



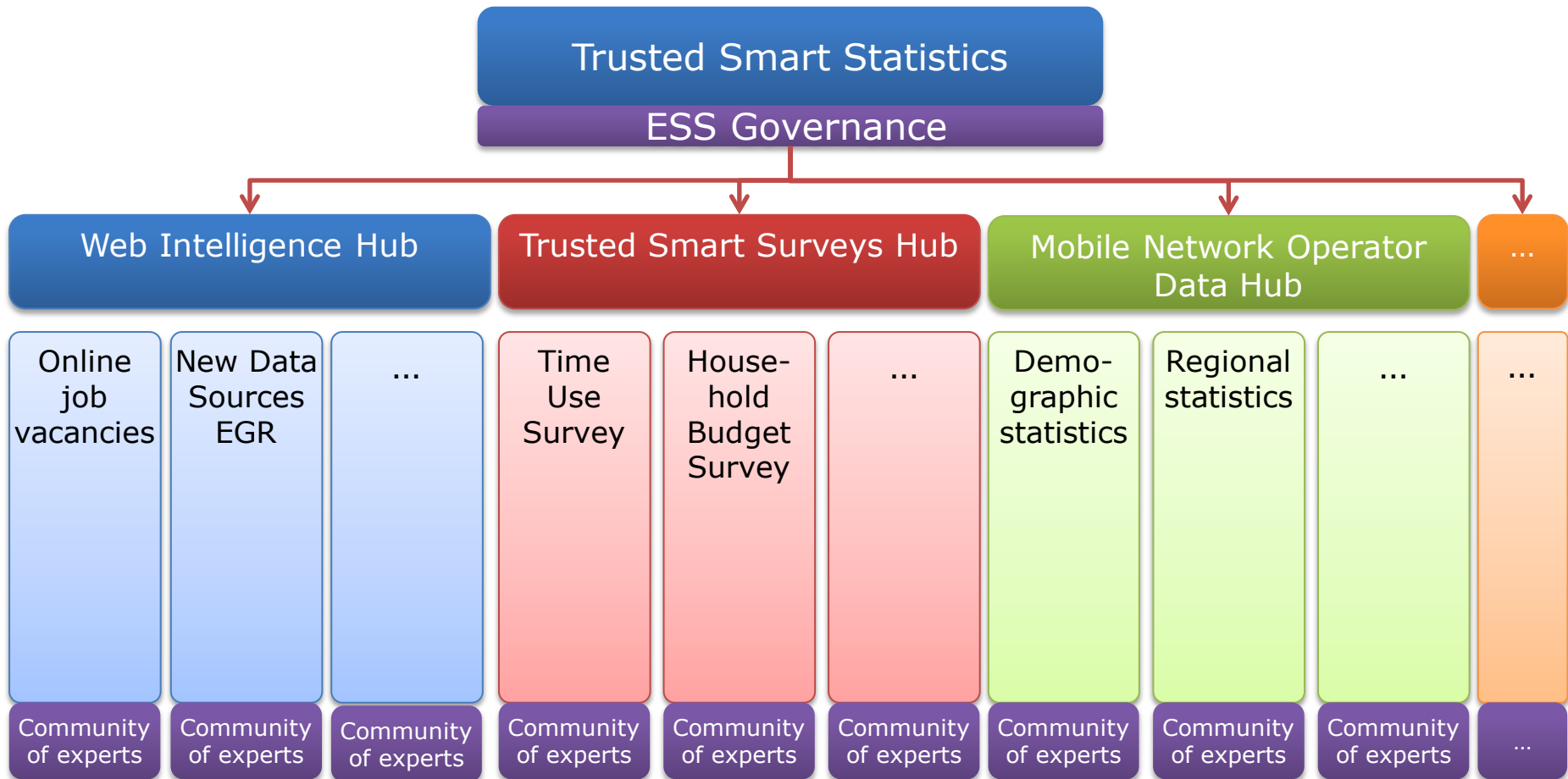
Trusted Smart Statistics - Web Intelligence Hub

Update on developments on the use of OJA in official statistics,

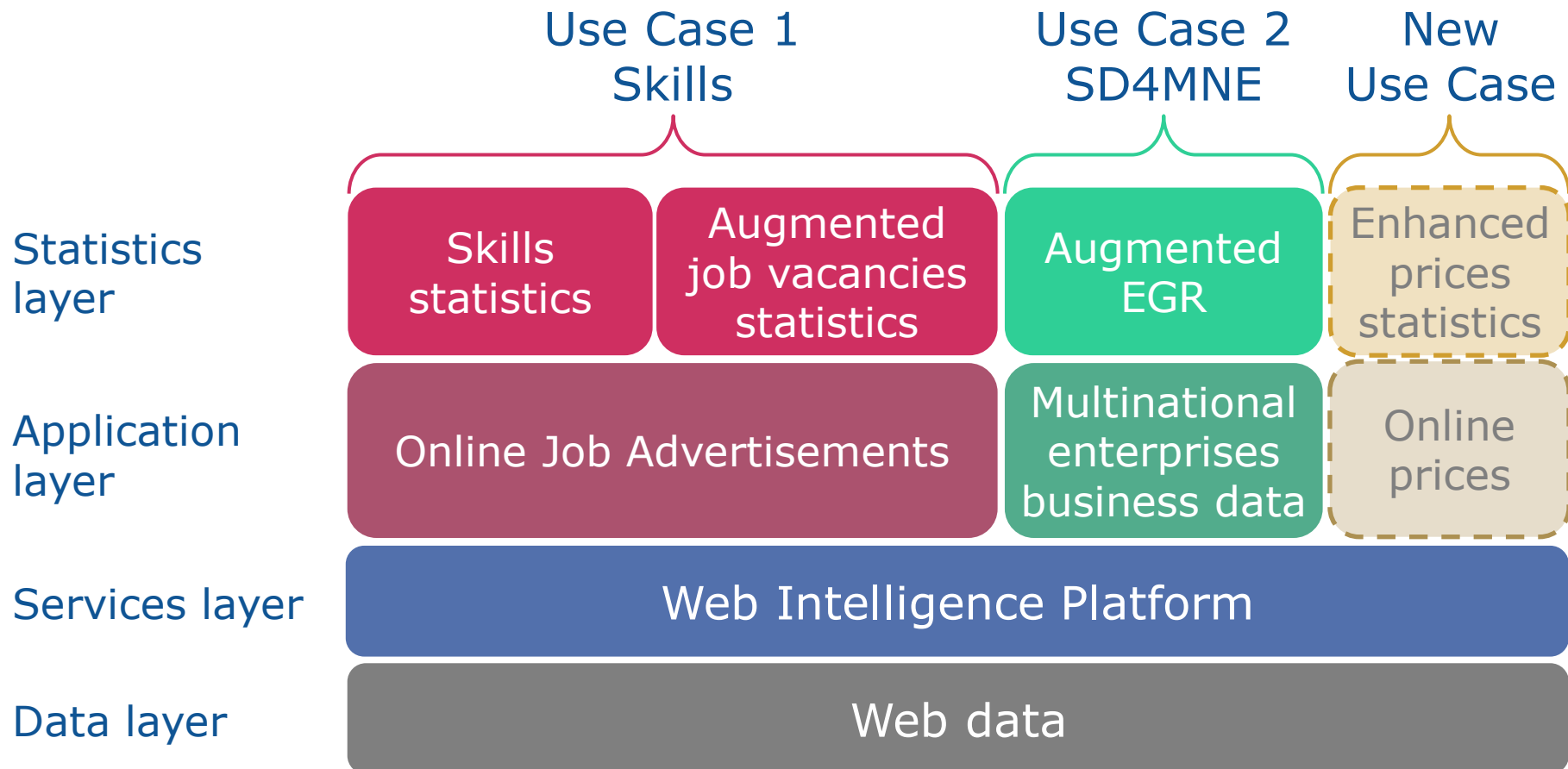
Big Data for Labour Market Intelligence

6 December 2022

Trusted Smart Statistics



Web Intelligence Hub



Web Intelligence Hub

- Know more about Trusted Smart Statistics and the Web Intelligence Hub:
 - **YouTube**
 - **ETF Open Space channel**
 - **Case study: Eurostat and Smart Statistics, by Fernando Reis (Eurostat), 8 June 2021, EN**
 - **<https://youtu.be/bAUT52D1MpM>**

Updates on OJA

- Labour market concentration
 - **Competition in urban hiring markets: evidence from online job advertisements**
 - **<https://ec.europa.eu/eurostat/web/products-statistical-working-papers/-/ks-01-21-430>**
- Validation and continuous improvement
- Linking to business register and NACE evaluation
- **OJA gold standard for occupation**
- **OJA time-series**



OJA GOLD STANDARD FOR OCCUPATION

Purpose of collecting labelled data

1

Monitor and improve the quality of the OJA data production

- Evaluation of classifiers → perform quality checks of the **data classified**
- Measurement of the accuracy of the **classifiers**
- Improvement of “ontologies” / dictionaries

2

Provide annotated OJA data



HOW?

- Gold standard: benchmarking annotators
- Monitor quality of the automatic classification process
- Train Machine Learning models

OJA-NLP dataflow

- **OJA-NLP data flow (released yearly):**
 - **Set of raw data extracted for a sample of ads:**
 - Job description full text
 - Job title
 - Classified data
 - Additional information used to classify (matched token, dictionary terms...)
 - **Stratified sample:**
 - for each **language** (25 languages, excluding very small languages,
 - at least **50 ads for each category**),
 - stratify by: occupation, contract, salary, working hours, education, economic activity, experience.
 - **Period covered: oct 2021 – Jan 2022**

Data annotation: Definition

- **Process of labelling the data → various formats: text, video or images.**
- **Refers to the human classification of raw data → annotated (labelled) data.**
- **Enrich a linguistic data collection with annotations (NLP, Semantic Web technologies, LOD Linked Open Data Cloud, ...) → machine learning, quality purposes**
 - *Corpora* - datasets of natural language
 - *Annotated corpus* - single set of data annotated with the same specification:
 1. audit samples;
 2. train, validation and testing of machine learning models;
 3. estimate the precision of the classifiers.

Data annotation: Quality

- Quality of annotated OJAs
 - **Measurements:**
 - ✓ **Accuracy** → how close a label is to the truth.
 - ✓ **Consistency** → degree to which multiple annotations on various training items agree with one another.
 - **Standard methods to assure annotated data of high quality:**
 - ✓ Use of gold standards, **consensus**, and **auditing**.

Data annotation: Gold standard

- Quality of data annotation
 - **Use of gold standards (or Benchmarks) method**
 - ✓ Gold standard: sample carefully built by experts with a very high precision.
 - ✓ Expensive: require the intensive team work of highly qualified experts.
 - ✓ Small in relative terms and are normally not be sufficient for training ML models.
 - ✓ Ideal for **benchmarking annotators** used to obtain other types of annotated data.

Gold standard for occupation of OJA data

- ❑ Prepare samples for annotation
 - Extract reproducible samples
 - (country, language) pair

- ❑ Variable to annotate: 'Occupation 4D' (and lower levels if necessary)

- ❑ Variables included in the sample:
 - *oja_id,*
 - *title,*
 - *full job description,*
 - *occupation,*
 - *match_token,*
 - *dictionary term*

Specific guidelines: How to label OJA data?

Labels used in the annotation process:

1. Labels = ISCO-08 labels (for occupation projects)
2. Metadata labels

- ✓ **Correct**
- ✓ **Incorrect**
- ✓ **Comment**
- ✓ **No reference to occupation in the description**
- ✓ **Impossible to classify at 4th level**
- ✓ **Wrong language**
- ✓ **Not a job ad**
- ✓ **Job description missing**
- ✓ **Multiple ISCO labels**
- ✓ **Misspelling**

First 4 countries: Preliminary results

- Nice set of countries-languages:
 - **AT, IT, SI, BG**
 - **It covers Latin, Germanic and Slavic languages**
 - **Results are broadly confirmed for a few more countries for which we recently received data**
 - **But they are still preliminary: typically one annotator per country, and dealing with a very complex classification system**
- Aggregate accuracy rate (correct/[correct+incorrect]) at ISCO-4d is 44.9%
 - **Not bad for a classification with 400 categories (a random assigning would score around 0.25%)**
 - **This is stratified, so rarer and more difficult occupations get more weight**
 - **The accuracy rate increases by some 10-15pp when weighting by frequency**



OJA TIME-SERIES

Relevance of OJA time series analysis

- Answering many relevant question: What was the impact of covid on the labour market? How is skill demand changing? Etc.
- Offering external validation to OJA indicators by testing its relationship to other relevant series
- Augmenting other indicators through nowcasting
 - **OJA data are more timely than other series, so they could be used for nowcasting**
 - **E.g. nowcasting of job vacancies in the latest quarter**
- Flows or stocks of OJA time series could be more relevant, depending on the type of analysis

Issues with OJA time series analysis

- **OJA data ingestion happens through scraping or through API download and its regularity is less than ideal.**
- **Scrapers can fail for reasons like changes in website structure or temporary overload;**
- **The market for online ads changes => need to update the list of data sources**
- **This leads to the following three problems / solutions:**
 1. Missing data -> Estimated through a survival analysis model
 2. 0-to-N peaks -> Even allocation of scraped ads to days within the scraping interval
 3. Incomparable source sets -> Chaining

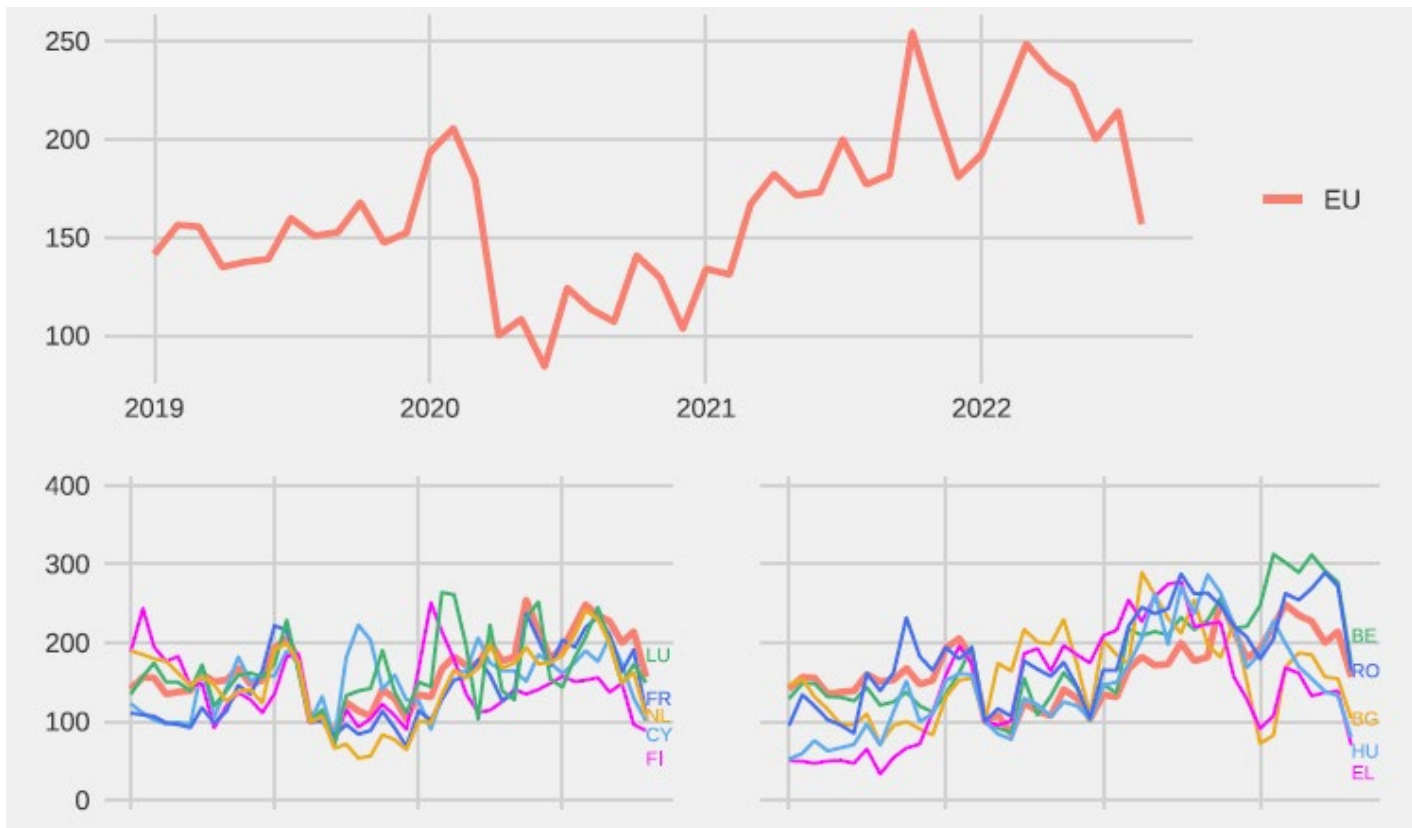
R implementation

- Queries are executed in SQL from the cloud by using the DBI and noctua R packages (Ascheri et al 2022, https://www.revistadestatistica.ro/wp-content/uploads/2022/03/RRS-1_2022_1.pdf)
- Processing is slow: 200m ads divided into some 30 countries X 40 occupations X 400 sources X 1600 dates \approx 800m cells
- The script is divided into 10 steps loading/saving intermediary input/output
- Other metadata, like discarded ads at each step and a “EU tracker” projecting the EU total at each step, are saved at each step

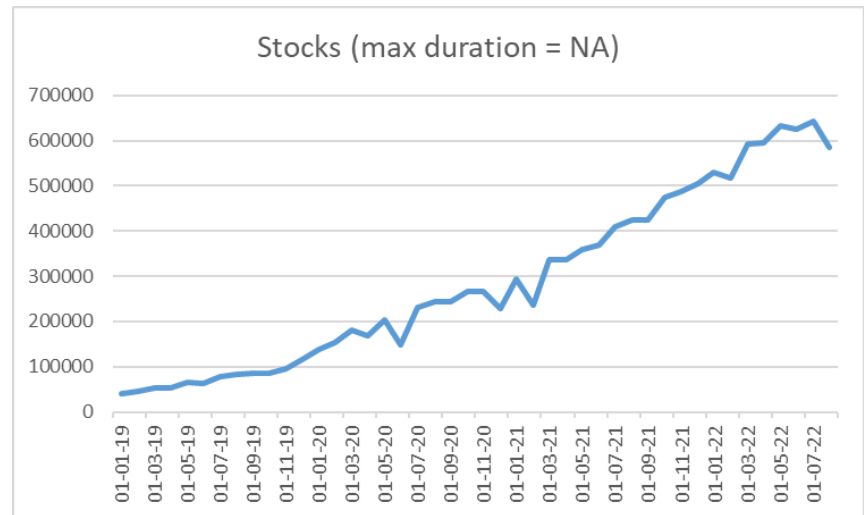
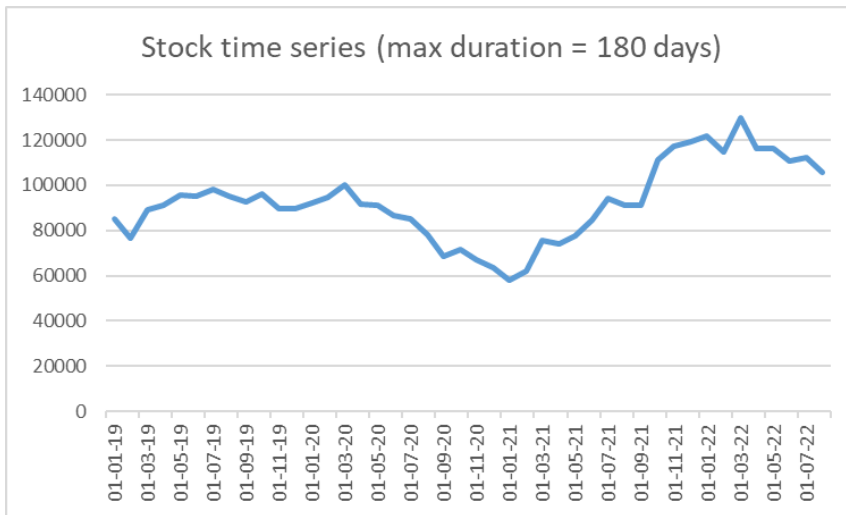
Implementation: Users

- A table with a set of weights by occupation / country / month will be made available to users
 - **So they can replicate easily the OJA time series and do their own analysis**
- The weight for each cell is calculated as the number of posted ads estimated through the time series method and the number of ads found in the OJA dataset
 - $w_{o,c,m} = N_{o,c,m}^{time-series} / N_{o,c,m}^{raw}$
- Users will only need to merge their OJA queries with the time-series-weight table and will be good to go

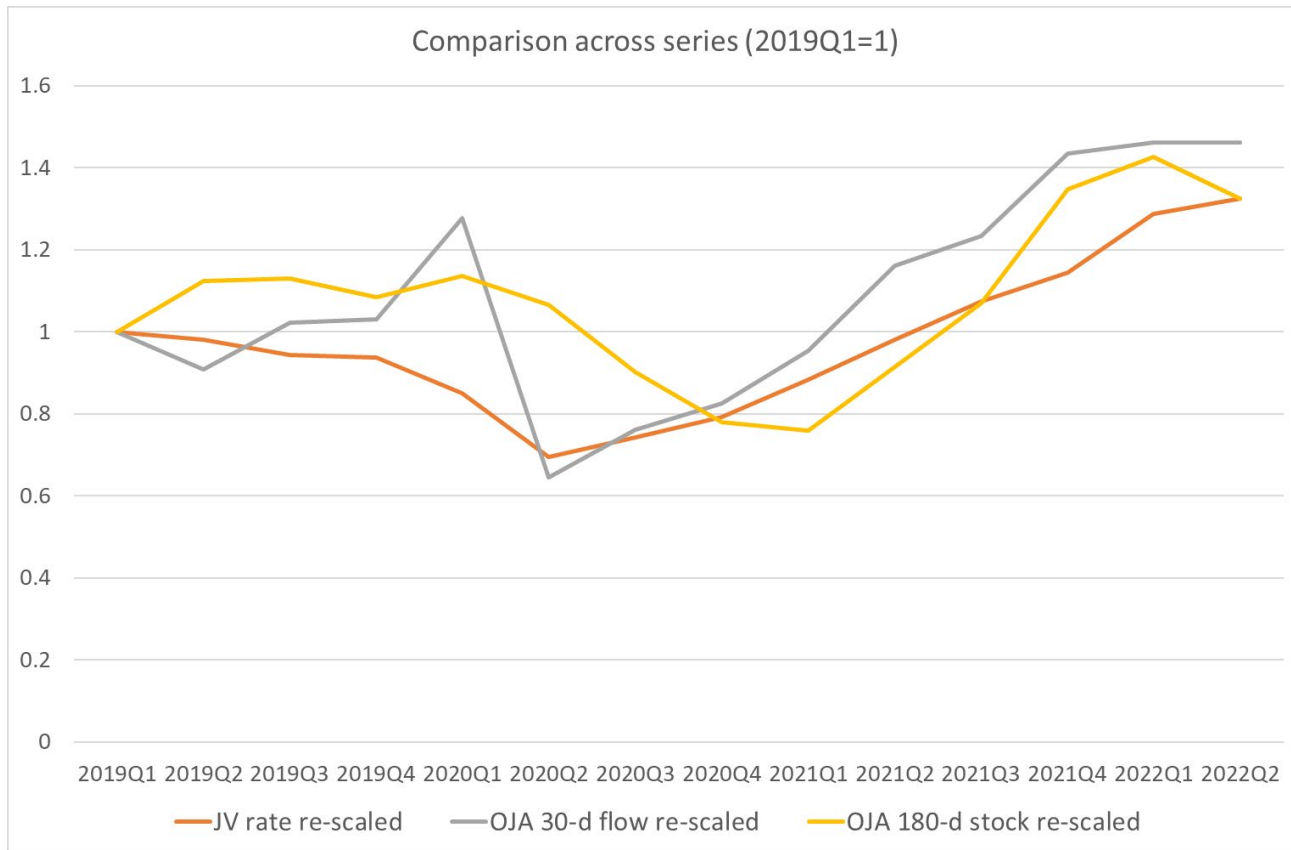
Country series: Flows



Stocks



Comparison of JV and OJA series



Results will be presented at NTTS

- New Technologies and techniques for Statistics 2023
- https://ec.europa.eu/eurostat/cros/content/NTTS2023_en



Thank you for your attention

Fernando Reis

Eurostat



fernando.reis@ec.europa.eu



<https://github.com/reisfe/>



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