tunisia	650.0	1000.0	2500.0	4000.0
Turkey	28880.0	48087.0	115292.0	287927.0
Turkmenistan	NaN	NaN	NaN	NaN
Uganda	NaN	596.0	993.0	2283.0
Ukraine	7009.0	22025.0	50053.0	100093.0
United Arab Emirates	NaN	2418.0	9287.0	86068.0
United Kingdom	599560.0	1099145.0	2398186.0	4307074.0
United States	12811919.0	24600557.0	44232691.0	58935678.0
Uruguay	1999.0	9997.0	59978.0	109954.0
Uzbekistan	NaN	350.0	1000.0	2503.0
Vanuatu	NaN	NaN	98.0	244.0
Venezuela	11968.0	26930.0	55857.0	89773.0
Vietnam	NaN	NaN	101.0	3031.0
Virgin Islands (U.S.)	1000.0	2999.0	4998.0	7497.0
West Bank and Gaza	NaN	NaN	NaN	NaN
Yemen, Rep.	NaN	NaN	98.0	2438.0
Zambia	589.0	783.0	831.0	879.0
Zimbabwe	199.0	898.0	1997.0	3998.0

	1998	1999	...	2001 \
Country			...	
Afghanistan	NaN	NaN	...	1118.0
Albania	2002.0	2502.0	...	10026.0
Algeria	6006.0	60055.0	...	200180.0
Angola	2424.0	9723.0	...	19570.0
Antigua and Barbuda	3021.0	4028.0	...	7045.0
Argentina	300195.0	1200184.0	...	3648448.0
Armenia	4000.0	30002.0	...	50006.0
Aruba	NaN	3985.0	...	15794.0
Australia	5772823.0	7727668.0	...	10212333.0
Austria	1230014.0	1840018.0	...	3149915.0
Azerbaijan	2990.0	7977.0	...	25030.0
Bahamas	6787.0	11067.0	...	35590.0
Bahrain	19671.0	29586.0	...	96625.0
Bangladesh	4599.0	46014.0	...	171275.0
Barbados	5240.0	6339.0	...	32013.0
Belarus	7503.0	50020.0	...	430483.0
Belgium	798809.0	1397655.0	...	3194941.0
Benin	2947.0	9805.0	...	24427.0
Bermuda	19993.0	24989.0	...	29984.0
Bhutan	NaN	764.0	...	5084.0
Bolivia	49939.0	79908.0	...	179745.0
Bosnia and Herzegovina	5000.0	7000.0	...	45000.0
Botswana	10214.0	19397.0	...	61222.0
Brazil	2503664.0	3505320.0	...	8009872.0
Brunei	19622.0	24521.0	...	43190.0

Bulgaria	150001.0	234602.0	...	605007.0
Burkina Faso	5279.0	7384.0	...	19951.0
Burundi	986.0	2463.0	...	6890.0
Cambodia	1957.0	3908.0	...	9738.0
Cameroon	1979.0	19778.0	...	44438.0
...
Sudan	1564.0	2935.0	...	49068.0
Suriname	7585.0	8711.0	...	14508.0
Swaziland	989.0	4934.0	...	13778.0
Sweden	2962762.0	3667832.0	...	4595249.0
Switzerland	1764011.0	2426492.0	...	3971601.0
Syria	9681.0	19356.0	...	58211.0
Tajikistan	NaN	2002.0	...	3195.0
Tanzania	2994.0	24941.0	...	59813.0
Thailand	672850.0	1514185.0	...	3550430.0
Togo	13752.0	27456.0	...	136739.0
Tonga	742.0	992.0	...	2782.0
Trinidad and Tobago	34916.0	74822.0	...	199533.0
Tunisia	10001.0	150033.0	...	410303.0
Turkey	431461.0	1436989.0	...	3349546.0
Turkmenistan	NaN	2003.0	...	7975.0
Uganda	14879.0	24784.0	...	59449.0
Ukraine	150117.0	200123.0	...	600158.0
United Arab Emirates	190358.0	432896.0	...	827413.0
United Kingdom	7995303.0	12493278.0	...	19786366.0
United States	83053031.0	100125498.0	...	140147752.0
Uruguay	229893.0	329828.0	...	369767.0
Uzbekistan	5008.0	7509.0	...	149646.0
Vanuatu	488.0	976.0	...	5365.0
Venezuela	324191.0	678327.0	...	1150224.0
Vietnam	10073.0	100422.0	...	1007834.0
Virgin Islands (U.S.)	9997.0	11996.0	...	19979.0
West Bank and Gaza	NaN	NaN	...	60339.0
Yemen, Rep.	3899.0	9745.0	...	16586.0
Zambia	2927.0	14625.0	...	24362.0
Zimbabwe	10012.0	20056.0	...	100581.0

	2002	2003	2004	2005 \
Country				
Afghanistan	1124.0	22569.0	28244.0	338045.0
Albania	12053.0	30194.0	75634.0	189887.0
Algeria	500441.0	700615.0	1501387.0	1921982.0
Angola	40260.0	57159.0	74173.0	188530.0
Antigua and Barbuda	10063.0	14079.0	20102.0	29133.0
Argentina	4097919.0	4527354.0	6148589.0	6854530.0
Armenia	60009.0	140027.0	150038.0	161054.0
Aruba	17784.0	20150.0	22781.0	25653.0
Australia	NaN	NaN	NaN	12854218.0

Austria	2955166.0	3473318.0	4443118.0	4774674.0
Azerbaijan	413989.0	NaN	NaN	689641.0
Bahamas	55044.0	62040.0	69243.0	79840.0
Bahrain	115898.0	139496.0	144151.0	154411.0
Bangladesh	187866.0	223717.0	275931.0	339713.0
Barbados	74828.0	106919.0	134434.0	151673.0
Belarus	891820.0	NaN	NaN	NaN
Belgium	4750690.0	5148463.0	5578276.0	6228738.0
Benin	48770.0	68162.0	87484.0	97027.0
Bermuda	32982.0	35979.0	38976.0	41971.0
Bhutan	10156.0	15217.0	20280.0	25364.0
Bolivia	269533.0	309315.0	398835.0	478149.0
Bosnia and Herzegovina	99995.0	149986.0	584937.0	806363.0
Botswana	61221.0	61221.0	61215.0	61195.0
Brazil	16403934.0	23989345.0	35071405.0	39099569.0
Brunei	52365.0	68342.0	105771.0	132418.0
Bulgaria	716654.0	943950.0	1412178.0	1545501.0
Burkina Faso	26160.0	50025.0	55200.0	66721.0
Burundi	7869.0	13765.0	24572.0	39313.0
Cambodia	29157.0	33937.0	39639.0	42386.0
Cameroon	59212.0	98623.0	167546.0	246216.0
...
Sudan	157404.0	197315.0	296864.0	496277.0
Suriname	19981.0	22976.0	29969.0	31970.0
Swaziland	19665.0	26533.0	35370.0	40848.0
Sweden	6282022.0	7070576.0	7531615.0	7659593.0
Switzerland	4454555.0	4756908.0	4990915.0	5197920.0
Syria	355051.0	594309.0	778094.0	1044003.0
Tajikistan	3486.0	4092.0	4952.0	19275.0
Tanzania	79722.0	249043.0	331609.0	1669734.0
Thailand	4868421.0	6078800.0	7053467.0	10022117.0
Togo	181911.0	190546.0	200009.0	216322.0
Tonga	2880.0	2976.0	3964.0	4953.0
Trinidad and Tobago	286329.0	339208.0	354186.0	381156.0
Tunisia	506073.0	631052.0	837019.0	957023.0
Turkey	7447774.0	8179652.0	9802998.0	10534937.0
Turkmenistan	13898.0	19762.0	35416.0	47348.0
Uganda	99075.0	123844.0	198148.0	495330.0
Ukraine	900071.0	1499871.0	1649659.0	1759537.0
United Arab Emirates	921614.0	1002533.0	1102215.0	1627740.0
United Kingdom	33515216.0	38635538.0	39295002.0	42141909.0
United States	169576672.0	179718040.0	190430149.0	201742976.0
Uruguay	379741.0	529611.0	566728.0	667436.0
Uzbekistan	273461.0	487511.0	666800.0	867554.0
Vanuatu	6829.0	7805.0	9757.0	10732.0
Venezuela	1241034.0	1930220.0	2201971.0	3347147.0
Vietnam	1493286.0	3076281.0	6286874.0	10594671.0
Virgin Islands (U.S.)	29951.0	29931.0	29915.0	29905.0

West Bank and Gaza	104051.0	141243.0	153305.0	569071.0
Yemen, Rep.	97699.0	117438.0	176475.0	216521.0
Zambia	51088.0	107248.0	225365.0	326878.0
Zimbabwe	503600.0	806572.0	826930.0	1007663.0

	2006	2007	2008	2009 \
Country				
Afghanistan	598865.0	553771.0	549056.0	1085510.0
Albania	303350.0	476594.0	759081.0	1315402.0
Algeria	2462986.0	3204578.0	3504773.0	3924904.0
Angola	324498.0	560812.0	829746.0	1113307.0
Antigua and Barbuda	53213.0	60206.0	65187.0	65149.0
Argentina	8166600.0	10214688.0	11164731.0	13621240.0
Armenia	172887.0	185104.0	191211.0	472002.0
Aruba	28759.0	32206.0	36090.0	40193.0
Australia	13691235.0	14667832.0	15418953.0	16262458.0
Austria	5261760.0	5764496.0	6078674.0	6147500.0
Azerbaijan	1043526.0	1282575.0	1527567.0	2484249.0
Bahamas	84257.0	88792.0	105237.0	114636.0
Bahrain	229174.0	304659.0	546701.0	619876.0
Bangladesh	1423535.0	2591222.0	3636958.0	4557934.0
Barbados	170769.0	175737.0	181000.0	187379.0
Belarus	1583845.0	1916461.0	2226559.0	2643159.0
Belgium	6722851.0	7300198.0	6997430.0	7462657.0
Benin	121055.0	145215.0	154586.0	192680.0
Bermuda	44965.0	47958.0	53213.0	53948.0
Bhutan	30484.0	40764.0	45939.0	51170.0
Bolivia	577075.0	993596.0	1041680.0	1397602.0
Bosnia and Herzegovina	950025.0	1055106.0	1308125.0	1421924.0
Botswana	81548.0	101774.0	122176.0	121867.0
Brazil	52963579.0	58609644.0	64799077.0	75791320.0
Brunei	156223.0	168673.0	176960.0	192000.0
Bulgaria	2083091.0	2570191.0	3011477.0	3394203.0
Burkina Faso	92516.0	112958.0	142740.0	180625.0
Burundi	49151.0	53954.0	64341.0	155246.0
Cambodia	63303.0	66982.0	70495.0	74083.0
Cameroon	364127.0	537656.0	637798.0	736321.0
...
Sudan	3185731.0	3496357.0	4207779.0	NaN
Suriname	47965.0	71980.0	108473.0	163028.0
Swaziland	41313.0	46453.0	78799.0	88794.0
Sweden	7978004.0	7513243.0	8313185.0	8473110.0
Switzerland	5653541.0	5806685.0	5997978.0	6196045.0
Syria	1481976.0	NaN	2757196.0	3469298.0
Tajikistan	246136.0	475361.0	587506.0	683087.0
Tanzania	2315569.0	2956909.0	3804090.0	4352474.0
Thailand	11545108.0	13579629.0	12424773.0	13809931.0
Togo	237786.0	254376.0	277288.0	300995.0

Tonga	5947.0	7342.0	8347.0	10352.0
Trinidad and Tobago	396175.0	428192.0	463202.0	592003.0
Tunisia	1300957.0	1732293.0	2821089.0	3531386.0
Turkey	12597189.0	20038925.0	24376486.0	26152021.0
Turkmenistan	63364.0	68333.0	86072.0	97104.0
Uganda	742880.0	1114070.0	2475812.0	3165582.0
Ukraine	2099484.0	3031481.0	10315912.0	15154526.0
United Arab Emirates	2424619.0	3297380.0	4468769.0	5204111.0
United Kingdom	41662350.0	45726882.0	48029802.0	51516674.0
United States	206493364.0	226713423.0	225691907.0	218457578.0
Uruguay	978271.0	1134080.0	1314965.0	1403389.0
Uzbekistan	1674617.0	1985297.0	2434558.0	4627564.0
Vanuatu	12682.0	15122.0	16577.0	17534.0
Venezuela	4130269.0	5747711.0	7261277.0	8898213.0
Vietnam	14506873.0	17643674.0	20559697.0	23072261.0
Virgin Islands (U.S.)	29902.0	29906.0	29914.0	29921.0
West Bank and Gaza	669699.0	789496.0	932070.0	1266856.0
Yemen, Rep.	265638.0	1099544.0	1558972.0	2323490.0
Zambia	488794.0	587097.0	687068.0	802868.0
Zimbabwe	1226884.0	1354215.0	1419476.0	1417045.0

2010

Country	
Afghanistan	1256470.0
Albania	1441928.0
Algeria	4433526.0
Angola	1908191.0
Antigua and Barbuda	70968.0
Argentina	14548455.0
Armenia	1360512.0
Aruba	45145.0
Australia	16923972.0
Austria	6102179.0
Azerbaijan	4226380.0
Bahamas	147437.0
Bahrain	694009.0
Bangladesh	5501609.0
Barbados	191878.0
Belarus	3041748.0
Belgium	8034050.0
Benin	277002.0
Bermuda	54687.0
Bhutan	98728.0
Bolivia	1985970.0
Bosnia and Herzegovina	1955277.0
Botswana	120417.0
Brazil	79245740.0
Brunei	199460.0

Bulgaria	3464630.0
Burkina Faso	230562.0
Burundi	176040.0
Cambodia	178142.0
Cameroon	783956.0
...	...
Sudan	NaN
Suriname	165733.0
Swaziland	95122.0
Sweden	8441718.0
Switzerland	6430363.0
Syria	4224995.0
Tajikistan	794483.0
Tanzania	4932535.0
Thailand	14653914.0
Togo	324296.0
Tonga	12487.0
Trinidad and Tobago	650611.0
Tunisia	3856984.0
Turkey	28969976.0
Turkmenistan	110924.0
Uganda	4178085.0
Ukraine	20451748.0
United Arab Emirates	5859118.0
United Kingdom	52730234.0
United States	229684122.0
Uruguay	1606911.0
Uzbekistan	5488940.0
Vanuatu	19172.0
Venezuela	10337115.0
Vietnam	24211031.0
Virgin Islands (U.S.)	NaN
West Bank and Gaza	1512273.0
Yemen, Rep.	2970485.0
Zambia	1308857.0
Zimbabwe	1445717.0

[196 rows x 21 columns]

In [90]: internet_users.min(axis=0)

Out[90]:

1990	0.0
1991	0.0
1992	0.0
1993	0.0
1994	0.0
1995	0.0
1996	0.0

```

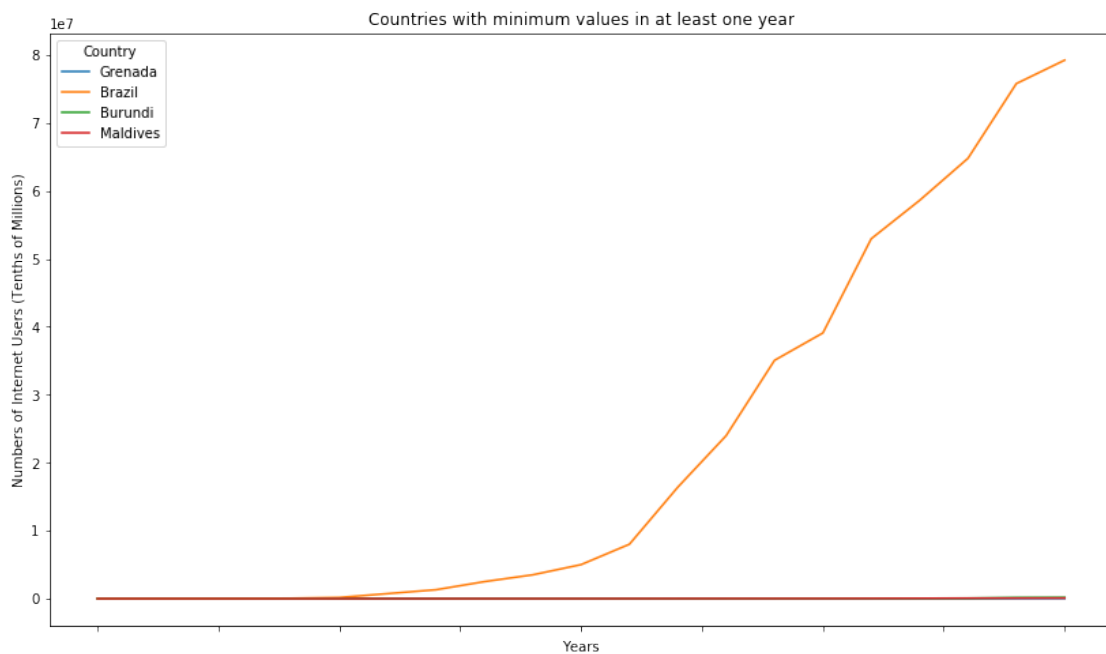
1997      0.0
1998      0.0
1999      0.0
2000      0.0
2001      0.0
2002      0.0
2003      0.0
2004      0.0
2005      0.0
2006      0.0
2007      0.0
2008      0.0
2009      0.0
2010    8959.0
dtype: float64

```

```

In [91]: list_countries_min_values_to_check = ['Grenada', 'Brazil', 'Burundi', 'Maldives']
min_values_internet_users = internet_users.loc[list_countries_min_values_to_check].T
min_values_internet_users.set(xlabel='Years', ylabel='Numbers of Internet Users (Tenths of Millions)')
plt.show()

```



We can also see the 5 smallest values in 2010 and see how they behaved over the years:

```

In [92]: internet_users['2010'].sort_values().head(5)

```

```

Out[92]: Country
Kiribati      8959.0

```

```

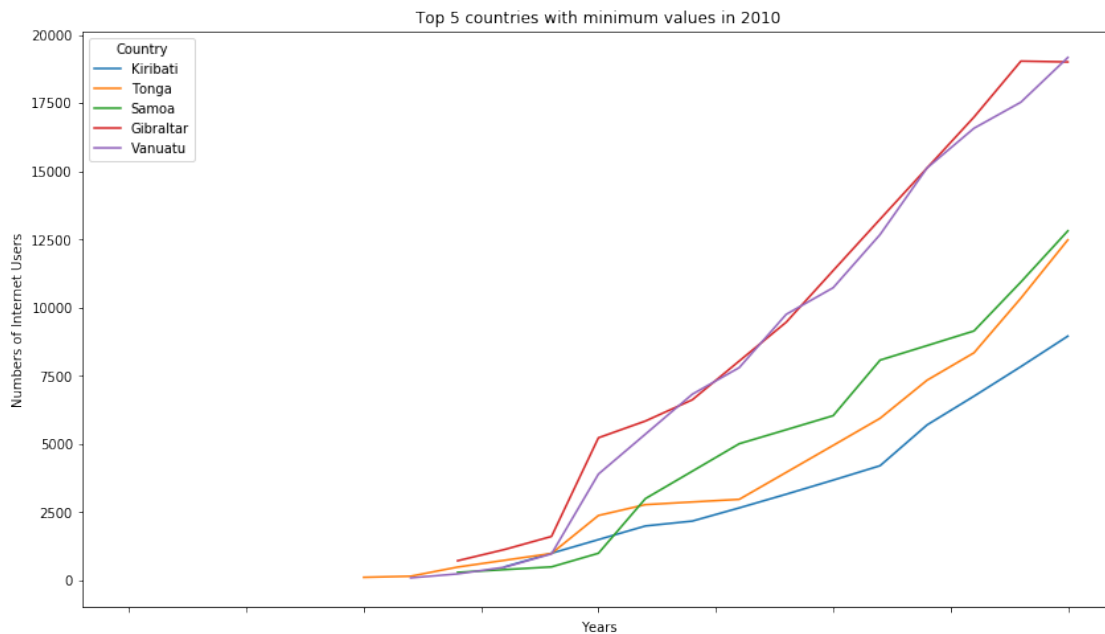
Tonga          12487.0
Samoa          12816.0
Gibraltar      19009.0
Vanuatu        19172.0
Name: 2010, dtype: float64

```

```

In [93]: list_countries_min_values_to_check = ['Kiribati', 'Tonga', 'Samoa', 'Gibraltar', 'Vanuatu']
min_values_internet_users = internet_users.loc[list_countries_min_values_to_check].tr
min_values_internet_users.set(xlabel='Years', ylabel='Numbers of Internet Users')
plt.show()

```



Also, we can check the top 5 in absolute numbers:

```

In [94]: internet_users['2010'].sort_values(ascending=False).head(5)

```

```

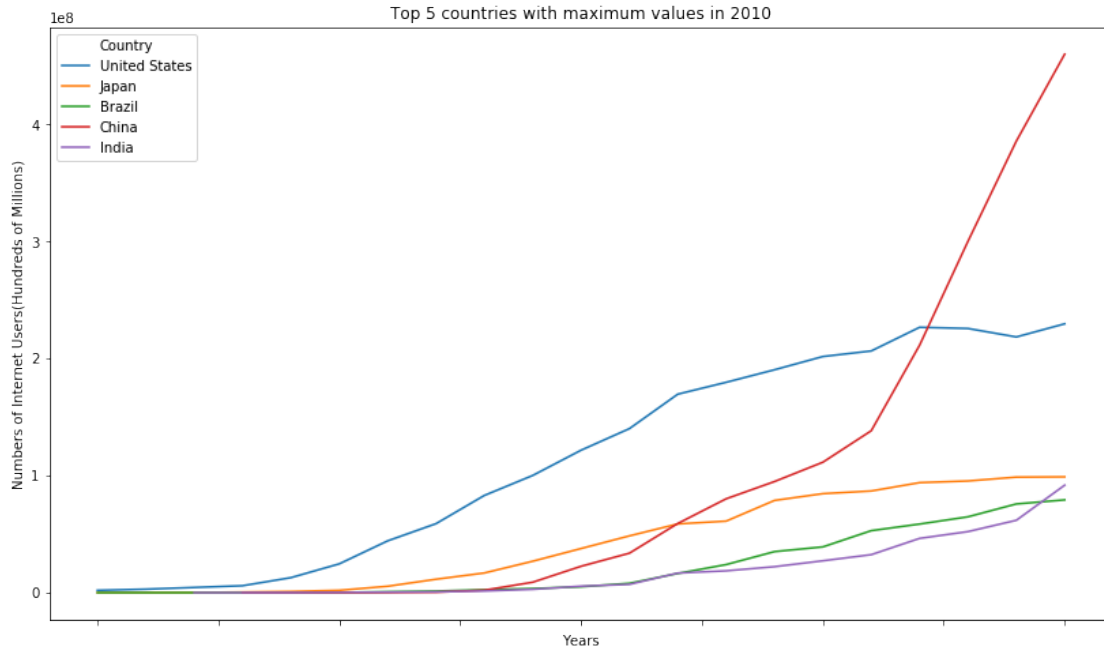
Out[94]: Country
China          460077957.0
United States  229684122.0
Japan          98951089.0
India          91846075.0
Brazil         79245740.0
Name: 2010, dtype: float64

```

```

In [95]: list_countries_max_values_to_check = ['United States', 'Japan', 'Brazil', 'China', 'India']
max_values_internet_users = internet_users.loc[list_countries_max_values_to_check].tr
max_values_internet_users.set(xlabel='Years', ylabel='Numbers of Internet Users(Hundr
plt.show()

```



1.2.3 Which country had the largest internet boom?

To answer this question we should calculate something like the derivate of the graph for each country (Newton and Leibniz say hello) and then find the largest value overall. To do this, unfortunately, we will have to deal with some unpleasant NaN mess which will prevent us from dealing with this "pandas style": we will have to use loops and ifs.

```
In [96]: #first will remove all values that are None in 2010
internet_users = internet_users[np.isfinite(internet_users['2010'])]
year_list = []
year_diff_list = []
for year in range(1990,2011):
    year_list.append(str(year))
for year in range(1990,2010):
    year_diff_list.append(str(year)+'-'+str(year+1))
internet_users_variation = pd.DataFrame(columns=year_diff_list)
for index,row in internet_users.iterrows():
    next_row = []
    i = 0
    while i < len(year_list)-1:
        if np.isnan(row[year_list[i+1]]) and not np.isnan(row[year_list[i]]) and i !=
            k=2
            stop = False
            #print('Entered in '+index+' with an i of ' + str(i))
            while k+i < len(year_list) and not stop:
                #print('First loop in '+index+' with an k of '+str(k))
```

```

if not np.isnan(row[year_list[i+k]]):
    #print('Found non None in '+index+' in column number '+str(i+k))
    stop = True
    diff = (row[year_list[i+k]] - row[year_list[i]])/k
    j=1
    while j<=k:
        #print('Second loop in '+index+' with an j of '+str(j)+' appended')
        next_row.append(j*diff)
        j+=1
    else:
        k += 1
i += k
#print('Value of i at the end of the first loop is '+ str(i))

else:
    #print('Adding simple difference with i of '+str(i))
    next_row.append(row[year_list[i+1]]-row[year_list[i]])
    i += 1
#print(len(next_row))
#print(index)
#print(len(year_diff_list))
#print(len(next_row))
#print('Appending row to '+index)
internet_users_variation.loc[index] = next_row
internet_users_variation.head(10)

```

Out[96]:

	1990-1991	1991-1992	1992-1993	1993-1994	\
Afghanistan	101.636364	203.272727	304.909091	406.545455	
Albania	70.200000	140.400000	210.600000	280.800000	
Algeria	25.000000	50.000000	75.000000	100.000000	
Angola	16.166667	32.333333	48.500000	64.666667	
Antigua and Barbuda	302.200000	604.400000	906.600000	1208.800000	
Argentina	502.000000	1004.000000	9030.000000	5010.000000	
Armenia	75.000000	150.000000	225.000000	300.000000	
Aruba	383.333333	766.666667	1150.000000	1533.333333	
Australia	90014.000000	120012.000000	39992.000000	49986.000000	
Austria	10000.000000	30000.000000	10000.000000	50000.000000	

	1994-1995	1995-1996	1996-1997	1997-1998	\
Afghanistan	508.181818	609.818182	7.114545e+02	8.130909e+02	
Albania	351.000000	651.000000	5.000000e+02	5.000000e+02	
Algeria	400.000000	0.000000	2.503000e+03	3.003000e+03	
Angola	80.833333	97.000000	6.290000e+02	1.698000e+03	
Antigua and Barbuda	1511.000000	503.000000	5.030000e+02	5.040000e+02	
Argentina	15028.000000	20020.000000	5.003100e+04	2.000720e+05	
Armenia	1400.000000	1300.000000	5.000000e+02	5.000000e+02	
Aruba	1916.666667	2300.000000	5.616667e+02	1.123333e+03	
Australia	99973.000000	100008.000000	2.432435e+06	2.740374e+06	

Austria	40000.000000	400001.000000	2.100050e+05	4.700080e+05	
	1998-1999	1999-2000	2000-2001	2001-2002	\
Afghanistan	9.147273e+02	1.016364e+03	1118.0	6.00	
Albania	5.000000e+02	1.003000e+03	6521.0	2027.00	
Algeria	5.404900e+04	9.008200e+04	50043.0	300261.00	
Angola	7.299000e+03	4.906000e+03	4941.0	20690.00	
Antigua and Barbuda	1.007000e+03	1.006000e+03	2011.0	3018.00	
Argentina	8.999890e+05	1.399252e+06	1049012.0	449471.00	
Armenia	2.600200e+04	1.000200e+04	10002.0	10003.00	
Aruba	1.685000e+03	9.955000e+03	1854.0	1990.00	
Australia	1.954845e+06	1.232838e+06	1251827.0	660471.25	
Austria	6.100040e+05	8.599820e+05	449915.0	-194749.00	
	2002-2003	2003-2004	2004-2005	2005-2006	2006-2007 \
Afghanistan	21445.0	5675.00	309801.0	260820.0	-45094.0
Albania	18141.0	45440.00	114253.0	113463.0	173244.0
Algeria	200174.0	800772.00	420595.0	541004.0	741592.0
Angola	16899.0	17014.00	114357.0	135968.0	236314.0
Antigua and Barbuda	4016.0	6023.00	9031.0	24080.0	6993.0
Argentina	429435.0	1621235.00	705941.0	1312070.0	2048088.0
Armenia	80018.0	10011.00	11016.0	11833.0	12217.0
Aruba	2366.0	2631.00	2872.0	3106.0	3447.0
Australia	1320942.5	1981413.75	2641885.0	837017.0	976597.0
Austria	518152.0	969800.00	331556.0	487086.0	502736.0
	2007-2008	2008-2009	2009-2010		
Afghanistan	-4715.0	536454.0	170960.0		
Albania	282487.0	556321.0	126526.0		
Algeria	300195.0	420131.0	508622.0		
Angola	268934.0	283561.0	794884.0		
Antigua and Barbuda	4981.0	-38.0	5819.0		
Argentina	950043.0	2456509.0	927215.0		
Armenia	6107.0	280791.0	888510.0		
Aruba	3884.0	4103.0	4952.0		
Australia	751121.0	843505.0	661514.0		
Austria	314178.0	68826.0	-45321.0		

To answer our question we simply perform a max_value search in the whole dataset.

```
In [97]: internet_users_variation.loc['China']
```

```
Out[97]: 1990-1991    6.683333e+02
         1991-1992    1.336667e+03
         1992-1993    2.005000e+03
         1993-1994    1.203000e+04
         1994-1995    4.611500e+04
         1995-1996    1.002490e+05
```

1996-1997	2.406010e+05
1997-1998	1.704102e+06
1998-1999	6.814362e+06
1999-2000	1.361895e+07
2000-2001	1.119377e+07
2001-2002	2.536553e+07
2002-2003	2.109288e+07
2003-2004	1.474971e+07
2004-2005	1.650925e+07
2005-2006	2.688585e+07
2006-2007	7.310170e+07
2007-2008	8.875316e+07
2008-2009	8.559838e+07
2009-2010	7.428931e+07

Name: China, dtype: float64

In [98]: internet_users_variation.max().idxmax()

Out [98]: '2007-2008'

In [99]: internet_users_variation.idxmax(axis=0)

Out [99]:

1990-1991	United States
1991-1992	United States
1992-1993	United States
1993-1994	United States
1994-1995	United States
1995-1996	United States
1996-1997	United States
1997-1998	United States
1998-1999	United States
1999-2000	United States
2000-2001	United States
2001-2002	United States
2002-2003	China
2003-2004	Japan
2004-2005	China
2005-2006	China
2006-2007	China
2007-2008	China
2008-2009	China
2009-2010	China

dtype: object

So China from 2007-2008 had the largest internet user growth of our dataset!

1.2.4 Did internet users got smaller in any country?

In order to answer this question, we list all minimum values of our table.


```

In [100]: internet_users_variation.min()
Out[100]: 1990-1991      0.0
          1991-1992      0.0
          1992-1993      0.0
          1993-1994      0.0
          1994-1995     -6.0
          1995-1996      0.0
          1996-1997   -1041.0
          1997-1998     -2.0
          1998-1999     -7.0
          1999-2000     -6.0
          2000-2001  -190011.0
          2001-2002 -194749.0
          2002-2003    -12.0
          2003-2004   -8744.0
          2004-2005 -399920.0
          2005-2006 -479559.0
          2006-2007 -464761.0
          2007-2008 -1154856.0
          2008-2009 -7234329.0
          2009-2010  -45321.0
          dtype: float64

In [101]: internet_users_variation.idxmin(axis=0)
Out[101]: 1990-1991      Botswana
          1991-1992      Burundi
          1992-1993      Burundi
          1993-1994      Burundi
          1994-1995    Greenland
          1995-1996      Algeria
          1996-1997      Bahamas
          1997-1998      Liberia
          1998-1999      Dominica
          1999-2000      Rwanda
          2000-2001      Israel
          2001-2002      Austria
          2002-2003      Liberia
          2003-2004        Oman
          2004-2005    South Africa
          2005-2006    United Kingdom
          2006-2007        Sweden
          2007-2008        Thailand
          2008-2009    United States
          2009-2010        Austria
          dtype: object

```

So, answering the question, yes, there are many occasions in which the number of internet users got smaller in a country, and in fact it happened even in developed countries such as

USA,UK, Sweden and Austria! We do not have enough info on the dataset however to know if this covers mobile internet usage or just cable/home internet plans.

Conclusions

We observe that overall internet users are growing worldwide, but there are some cases in which the number has decreased. One huge example is in USA during 2008 and 2009, probably because of the economic crisis, but we cannot asser it for sure because we could not determine if this database consolidates all internet users or only cable/home internet plans.

One noticeable country is China: if plotted against USA it is incredible to see the huge explosion of internet users and how it seems to never stop growing.