



## WELCOME.







# Statistics and Internet CPS70 Quality Aspects of Web Data based on the Experiences of the ESSnet Trusted Smart Statistics – Web Intelligence Network

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#### **Outline/Content**



- ESSnet Trusted Smart Statistics
   Web Intelligence Network (WIN)
- "Minimal Guidelines and Recommendations for Implementation" – Structure of document and examples
- Ongoing quality work in the ESSnet:
  - Landscaping and selection of websites
  - Quality measures for hierarchical classifications

## ESSnet Web Intelligence Network (WIN)



ESSnet WIN (2021 – 2025):

Consortium of 17 organizations from 14 European countries

4 Work Packages (WP)

- WP1 Coordination, support an dissemination
- WP2 Online Job Advertisements (OJA)
   Online Based Enterprise Characteristics (OBEC)
- WP3 New use cases: real estate, construction activities, prices of household appliances, hotel prices, business register enhancement
- WP4 Methodology and Quality

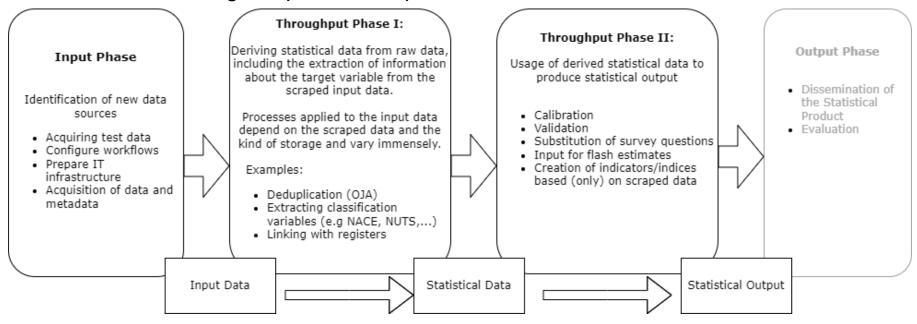
More information about WIN including a Blog:

https://cros-legacy.ec.europa.eu/content/work-package-3-%E2%80%93-new-use-cases\_en

#### "Minimal Guidelines and Recommendations for Implementation"



- Based on existing Quality Guidelines from ESSNet Big Data II
- Structured along the production process:



https://cros-legacy.ec.europa.eu/content/deliverable-41-minimal-guidelines-and-recommendations-implementation\_en





# Guidelines for Throughput Phase I include the following **Quality Aspects**:

- Coverage
- Comparability over time
- Measurement errors
- Model errors (from raw data to statistical data)
- Processing errors (from raw data to statistical data)

## **Examples for Quality Guidelines**



#### **Guidelines with respect to Coverage and Representativeness**

- Try to estimate the population size and compare with traditional data.
  For example, when you are scraping enterprise characteristics, try to
  count the number of websites that are accessible and can be used for
  web scraping. Compare this number with the data from your business
  register.
- Make a pilot web scraping to assess what information is included on the websites.
  - Check if specific information, e.g. territorial unit or industrial sector, can be extracted from the website. When information on the website is limited, it is also not very likely to monitor enterprise activity (e.g., innovations in enterprises) on the website.

## **Examples for Quality Guidelines**



#### **Guidelines with respect to Comparability over Time**

- Check if the modification/update date can be extracted from the website.
- When web-scraping specific information from the website (e.g. job vacancies), try to extract the date of publishing this information.
- When the website is not up to date it is unlikely to detect enterprise activity in longer time series.

#### Ongoing quality work of WP4: Landscaping



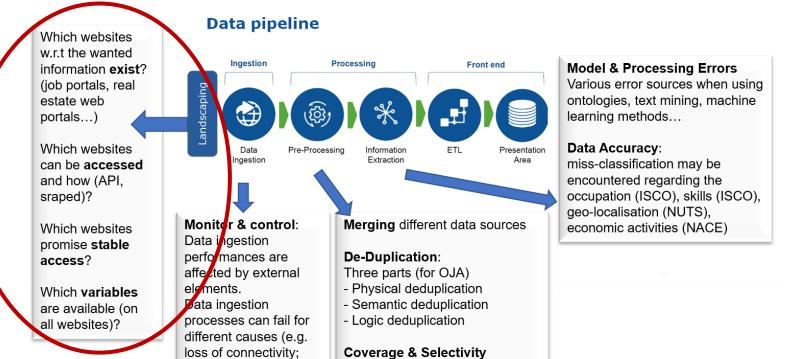


Figure: Data pipeline and respective quality aspects for OJA, WIH

internal server errors

from job portals:

websites changes)

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Penetration of OJA markets

and may change over time

varies in and across countries

## Ongoing quality work: Landscaping



Definition: Landscaping refers to the cataloguing and measurement of all web-based data sources relevant for the topic of interest.

The effort of landscaping varies depending on the topic of interest:

- All needed data might be available on one website Example: satellite data
- The great extent of existing websites and the impossibility to scrape and combine them all makes it necessary to **select websites**Examples: online job advertisements, real estate prices or price statistics
- All websites w.r.t. topic of interest should be scraped, combination of ingested information is possible Example: enterprise characteristics

## Ongoing quality work: Selection of websites



Which websites to scrape?

- -> Most important ones? Highest quality?
- -> **Score** is needed

Three groups of information to take into account:

- Information from the website (stop criteria, mandatory variables, optional variables..)
- Information about the website (e.g. market share, rank of Google search, coverage of niche markets, reliable owner of website,...)
- Experience (test scraping, prior rounds of scraping)

#### Ongoing quality work: Selection of websites



#### Course of action:

- Decide which groups of information and which criteria to take into account
- Choose a model to incorporate all selected criteria to calculate a score
- Calculate score and rank all respective websites
- Scrape the best-ranked websites
- Document each step and re-evaluate after some time

#### Examples

- Members of WP3 "New use cases" agreed on a score
   assessed real-estate
   based only on information from website
   https://cros-legacy.ec.europa.eu/content/wp3-deliverable-31-wp3-1st-interim-technical-report-20220330 en
- Eurostat's score for ranking OJA websites inlcuded also metainformation and expertise from country experts

Table 2.1.1-3: Assessed real estate portals

Web portal	Score (maximum = 100)
clever-immobilien.de	83
sparkasse.de	83
Immmobase.de	80
hermann-immobilien.de	76
bonava.de	76
ohne-makler.net	73
1a-immobilienmarkt.de	0
de.trovit.com	0
deinneueszuhause.de	0
immo4trans.de	0
ebay-kleinanzeigen.de	0
immobilien.de	0
immobilo.de	0
immonet.de	0
wohnen-in-hessen.de	0
kip.net	0

Table from Del.3\_1, UC1, Score for assessed real-estate portals for Germany

#### Ongoing quality work: Hierarchical classifications



Which websites w.r.t the wanted information exist? (job portals, real estate web portals...)

Which websites can be **accessed** and how (API, sraped)?

Which websites promise stable access?

Which **variables** are available (on all websites)?

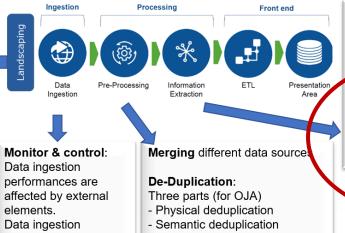
#### Data pipeline

processes can fail for

different causes (e.g. loss of connectivity:

internal server errors

from job portals; websites changes)



Model & Processing Errors

Various error sources when using ontologies, text mining, machine learning methods

Data Accuracy:

miss-classification may be encountered regarding the occupation (ISCO), skills (ISCO), geo-localisation (NUTS), economic activities (NACE)

Penetration of OJA markets varies in and across countries and may change over time

Logic deduplication

Coverage & Selectivity

#### Accuracy for hierarchical classifications



Extracting hierarchical classification variables from scraped information is an essential process step. Typical classifications: NACE, NUTS, ISCO ...

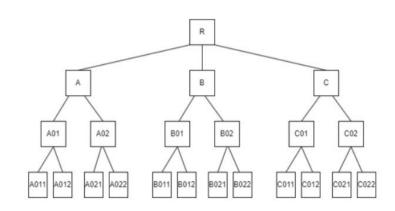
Traditionally, **accuracy** is measured for **flat classifications**: predicted value is true or false

-> Need for an evaluation measure, which takes into account how close the predicted and the true value are!

## Overview of recommended evaluation measures:

https://cros-legacy.ec.europa.eu/content/issue-13-data-accuracy-hierarchical-classification\_en

Hierarchical classification







# THANK YOU.

