



Merging JsonStat and GeoJson formatted data to create a GeoDataFrame, it's visualisation, and writing it to an Esri Shapefile

An example from Ireland using Python

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1. Introduction

In this tutorial we will discuss how to extract and join statistical data from the CSO's online database, StatBank, with geographic data, namely polygons, extracted from the OSi open data portal to create a GeoDataFrame, it's visualisation, and write it to an Esri Shapefile

2. Data Repository - Statbank

StatBank is the CSO's online database of Official Statistics . This database contains current and historical data series compiled from CSO statistical releases and is accessed at

<http://www.cso.ie/px/pxeirestat/statire/SelectTable/Omrade0.asp?PLanguage=0>

3. Geographic data repository – Open Data Portal

The OSi provide Open Geographic Data is data that can be freely used, re-used and redistributed by anyone - subject only, at most, to the requirement that the source of the information is to attributed and accessed at <https://data-osi.opendata.arcgis.com/>.

4. JSON-stat format

The **JSON-stat format** is a simple lightweight JSON¹ format for data dissemination. It is based in a cube model that arises from the evidence that the most common form of data dissemination is the tabular form. In this cube model, **datasets** are organized in **dimensions**. Dimensions are organized in **categories**.

Data dissemination is not the business of a few anymore. Even though the **JSON-stat format** can be the perfect companion for the **open data initiatives** of National Statistical Offices, it is suitable for all kinds of data disseminators because it has been designed with simplicity in mind.

¹ **JSON** (JavaScript Object Notation) is a lightweight data-interchange format. It is easy for humans to read and write. It is easy for machines to parse and generate. It is based on a subset of the [JavaScript Programming Language, Standard ECMA-262 3rd Edition - December 1999](#). JSON is a text format that is completely language independent but uses conventions that are familiar to programmers of the C-family of languages, including C, C++, C#, Java, JavaScript, Perl, Python, and many others. These properties make JSON an ideal data-interchange language.

5. GeoJSON

GeoJSON is a format for encoding a variety of geographic data structures. GeoJSON supports the following geometry types:

- Point,
- LineString,
- Polygon,
- MultiPoint,
- MultiLineString, and
- MultiPolygon.

Geometric objects with additional properties are Feature objects. Sets of features are contained by FeatureCollection objects.

6. Python

Python is an interpreted, high-level, general-purpose programming language.

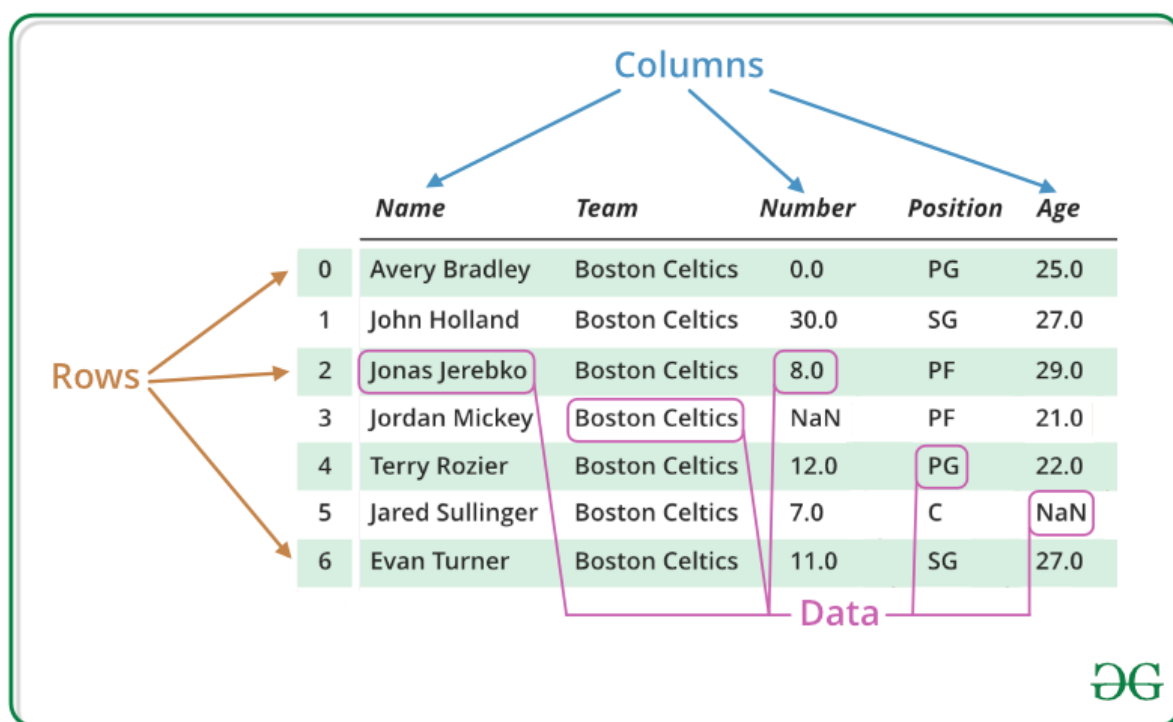
Python can be downloaded and installed from either

<https://www.python.org/downloads/>, or

<https://www.anaconda.com/>

6.1 Pandas DataFrame

A **Pandas DataFrame** is two-dimensional size-mutable, potentially heterogeneous tabular data structure with labeled axes (rows and columns). A Data frame is a two-dimensional data structure, i.e., data is aligned in a tabular fashion in rows and columns. Pandas DataFrame consists of three principal components, the **data**, **rows**, and **columns**.



The diagram illustrates a Pandas DataFrame with the following structure:

	<i>Name</i>	<i>Team</i>	<i>Number</i>	<i>Position</i>	<i>Age</i>
0	Avery Bradley	Boston Celtics	0.0	PG	25.0
1	John Holland	Boston Celtics	30.0	SG	27.0
2	Jonas Jerebko	Boston Celtics	8.0	PF	29.0
3	Jordan Mickey	Boston Celtics	NaN	PF	21.0
4	Terry Rozier	Boston Celtics	12.0	PG	22.0
5	Jared Sullinger	Boston Celtics	7.0	C	NaN
6	Evan Turner	Boston Celtics	11.0	SG	27.0

Annotations in the diagram:

- Columns:** A blue arrow points to the header row.
- Rows:** Three orange arrows point to the first three rows of the data.
- Data:** A pink box highlights the data cells for the second row (Jonas Jerebko), specifically the values '8.0', 'NaN', 'PG', and 'NaN'.

Figure 1: Pandas DataFrame

6.2 GeoDataFrame

A GeoDataFrame object is a pandas.DataFrame that has a column with geometry.

6.3 Python libraries

Within Python there are numerous libraries which are collections of functions and methods that allows one to perform many actions without writing your own code.

In this tutorial several libraries are installed, imported and used.

7. Python code

There are several steps in the process.

7.1 Installing the libraries

In this tutorial the following libraries are installed.

```
!pip install jsonstat.py
```

```
!pip install pyjstat
```

It should be noted that when installing the geopandas library one has to install several dependency libraries. A number of these libraries may not install directly and one may have to go to this website and download the .whl file and manually load using Anaconda prompt. <https://www.lfd.uci.edu/~gohlke/pythonlibs/#shapely>

```
!pip install numpy
```

```
!pip install GDAL # had to be manually loaded
```

```
!pip install pyshp
```

```
!pip install Shapely # had to be manually loaded
```

```
!pip install Fiona # had to be manually loaded
```

```
!pip install geopy
```

```
!pip install pyproj
```

```
!pip install geopandas # had to be manually loaded
```

```
!pip install descartes
```

If one wishes to visualise a GeoDataFrame then install these libraries.

```
!pip install matplotlib
```

```
!pip install mplleaflet
```

7.2 Importing and initialising the main python libraries

To activate the libraries, one must import and initialise them.

```
import geopandas as gpd
import pandas as pd      # using panda to convert jsonstat data set to pandas dataframe
import os                # The functions that the OS module provides allows you to
                           # interface with the underlying operating system that Python is
                           # running on – be that Windows, etc.

from pyjstat import pyjstat
import numpy as np
import shapefile as shp
import jsonstat         # import jsonstat.py package
import matplotlib.pyplot as plt # for plotting
import requests
import seaborn as sns
```

8. Setting current directory and downloading JsonStat file from the CSO's Statbank API

The following code allows one to identify and set and current working directory

```
os.getcwd()                # curent working directory 'C:\\XXXX'
os.chdir('C:\\XXXX\\Python') # changing current directory
```

We will now download the JsonStat QLF07 file, also known as a collection, from CSO Statbank's API, which contains data for Persons aged 15 years and over in Employment by Sex, NACE Rev 2 Economic Sector, Region and Quarter obtained from the CSO's Quarterly Labour Force Survey

First, we will define a url and file_name as follows

```
url = 'https://www.cso.ie/StatbankServices/StatbankServices.svc/jsonservice/responseinstance/QLF07'
file_name = "QLF07.json"
```


9. Extracting the JsonStatCollection

We will extract the JsonStatCollection and create a directory for data

```
jsonstat.from_url(url, file_name)  
file_path = (os.path.join('C:\\XXXX', file_name))
```

10. Initialising the JsonStatCollection from the file

We will now Initialise JsonStatCollection from the file and print the list of dataset contained into the collection.

```
collection = jsonstat.from_file(file_path)
```

To identify the number of dataset contained in this collection

```
len(collection)
```

The value '1' is returned indicating that there is one dataset contained in this collection and this can be confirmed by running the command,

```
collection
```

which provides this output,

JsonstatCollection contains the following JsonStatDataSet:

```
+----+-----+  
| pos | dataset |  
+----+-----+  
| 0  | 'dataset' |  
+----+-----+
```



11. Selecting and extract dataset contained into collection

We select, and extract dataset contained in the collection and name it QLF07 as follows:

```
QLF07 = collection.dataset(0)
```

To understand what is contained in the dataset simply run the code,

```
QLF07
```

Which provides the following output.

```
name: 'dataset'
label: 'Persons aged 15 years and over in Employment by Sex, NACE Rev 2 Economic Sector, Region and Quarter'
source: 'Persons aged 15 years and over in Employment by Sex, NACE Rev 2 Economic Sector, Region and Quarter'
size: 13122
```

pos	id	label	size	role
0	Sex	Sex	3	
1	NACE Rev 2 Economic Sector	NACE Rev 2 Economic Sector	18	
2	Region	Region	9	
3	Quarter	Quarter	27	time
4	Statistic	Statistic	1	metric

12. Exploring the dimensions of the JsonStatDataSet:

One can obtain some details about the dimensions, i.e categories within a dimension as follows.

```
QLF07.dimension('dataset')
```

Which provides the following output with the name of the dimensions identified.

JsonStatException: "dataset 'dataset': unknown dimension 'dataset' know dimensions
ids are:
Sex, NACE Rev 2 Economic Sector, Region, Quarter, Statistic"

To investigate the dimension region, for example, use the following command.

```
QLF07.dimension('Region')
```

Which returns the following.

```
+-----+-----+-----+
| pos | idx  | label      |
+-----+-----+-----+
| 0  | '-'  | 'State'    |
| 1  | 'IE041' | 'Border'   |
| 2  | 'IE042' | 'West'     |
| 3  | 'IE051' | 'Mid-West' |
| 4  | 'IE052' | 'South-East' |
| 5  | 'IE053' | 'South-West' |
| 6  | 'IE061' | 'Dublin'   |
| 7  | 'IE062' | 'Mid-East' |
| 8  | 'IE063' | 'Midland'  |
+-----+-----+-----+
```

The following code provides details on the Nace Rev 2 Economic Sector dimension.

```
QLF07.dimension('NACE Rev 2 Economic Sector')
```

It can also be referenced as dimension 1, as follow.

```
QLF07.dimension(1)
```

Which returns the following.

```
# dimension 1
```



pos	idx	label
0	'-'	'All NACE economic sectors'
1	'A'	'Agriculture, forestry and fishing (A)'
2	'F'	'Construction (F)'
3	'G'	'Wholesale and retail trade, repair of motor vehicles and motorcycles (G)'
4	'H'	'Transportation and storage (H)'
5	'I'	'Accommodation and food service activities (I)'
6	'J'	'Information and communication (J)'
7	'M'	'Professional, scientific and technical activities (M)'
8	'N'	'Administrative and support service activities (N)'
9	'O'	'Public administration and defence, compulsory social security (O)'
10	'P'	'Education (P)'
11	'Q'	'Human health and social work activities (Q)'
12	'Y0900'	'Industry (B to E)'
13	'Y1000'	'Industry and Construction (B to F)'
14	'Y3000'	'Services (G to U)'
15	'Y3500'	'Financial, insurance and real estate activities (K,L)'
16	'Y7500'	'Other NACE activities (R to U)'
17	'ZXD400'	'Not stated'

We continue checking the other dimensions, with

QLF07.dimension('Sex')

returning:

pos	idx	label
0	'-'	'Both sexes'
1	'1'	'Male'
2	'2'	'Female'

and



QLF07.dimension('Quarter')

returning:

```

#-----+-----+-----+ +-----+-----+
#| pos | idx      | label    | | pos | idx      | label    |
#-----+-----+-----+ +-----+-----+
#| 0   | '2012Q1' | '2012Q1' | | 14  | '2015Q3' | '2015Q3' |
#| 1   | '2012Q2' | '2012Q2' | | 15  | '2015Q4' | '2015Q4' |
#| 2   | '2012Q3' | '2012Q3' | | 16  | '2016Q1' | '2016Q1' |
#| 3   | '2012Q4' | '2012Q4' | | 17  | '2016Q2' | '2016Q2' |
#| 4   | '2013Q1' | '2013Q1' | | 18  | '2016Q3' | '2016Q3' |
#| 5   | '2013Q2' | '2013Q2' | | 19  | '2016Q4' | '2016Q4' |
#| 6   | '2013Q3' | '2013Q3' | | 20  | '2017Q1' | '2017Q1' |
#| 7   | '2013Q4' | '2013Q4' | | 21  | '2017Q2' | '2017Q2' |
#| 8   | '2014Q1' | '2014Q1' | | 22  | '2017Q3' | '2017Q3' |
#| 9   | '2014Q2' | '2014Q2' | | 23  | '2017Q4' | '2017Q4' |
#| 10  | '2014Q3' | '2014Q3' | | 24  | '2018Q1' | '2018Q1' |
#| 11  | '2014Q4' | '2014Q4' | | 25  | '2018Q2' | '2018Q2' |
#| 12  | '2015Q1' | '2015Q1' | | 26  | '2018Q3' | '2018Q3' |
#| 13  | '2015Q2' | '2015Q2' |
#-----+-----+-----+ +-----+-----+
    
```

And for the Statistic dimension, number 4,

QLF07.dimension(4)

returns

```

+-----+-----+-----+-----+
| pos | idx      | label                                                                 |
+-----+-----+-----+-----+
| 0   | 'QLF07C01' | 'Persons aged 15 years and over in Employment (Thousand)' |

```

--	--	--

13. Accessing a value in the `JsonStatDataSet`, “dataset”

To access a value in the dataset one references the dimension values such as:

```
QLF07.data(Region = 'IE041', Sex = '1', Quarter = '2018Q3')
```

Which returns

```
JsonStatValue(idx=4427, value=97.5, status=None)
```

It should be noted that the dimension name NACE Rev 2 Economic Sector cannot be used as it contains spaces, simply use the dimension number instead.

14. Transforming the `JsonStatDataSet`, “dataset”, into a pandas `DataFrame`

The following code transforms the `JsonStatDataSet`, “dataset”, into a pandas `DataFrame`

```
df_QLF07 = QLF07.to_data_frame(content='idx')
```

Note that the `content = 'idx'` is used to grab the geography category values, which we will call `GEOGID`, from the regional dimension

It is also possible to grab the label, by replacing `content='idx'` with `content = 'label'`

```
df_QLF07A = QLF07.to_data_frame(content='label')
```

Again, one can display the dataframe.

```
display(df_QLF07)
```

Which provides the following output.



Sex	NACE Rev 2	Economic Sector	Region	Quarter	Statistic	Value
0	-	-	-	2012Q1	QLF07C01	1863.2
1	-	-	-	2012Q2	QLF07C01	1878.0
2	-	-	-	2012Q3	QLF07C01	1887.0
3	-	-	-	2012Q4	QLF07C01	1893.6
4	-	-	-	2013Q1	QLF07C01	1892.0
5	-	-	-	2013Q2	QLF07C01	1926.3
6	-	-	-	2013Q3	QLF07C01	1961.8
7	-	-	-	2013Q4	QLF07C01	1971.0
8	-	-	-	2014Q1	QLF07C01	1950.7
9	-	-	-	2014Q2	QLF07C01	1970.3
10	-	-	-	2014Q3	QLF07C01	2008.9
11	-	-	-	2014Q4	QLF07C01	2025.2
12	-	-	-	2015Q1	QLF07C01	2014.4
13	-	-	-	2015Q2	QLF07C01	2049.7
14	-	-	-	2015Q3	QLF07C01	2079.9
15	-	-	-	2015Q4	QLF07C01	2085.4
16	-	-	-	2016Q1	QLF07C01	2080.8
17	-	-	-	2016Q2	QLF07C01	2126.7
18	-	-	-	2016Q3	QLF07C01	2158.0
19	-	-	-	2016Q4	QLF07C01	2163.5
20	-	-	-	2017Q1	QLF07C01	2158.4
21	-	-	-	2017Q2	QLF07C01	2180.9
22	-	-	-	2017Q3	QLF07C01	2206.5
23	-	-	-	2017Q4	QLF07C01	2230.8
24	-	-	-	2018Q1	QLF07C01	2220.7
25	-	-	-	2018Q2	QLF07C01	2255.0
26	-	-	-	2018Q3	QLF07C01	2273.2
27	-	-	IE041	2012Q1	QLF07C01	145.1
28	-	-	IE041	2012Q2	QLF07C01	146.3
29	-	-	IE041	2012Q3	QLF07C01	145.3

One can display the dataframe with the labels.

```
display(df_QLF07A)
```

#	Sex	NACE Rev 2 Economic Sector	Region	Quarter	Statistic	Value
#0	Both sexes	All NACE economic sectors	State	2012Q1	Persons aged 15 years and over in Employment (...)	1863.2
#1	Both sexes	All NACE economic sectors	State	2012Q2	Persons aged 15 years and over in Employment (...)	1878.0
#2	Both sexes	All NACE economic sectors	State	2012Q3	Persons aged 15 years and over in Employment (...)	1887.0
#3	Both sexes	All NACE economic sectors	State	2012Q4	Persons aged 15 years and over in Employment (...)	1893.6
#4	Both sexes	All NACE economic sectors	State	2013Q1	Persons aged 15 years and over in Employment (...)	1892.0
#5	Both sexes	All NACE economic sectors	State	2013Q2	Persons aged 15 years and over in Employment (...)	1926.3
#6	Both sexes	All NACE economic sectors	State	2013Q3	Persons aged 15 years and over in Employment (...)	1961.8
#7	Both sexes	All NACE economic sectors	State	2013Q4	Persons aged 15 years and over in Employment (...)	1971.0
#8	Both sexes	All NACE economic sectors	State	2014Q1	Persons aged 15 years and over in Employment (...)	1950.7
#9	Both sexes	All NACE economic sectors	State	2014Q2	Persons aged 15 years and over in Employment (...)	1970.3
#10	Both sexes	All NACE economic sectors	State	2014Q3	Persons aged 15 years and over in Employment (...)	2008.9
#11	Both sexes	All NACE economic sectors	State	2014Q4	Persons aged 15 years and over in Employment (...)	2025.2
#12	Both sexes	All NACE economic sectors	State	2015Q1	Persons aged 15 years and over in Employment (...)	2014.4
#13	Both sexes	All NACE economic sectors	State	2015Q2	Persons aged 15 years and over in Employment (...)	2049.7
#14	Both sexes	All NACE economic sectors	State	2015Q3	Persons aged 15 years and over in Employment (...)	2079.9
#15	Both sexes	All NACE economic sectors	State	2015Q4	Persons aged 15 years and over in Employment (...)	2085.4
#16	Both sexes	All NACE economic sectors	State	2016Q1	Persons aged 15 years and over in Employment (...)	2080.8
#17	Both sexes	All NACE economic sectors	State	2016Q2	Persons aged 15 years and over in Employment (...)	2126.7
#18	Both sexes	All NACE economic sectors	State	2016Q3	Persons aged 15 years and over in Employment (...)	2158.0
#19	Both sexes	All NACE economic sectors	State	2016Q4	Persons aged 15 years and over in Employment (...)	2163.5
#20	Both sexes	All NACE economic sectors	State	2017Q1	Persons aged 15 years and over in Employment (...)	2158.4
#21	Both sexes	All NACE economic sectors	State	2017Q2	Persons aged 15 years and over in Employment (...)	2180.9
#22	Both sexes	All NACE economic sectors	State	2017Q3	Persons aged 15 years and over in Employment (...)	2206.5
#23	Both sexes	All NACE economic sectors	State	2017Q4	Persons aged 15 years and over in Employment (...)	2230.8
#24	Both sexes	All NACE economic sectors	State	2018Q1	Persons aged 15 years and over in Employment (...)	2220.7
#25	Both sexes	All NACE economic sectors	State	2018Q2	Persons aged 15 years and over in Employment (...)	2255.0
#26	Both sexes	All NACE economic sectors	State	2018Q3	Persons aged 15 years and over in Employment (...)	2273.2
#27	Both sexes	All NACE economic sectors	Border	2012Q1	Persons aged 15 years and over in Employment (...)	145.1
#28	Both sexes	All NACE economic sectors	Border	2012Q2	Persons aged 15 years and over in Employment (...)	146.3
#29	Both sexes	All NACE economic sectors	Border	2012Q3	Persons aged 15 years and over in Employment (...)	145.3

15. Renaming columns in a dataframe

To confirm the column list in the dataframes use the following code.

```
df_QLF07.columns
```

which returns,

```
Out[79]: Index(['Sex', 'NACE Rev 2 Economic Sector', 'Region', 'Quarter', 'Statistic',  
              'Value'], dtype='object')
```

and,


```
df_QLF07A.columns
```

which returns

```
Out[80]: Index(['Sex', 'NACE Rev 2 Economic Sector', 'Region', 'Quarter', 'Statistic',  
              'Value'], dtype='object')
```

We can now rename columns *Region* to *NUTS3* in *df_QLF07* dataframe and to *NUTS3NAME* in *df_QLF07A* dataframe, in order to join with the GeoJson file below

```
df_QLF07.rename(columns={'Region':'NUTS3'}, inplace=True)  
df_QLF07A.rename(columns={'Region':'NUTS3NAME'}, inplace=True)
```

We will now list the of columns in each dataframe to confirm the change of name.

```
df_QLF07.columns
```

returns,

```
Out[83]: Index(['Sex', 'NACE Rev 2 Economic Sector', 'NUTS3', 'Quarter', 'Statistic',  
              'Value'], dtype='object')
```

and,

```
df_QLF07A.columns
```

returns,

```
Out[84]: Index(['Sex', 'NACE Rev 2 Economic Sector', 'NUTS3NAME', 'Quarter',  
              'Statistic', 'Value'], dtype='object')
```

16. Creating a GeoDataFrame directly from GeoJson dataset on OSi's open portal

We will now create a dataframe directly from a GeoJson data for the NUTS3 Regions from the Ordnance Survey Ireland's (OSi) open data portal

Again we will define a url as follows:

```
url1 = "http://data-osi.opendata.arcgis.com/datasets/8e1da9ca81cb478d8146580b130abe08_2.geojson"
```

Then create the data frame.

```
dfa = gpd.read_file(url1)
```

To ensure that the a GeoDataFrame has been created, we can simply run the code.

```
type(dfa)
```

which returns,

```
geopandas.geodataframe.GeoDataFrame
```

We can list the columns in the GeoDataFrame,

```
dfa.columns
```

which returns,

```
Index(['OBJECTID', 'NUTS1', 'NUTS1NAME', 'NUTS2', 'NUTS2NAME', 'NUTS3',  
'NUTS3NAME', 'GUID', 'Shape__Area', 'Shape__Length', 'geometry'],  
      dtype='object')
```

We can explore this GeoDataFrame by referencing the columns.

```
dfa.GUID
```



returns,

```
0 B1A65D7C-1984-4A87-AD58-0E846812C992
1 A69CA800-8D87-4920-A7C1-50426A1D39B4
2 42C5C2A5-2D71-4BD1-BDB5-BD7D3198CD78
3 27C93D4E-AD0E-4B0C-8FA8-3566AEEDA5CC
4 604546A1-A856-4B9B-AD46-E88B27C27155
5 F97E459B-57ED-49C0-8A28-2BC1C7F08E88
6 8E4862CC-7E43-4BF5-A4EF-B2D5ECBA61EF
7 B26C8BAA-F3C5-49A9-B74E-D7FED1823E65
Name: GUID, dtype: object
```

With

```
dfa.NUTS3
```

returning,

```
0 IE041
1 IE042
2 IE051
3 IE052
4 IE053
5 IE061
6 IE062
7 IE063
Name: NUTS3, dtype: object
```

While,

```
dfa.NUTS3NAME
```

returns,

```
0 Border
1 West
```

- 2 Mid-West
- 3 South-East
- 4 South-West
- 5 Dublin
- 6 Mid-East
- 7 Midlands

Name: NUTS3NAME, dtype: object

17. Merging the Data and GeoData Frames

We will create a new GeoDataFrame by merging the Data and Geodata frames on the NUTS3 column.

```
dfb = dfa.merge(df_QLF07, on='NUTS3')
```

We will now list the of columns in the new GeoDataFrame to confirm the merge was successful.

```
dfb.columns
```

which returns,

```
Index(['OBJECTID', 'NUTS1', 'NUTS1NAME', 'NUTS2', 'NUTS2NAME', 'NUTS3', '#  
'NUTS3NAME', 'GUID', 'Shape__Area', 'Shape__Length', 'geometry', 'Sex', 'NACE  
Rev 2 Economic Sector', 'Quarter', 'Statistic', 'Value'], dtype='object')
```

We can use head and tail commands to quickly view the new GeoDataFrame, **dfb**.



The head command,

```
dfb.head()
```

will output the 1st five lines of the GeoDataFrame as follows, (not all columns are shown by default):

OBJECTID	NUTS1	NUTS1NAME	NUTS2	NUTS2NAME	NUTS3	NUTS3NAME	GUID	Shape_Area	Shape_Length	geometry	Quarter	Statistic	Value	
0	1	IE	Ireland	IE04	Northern and Western	IE041	Border B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	-	2012Q1	QLF07C01	145.1
1	1	IE	Ireland	IE04	Northern and Western	IE041	Border B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	-	2012Q2	QLF07C01	146.3
2	1	IE	Ireland	IE04	Northern and Western	IE041	Border B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	-	2012Q3	QLF07C01	145.3
3	1	IE	Ireland	IE04	Northern and Western	IE041	Border B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	-	2012Q4	QLF07C01	147.7
4	1	IE	Ireland	IE04	Northern and Western	IE041	Border B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	-	2013Q1	QLF07C01	150.9

and, the tail command,

```
dfb.tail()
```

will output the last five lines of the GeoDataFrame as follows, (not all columns are shown by default):

OBJECTID	NUTS1	NUTS1NAME	NUTS2	NUTS2NAME	NUTS3	NUTS3NAME	GUID	Shape_Area	Shape_Length	geometry	NACE Rev 2 Economic Sector	Quarter	Statistic	Value	
11659	8	IE	Ireland	IE06	Eastern and Midland	IE063	Midlands B26C8BAA-F3C5-49A9-B74E-D7FED1823E65	0.896836	7.564148	POLYGON ((-7.64673135464132 53.9415530759351, ...	2	ZXD400	2017Q3	QLF07C01	NaN
11660	8	IE	Ireland	IE06	Eastern and Midland	IE063	Midlands B26C8BAA-F3C5-49A9-B74E-D7FED1823E65	0.896836	7.564148	POLYGON ((-7.64673135464132 53.9415530759351, ...	2	ZXD400	2017Q4	QLF07C01	NaN
11661	8	IE	Ireland	IE06	Eastern and Midland	IE063	Midlands B26C8BAA-F3C5-49A9-B74E-D7FED1823E65	0.896836	7.564148	POLYGON ((-7.64673135464132 53.9415530759351, ...	2	ZXD400	2018Q1	QLF07C01	NaN
11662	8	IE	Ireland	IE06	Eastern and Midland	IE063	Midlands B26C8BAA-F3C5-49A9-B74E-D7FED1823E65	0.896836	7.564148	POLYGON ((-7.64673135464132 53.9415530759351, ...	2	ZXD400	2018Q2	QLF07C01	NaN
11663	8	IE	Ireland	IE06	Eastern and Midland	IE063	Midlands B26C8BAA-F3C5-49A9-B74E-D7FED1823E65	0.896836	7.564148	POLYGON ((-7.64673135464132 53.9415530759351, ...	2	ZXD400	2018Q3	QLF07C01	NaN



18. Creating additional GeoDataFrames from dfb

We will now create individual GeoDataFrames for males, females and total, by parsing the GeoDataFrame, dfb, where:

- Sex: 1 = Males, 2 = Females, - = Total,
- NACE Rev 2 Economic Sector = F, Construction, and
- Quarter = 2018Q3

```
dfbm = dfb.loc[(dfb['Sex'] == '1') & (dfb['NACE Rev 2 Economic Sector'] == 'F') & (dfb['Quarter'] == '2018Q3')]
dfbf = dfb.loc[(dfb['Sex'] == '2') & (dfb['NACE Rev 2 Economic Sector'] == 'F') & (dfb['Quarter'] == '2018Q3')]
dfbt = dfb.loc[(dfb['Sex'] == '-') & (dfb['NACE Rev 2 Economic Sector'] == 'F') & (dfb['Quarter'] == '2018Q3')]
```

We can output the 1st five lines of each GeoDataFrame, for male, female and total.

```
dfbm.head()
```

OBJECTID	NUTS1	NUTS1_NAME	NUTS2	NUTS2NAME	NUTS3	NUTS3NAME	GUID	Shape__Area	Shape__Length	geometry	Sex	Quarter	Statistic	Value
566	1	IE Ireland	IE04	Northern and Western	IE041	Border	B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	1	2018Q3	QLF07C01	13.9
2024	2	IE Ireland	IE04	Northern and Western	IE042	West	A69CA800-8D87-4920-A7C1-50426A1D39B4	1.940981	38.200397	(POLYGON ((-9.524108481921839 53.06827438179, ...	1	2018Q3	QLF07C01	13.3
3482	3	IE Ireland	IE05	Southern	IE051	Mid-West	42C5C2A5-2D71-4BD1-BDB5-BD7D3198CD78	1.396164	11.020233	(POLYGON ((-9.935623142260511 52.5611963634314,...	1	2018Q3	QLF07C01	14.5
4940	4	IE Ireland	IE05	Southern	IE052	South-East	27C93D4E-AD0E-4B0C-8FA8-3566AEEDA5CC	0.951352	11.412129	(POLYGON ((-7.76097866504068 51.9451840492602,...	1	2018Q3	QLF07C01	12.3
6398	5	IE Ireland	IE05	Southern	IE053	South-West	604546A1-A856-4B9B-AD46-E88B27C27155	1.610451	32.378775	(POLYGON ((-9.472006032876481 51.4533061425, -...	1	2018Q3	QLF07C01	20.5



dfbf.head()

OBJECTID	NUTS1	NUTS1NAME	NUTS2	NUTS2NAME	NUTS3	NUTS3NAME	GUID	Shape_Area	Shape_Length	geometry	Sex	Quarter	Statistic	Value	
1052	1	IE	Ireland	IE04	Northern and Western	IE041	Border	B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	2	2018Q3	QLF07C01	NaN
2510	2	IE	Ireland	IE04	Northern and Western	IE042	West	A69CA800-8D87-4920-A7C1-50426A1D39B4	1.940981	38.200397	(POLYGON ((-9.524108481921839 53.06827438179, ...	2	2018Q3	QLF07C01	NaN
3968	3	IE	Ireland	IE05	Southern	IE051	Mid-West	42C5C2A5-2D71-4BD1-BDB5-BD7D3198CD78	1.396164	11.020233	(POLYGON ((-9.935623142260511 52.5611963634314...	2	2018Q3	QLF07C01	NaN
5426	4	IE	Ireland	IE05	Southern	IE052	South-East	27C93D4E-AD0E-4B0C-8FA8-3566AEEDA5CC	0.951352	11.412129	(POLYGON ((-7.76097866504068 51.9451840492602,...	2	2018Q3	QLF07C01	NaN
6884	5	IE	Ireland	IE05	Southern	IE053	South-West	604546A1-A856-4B9B-AD46-E88B27C27155	1.610451	32.378775	(POLYGON ((-9.472006032876481 51.4533061425, -...	2	2018Q3	QLF07C01	NaN

dfbt.head()

OBJECTID	NUTS1	NUTS1NAME	NUTS2	NUTS2NAME	NUTS3	NUTS3NAME	GUID	Shape_Area	Shape_Length	geometry	Sex	Quarter	Statistic	Value	
80	1	IE	Ireland	IE04	Northern and Western	IE041	Border	B1A65D7C-1984-4A87-AD58-0E846812C992	1.594889	31.557961	(POLYGON ((-7.25362176708799 55.4458025525653,...	-	2018Q3	QLF07C01	14.9
1538	2	IE	Ireland	IE04	Northern and Western	IE042	West	A69CA800-8D87-4920-A7C1-50426A1D39B4	1.940981	38.200397	(POLYGON ((-9.524108481921839 53.06827438179, ...	-	2018Q3	QLF07C01	14.4
2996	3	IE	Ireland	IE05	Southern	IE051	Mid-West	42C5C2A5-2D71-4BD1-BDB5-BD7D3198CD78	1.396164	11.020233	(POLYGON ((-9.935623142260511 52.5611963634314...	-	2018Q3	QLF07C01	15.2
4454	4	IE	Ireland	IE05	Southern	IE052	South-East	27C93D4E-AD0E-4B0C-8FA8-3566AEEDA5CC	0.951352	11.412129	(POLYGON ((-7.76097866504068 51.9451840492602,...	-	2018Q3	QLF07C01	12.7
5912	5	IE	Ireland	IE05	Southern	IE053	South-West	604546A1-A856-4B9B-AD46-E88B27C27155	1.610451	32.378775	(POLYGON ((-9.472006032876481 51.4533061425, -...	-	2018Q3	QLF07C01	21.8

19. Writing a GeoDataFrame to an Esri shape file

We will now write GeoDataFrame for containing the data for males, *dfbm*, to an Esri shape file for visualisation, as follows:

```
dfbm.to_file('NUTS3__Generalised_20mA', driver='ESRI Shapefile')
```

A *NUTS3__Generalised_20mA* folder is created in the working directory 'C:\XXXX\Python'.

20. Visualisation with python

One can also visualise a GeoDataFrame in Python using *matplotlib*.

Import the library.

```
import matplotlib.pyplot as plt
```

One can then preview the GeoDataFrame, *dfa*, as follows:

```
dfa.plot()
```

Which returns the NUTS3 outlines, as it contains no data.

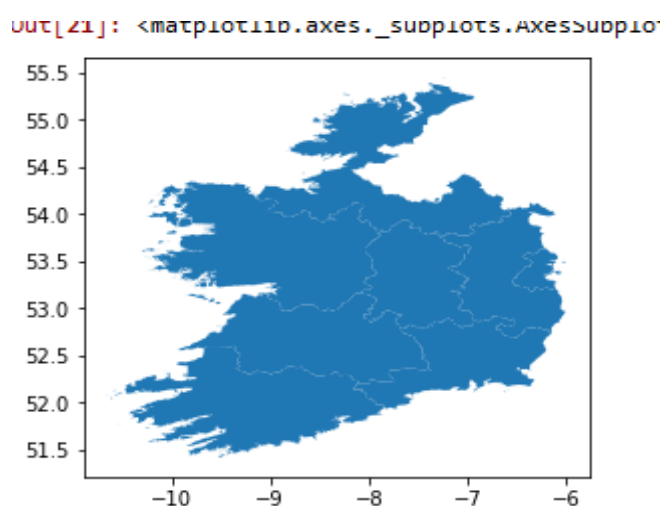


Fig 1: NUTS 3 outline

We can now visualise the GeoDataFrame, *dfbm*, which contains data on the number of Males in the Construction Industry for Quarter 3 of 2018..

We set the column that will be called to be visualised on the map.

```
variable = 'Value'
```

Next, we set up the figure and axis.

```
fig, ax = plt.subplots(1, figsize=(10,6))
```

We now ADD layer of polygons on the axis

```
dfbmpp = dfbm.plot(column=variable,  
                    ax=ax,  
                    alpha = 0.8, #intensity of colour of a polygon  
                    cmap="Reds', #colour of the polygon  
                    legend='True',  
                    linewidth=0.3,  
                    edgecolor='0.8',  
                    label = 'Males')
```

We will also turn off the axis seen in **Figure 1**, as follows.

```
ax.axis('off')
```

A figure title is added as follows,

```
ax.set_title('Numbers of Males at work in the Construction Industry, Q3 2018')
```

and a legend title is set.

```
ax.legend(title='000')
```

Now we can view the map.

`ax.figure`

which outputs:

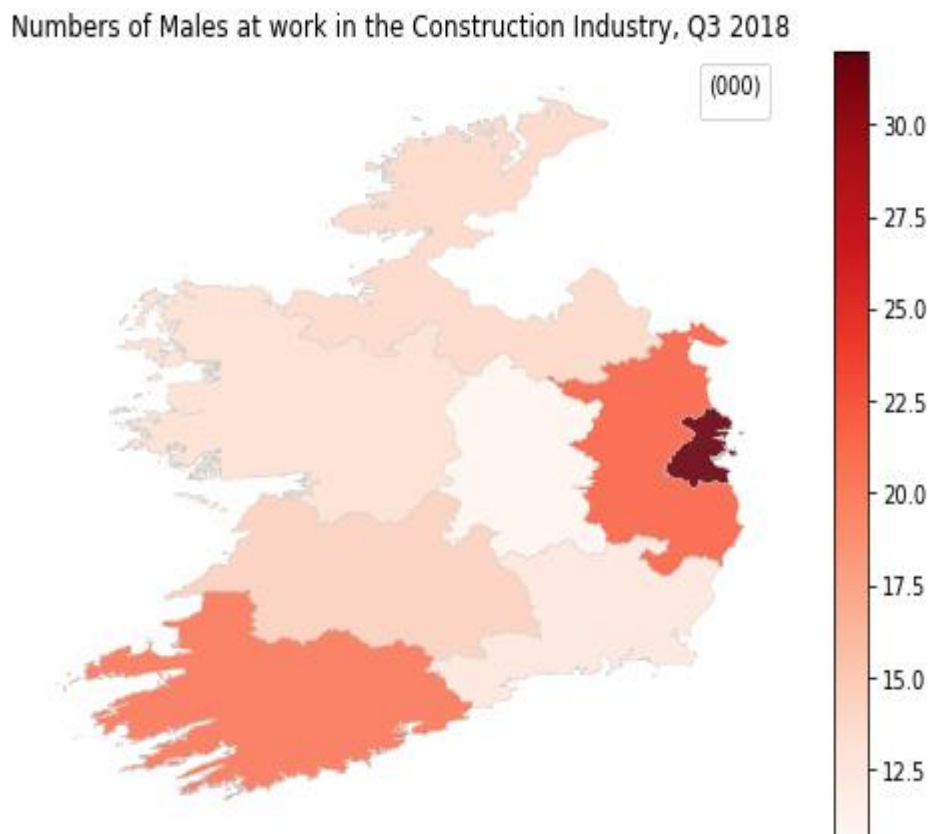


Fig 2: 'Numbers of Males at work in the Construction Industry, Q3 2018'

21 Converting matplotlib plots from Python into interactive Leaflet web maps.

One can convert matplotlib plots from Python into interactive Leaflet web maps using the *mplleaflet* library.

Leaflet does not readily support the use of legends and requires very dark Choropleth map colours. We will return the code above leaving out the legend and changing the colour to 'Dark2_r' to produce a new *ax.figure*.

```
variable = 'Value'  
fig, ax = plt.subplots(1, figsize=(10,6))  
dfbmpp = dfbm.plot(column=variable,  
                    ax=ax,  
                    alpha = 0.8, #intensity of colour of a polygon  
                    cmap="Dark2_r, #colour of the polygon  
                    legend='True',  
                    linewidth=0.3,  
                    edgecolor='0.8',  
                    label = 'Males')  
ax.axis('off')  
ax.set_title('Numbers of Males at work in the Construction Industry, Q3 2018')  
ax.figure
```

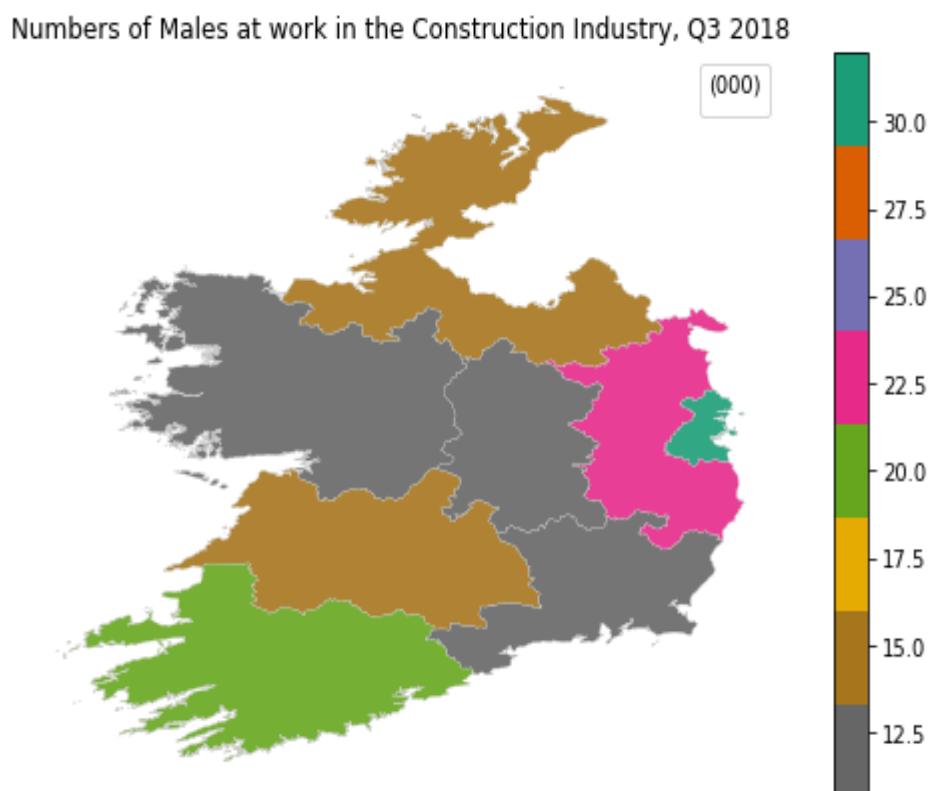


Fig 2: 'Numbers of Males at work in the Construction Industry, Q3 2018'

We now Import the mplleaflet library,

```
import mplleaflet
```

and Output `_map.html` file and display it in your browser.

```
mplleaflet.show(fig=ax.figure)
```



Fig 3: html map, displayed in browser.