

Dan_UK_Population_Study

April 3, 2019

1 Study of UK population dynamics

1.1 Introduction

In this document I will analyze evolution of UK population accross: * Time * Region * Gender * Age

1.2 Data Loading

First of all, I trasform data collected from the given url into a "tidy" dataframe where * Each variable forms a column * Each observation forms a row * Each measure forms a dataframe * There is not "Total" element and dimensions elements are partition of the total

```
In [1]: import requests
import json
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

In [2]: url="https://www.nomisweb.co.uk/api/v01/dataset/NM_31_1.jsonstat.json"
r = requests.get(url)
data_json = json.loads(r.content.decode())
size=data_json['size']

In [3]: #initialize empty columns
time=[]
region=[]
gender=[]
age=[]
meas=[]

In [4]: # create every column separately to have better performances
for time_index in range(size[0]):
    time=time+[list(data_json['dimension']['time']['category']['label'].
                    values())[time_index])*size[1]*size[2]*size[3]*size[4]
    for geo_index in range(size[1]):
        region=region+[list(data_json['dimension']['geography']['category']['label'].
                             values())[geo_index])*size[2]*size[3]*size[4]
```

```

for sex_index in range(size[2]):
    gender=gender+[list(data_json['dimension']['sex']['category']['label'].
                        values())[sex_index])*size[3]*size[4]
for age_index in range(size[3]):
    age=age+[list(data_json['dimension']['age']['category']['label'].
                  values())[age_index])*size[4]
for meas_index in range(size[4]):
    meas=meas+[list(data_json['dimension']['measures']['category']
                    ['label'].values())[meas_index]]

```

In [5]: *#create the complete DataFrame from json*

```

data_df=pd.DataFrame(
    {'Year':time,
     'Region':region,
     'Gender':gender,
     'Age':age,
     'Measure':meas,
     'Value':data_json['value']}
)

```

In [6]: *#delete "Total" values and all elements that
#do not create a partition like "Aged 16 - 59/64"*

```

df_stage1=data_df[
    (data_df.Age!="All ages") &
    (data_df.Age!="Aged 16 - 59/64") &
    (data_df.Age!="Aged 18 - 24") &
    (data_df.Age!="Aged 16 - 64") &
    (data_df.Age!="Aged 0 - 15") &
    (data_df.Age!="Aged 65 and over") &
    (data_df.Gender!="Total") &
    (data_df.Measure=="Value")
][['Year', 'Region', 'Gender', 'Age', 'Value']]
df_stage1=df_stage1.dropna() #remove NAN rows of NI from 1981 to 1991
df_stage1['Value']=df_stage1['Value'].astype(int)
df_stage1['Year']=df_stage1['Year'].astype(int)

#create the England data by subtracting "Wales" from "England and Wales"
df_wales=df_stage1[df_stage1.Region=="Wales"]
df_engl_wales=df_stage1[df_stage1.Region=="England and Wales"]
df_engl=pd.merge(df_wales,df_engl_wales,on=["Year", "Gender", "Age"],how="inner")
df_engl['Value']=df_engl["Value_y"]-df_engl["Value_x"]
df_engl["Region"]="England"
df_engl=df_engl[["Year", "Region", "Gender", "Age", "Value"]]
df_pop=pd.concat([df_stage1[df_stage1.Region!="England and Wales"],df_engl])

#new age binning
df_pop=df_pop.replace(df_pop.Age.unique(),
                      ["Aged 0-24"]*6+["Aged 25-50"]*5+

```

```

["Aged 51-74"]*5+["Aged 75+"]*3
)
df_pop.head()

```

```

Out [6]:      Year Region Gender      Age  Value
2    1981  Wales   Male  Aged 0-24  18300
4    1981  Wales   Male  Aged 0-24  70200
6    1981  Wales   Male  Aged 0-24  95300
8    1981  Wales   Male  Aged 0-24 113500
10   1981  Wales   Male  Aged 0-24 118700

```

1.3 Data Exploration

Starting from the created dataframe, I study the population progression by the following dimensions: * Number of people * Geographic distribution * Age distribution ### Table with gender distribution in the last year As requested I created a dataframe showing male, female and total population in columns, per UK region in rows, as well as the UK total, for the most recent year.

```

In [8]: most_recent_year=df_pop.Year.max() #compute the most recent year
df_pop_ry=df_pop[df_pop.Year==most_recent_year].groupby(
    ['Region', 'Gender']).sum().reset_index()
df_pop_ry=df_pop_ry.pivot_table(
    index='Region', columns='Gender', values='Value',aggfunc='sum',margins=True)
df_pop_ry

```

```

Out [8]: Gender      Female      Male      All
Region
England      28138500  27481400  55619900
Northern Ireland  950500   920400   1870900
Scotland      2784800  2640400   5425200
Wales         1584900  1540100   3125000
All           33458700  32582300  66041000

```

1.3.1 Number of people analysis

I analyze the number of people trend from 1991 to 2017 because we do not have data of Northern Ireland before 1991.

```

In [9]: df_pop_np=df_pop[df_pop.Year>1991].groupby(
    ['Year', 'Region']).sum().reset_index()
df_pop_np=df_pop_np.pivot_table(
    index='Year', columns='Region', values='Value',aggfunc='sum',margins=False)
df_pop_np.head()

```

```

Out [9]: Region  England  Northern Ireland  Scotland  Wales
Year
1992      47998000           1623400    5085500  2877500
1993      48102700           1635800    5092400  2883400
1994      48228700           1643500    5102400  2887600

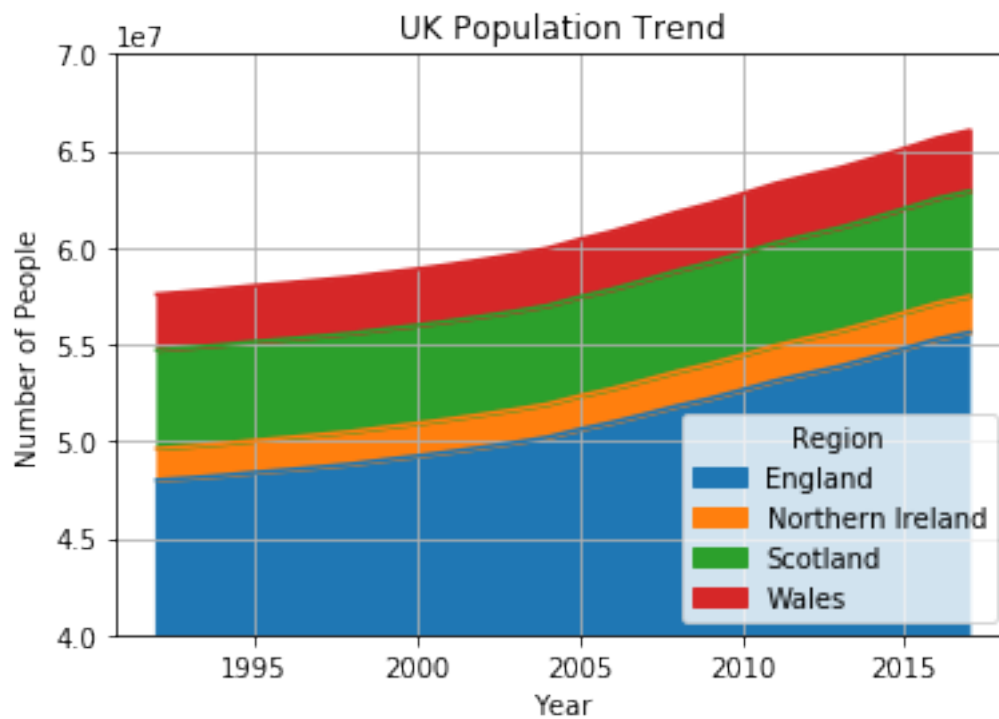
```

1995	48383400	1649100	5104000	2888600
1996	48519100	1661800	5092400	2891200

```
In [10]: pop_increase_rate=(df_pop_np.sum(axis=1)[2017]-df_pop_np.sum(axis=1)[2012])/5
pop_increase_rate
```

```
Out[10]: 467180.0
```

```
In [11]: df_pop_np.plot(kind="area")
plt.ylabel("Number of People")
plt.title("UK Population Trend")
plt.ylim([40000000,70000000])
plt.grid()
plt.savefig('np.png')
```



1.3.2 Region analysis

I analyze the percentage of people in each region from 1991 to 2017.

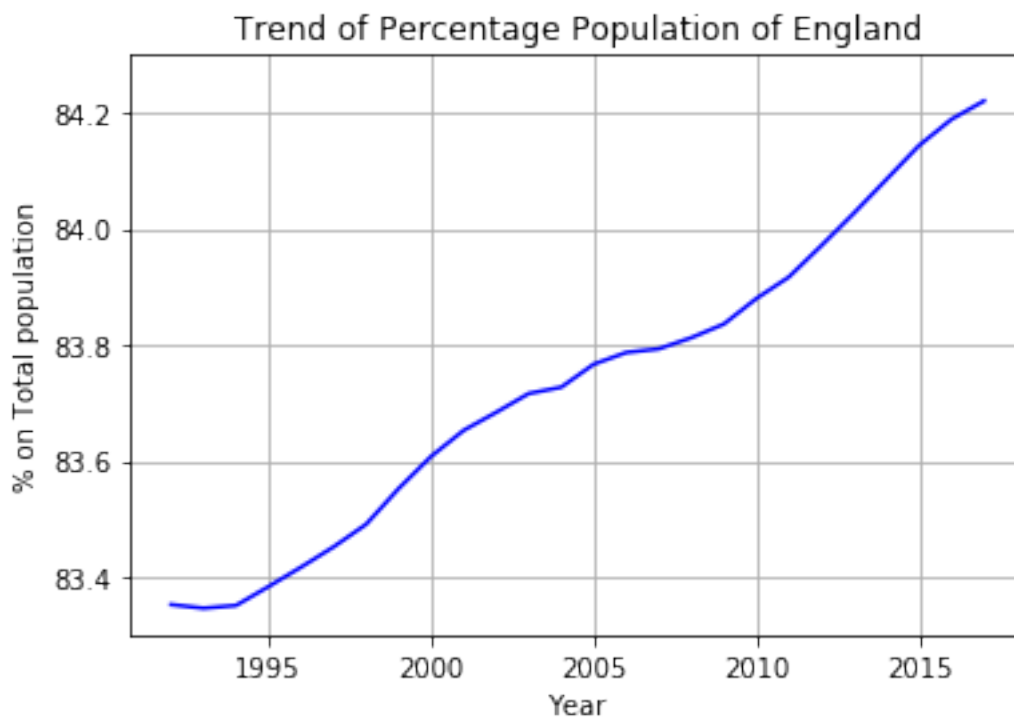
```
In [12]: df_pop_reg=pd.merge(
df_pop[df_pop.Year>1991].groupby(['Year','Region']).sum().reset_index(),
df_pop[df_pop.Year>1991].groupby(['Year']).sum().reset_index(),
on='Year',
how='inner')
```

```
)
df_pop_reg['Percentage']=df_pop_reg.Value_x/df_pop_reg.Value_y*100
df_pop_reg=df_pop_reg.pivot_table(
    index='Year',columns='Region', values='Percentage',aggfunc='sum',margins=False)
df_pop_reg.head()
```

```
Out[12]:
```

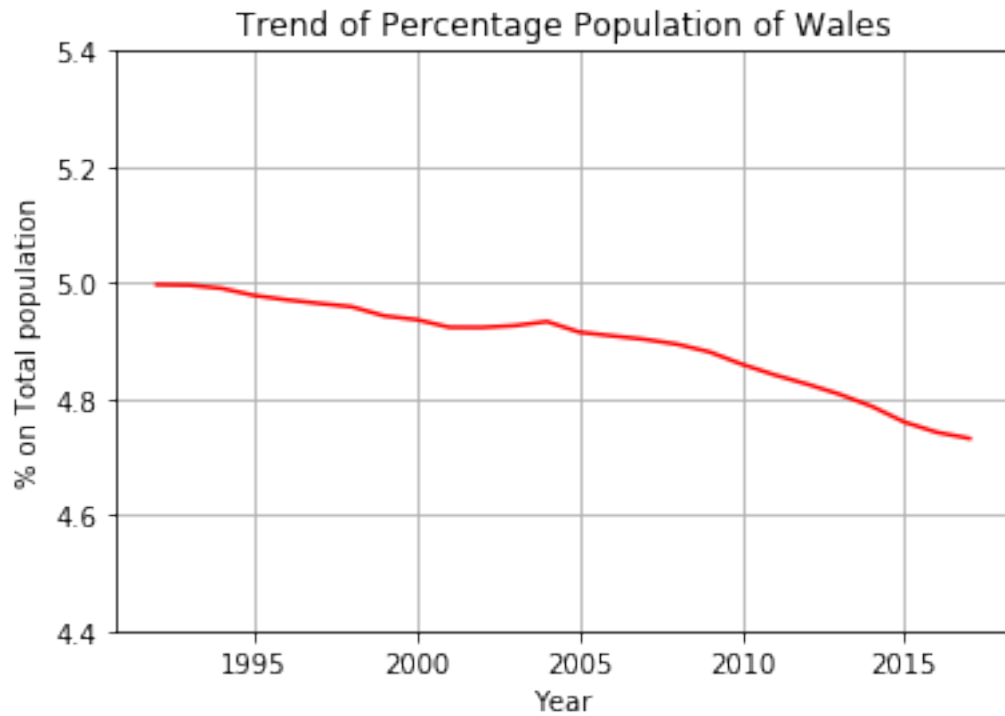
Region	England	Northern Ireland	Scotland	Wales
Year				
1992	83.352436	2.819166	8.831385	4.997013
1993	83.346242	2.834306	8.823463	4.995989
1994	83.350961	2.840369	8.818192	4.990477
1995	83.383570	2.842046	8.796193	4.978190
1996	83.417033	2.857069	8.755169	4.970730

```
In [13]: df_pop_reg['England'].plot(kind="line",color="blue")
plt.ylabel("% on Total population")
plt.title("Trend of Percentage Population of England")
plt.ylim([83.3,84.3]) #same scale for every chart (1%)
plt.grid()
plt.savefig('reg1.png')
```

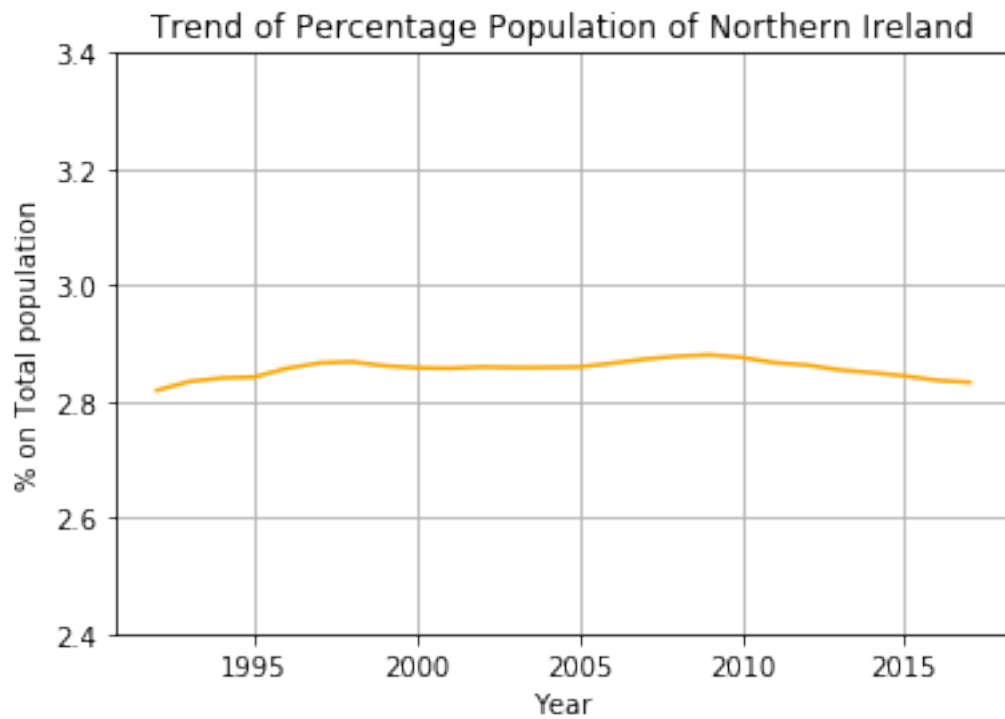


```
In [14]: df_pop_reg['Wales'].plot(kind="line",color="red")
plt.ylabel("% on Total population")
plt.title("Trend of Percentage Population of Wales")
```

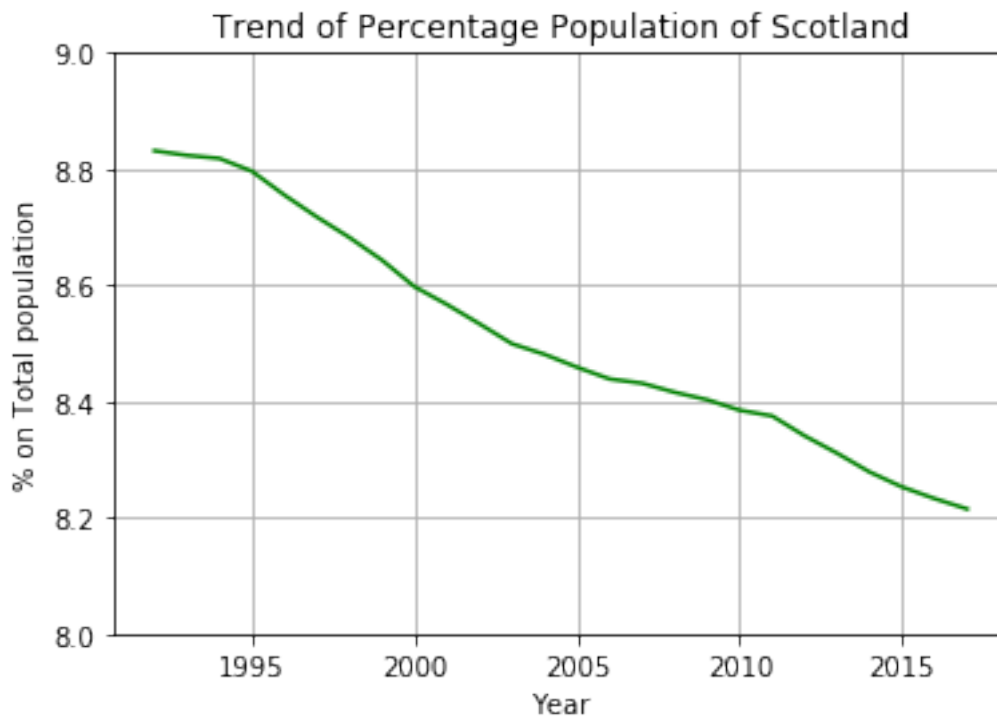
```
plt.ylim([4.4,5.4]) #same scale for every chart (1%)
plt.grid()
plt.savefig('reg2.png')
```



```
In [15]: df_pop_reg['Northern Ireland'].plot(kind="line",color="orange")
plt.ylabel("% on Total population")
plt.title("Trend of Percentage Population of Northern Ireland")
plt.ylim([2.4,3.4]) #same scale for every chart (1%)
plt.grid()
plt.savefig('reg3.png')
```



```
In [16]: df_pop_reg['Scotland'].plot(kind="line",color="green")
plt.ylabel("% on Total population")
plt.title("Trend of Percentage Population of Scotland")
plt.ylim([8,9]) #same scale for every chart (1%)
plt.grid()
plt.savefig('reg4.png')
```



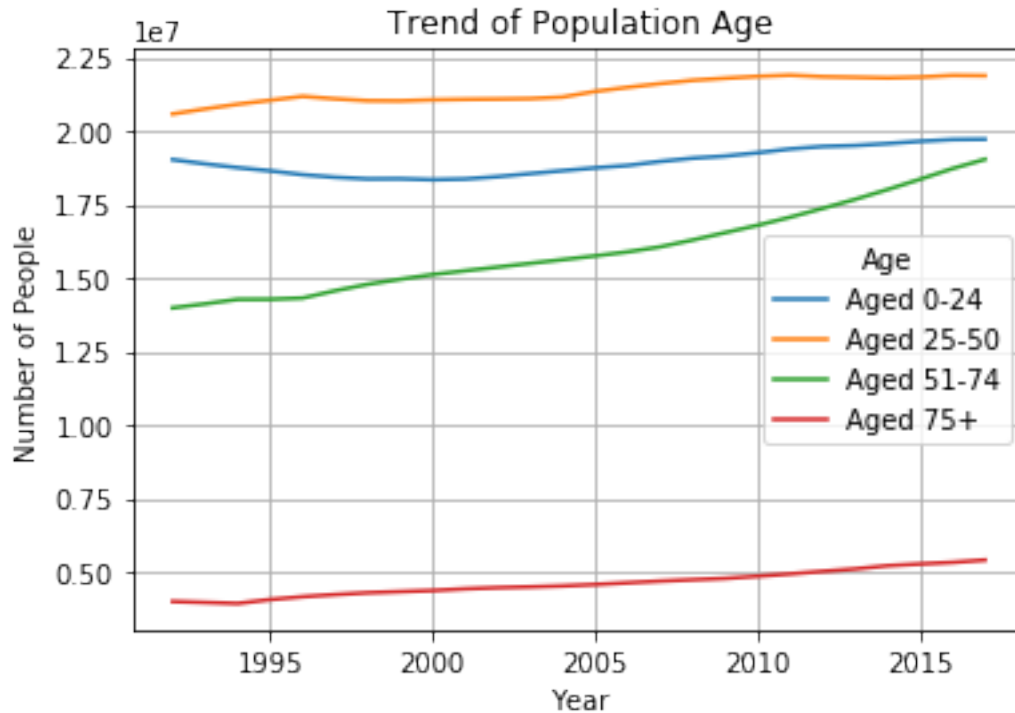
1.3.3 Region analysis

I analyze the percentage of people in the group 0-25/ 25-50/ 50-75/ 75+ from 1991 to 2017.

```
In [17]: df_pop_age_abs=df_pop[df_pop.Year>1991].pivot_table(
        index='Year',columns='Age', values='Value',aggfunc='sum',margins=False)
df_pop_age_abs.head()
```

```
Out[17]: Age    Aged 0-24    Aged 25-50    Aged 51-74    Aged 75+
Year
1992    19026200    20579800    13982800    3995600
1993    18881400    20751500    14119700    3961700
1994    18753000    20910100    14272200    3926900
1995    18644300    21045000    14275700    4060100
1996    18513300    21178300    14316000    4156900
```

```
In [18]: df_pop_age_abs.plot(kind="line")
plt.ylabel("Number of People")
plt.title("Trend of Population Age")
plt.grid()
plt.savefig('age1.png')
```

```
In [19]: df_pop_perc=pd.merge(
        df_pop[df_pop.Year==2017].groupby(['Year','Age','Region']).sum().reset_index(),
        df_pop[df_pop.Year==2017].groupby(['Year','Region']).sum().reset_index(),
        on=['Year','Region'],
        how='inner'
    )
    df_pop_perc['Percentage']=df_pop_perc.Value_x/df_pop_perc.Value_y*100
    df_pop_perc=df_pop_perc.pivot_table(
        index='Region',columns='Age', values='Percentage',aggfunc='sum',margins=False)
    df_pop_perc
```

```
Out[19]: Age          Aged 0-24  Aged 25-50  Aged 51-74  Aged 75+
Region
England          30.017134   33.305526   28.523065   8.154276
Northern Ireland  32.107542   32.968090   27.729970   7.194398
Scotland          27.847821   32.931505   30.961071   8.259603
Wales             29.190400   30.499200   31.142400   9.168000
```

```
In [20]: df_pop_perc.plot(kind="bar")
        plt.ylabel("% on Region population")
        plt.title("Age Percentage by Region in 2017")
        plt.grid()
        plt.legend(loc="lower right")
        plt.savefig('age2.png')
```

