

Session 5. A date with EDA: Bring your own data

2026-02-18

Highlights:

This is my mini-reflection. Paragraphs must be indented.
It can contain multiple paragraphs.

“Errors using inadequate data are much less than those using no data at all.”

— Charles Babbage

“Data that is loved tends to survive.”

— Kurt Bollacker

Session outline

- Reading external data
- Hands-on practice
- Some additional notes on working with Quarto / Jupyter

Reminder

The human brain is a highly evolved machine with a specialization in visual recognition of spatial patterns, including shapes, areas, lengths, directions, and colors. This is why well-designed statistical plots are so effective for conveying complex information.

Preliminaries

Load packages. Remember, packages are units of shareable code that augment the functionality of base Python. For this session, the following package/s is/are used:

```
import pandas as pd
import geopandas as gpd
import matplotlib.pyplot as plt

# Set display options to show all rows and columns
pd.set_option('display.max_rows', None)      # Show all rows
pd.set_option('display.max_columns', None)    # Show all columns
pd.set_option('display.width', None)         # Auto-detect width
pd.set_option('display.max_colwidth', None)   # Show full column content
```

We also will utilize some data from the edashop R package. To convert these R data files to Python files, we will use the `reticulate` package:

```
library(edashop) # A Package for a Workshop on Exploratory Data Analysis  
library(reticulate)
```

Reading external data

In this workshop we have used data that was meticulously prepared and documented to make it analysis-ready. But often we need to read data from so-called external sources, that is, files that are not in native R format, including Excel files, csv files, Stata files, shape files, and so on.

The original files of the data sets provided in {edashop} are shared with the package. For example, check the following help file:

```
?i40_index_rank
```

This is one of two data frames created using original files from Honti et al. (2020).

The path to the source CSV file can be obtained in R and then passed to Python:

```
csv_path <- system.file("extdata",  
                        "rankings.csv",  
                        package = "edashop")
```

You can check that this object is just a string with the path to the file:

```
csv_path = r.csv_path  
print(csv_path)
```

```
C:/Users/brdia/AppData/Local/R/cache/R/renv/library/edashop-3dfdfba5/windows/R-4.4/x86_64-w64-mingw32/ed
```

One way to read files in CSV format is with `pandas.read_csv()`:

```
imported_csv = pd.read_csv(csv_path)  
print(imported_csv.head())
```

	Regio	GDPRank	PrometheeRank	RII
0	UKN0	68.0	149.0	NaN
1	UKM9	165.0	193.0	NaN
2	UKM8	26.0	116.0	NaN
3	UKM7	16.0	70.0	NaN
4	UKM6	149.0	107.5	NaN

Once the file is imported it can be saved as a file of type rda for ease of access in the future. For reproducibility purposes, it is highly recommended to keep the source files intact and document any data processing done in a script or Quarto document. For example, here we change the names of the columns in the table we just imported:

```
imported_csv = imported_csv.rename(columns = {  
    'Regio': 'NUTS_ID',  
    'GDPRank': 'gdp_rank',  
    'PrometheeRank': 'promethee_rank',  
    'RII': 'regional_innovation_index'  
})  
print(imported_csv.head())
```

```

NUTS_ID gdp_rank promethee_rank regional_innovation_index
0    UKN0      68.0      149.0                 NaN
1    UKM9     165.0      193.0                 NaN
2    UKM8      26.0      116.0                 NaN
3    UKM7      16.0       70.0                 NaN
4    UKM6     149.0      107.5                 NaN

```

Then we save the file (optional):

```

# Save to a pickle file for faster loading later
imported_csv.to_pickle("external_csv.pkl")

# Or save as CSV
imported_csv.to_csv("external_csv.csv", index = False)

```

This makes it easy to retrieve our prepared data.

An external file in Stata's dat format is also shared:

```
dta_path <- system.file("extdata", "phd_italy.dta", package = "edashop")
```

As before `dta_path` is a string with the path to the file:

```
dta_path = r.dta_path
print(dta_path)
```

```
C:/Users/brdia/AppData/Local/R/cache/R/renv/library/edashop-3dfdfba5/windows/R-4.4/x86_64-w64-mingw32/edashop
```

pandas can read Stata files directly with `read_stata()`:

```
imported_dta = pd.read_stata(dta_path)
print(imported_dta.head())
```

```

      id q01_annodinascita q02_sesso q03_cittadinanzaest q04_provvive \
0  QSBBF760BN           1985      0.0            0.0          RM
1  QS1T4P4F5E           1982      0.0            0.0          CA
2  QSDPZCAF0M           1960      0.0            0.0          FI
3  QS07842WKF           1981      1.0            0.0        ESTERO
4  QS6DGKNLT2           1984      1.0            0.0          RO

      q05_genitorelaureato q06_genitoreaccad q07_genitoreimpr  concludott_cert \
0             1.0            0.0            0.0            1.0
1             1.0            0.0            0.0            1.0
2             0.0            0.0            0.0            0.0
3             1.0            0.0            0.0            0.0
4             1.0            1.0            0.0            1.0

      q08_phd_clean q08a_adj1 q08a_adj2 q08a_adj3 q08a_adj4 q08a_adj5 \
0            1.0      0.0      0.0      0.0      0.0      0.0
1            1.0      0.0      0.0      0.0      0.0      0.0
2            0.0      1.0      0.0      0.0      0.0      0.0
3            0.0      1.0      0.0      0.0      0.0      0.0
4            1.0      0.0      0.0      0.0      0.0      0.0

      q08a_adj6 q08a_adj7 q08a_adj8         struttura code_un \

```

0	1.0	0.0	0.0	Univ. ROMA LA SAPIENZA	26	
1	0.0	0.0	1.0	Univ. CAGLIARI	04	
2	0.0	0.0	0.0	Univ. FIRENZE	10	
3	0.0	0.0	0.0	Univ. PALERMO	20	
4	0.0	1.0	0.0	Univ. ROMA TOR VERGATA	27	
	nome_provincia	nome_regione	q13a_phdgiudizio	q13b_phdgiudizio	\	
0	Roma	Lazio	5.0	4.0		
1	Cagliari	Sardegna	6.0	5.0		
2	Firenze	Toscana	6.0	6.0		
3	Palermo	Sicilia	5.0	6.0		
4	Roma	Lazio	6.0	5.0		
	q13c_phdgiudizio	q13d_phdgiudizio	q13e_phdgiudizio	q13f_phdgiudizio	\	
0	4.0	3.0	4.0	3.0		
1	3.0	4.0	4.0	6.0		
2	3.0	3.0	5.0	NaN		
3	4.0	5.0	6.0	5.0		
4	5.0	5.0	5.0	4.0		
	q13g_phdgiudizio	q13h_phdgiudizio	q13i_phdgiudizio	q14_phdfinpriv	\	
0	1.0	4.0	3.0	0.0		
1	5.0	5.0	3.0	0.0		
2	4.0	NaN	NaN	0.0		
3	6.0	4.0	6.0	0.0		
4	1.0	6.0	5.0	0.0		
	q18_phdimpresa	q21_ricbase	q29_brevettisn	q29a_brevettin	q30_papernaz2	\
0	0.0	100.0	0.0	0	NaN	
1	0.0	100.0	0.0	0	NaN	
2	0.0	100.0	0.0	0	NaN	
3	0.0	10.0	0.0	0	NaN	
4	0.0	50.0	0.0	0	9.0	
	q31_paperint2	q371_lavora	q372_impresta	q372a_impattiva	\	
0	3.0	1.0	0.0	0.0		
1	2.0	1.0	0.0	0.0		
2	NaN	1.0	0.0	0.0		
3	NaN	0.0	0.0	0.0		
4	NaN	1.0	0.0	0.0		
	q373_impabbandono	q375_posizuni	q377_posizauto	q3711_impprov	\	
0	NaN	1.0	0.0			
1	0.0	1.0	0.0			
2	NaN	0.0	0.0			
3	0.0	0.0	0.0			
4	0.0	1.0	1.0			
	q3713_impcorso	q3715_impaddetti	q379_impanni	q53a_uniimp	q53b_uniimp	\
0	0.0	NaN	NaN	NaN	NaN	
1	0.0	NaN	NaN	NaN	NaN	
2	0.0	NaN	NaN	NaN	NaN	
3	0.0	NaN	NaN	6.0	5.0	
4	0.0	NaN	NaN	NaN	NaN	

```

q53c_uniimp  q53d_uniimp  q53e_uniimp  q53f_uniimp  q53g_uniimp  \
0      NaN        NaN        NaN        NaN        NaN
1      NaN        NaN        NaN        NaN        NaN
2      NaN        NaN        NaN        NaN        NaN
3      4.0        6.0        4.0        5.0        6.0
4      NaN        NaN        NaN        NaN        NaN

q53h_uniimp  q53i_uniimp  dimensione  med_school  polytech  geo_n  geo_c  \
0      NaN        NaN        4.0        1.0        0.0        0.0        1.0
1      NaN        NaN        3.0        1.0        0.0        0.0        0.0
2      NaN        NaN        4.0        1.0        0.0        0.0        1.0
3      5.0        6.0        4.0        1.0        0.0        0.0        0.0
4      NaN        NaN        3.0        1.0        0.0        0.0        1.0

geo_s  public_uni  vqr_average  utt_sn2006  utt_mission_d2006  disocc2006  \
0      0.0        1.0        0.99        1.0        1.0        7.188093
1      1.0        1.0        0.78        1.0        1.0        11.048825
2      0.0        1.0        1.05        1.0        1.0        4.425923
3      1.0        1.0        0.80        1.0        1.0        18.300277
4      0.0        1.0        0.83        1.0        1.0        7.188093

area1  area2  area3  area4  area5  area6  area7  area8  area9  area10  \
0      0        0        1        0        0        0        0        0        0
1      0        0        0        0        0        0        0        1        0
2      0        0        0        0        0        0        0        0        1
3      0        0        0        0        0        0        0        0        1
4      0        0        0        0        0        0        0        0        0

area11  area12  area13  area14  q378_imptipo_clean  so_reg2006  spinoff0506
0      0        0        0        0                    1.0        0.0
1      0        0        0        0                    1.0        3.0
2      0        0        0        0                    0.0        3.0
3      0        0        0        0                    0.0        0.0
4      0        1        0        0                    1.0        2.0

```

Package `pandas` also has utilities to read Excel files (requires `openpyxl` or `xlrd`):

```
xlsx_path <- system.file("extdata", "italy-nuts-codes.xlsx", package = "edashop")
```

```

xlsx_path = r.xlsx_path
imported_xlsx = pd.read_excel(xlsx_path)
print(imported_xlsx.head())

```

	NUTS_3	N3_Code	NUTS_2	N2_Code	NUTS_1	N1_Code
0	Torino	ITC11	Piemonte	ITC1	Northwest Italy	ITC
1	Vercelli	ITC12	Piemonte	ITC1	Northwest Italy	ITC
2	Biella	ITC13	Piemonte	ITC1	Northwest Italy	ITC
3	Verbano-Cusio-Ossola	ITC14	Piemonte	ITC1	Northwest Italy	ITC
4	Novara	ITC15	Piemonte	ITC1	Northwest Italy	ITC

As a last example, we use `geopandas` to read a shapefile (which requires `geopandas` and its dependencies `fiona`, `shapely`, etc.):

```
shp_path <- system.file("extdata", "nuts2.shp", package = "edashop")
```

```
shp_path = r.shp_path
imported_shp = gpd.read_file(shp_path)
print(imported_shp.head())
```

NUTS_ID	LEVL_CODE	CNTR_CODE		NAME_LATN		NUTS_NAME	FID	\
0	AT11	2	AT	Burgenland (AT)	Burgenland (AT)	Burgenland (AT)	AT11	
1	AT12	2	AT	Niederösterreich	Niederösterreich	Niederösterreich	AT12	
2	AT13	2	AT	Wien		Wien	AT13	
3	AT21	2	AT	Kärnten		Kärnten	AT21	
4	AT22	2	AT	Steiermark		Steiermark	AT22	

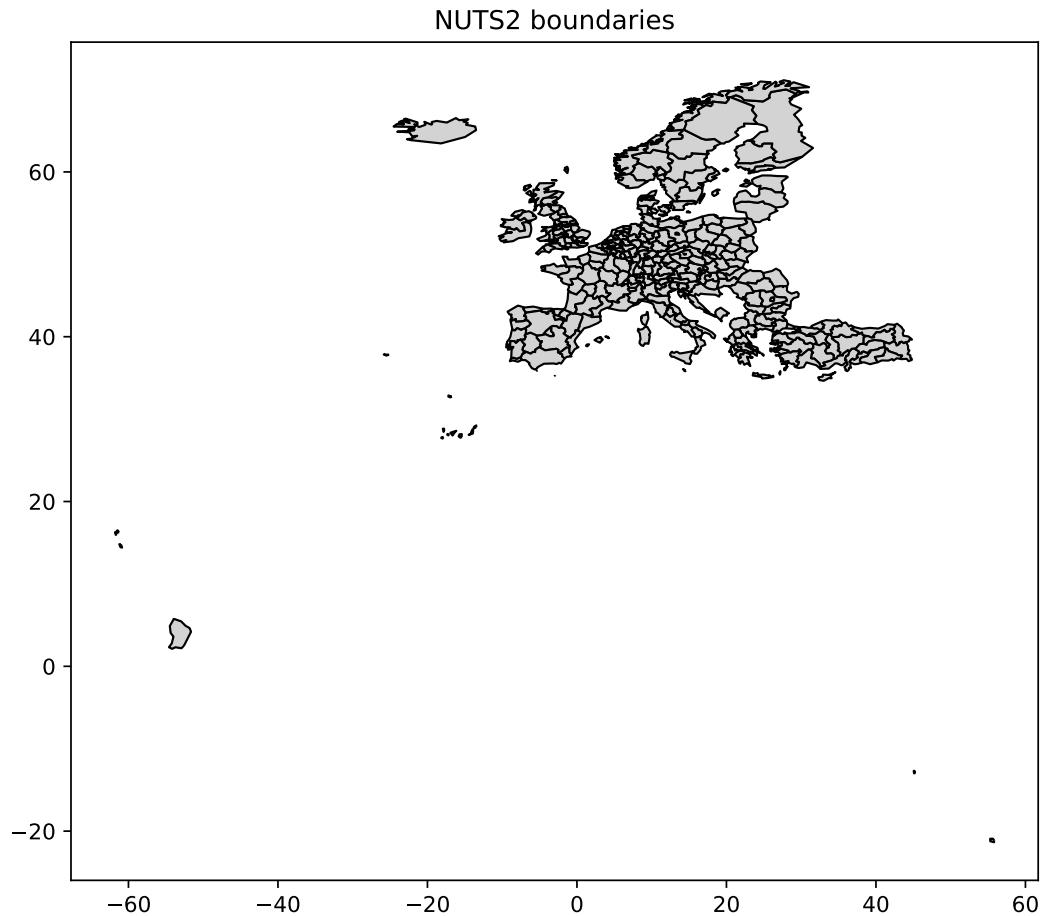
```
0
1 POLYGON ((15.54200 48.90800, 15.75400 48.85200, 16.94000 48.61700, 16.95000 48.53600, 16.85100 48.43300, 16.75400 48.35200, 15.54200 48.90800))
2
3
4 POLYGON ((16.17200 47.85000, 16.25000 47.85000, 16.25000 47.75000, 16.17200 47.75000, 16.17200 47.85000))
```

Hands-on practice

If you brought a data file of your own in an external format, try reading it here. Then, you can play with it.

As an example, I am going to plot the shapefile that we just read. `geopandas` objects can be plotted directly with `.plot()`:

```
fig, ax = plt.subplots(figsize = (8, 8))
imported_shp.plot(ax = ax, color = 'lightgrey', edgecolor = 'black')
ax.set_title("NUTS2 boundaries")
plt.show()
```



Next we can join the geometry of the boundaries to the rankings of industrial readiness, and convert to geopandas GeoDataFrame:

```
# Merge the CSV data with the shapefile on NUTS_ID
i40_rankings = imported_shp.merge(imported_csv, on = 'NUTS_ID', how = 'left')

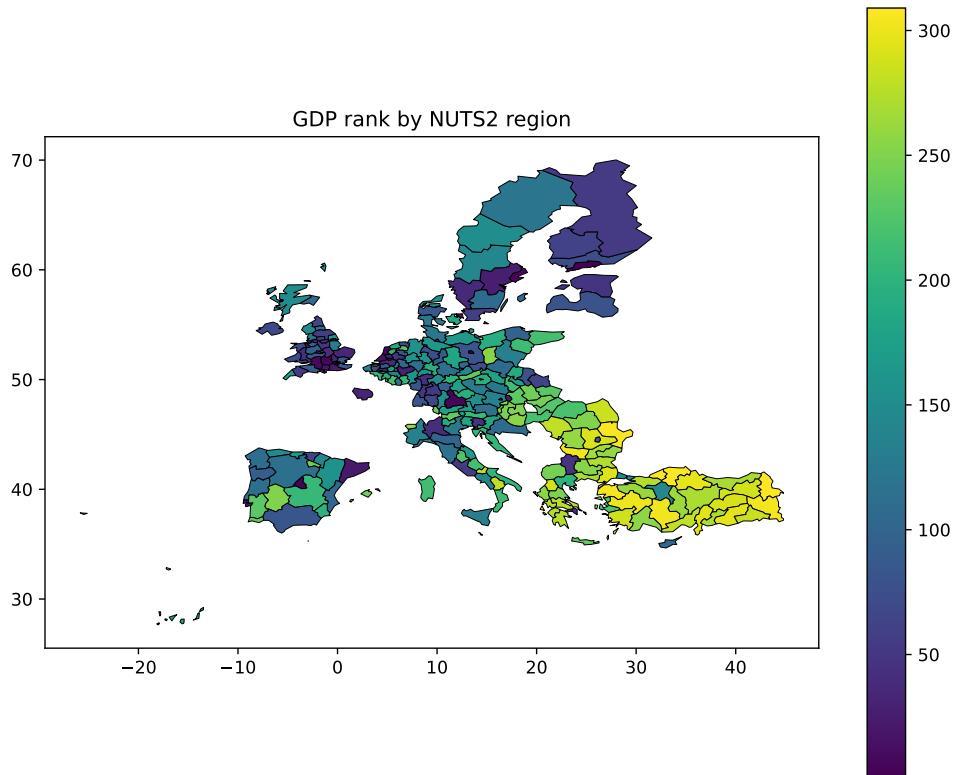
# Convert to GeoDataFrame
i40_rankings = gpd.GeoDataFrame(i40_rankings, geometry = 'geometry')
```

Now we can map something else, for instance, by encoding it to the color of the polygons (choropleth map):

```

fig, ax = plt.subplots(figsize = (10, 8))
i40_rankings.plot(column = 'gdp_rank', cmap = 'viridis', legend = True,
                    edgecolor = 'black', linewidth = 0.3, ax = ax)
ax.set_title("GDP rank by NUTS2 region")
plt.show()

```



A great resource to learn about working with spatial data in Python is the Geopandas documentation and the book Geographic Data Science with Python by Rey, Arribas-Bel, and Wolf.

Some additional notes on working with Rmarkdown

Quarto is a great invention. It implements principles of literate programming and enforces good discipline when writing and documenting not only code but also analysis processes.

Here is some additional information of value. The chunks of code that we use to communicate with the computer accept a number of options that stipulate the behavior of the chunk. Chunk options are written inside the curly brackets. They can also be named:

```

# Option eval controls whether the chunk is evaluated (i.e., run).
# This chunk of code can only be run manually, by clicking the play icon,

```

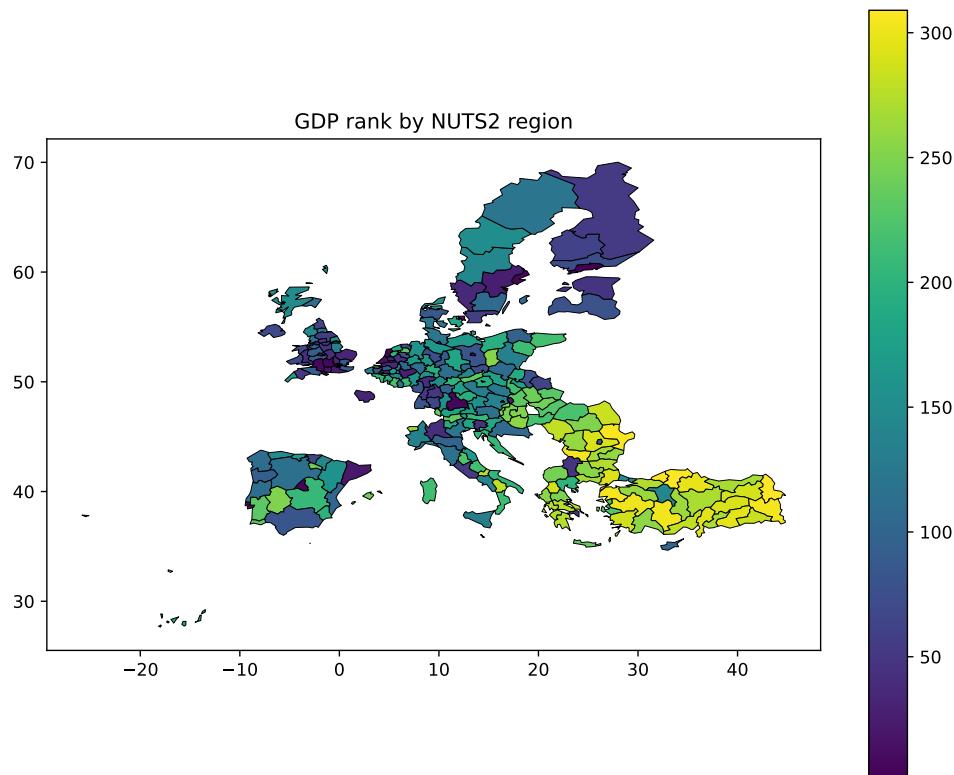
```
# but will be ignored when running the document or when knitting  
help(pd.read_csv)
```

Some chunks, like eval, control whether the code is run. Others, like echo, control whether the chunk itself is printed in the output document when knitting:

This chunk of code will not be printed in the output. The result of running the code will

Chunk include is used to include code and results of the code, or to suppress all. For example, when reading files or loading data, we may not wish to have those things printed in the output document:

For example, let us say that we wish to show a plot in a paper or report, but we do not need to show the code:



Quarto is a great format for creating reproducible research reports and papers. These are some examples (originally in R, but the same principles apply in Python with Quarto):

- Paez, Antonio, et al. “A spatio-temporal analysis of the environmental correlates of COVID-19 incidence in Spain.” Geographical analysis 53.3 (2021): 397-421. The repository with the reproducible document is here: <https://github.com/paezha/covid19-environmental-correlates>
- Paez A, Higgins CD (2021) The Accessibility Implications of a Pilot COVID-19 Vaccination Program in Hamilton, Ontario. Findings (doi.org/10.32866/001c.24082). The repository with the reproducible document is here: <https://github.com/paezha/Accessibility-Pharmacies-Hamilton-Vaccines>

- Higgins, C.D., Páez, A., Kim, G., Wang, J. (2021) Changes in accessibility to emergency and community food services during COVID-19 and implications for low income populations in Hamilton, Ontario. Social Science and Medicine, 291:114442 (doi.org/10.1016/j.socscimed.2021.114442). The repository with the reproducible document is here: <https://github.com/paezha/Accessibility-Food-Banks-Hamilton>

For your practice, you can create a new **Quarto** document and write a small exploratory data analysis report using your own data set or one of the data sets provided.

Happy analysis!

Practice

1. Create a new Quarto document (**File > New File > Quarto Document** in RStudio, or use the command line with `quarto create project`). Choose a template of your liking.
2. Transfer your data analysis exercise to the template and write a small report.