

Quality for new data sources: progress, challenges and directions of work for the European Statistical System

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Unit A5 – Methodology; Innovation in Official Statistics

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progress, challenges and directions of work
for the European Statistical System**

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Quality in OS - What?



Institutional environment

- Principe 1 : Professional Independence
- Principe 1bis : Coordination and cooperation
- Principe 2 : Mandate for Data Collection and Access to Data
- Principe 3 : Adequacy of Resources
- Principe 4 : Commitment to Quality
- Principe 5 : Statistical Confidentiality
- Principe 6 : Impartiality and Objectivity

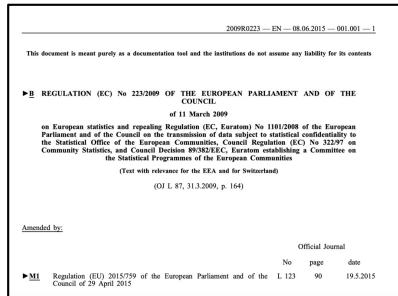
Statistical processes

- Principe 7 : Sound Methodology
- Principe 8 : Appropriate Statistical Procedures
- Principe 9 : Non – excessive Burden on Respondents
- Principe 10 : Cost effectiveness

Statistical output

- Principe 11 : Relevance
- Principe 12 : Accuracy and Reliability
- Principe 13 : Timeless and Punctuality
- Principe 14 : Coherence and Comparability
- Principe 15 : Accessibility

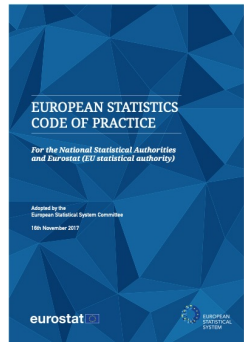
and Quality in OS evolve ...



Regulation 223/2009 on European Statistics

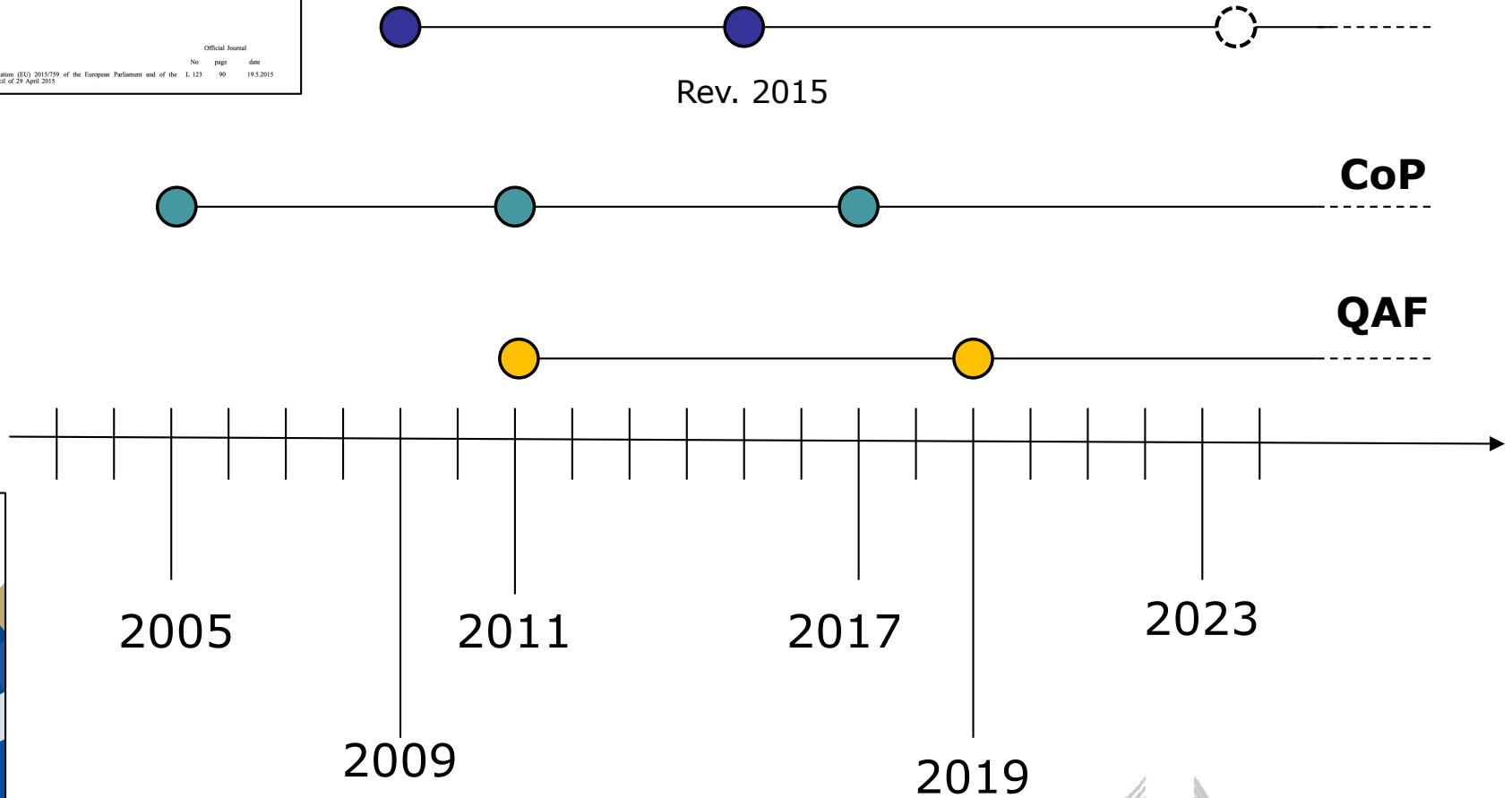
Proposal
Rev. 2023

Rev. 2015



CoP

QAF



Quality in OS - Why?

- *OS influence opinions, decisions ...*
 - ... by policymakers, citizens, businesses, researchers...
- **Quality** as differentiator of OS vs. other stats
 - OS ≠ commercial statistics, other public statistics
 - OS ≠ "experimental statistics"
 - One-off case study or short series, **no commitment** to regular production
 - Partial fulfillment of quality criteria, no full compliance
- **Official Statistics = Quality Statistics**

QS

New data sources in OS: Opportunities...

Institutional environment
Principe 1 : Professional Independence.....
Principe 1bis : Coordination and cooperation.....
Principe 2 : Mandate for Data Collection and Access to Data .
Principe 3 : Adequacy of Resources
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Statistical processes
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Principe 10 : Cost effectiveness.....
Statistical output
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The potential gains

More, better, richer, timelier
statistics than what would be possible or feasible without new data sources

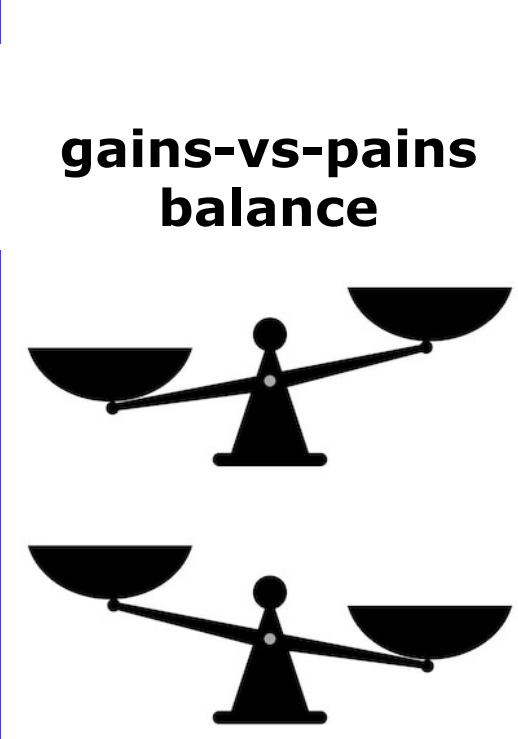
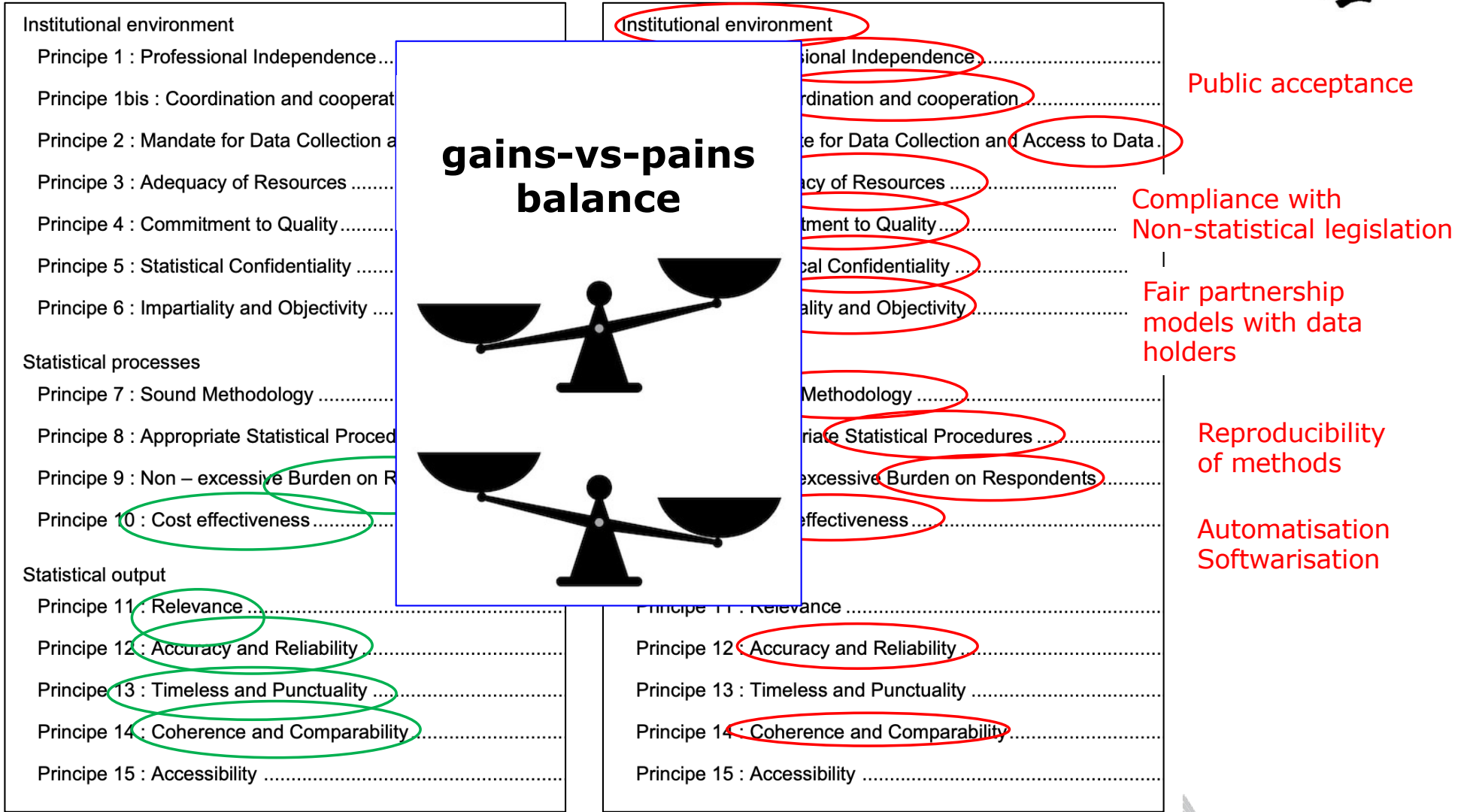
augmenting, not replacing
OS based on traditional data

New data sources in OS: Challenges...

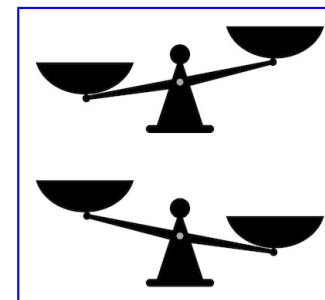
The actual pains

Institutional environment	
Principe 1 : Professional Independence	Public acceptance
Principe 1bis : Coordination and cooperation	
Principe 2 : Mandate for Data Collection and Access to Data	
Principe 3 : Adequacy of Resources	Compliance with Non-statistical legislation
Principe 4 : Commitment to Quality	
Principe 5 : Statistical Confidentiality	
Principe 6 : Impartiality and Objectivity	Fair partnership models with data holders
Statistical processes	
Principe 7 : Sound Methodology	Reproducibility of methods
Principe 8 : Appropriate Statistical Procedures	
Principe 9 : Non – excessive Burden on Respondents	
Principe 10 : Cost effectiveness	Automatisation Softwarisation
Statistical output	
Principe 11 : Relevance	
Principe 12 : Accuracy and Reliability	
Principe 13 : Timeless and Punctuality	
Principe 14 : Coherence and Comparability	
Principe 15 : Accessibility	

New data sources in OS – which ones?



Selection of data sources



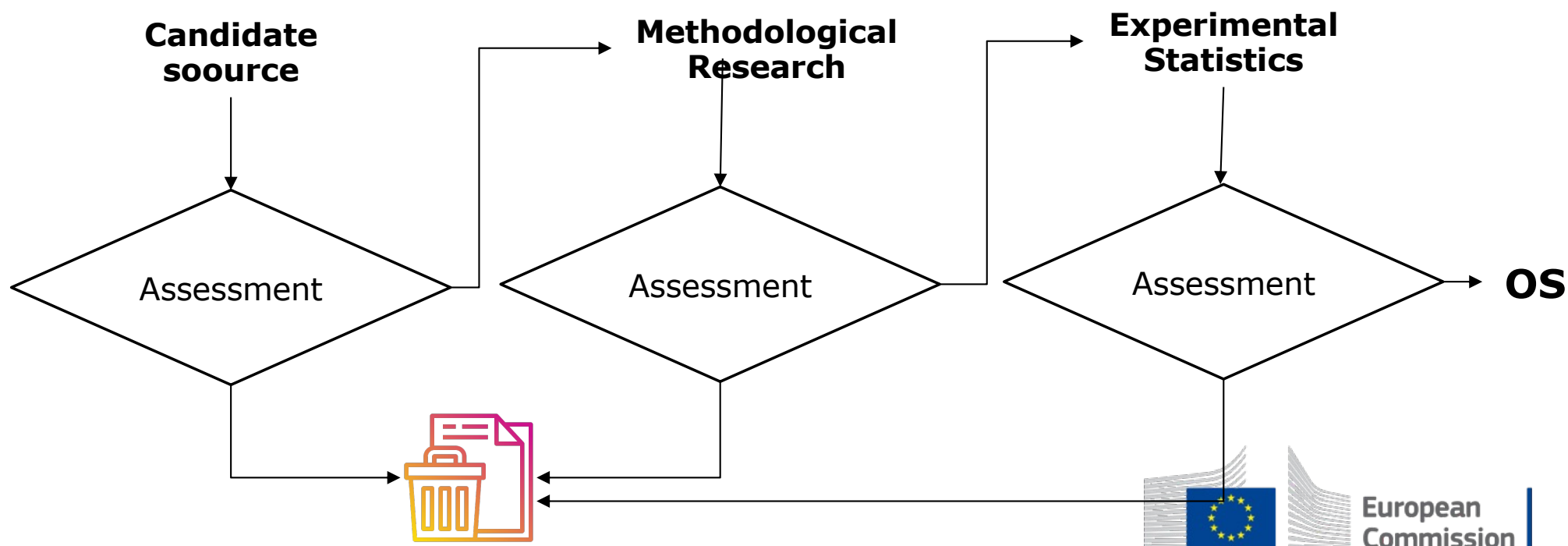
Be selective !

*New non-statistical data sources "may be reused for statistics" ..
but also **may NOT!***

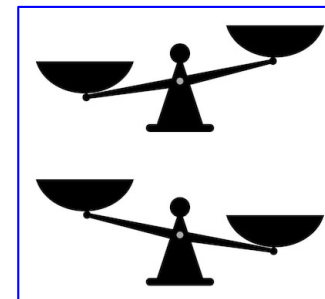
Decision based on gains-vs-pains balance ...

... for some data sources the balance is negative ...

... or is negative today and could turn positive tomorrow ...

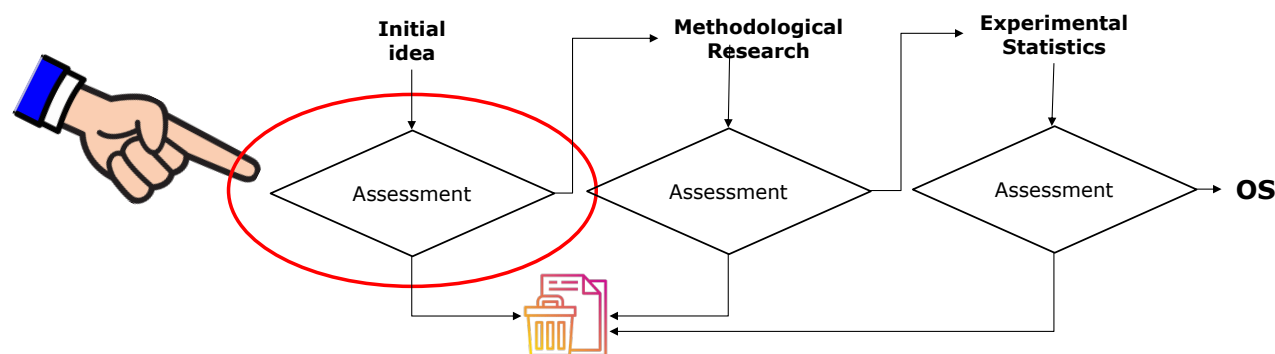


Selection of data sources

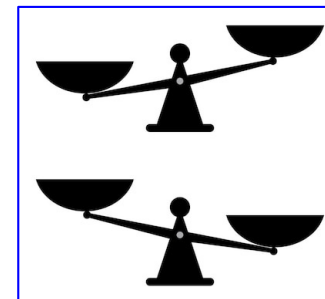


Dimensions to be considered (among others)

- *Technological and/or market **penetration***
 - How many “statistical units” can be reached?
 - What spatial coverage at European level?
- *Technological and/or market **stability***
 - Can we expect this data source to be there in 5-10-15 years?
 - How different will the data be in 5-10-15 years?
- ***Fragmentation/heterogeneity** of data formats/semantics*
 - Are there *de facto* or *de jure* standards?
 - Shall we deal with 1x, 10x, 100x or 1000x formats/APIs?



Selection of data sources

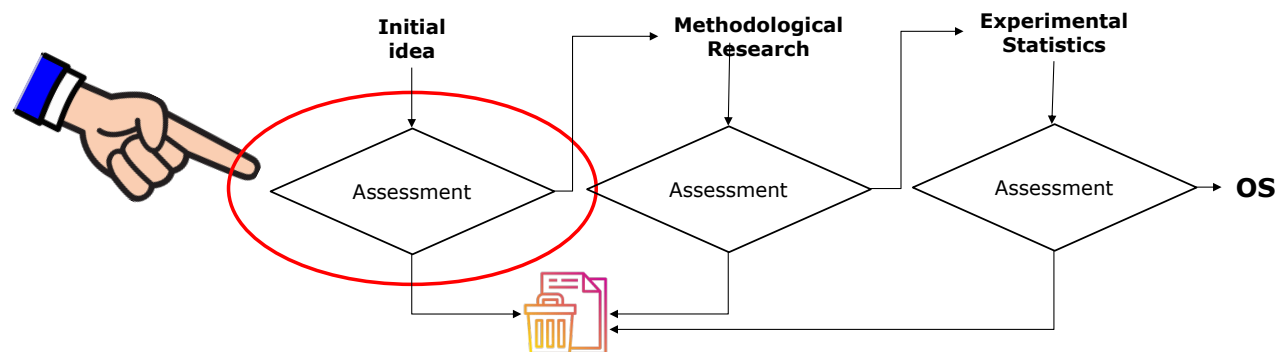


These dimensions depend on market and technology aspects, e.g.:

- Market **structure** – vendors, integrators, adopters– who determines the data content and formats?
- Market **concentration** – one player, few players, many players?

Do NOT depend on methodology or feedback from statistical users

→ it can be assessed at the beginning, even before research/experimental activities (landscaping analysis)



Classes of data sources

- *Methodologies and quality aspects cannot encompass whole set of "new data sources" but must be specific to "classes" of data sources*
- *ESSnet Big Data II – WP K (*)*

*"We believe that it is barely possible to write down meaningful **quality guidelines**, which can be applied to all kinds of new data sources. Instead, we group [...] according to so called **data classes**, for which we write down data-class specific quality guidelines." (*)*

*[...] with new data sources, it becomes more important to do a data-class-wise quality assessment than to go through one general error type after the other. The reason is that **processes diverge hugely across data classes**, and so do **potential errors**.*

(*) ESSnet Big data II – WP K – deliverable K3 - 2020

https://cros-legacy.ec.europa.eu/content/wpk-milestones-and-deliverables_en



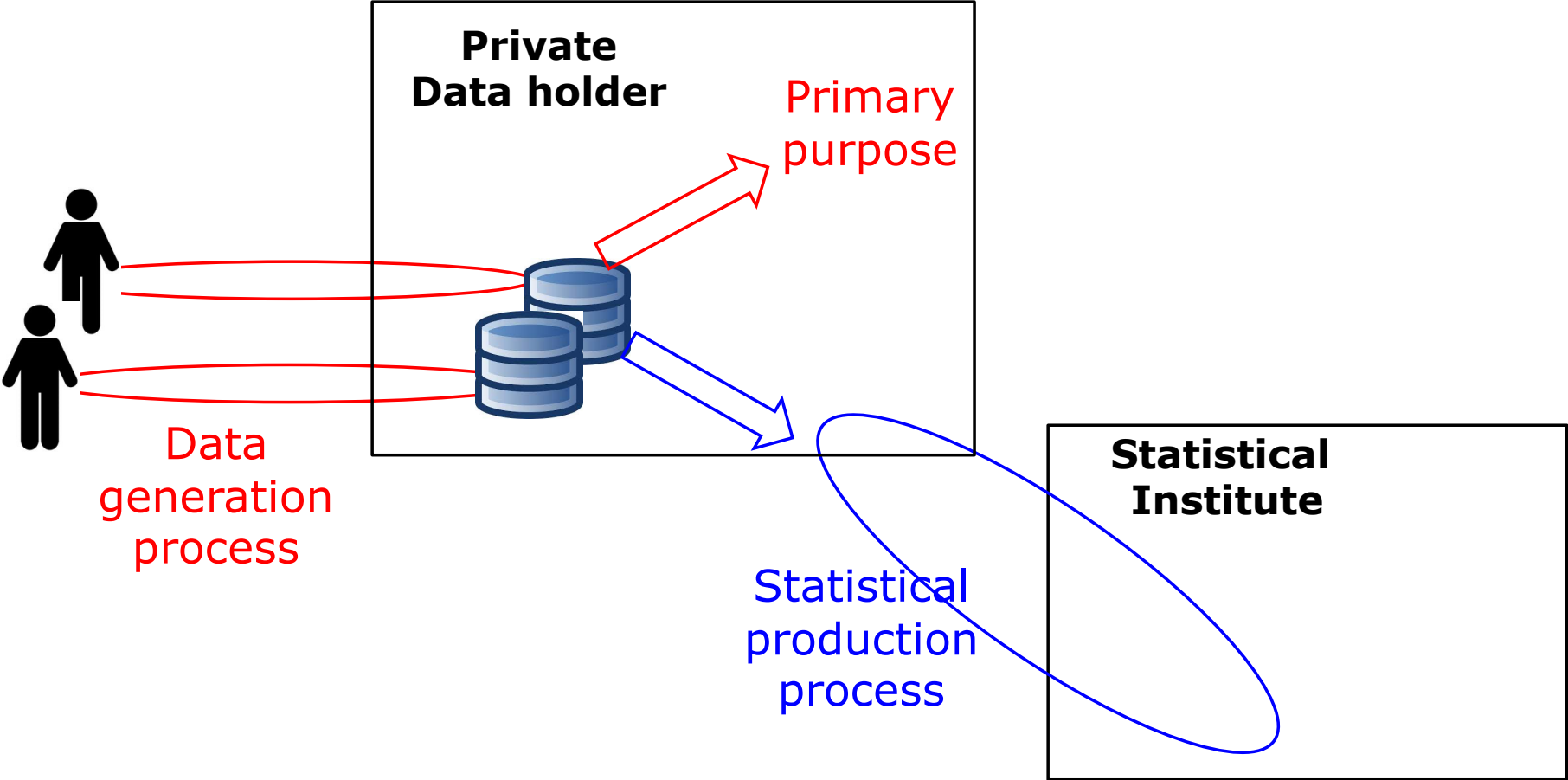
Focus on privately held data

Focus of this presentation on:

- *Digital traces, granular, sub-micro (nano-data)*
- *Generated primarily for non-statistical purposes*
- *Collected often by private companies (Privately Held Data)*
- *May be reused for statistics*

*Mutatis mutandis part of the following considerations apply also to (or can be inspirational for) other "**classes**" of new data sources (e.g. Earth Observation data, Internet data, ...)*

Reusing data held by somebody else is (also) a matter of processes ...



... and of operational meta-data about the process and it's performance ...

- *The data generation process is "dynamic"*
 - Changes in business, changes in technology → change in data
 - Customer churning, change in customer behaviour → change in data
 - Planned changes, e.g. system upgrade, new devices, new capabilities, new tariffs → change in data
 - Unplanned changes, e.g., system outage, errors, anomalies ...
 - (NB: the above applies to the data generation, not to the data processing, and holds true regardless of how the data processing is split between PDH and NS)

All these "facts" and "events" affect input data and their quality, therefore must be identified and reported proactively (when possible) or at least detected and reported reactively

OPERATIONAL METADATA

Operational metadata are metadata that describe the expected or actual outcomes of a process using evaluable and operational metrics."

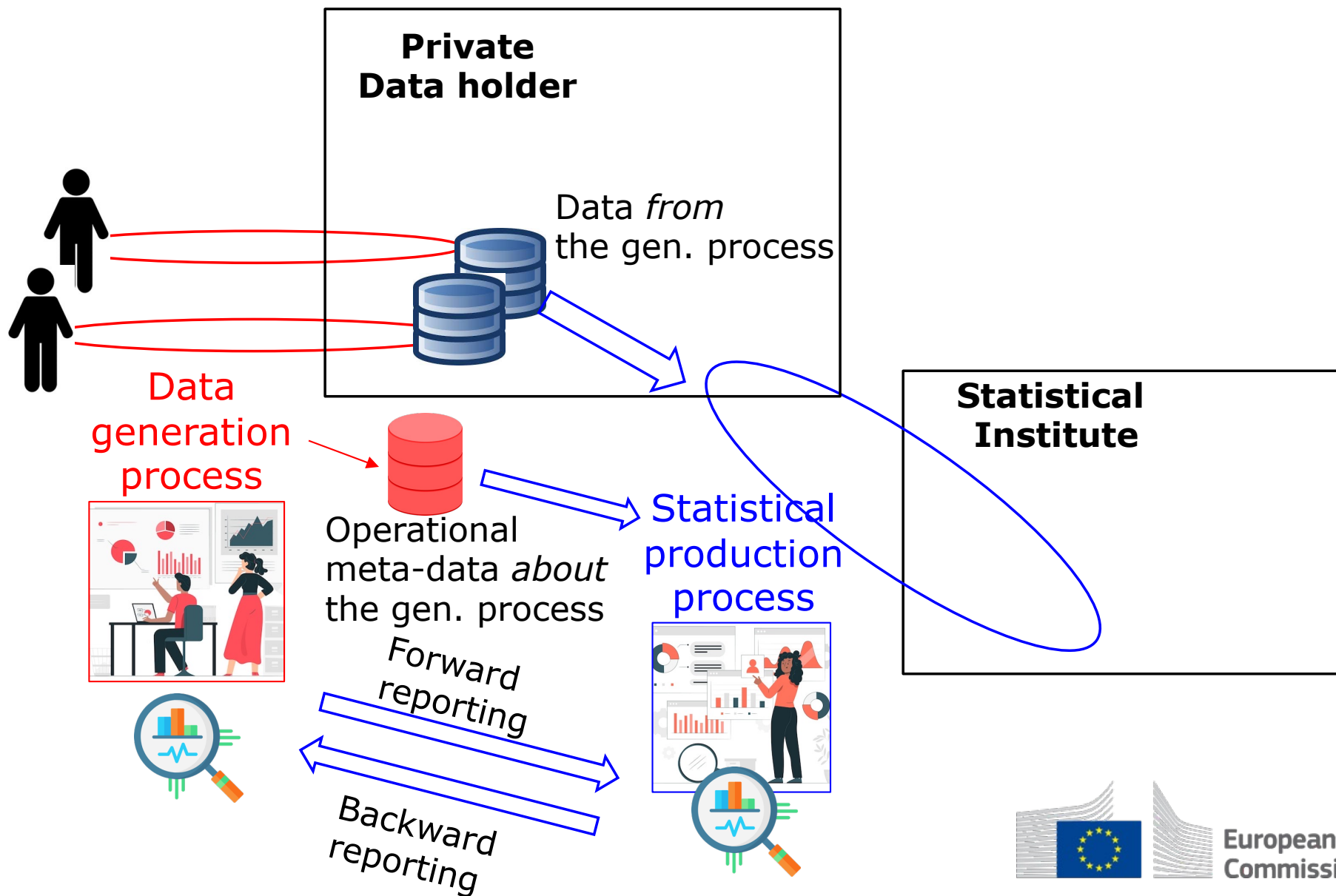
Operational metadata are a type of *reference metadata*. They include *quality metadata* and metadata measuring performance.

An alternative name is *paradata*.

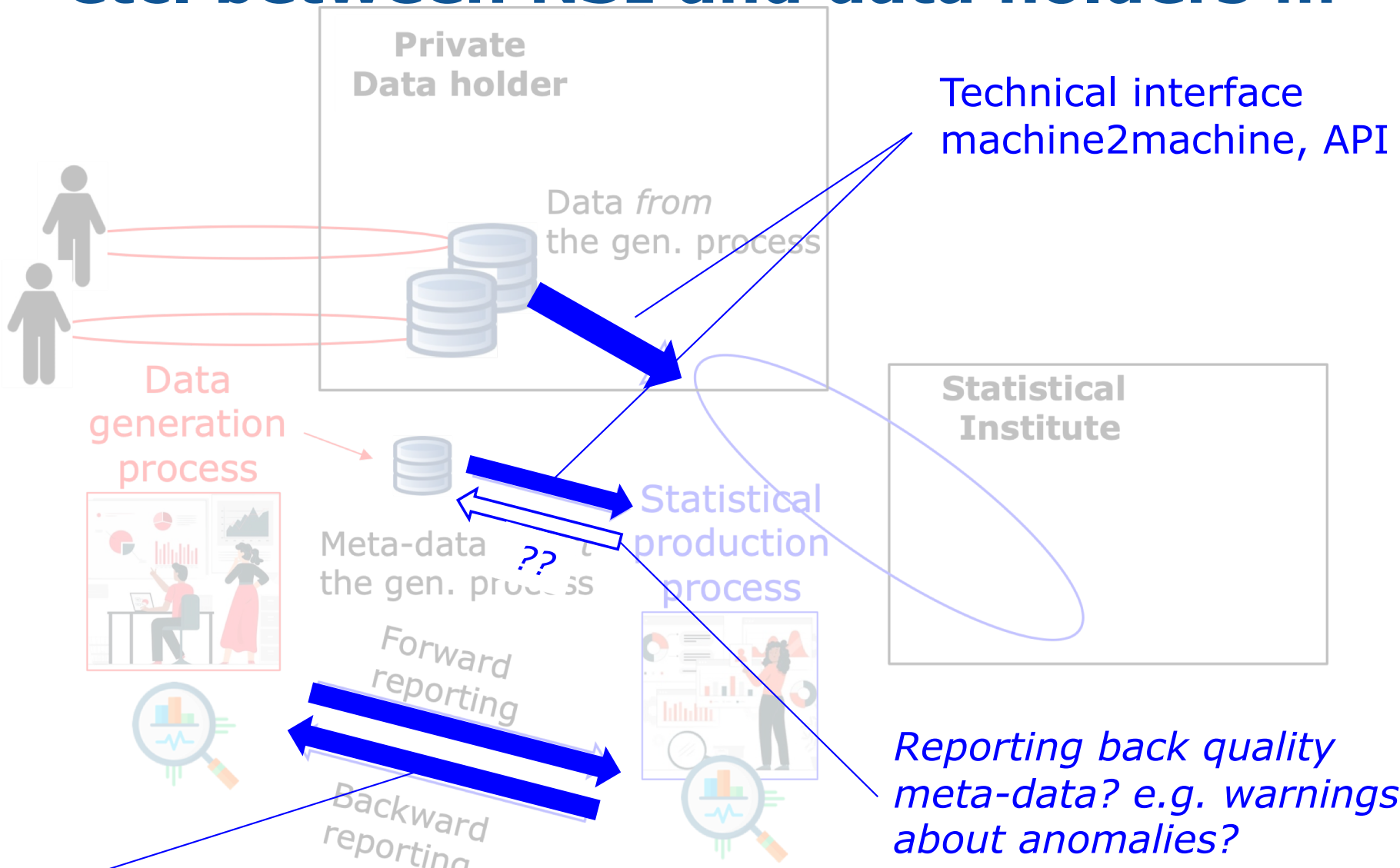
Source: ESS Handbook for quality and metadata reports
<https://europa.eu/!3cWkFk>



... and of operational meta-data about the process and it's performance ...



... and interfaces, protocols, policies, etc. between NSI and data holders ...



Technical interface
machine2machine, API

Reporting back quality
meta-data? e.g. warnings
about anomalies?

Organisational interface – **bidirectional (!)**
Human2human: **forms – policies**

Communicating process meta-data

- *Communicate in both directions !!*
 - PDH-to-NSI: "Next week we plan a large system upgrade, will cause outages or generation of spurious data from time X to Y in region Z"
 - NSI-to-PDH "We detected an anomalous data pattern starting around time X that apparently affects region Z, can you please help us determine what is happening"?
- *Balance between under-reporting (bad for accuracy) and over-reporting (bad for burden)*

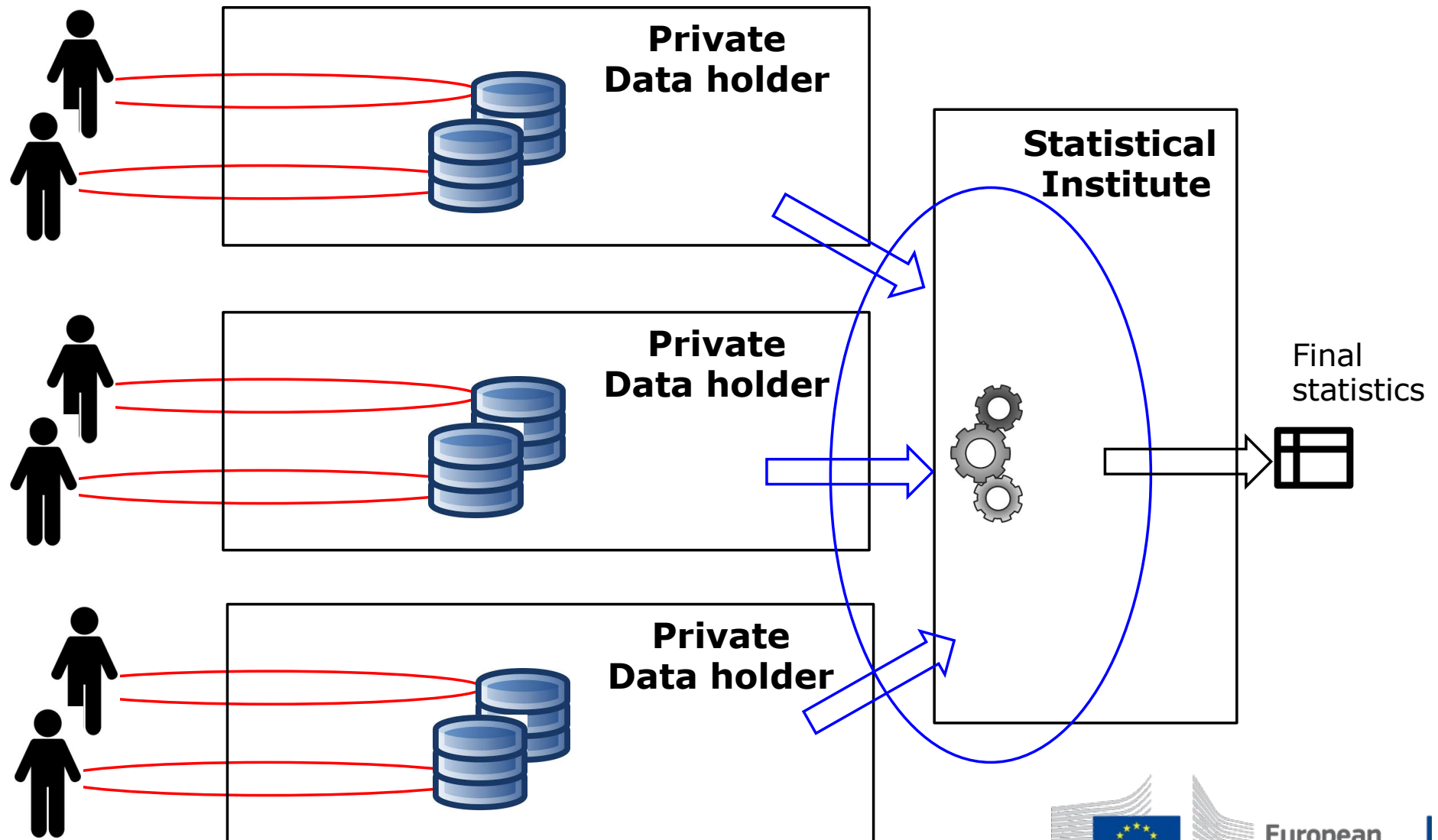
How to motivate the data holders?

- ***Incentives?***
- ***Legislative obligations?***
- ***Cost compensation?***

- *All of the above*
 - according to the Expert Group on facilitating the use of new data sources for official statistics (2022)
<https://europa.eu/!JGR3Gx>



Integration of data from multiple data holders improves quality



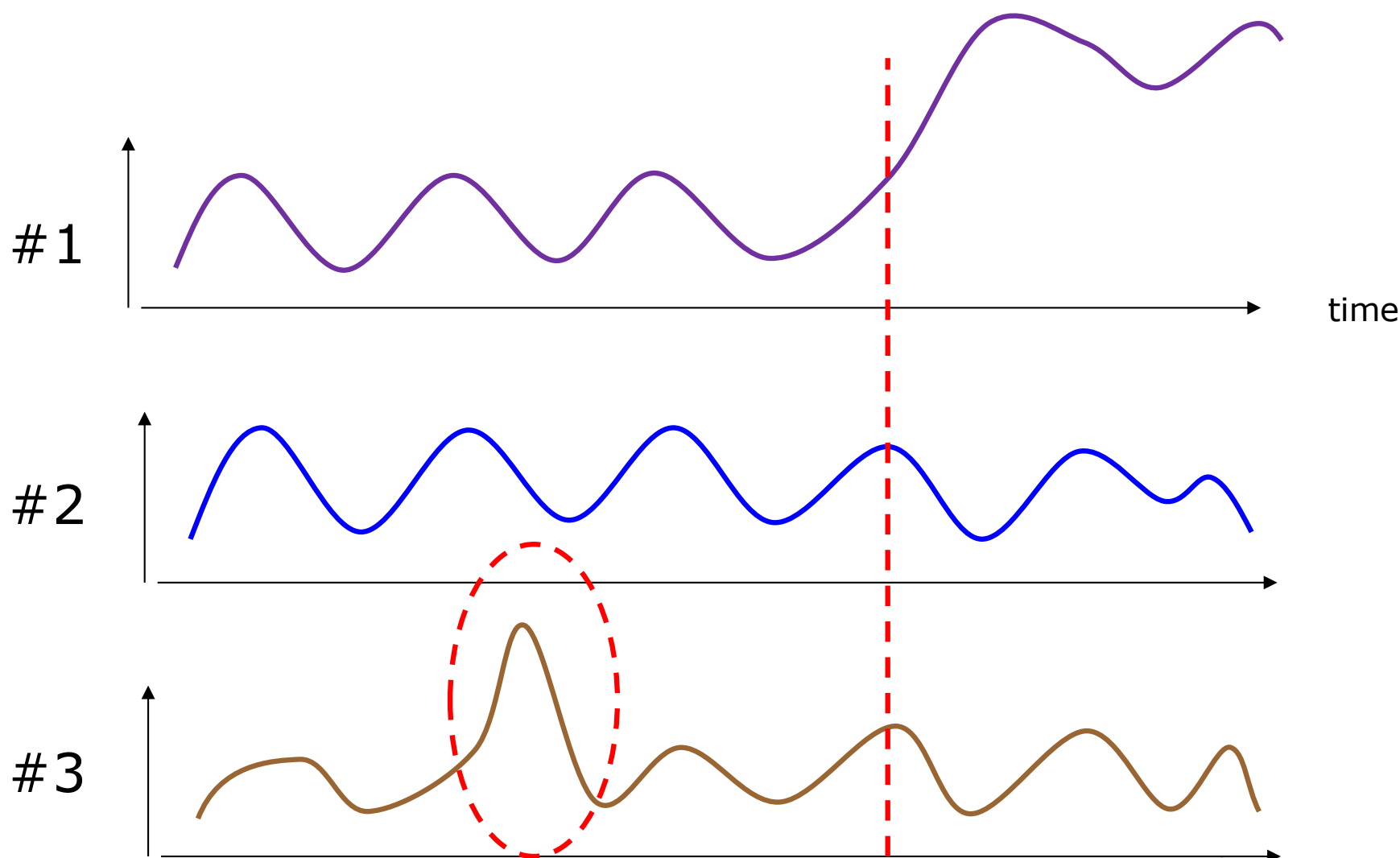
Integration of data from multiple data holders improves quality

- *Better representativeness of the total population, mitigate effects of population coverage bias in the final statistics*
 - Privately Held Data refer to “customers”, not citizens
- *Improved temporal stability, mitigation of customer churning*
- *Mitigate sensitivity to provider-specific aspects of data generation*
- *Improved robustness to anomalies, outages, partial or complete disruptions of data provision*
- *Equal treatment of all data providers in the same business sector (“level playing field”)*
- *Easier protection of business-sensitive information from individual data providers in the final statistics*

Source: Position paper by ESS Task Force on MNO data “Reusing Mobile Network Operator data for Official Statistics: the case for a common methodological framework for the European Statistical System”, 2023 <https://europa.eu/!KbdVG4>



Consistency and plausibility checks across data providers



Integration of non-statistical “big” data and statistical data

- *Data produced for business purposes refer to “customers” and oftent to “customers’ devices”*
 - Observed population does not map 1:1 to target population
 - Observed population is not a representative sample of whole population
 - Coverage gaps, multiple counting, coverage bias
- *Mapping not static*
 - customer churning
 - change in user behaviour
- *Statistics based solely on a single “big data” source maybe inaccurate and unstable*
- *Big Data may lack variables of interest for statistics*

Integration of non-statistical “big” data and statistical data

- *Combination of non-statistical “big” data sources with (small?) statistical data may deliver the best of both*
 - From “big data”: timeliness (near real-time), spatio/temporal detail (“interpolation”), temporal continuity and spatial coverage, variables derived from “objective” observations
 - From statistical data: correct projection to target population (mitigation of bias, multiple counting, coverage gaps), additional variables of interest not observed by big data



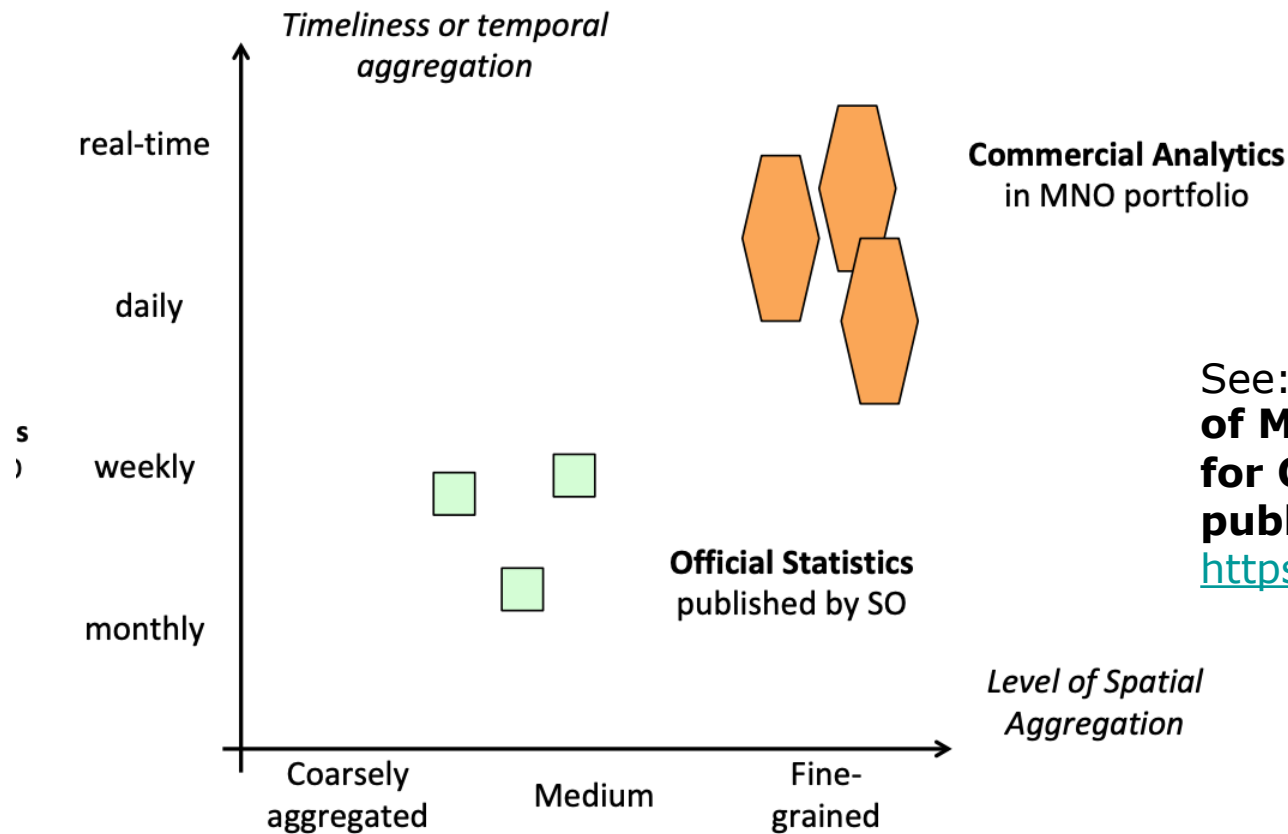
Fertilise big data by statistical data

Augment statistical data with big data

Integration of non-statistical “big” data and statistical data

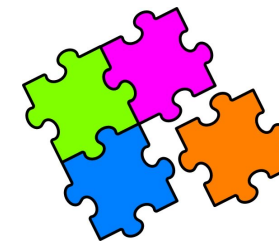
- *More balanced positioning of NSI and private data holders*
 - Not provider-consumer, but win-win partnership
- *No derogations to statistical confidentiality!*
 - Statistical data cannot be used for non-statistical purposes
 - Statistics integrating data from **multiple providers and statistical data** produces an authoritative final statistics that may serve as “calibration reference” also for individual providers
 - Intermediate aggregate non-personal data may be shareable back with individual providers
 - Privacy-Enhancing Technologies may help to protect data confidentiality
- *Reassert the role of NSIs and statistical surveys in the new data-rich ecosystem*
 - Survey still needed, but smaller and less burdensome if combined with big data, for better/richer/timelier final statistics

Commercial Analytics may coexist with, and even get reinforced by Official Statistics



See: DGINS 2018 paper **Processing of Mobile Network Operator data for Official Statistics: the case for public-private partnerships**
<https://zenodo.org/records/10246468>

Integration of MNO and non-MNO data



- *ESSnet METH-TOO: **research grant** focusing on the combination of MNO and non-MNO data*
 - Started officially on 1st November 2023 for 2 years, 900 keur
 - Consortium of 10x NSI, ISTAT coordinator
- *Project Objectives*
 - *WP1 - Landscaping analysis of candidate non-MNO data sources* lead FR
 - *WP2 - Developing formal methods and open-source tools for integration of MNO and non-MNO data* lead NO
 - *WP3- Proof of concept of ad-hoc survey to improve MNO data* lead AT

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DESTATIS	STATISTISCHES BUNDESAMT	DE
INE	INSTITUTO NACIONAL DE ESTADISTICA	ES
INSEE	MINISTERE DE L'ECONOMIE DES FINANCES ET DE LA RELANCE	FR
CBS	CENTRAAL BUREAU VOOR DE STATISTIEK	NL
SSB	STATISTISK SENTRALBYRAA	NO
INS	NATIONAL INSTITUTE FOR STATISTICS	RO
SCB	STATISTISKA CENTRALBYRAN	SE
INE-PT	INSTITUTO NACIONAL DE ESTATISTICA PORTUGAL	PT

Softwarization of statistical methods

- ***Volume and complexity*** of granular data
- ***Automation*** of data processing
 - Human work shifts from 'executing instructions' to 'formulating instructions' to be executed by machines
- ***Data processing to be represented in formal languages***
 - Programming languages, schema, ontologies ... code!

Automatization → ***Softwarization*** of statistical methodologies*

(*) A reflection on methodological sensitivity, quality and transparency in the processing of new 'big' data sources, Q2022 conference, Vilnius
<https://zenodo.org/records/10246446>



Softwarization of statistical methods

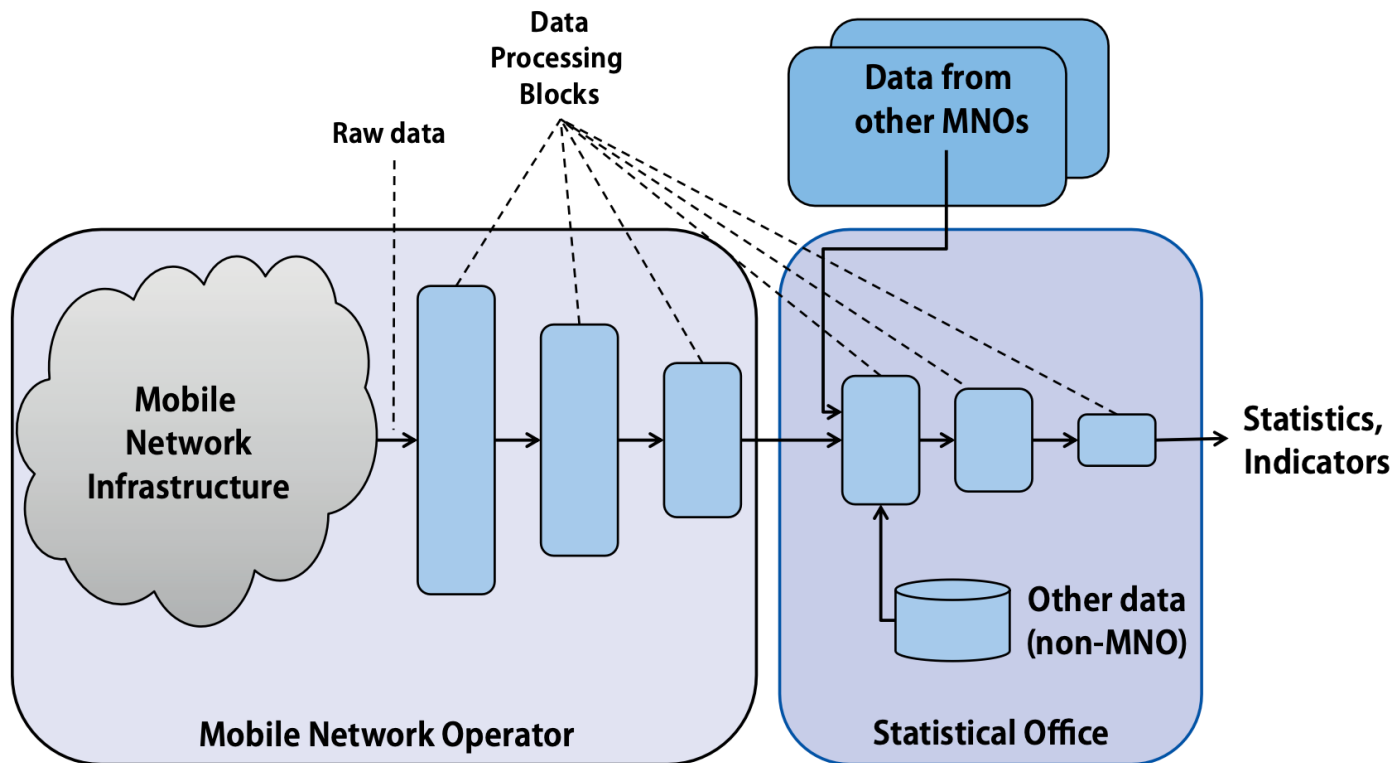
- *Opportunities and implications*
 - Source code as reference metadata !
 - Open-source code release to abide to methodological transparency
 - Reproducibility of methods (\neq replicability of results)
 - Independent auditability of methods, collaborative improvement
 - Methodological development may learn from (and import) established practices in complex software development (e.g. modularity, collaborative development, versioning)
 - Quality of (reference) software as natural component of methodological quality
 - Sharing code with other NSIs and/or data providers, ease harmonisation, pool resources

See also “Standardisation of methods and processes” presentation given at ISTAT Workshop on Methodologies for Official Statistics, Dec’22

https://www.istat.it/it/files/2022/11/4_3_Slide_Ricciato.pdf

Standardisation of end-to-end data processing workflows

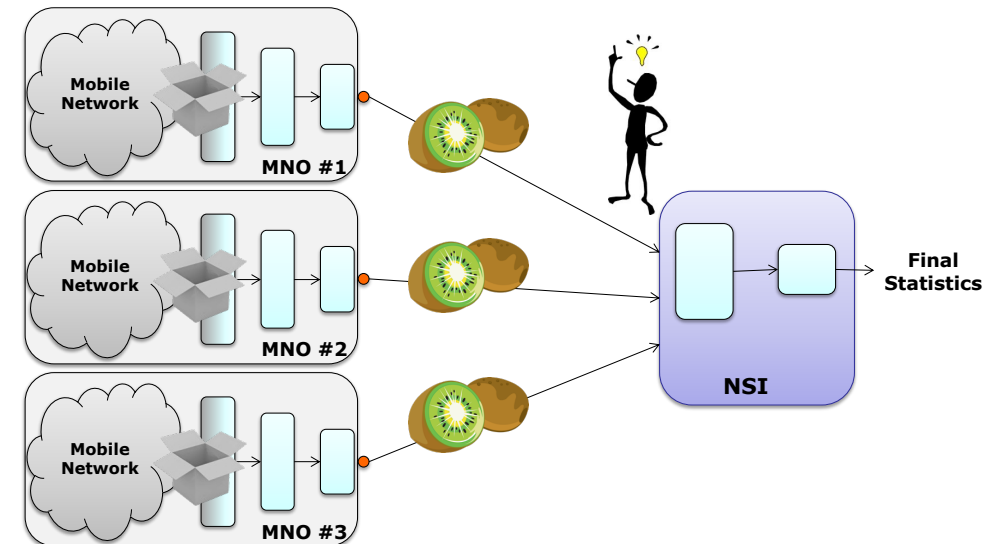
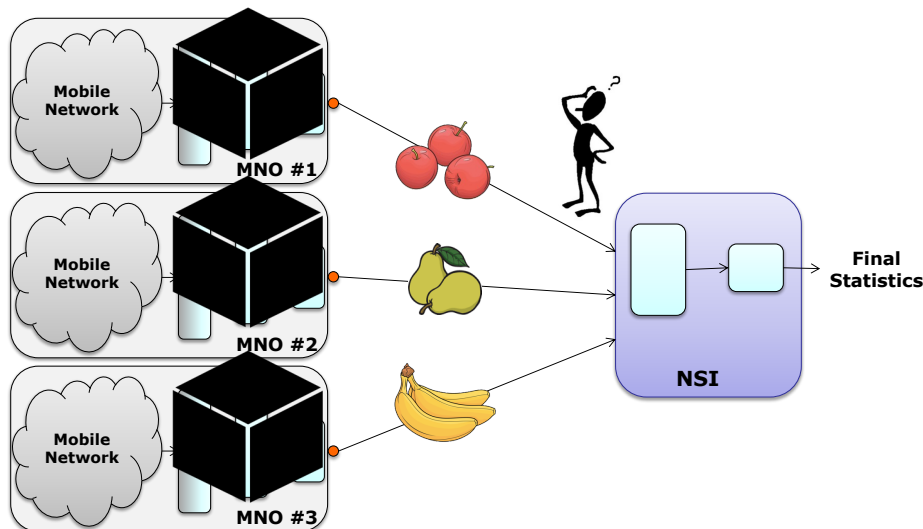
- *Detailed and non-ambiguous representation of operations and data structures in formal language ...*
- *... (co)defined by NSI even if they are executed at data provider's premises*



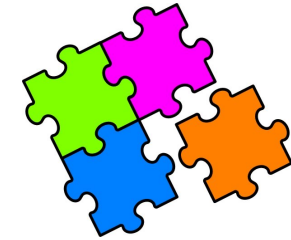
Softwarisation & standardisation



- *Common/standard methodologies and definitions necessary to achieve **comparable** and **combinable** results*
- *Softwarisation and Standardisation reinforce each other*

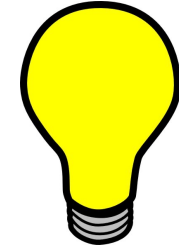


Multi-MNO project



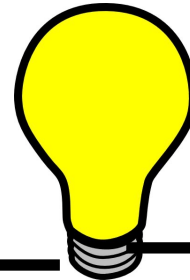
- ***Co-development partnership***
involving NSI experts and industry experts working together
 - Started in Jan'23 for 2.5 years, until mid 2025
 - Public procurement procedure based on open call for tenders, 1.2 Mio
- ***Project Objectives***
 - *Develop a first proposal for an open end-to-end **methodological framework and associated **quality framework and guidelines**, with focus on an initial selection of use-cases***
 - *Open-source reference software pipeline implementing the proposed methodological framework;*
 - *Practical demonstration of the processing pipeline on real-world data across 5x MNOs in 4x EU countries*
- ***Consortium***
 - *GOPA (Germany, consortium leader)*
 - *2x Industry partners: NOMMON (Spain), POSITIUM (Estonia)*
 - *2x NSI: CBS (Netherlands), ISTAT (Italy)*
 - *5X MNOs: Orange Spain, Vodafone Spain, Vodafone Italy, A1 Slovenia, POST Lux.*

The vision

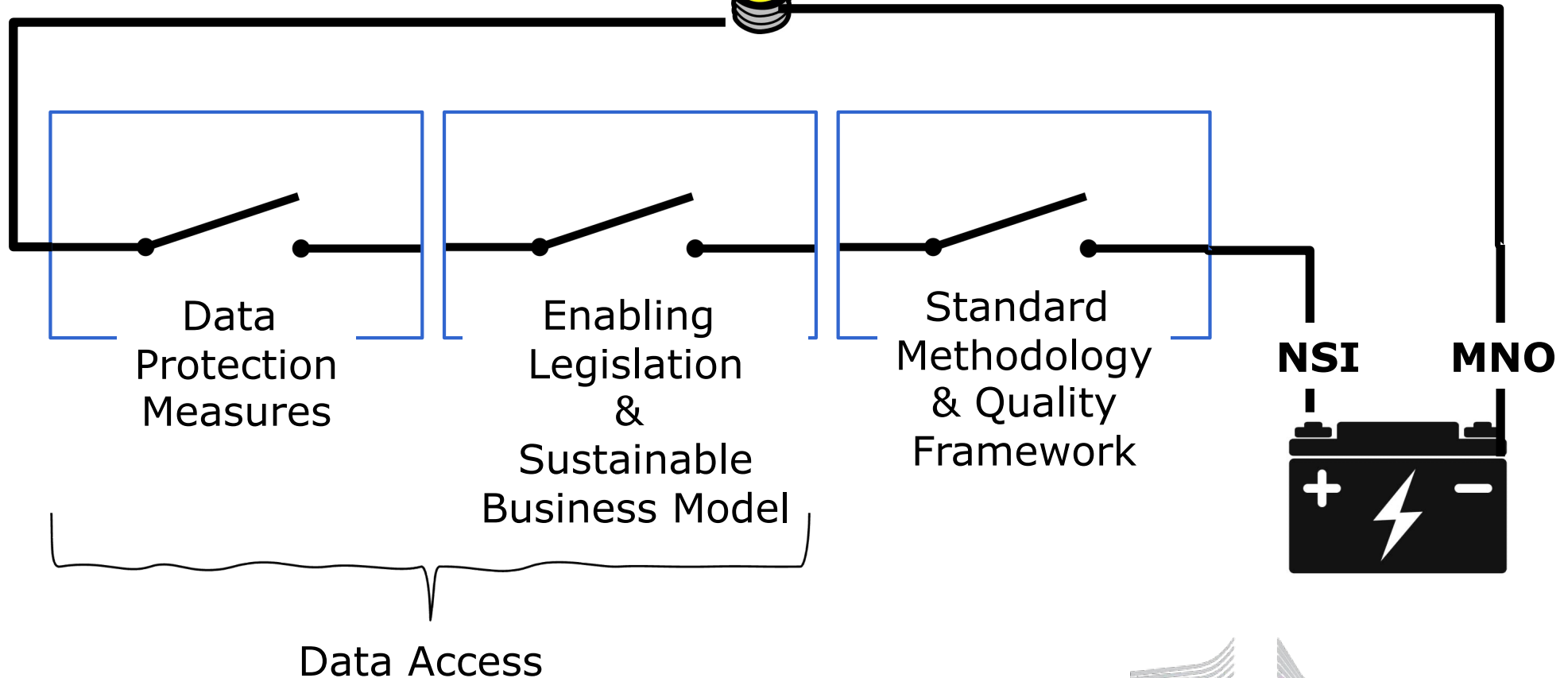


- *In 202x MNO data are (re)used for regular production of **official statistics***
 - Not merely “experimental statistics”...
 - Data from multiple MNOs in each country and across countries - Multi-MNO
 - Combined with statistical data
 - Processed according to **standard methodologies** and **transparent quality criteria** defined at EU level, by the ESS in collaboration with industry
 - Evolvable methodological framework
 - Processed (at least partly) at MNOs premises
 - Built-in privacy protection measures defined at EU level in consultation with EDPS/EDPB

Series of challenges need to be worked out in parallel



Regular production of **Official** Statistics based on MNO data



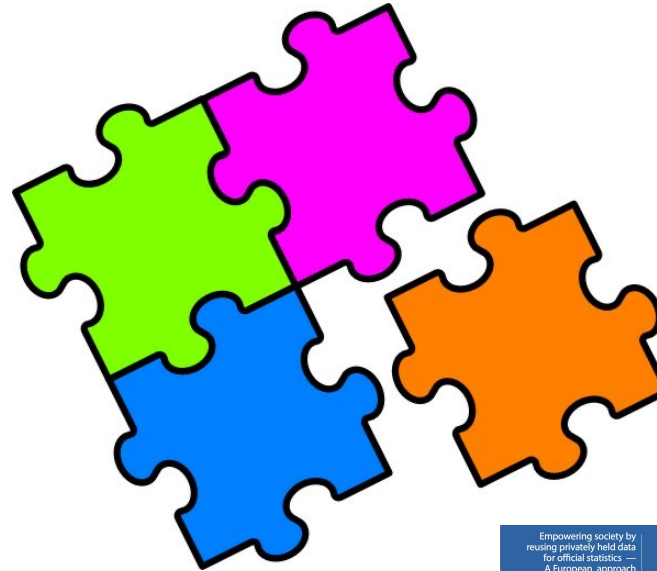
Composing the puzzle

ESSnet Big Data II – WPI & WPK



TF MNO

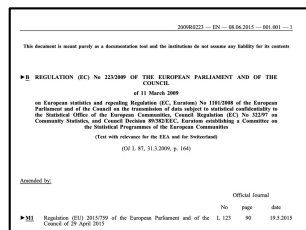
Multi-MNO project



Other projects on Privacy-Enhancing Technologies

Research Grant
ESSnet METH-TOO

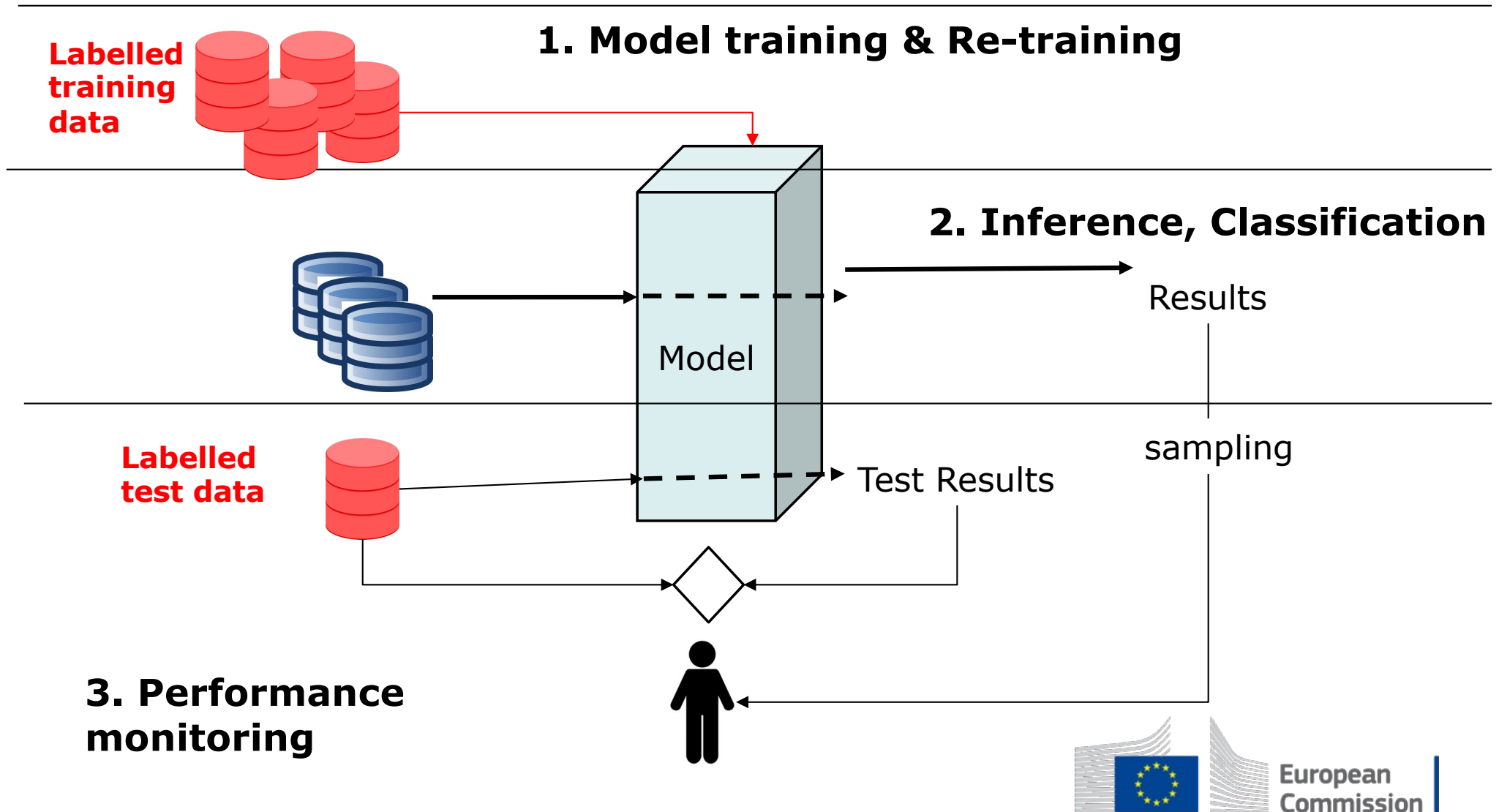
Rev. 223



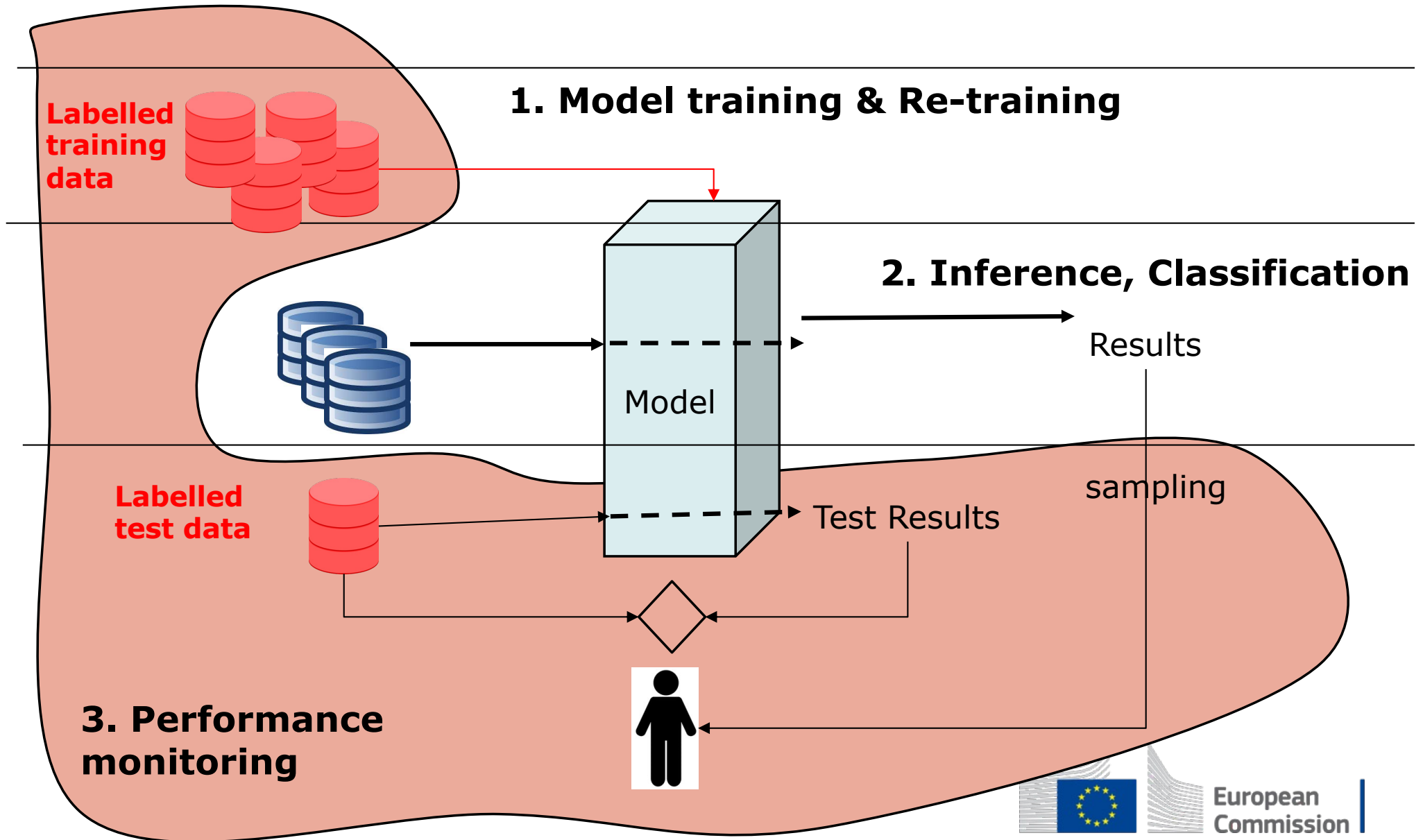
EG B2G4S



A few words on Quality aspects of supervised ML in OS



A few words on Quality aspects of supervised ML in OS



A few words on Quality aspects of supervised ML in OS

- ***Quality framework must cover processes and data related to (re)training and monitoring***
 - Quality of training data → quality of ML model → quality of ML results
- *Examples of issues to be addressed (guidelines, criteria)*
 - How do you assess/monitor/audit the performances of the ML?
Does that involve human inspection?
 - What conditions/criteria will indicate that a fresh re-training is needed?
E.g., drift of performance metrics?
 - Provenance of training data: How do you produce training data? Who labels them? How many people and with what skills will be in charge of labelling? (NB: there are cost-accuracy tradeoffs)
 - Can (or should) you publish the training data for the sake of methodological transparency?
 - Versioning ML models; versioning of training data
 - Is the energy consumption significant at any stage?
 - ...

Conclusions

- Transition from **experimental statistics** to **official statistics** requires walking the whole quality path
- Developments in **methodologies** and **quality aspects** must go hand-in-hand
- Mind the strategic implications of methodology/quality choices, e.g. NSI vs data holder(s)
- Concrete guidelines and operational criteria must be specific to data classes, but advancements on one data class (e.g., MNO data) will be useful (or at least inspirational) for other classes

Thanks for your attention

