# Country\_Energy\_GDP

February 12, 2020

You are currently looking at **version 1.5** of this notebook. To download notebooks and datafiles, as well as get help on Jupyter notebooks in the Coursera platform, visit the Jupyter Notebook FAQ course resource.

# 1 Pandas use assignment: data cleaning, merging, and profiling

Check Python and Import search directories, Install and Import needed modules

```
[]: import sys
[]: print(sys.executable)
[]: print(sys.path)
[]: !pip3 install xlrd
     !pip3 install pandas
     !pip3 install numpy
     !pip3 install re
     !pip3 install matplotlib
[]: !pip3 install --upgrade pip
[2]: import pandas as pd
     import numpy as np
     import matplotlib
     from matplotlib import pyplot as plt
     import re
     from pandas import ExcelWriter
     from pandas import ExcelFile
     from functools import reduce
     import operator
```

# 1.0.1 Data Preparation

1) Load the energy data from the file Energy Indicators.xls, which is a list of indicators of energy supply and renewable electricity production from the United Nations for the year

2013, and should be put into a DataFrame with the variable name of energy.

Remaining column labels should be:

```
['Country', 'Energy Supply', 'Energy Supply per Capita', '% Renewable']
```

- 2) Convert Energy Supply to gigajoules (there are 1,000,000 gigajoules in a petajoule). For all countries which have missing data (e.g. data with "...") make sure this is reflected as np.NaN values.
- 3) Rename the following list of countries (for use in later questions):

```
"Republic of Korea": "South Korea", "United States of America": "United States", "United Kingdom of Great Britain and Northern Ireland": "United Kingdom", "China, Hong Kong Special Administrative Region": "Hong Kong"
```

4) Remove numbers and/or parenthesis from country names that contain them, e.g.

```
'Bolivia (Plurinational State of)' should be 'Bolivia',
```

'Switzerland17' should be 'Switzerland'.

- 5) Next, load the GDP data from the file world\_bank.csv, which is a csv containing countries' GDP from 1960 to 2015 from World Bank. Call this DataFrame GDP.
- 6) Rename the following list of countries:

```
"Korea, Rep.": "South Korea", "Iran, Islamic Rep.": "Iran", "Hong Kong SAR, China": "Hong Kong"
```

- 7) Finally, load the Sciamgo Journal and Country Rank data for Energy Engineering and Power Technology from the file scimagojr-3.xlsx, which ranks countries based on their journal contributions in the aforementioned area. Call this DataFrame ScimEn.
- 8) Join the three datasets: GDP, Energy, and ScimEn into a new dataset (using the intersection of country names). Use only the last 10 years (2006-2015) of GDP data and only the top 15 countries by Scimagojr 'Rank' (Rank 1 through 15).

The index of this DataFrame should be the name of the country, and the columns should be ['Rank', 'Documents', 'Citable documents', 'Citations', 'Self-citations', 'Citations per document', 'H index', 'Energy Supply', 'Energy Supply per Capita', '% Renewable', '2006', '2007', '2008', '2009', '2010', '2011', '2012', '2013', '2014', '2015'].

This function should return a DataFrame with 20 columns and 15 entries.

# 1.0.2 Read-in and clean energy dataset

```
[6]: def energy():
    energy = pd.ExcelFile('Energy Indicators.xls')

# parsing first (index: 0) sheet
    total_rows = energy.book.sheet_by_index(0).nrows
    #print(f"total rows:{total_rows}")
    ##skiprows indexing starts at 1
    header = 17
```

```
## 244 instead of the 245 rows of interest (with header/skiprows exception),
\rightarrow since indexing starts at 0
   nrows = 244 - header
   footer = total rows - 244
   #print(f"header:{header}")
   #print(f"nrows:{nrows}")
    #print(f"footer:{footer}")
# (-1) in order to allow for the column label row
   skipfooter = total_rows - nrows - header - 1
    #print(f"skipfooter:{skipfooter}")
   energy = energy.parse(0, skiprows = header, skipfooter = skipfooter)
   energy = energy.iloc[:, 2:].copy()
#header indexing starts at 0
   #print(energy.head())
   column_names = ['Country', 'Energy Supply', 'Energy Supply per Capita', '% |
→Renewable'
   energy.columns = column_names
   energy = energy.replace("...",np.nan)
   energy['Energy Supply'] = energy['Energy Supply']*1000000
   energy['Country'] = energy['Country'].map({'Republic of Korea': 'Southu
→Korea', "United States of America20": "United States", "United Kingdom of
→Great Britain and Northern Ireland19": "United Kingdom", "China, Hong Kong
 →Special Administrative Region3": "Hong Kong"}).fillna(energy['Country'])
    #print('Hong Kong' in energy['Country'].unique())
    #print(energy['Country'])
   #preceeding mapper replaces following clunky code:
   #i = 0
   #for country in energy['Country']:
     # if country == "Republic of Korea":
             energy['Country'][i] = "South Korea"
       # elif country == "United States of America20":
            energy['Country'][i] = "United States"
        #elif country == "United Kingdom of Great Britain and Northern_
\rightarrow Ireland19":
        # energy['Country'][i] = "United Kingdom"
        #elif country == "China, Hong Kong Special Administrative Region3":
         # energy['Country'][i] = "Hong Kong"
        #i+= 1
#print(energy["Country"])
   energy['Country'] = energy['Country'].apply(lambda x: ''.join([e for e in x_
→if not e.isdigit()])).fillna(energy['Country'])
    #preceeding lambda replaces following code:
```

```
#i = 0
#for country in energy['Country']:
# L = ''.join([c for c in country if not c.isdigit()])
##L = country.str.findall('(\d+)', expand=False).astype(int).tostring()
# energy['Country'][i] = energy['Country'][i].replace(country, L)
# i+=1

## replace '(' and ')' with empty string (i.e., eliminate these special_
chars) from country names, if present
energy['Country'] = energy['Country'].apply(lambda x: re.sub(r"\s\(.*\)",__
"", x))
#print('Bolivia' in energy['Country'].unique())
return energy
energy()

Country Energy Supply Energy Supply per Capita \
```

```
[6]:
     0
                         Afghanistan
                                       3.210000e+08
                                                                           10.0
     1
                             Albania
                                       1.020000e+08
                                                                           35.0
     2
                             Algeria
                                       1.959000e+09
                                                                           51.0
     3
                      American Samoa
                                                 NaN
                                                                            NaN
     4
                             Andorra
                                       9.000000e+06
                                                                          121.0
     222
                            Viet Nam
                                       2.554000e+09
                                                                           28.0
                                                                           26.0
     223
         Wallis and Futuna Islands
                                       0.000000e+00
     224
                               Yemen
                                       3.440000e+08
                                                                           13.0
     225
                              Zambia
                                       4.000000e+08
                                                                           26.0
     226
                            Zimbabwe
                                       4.800000e+08
                                                                           32.0
          % Renewable
            78,669280
     0
     1
           100.000000
     2
             0.551010
     3
             0.641026
            88.695650
     4
     . .
            45.321520
     222
     223
             0.000000
     224
             0.00000
     225
            99.714670
     226
            52.536120
```

[227 rows x 4 columns]

### 1.0.3 Read-in and clean GDP dataset

```
[7]: def gdp():
         GDP = pd.read_csv('world_bank.csv', header = 4)
         #GDP['Country Name'] = GDP['Country Name'].map({"Korea, Rep.": 'South⊔
      →Korea', "Iran, Islamic Rep.": 'Iran', "Hong Kong SAR, China": 'Hong Kong'}).
      → fillna(GDP['Country Name'])
         countries_dict = {"Korea, Rep.": 'South Korea', "Iran, Islamic Rep.": __
      →'Iran', "Hong Kong SAR, China": 'Hong Kong'}
         #no need for fillna call at end of method chain when using conditional
      \rightarrow lambda inside map call.
         GDP['Country Name'] = GDP['Country Name'].map(lambda x: x.replace(x, __
      →countries_dict[x] if x in countries_dict else x))
         #print('Iran' in GDP['Country Name'].unique())
         #preceeding lambda replaces the following otherwise lengthier code:
         #for i, row in GDP.iterrows():
          # if row[0] == "Korea, Rep.":
                  GDP['Country Name'][i] = 'South Korea'
             ##print (i, row)
             #if row[0] == "Iran, Islamic Rep.":
              # GDP['Country Name'][i] = 'Iran'
             #if row[0] == "Hong Kong SAR, China":
                GDP["Country Name"][i] = 'Hong Kong'
         GDP.rename(columns = {'Country Name': 'Country'}, inplace = True)
         GDP = GDP.loc[:, "Country":"Country Code"].join(GDP.loc[:,'2006':'2015'])
         GDP.drop(["Country Code"], axis = 1, inplace = True)
         return GDP
     gdp()
```

[7]:		Country	2006	2007	2008	2009	\
	0	Aruba	NaN	NaN	NaN	NaN	
	1	Andorra	4.018196e+09	4.021331e+09	3.675728e+09	3.535389e+09	
	2	Afghanistan	1.030523e+10	1.172119e+10	1.214448e+10	1.469733e+10	
	3	Angola	5.581103e+10	6.842044e+10	7.787420e+10	7.975320e+10	
	4	Albania	9.771760e+09	1.034829e+10	1.112752e+10	1.150029e+10	
		•••	•••	•••	***	•••	
	259	Yemen, Rep.	2.672565e+10	2.761787e+10	2.872656e+10	2.991436e+10	
	260	South Africa	3.402852e+11	3.585261e+11	3.699668e+11	3.642764e+11	
	261	Congo, Dem. Rep.	1.650894e+10	1.754232e+10	1.863448e+10	1.916651e+10	
	262	Zambia	1.440569e+10	1.560892e+10	1.682234e+10	1.837342e+10	
	263	Zimbabwe	1.006276e+10	9.695130e+09	7.982103e+09	8.459783e+09	
		2010	2011	2012	2013	2014 \	
	0	2.467704e+09	NaN	NaN	NaN	NaN	

```
1
    3.346317e+09 3.185605e+09 3.129538e+09 3.127550e+09
                                                                   NaN
2
    1.593680e+10 1.691113e+10 1.935220e+10 1.973134e+10 1.999032e+10
3
    8.247091e+10 8.570262e+10 9.012096e+10 9.626143e+10 1.008863e+11
    1.192695e+10 1.223109e+10 1.240477e+10 1.254247e+10 1.279331e+10
259 3.090675e+10 2.624342e+10 2.689160e+10 2.800914e+10
                                                                   NaN
260 3.753494e+11 3.874074e+11 3.960071e+11 4.047682e+11 4.110369e+11
261 2.052329e+10 2.193213e+10 2.350200e+10 2.550050e+10 2.778776e+10
262 2.026555e+10 2.140358e+10 2.302438e+10 2.420595e+10 2.542227e+10
263 9.422161e+09 1.054391e+10 1.165789e+10 1.218064e+10 1.264939e+10
            2015
0
             NaN
1
             NaN
2
    2.029415e+10
3
    1.039106e+11
4
    1.312082e+10
. .
259
             NaN
260 4.163117e+11
261 2.970961e+10
262 2.624127e+10
263 1.278517e+10
[264 rows x 11 columns]
```

### 1.0.4 Read-in SciMen dataset then merge all datasets

[8]:		Rank	Documents	Citable documents	Citations \	<b>\</b>
	Country					
	China	1	127050	126767	597237	
	United States	2	96661	94747	792274	
	Japan	3	30504	30287	223024	
	United Kingdom	4	20944	20357	206091	
	Russian Federation	5	18534	18301	34266	
	Canada	6	17899	17620	215003	
	Germany	7	17027	16831	140566	
	India	8	15005	14841	128763	
	France	9	13153	12973	130632	
	South Korea	10	11983	11923	114675	
	Italy	11	10964	10794	111850	
	Spain	12	9428	9330	123336	
	Iran	13	8896	8819	57470	
	Australia	14	8831	8725	90765	
	Brazil	15	8668	8596	60702	
		0.76				,
	Q	Seli-	citations	Citations per docume	ent H index	\
	Country		444600	4	70 400	
	China		411683	<del>-</del> ·	.70 138	
	United States		265436		.20 230	
	Japan		61554		.31 134	
	United Kingdom		37874		.84 139	
	Russian Federation		12422		.85 57	
	Canada		40930		.01 149	
	Germany		27426		. 26 126	
	India		37209		.58 115	
	France		28601		.93 114	
	South Korea		22595		.57 104	
	Italy		26661	10.	.20 106	

Spain	23964	4	13.08	115	
Iran	1912	5	6.46	72	
Australia	1560	6	10.28	107	
Brazil	14396	5	7.00	86	
	Energy Supply	Energy Suppl	y per Capita	% Renewable '	\
Country					
China	1.271910e+11		93.0	19.754910	
United States	9.083800e+10		286.0	11.570980	
Japan	1.898400e+10		149.0	10.232820	
United Kingdom	7.920000e+09		124.0	10.600470	
Russian Federation	3.070900e+10		214.0	17.288680	
Canada	1.043100e+10		296.0	61.945430	
Germany	1.326100e+10		165.0	17.901530	
India	3.319500e+10		26.0	14.969080	
France	1.059700e+10		166.0	17.020280	
South Korea	1.100700e+10		221.0	2.279353	
Italy	6.530000e+09		109.0	33.667230	
Spain	4.923000e+09		106.0	37.968590	
Iran	9.172000e+09		119.0	5.707721	
Australia	5.386000e+09		231.0	11.810810	
Brazil	1.214900e+10		59.0	69.648030	
	2006	2007	2008	2009	\
Country					
China	3.992331e+12	4.559041e+12	4.997775e+12	5.459247e+12	
United States	1.479230e+13	1.505540e+13	1.501149e+13	1.459484e+13	
Japan	5.496542e+12	5.617036e+12	5.558527e+12	5.251308e+12	
United Kingdom	2.419631e+12	2.482203e+12	2.470614e+12	2.367048e+12	
Russian Federation	1.385793e+12	1.504071e+12	1.583004e+12	1.459199e+12	
Canada	1.564469e+12	1.596740e+12	1.612713e+12	1.565145e+12	
Germany	3.332891e+12	3.441561e+12	3.478809e+12	3.283340e+12	
India	1.265894e+12	1.374865e+12	1.428361e+12	1.549483e+12	
France	2.607840e+12	2.669424e+12	2.674637e+12	2.595967e+12	
South Korea	9.410199e+11	9.924316e+11	1.020510e+12	1.027730e+12	
Italy	2.202170e+12	2.234627e+12	2.211154e+12	2.089938e+12	
Spain	1.414823e+12	1.468146e+12	1.484530e+12	1.431475e+12	
Iran	3.895523e+11	4.250646e+11	4.289909e+11	4.389208e+11	
Australia	1.021939e+12	1.060340e+12	1.099644e+12	1.119654e+12	
Brazil	1.845080e+12	1.957118e+12	2.056809e+12	2.054215e+12	
	2010	2011	2012	2013	\
Country					
China	6.039659e+12	6.612490e+12	7.124978e+12	7.672448e+12	
United States	1.496437e+13	1.520402e+13	1.554216e+13	1.577367e+13	
Japan	5.498718e+12	5.473738e+12	5.569102e+12	5.644659e+12	
United Kingdom	2.403504e+12	2.450911e+12	2.479809e+12	2.533370e+12	

```
Russian Federation 1.524917e+12 1.589943e+12 1.645876e+12 1.666934e+12
Canada
                   1.613406e+12 1.664087e+12 1.693133e+12 1.730688e+12
Germany
                   3.417298e+12 3.542371e+12 3.556724e+12 3.567317e+12
India
                   1.708459e+12 1.821872e+12 1.924235e+12 2.051982e+12
France
                   2.646995e+12 2.702032e+12 2.706968e+12 2.722567e+12
South Korea
                   1.094499e+12 1.134796e+12 1.160809e+12 1.194429e+12
Italy
                   2.125185e+12 2.137439e+12 2.077184e+12 2.040871e+12
Spain
                   1.431673e+12 1.417355e+12 1.380216e+12 1.357139e+12
Iran
                   4.677902e+11 4.853309e+11 4.532569e+11 4.445926e+11
Australia
                   1.142251e+12 1.169431e+12 1.211913e+12 1.241484e+12
Brazil
                   2.208872e+12 2.295245e+12 2.339209e+12 2.409740e+12
                           2014
                                         2015
Country
China
                   8.230121e+12 8.797999e+12
United States
                   1.615662e+13 1.654857e+13
                   5.642884e+12 5.669563e+12
Japan
United Kingdom
                   2.605643e+12 2.666333e+12
Russian Federation 1.678709e+12 1.616149e+12
Canada
                   1.773486e+12 1.792609e+12
Germany
                   3.624386e+12 3.685556e+12
India
                   2.200617e+12 2.367206e+12
France
                   2.729632e+12 2.761185e+12
South Korea
                   1.234340e+12 1.266580e+12
Italy
                   2.033868e+12 2.049316e+12
Spain
                   1.375605e+12 1.419821e+12
Iran
                   4.639027e+11
Australia
                   1.272520e+12 1.301251e+12
Brazil
                   2.412231e+12 2.319423e+12
```

# 1.0.5 Question 2

The previous question joined three datasets then reduced this to just the top 15 entries. When you joined the datasets, but before you reduced this to the top 15 items, how many entries did you lose?

This function should return a single number.

```
return diff
answer_two()
```

[9]: 156

1.1 Answer the following questions in the context of only the top 15 countries by Scimagojr Rank (aka the DataFrame returned by answer\_one())

### 1.1.1 Question 3

What is the average GDP over the last 10 years for each country? (exclude missing values from this calculation.)

This function should return a Series named avgGDP with 15 countries and their average GDP sorted in descending order.

```
def answer_three():
    Top15 = answer_one()
    Top15['AvgGDP'] = Top15[list(Top15.loc[:,'2006':'2015'])].mean(axis = 1)
    Top15.sort_values('AvgGDP', ascending = False, inplace = True)
    #print(Top15['2006'].dtype)
    avgGDP = Top15['AvgGDP']
    return avgGDP
answer_three()
```

# [10]: Country

```
United States
                      1.536434e+13
China
                      6.348609e+12
Japan
                      5.542208e+12
Germany
                      3.493025e+12
France
                      2.681725e+12
United Kingdom
                      2.487907e+12
Brazil
                      2.189794e+12
                      2.120175e+12
Italy
India
                      1.769297e+12
Canada
                      1.660647e+12
Russian Federation
                      1.565459e+12
Spain
                      1.418078e+12
Australia
                      1.164043e+12
South Korea
                      1.106715e+12
                      4.441558e+11
Name: AvgGDP, dtype: float64
```

# 1.1.2 Question 4

By how much had the GDP changed over the 10 year span for the country with the 6th largest average GDP?

This function should return a single number.

```
[11]: def answer_four():
    Top15 = answer_one()
    gdpDiff = Top15.loc['United Kingdom', '2015'] - Top15.loc['United Kingdom',
    '2006']
    return gdpDiff
answer_four()
```

[11]: 246702696075.3999

# 1.1.3 Question 5

What is the mean Energy Supply per Capita?

This function should return a single number.

```
[12]: def answer_five():
    Top15 = answer_one()
    meanEn = Top15['Energy Supply per Capita'].mean()
    return meanEn
    answer_five()
```

[12]: 157.6

# 1.1.4 Question 6

What country has the maximum \% Renewable and what is the percentage?

This function should return a tuple with the name of the country and the percentage.

```
[13]: def answer_six():
    Top15 = answer_one()
    myList = []
    for i, row in Top15.iterrows():
        myList.append((i, Top15.loc[i, "% Renewable"]))
    maxPercentRenewable = max(myList, key = operator.itemgetter(1))
    #print(type(maxPercentRenewable))
    return maxPercentRenewable
answer_six()
```

[13]: ('Brazil', 69.64803)

#### 1.1.5 Question 7

Create a new column that is the ratio of Self-Citations to Total Citations. What is the maximum value for this new column, and what country has the highest ratio?

This function should return a tuple with the name of the country and the ratio.

```
[14]: def answer_seven():
    Top15 = answer_one()
    Top15['Ratio'] = Top15['Self-citations']/Top15['Citations']
    newList = []

    for i, row in Top15.iterrows():
        newList.append((i, Top15.loc[i, 'Ratio']))

    maxRatio = max(newList, key = operator.itemgetter(1))
    return maxRatio
    answer_seven()
```

[14]: ('China', 0.6893126179389422)

# 1.1.6 Question 8

Create a column that estimates the population using Energy Supply and Energy Supply per capita. What is the third most populous country according to this estimate?

This function should return a single string value.

[15]: 'United States'

### 1.1.7 Question 9

Create a column that estimates the number of citable documents per person. What is the correlation between the number of citable documents per capita and the energy supply per capita? Use the .corr() method, (Pearson's correlation).

This function should return a single number.

```
[16]: def answer_nine():
    Top15 = answer_one()
    Top15['PopEst'] = Top15['Energy Supply'] / Top15['Energy Supply per Capita']
    Top15['Citable docs per Capita'] = Top15['Citable documents'] /
    →Top15['PopEst']
```

```
corr = Top15['Citable docs per Capita'].corr(Top15['Energy Supply per⊔

    Gapita'])

    return corr
answer nine()
```

# [16]: 0.7940010435442946

# 1.1.8 Question 10

Create a new column with a 1 if the country's % Renewable value is at or above the median for all countries in the top 15, and a 0 if the country's % Renewable value is below the median.

This function should return a series named HighRenew whose index is the country name sorted in ascending order of rank.

```
[17]: def answer_ten():
          Top15 = answer one()
          Top151 = Top15[Top15['% Renewable'] >= Top15['% Renewable'].median()].copy()
          Top151['HighRenew'] = 1
          Top152 = Top15[Top15['% Renewable'] < Top15['% Renewable'].median()].copy()</pre>
          Top152['HighRenew'] = 0
          HighRenew = Top151[['HighRenew', 'Rank']].append(Top152[['HighRenew', L
       → 'Rank']])
          HighRenew.sort_values('Rank', inplace= True)
          HighRenew = HighRenew['HighRenew']
          #print(Top15['HighRenew'].dtype)
          return HighRenew
      answer_ten()
```

### [17]: Country

China 1 United States 0 Japan United Kingdom Russian Federation 1 Canada 1 Germany 1 India 0 France 1 South Korea 0 Italy Spain 1 Iran 0 Australia 0 Brazil

Name: HighRenew, dtype: int64

# 1.1.9 Question 11

Use the following dictionary to group the Countries by Continent, then create a dateframe that displays the sample size (the number of countries in each continent bin), and the sum, mean, and std deviation for the estimated population of each country.

This function should return a DataFrame with index named Continent ['Asia', 'Australia', 'Europe', 'North America', 'South America'] and columns ['size', 'sum', 'mean', 'std']

```
[18]: def answer_eleven():
          Top15 = answer_one()
          Top15['PopEstimate'] = Top15['Energy Supply']/Top15['Energy Supply per_

    Gapita']

          ContinentDict = {'China':'Asia',
                         'United States': 'North America',
                         'Japan':'Asia',
                         'United Kingdom': 'Europe',
                         'Russian Federation': 'Europe',
                         'Canada':'North America',
                         'Germany': 'Europe',
                         'India':'Asia',
                         'France': 'Europe',
                         'South Korea': 'Asia',
                         'Italy': 'Europe',
                         'Spain': 'Europe',
                         'Iran':'Asia',
                         'Australia': 'Australia',
                         'Brazil': 'South America'}
          index1, sizeColumn, sumColumn, meanColumn, stdColumn = [],[],[],[],
          for index, sub_df in Top15.groupby(ContinentDict):
              index1.append(index)
              sizeColumn.append(len(sub_df.index))
```

[18]:		size	sum	mean	std
	Continent				
	Asia	5	2.898666e+09	5.797333e+08	6.790979e+08
	Australia	1	2.331602e+07	2.331602e+07	NaN
	Europe	6	4.579297e+08	7.632161e+07	3.464767e+07
	North America	2	3.528552e+08	1.764276e+08	1.996696e+08
	South America	1	2.059153e+08	2.059153e+08	NaN

# 1.1.10 Question 12

Cut % Renewable into 5 bins. Group Top15 by the Continent, as well as these new % Renewable bins. How many countries are in each of these groups?

This function should return a **Series** with a MultiIndex of **Continent**, then the bins for **%** Renewable. Do not include groups with no countries.

```
'South Korea': 'Asia',
                  'Italy': 'Europe',
                  'Spain':'Europe',
                  'Iran':'Asia',
                  'Australia': 'Australia',
                  'Brazil':'South America'}
   mySeries = pd.Series()
    #Top15.set_index([ContinentDict, 'Bins'])
   thg = Top15.groupby([ContinentDict, 'Bins']).size()
   thg.index.names = ['Continent', 'Bins']
    #df.groupby(['col1','col2']).size()
    #df.reset_index(inplace = True)
    #Top15.rename(columns = {'index': 'Country'}, inplace = True)
   myList = []
    #for indices, sub_df in df:
     # mySeries = [(indices, sub_df.index)]
      # myList.append(mySeries)
       # print(indices)
        #print("----")
        #print(sub_df)
   return thg
answer_twelve()
```

```
[19]: Continent
                     Bins
                     (2.212, 15.753]
      Asia
                     (15.753, 29.227]
                                          1
      Australia
                     (2.212, 15.753]
                                         1
                     (2.212, 15.753]
      Europe
                                         1
                     (15.753, 29.227]
                                         3
                     (29.227, 42.701]
                                         2
      North America (2.212, 15.753]
                                         1
                     (56.174, 69.648]
                                          1
      South America (56.174, 69.648]
                                         1
      dtype: int64
```

# 1.1.11 Question 13

Convert the Population Estimate series to a string with thousands separator (using commas). Do not round the results.

```
e.g. 317615384.61538464 \rightarrow 317,615,384.61538464
```

This function should return a Series PopEst whose index is the country name and whose values are the population estimate string.

```
[20]: def answer_thirteen():
    Top15 = answer_one()
    Top15['PopEst'] = Top15['Energy Supply'] / Top15['Energy Supply per Capita']
    PopEst = Top15['PopEst'].apply(lambda x: '{:,}'.format(x))
    return PopEst
answer_thirteen()
```

# [20]: Country

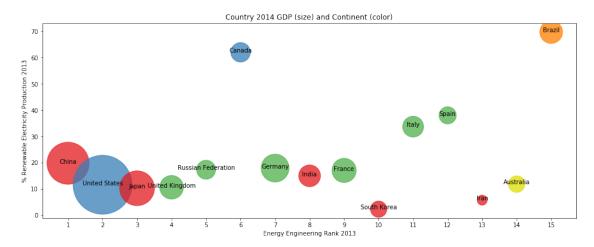
China	1,367,645,161.2903225
United States	317,615,384.61538464
Japan	127,409,395.97315437
United Kingdom	63,870,967.741935484
Russian Federation	143,500,000.0
Canada	35,239,864.86486486
Germany	80,369,696.96969697
India	1,276,730,769.2307692
France	63,837,349.39759036
South Korea	49,805,429.864253394
Italy	59,908,256.880733944
Spain	46,443,396.2264151
Iran	77,075,630.25210084
Australia	23,316,017.316017315
Brazil	205,915,254.23728815
Name: PopEst, dtype:	object

1 , 11 3

# 1.1.12 Visualization

Plotting a 'bubble plot' scatterplot to see an example visualization.

This bubble chart illustrates a distribution of a categorical and continuous variable along two dimensions of data; country's respective continent (color of bubble) and 2014 GDP (size of bubble) across % Renewable vs. Rank dimensions.



[]: