

Integration of alternative data into consumer price statistics: the UK approach

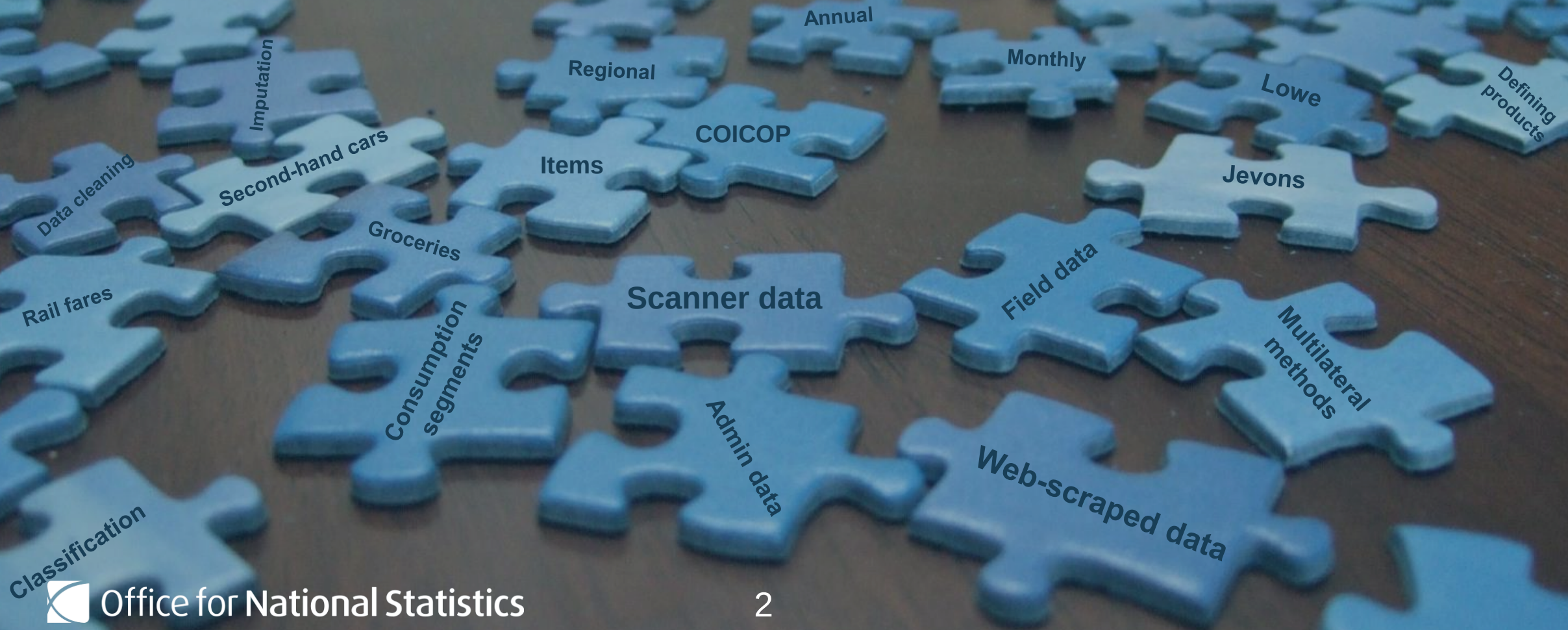
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Scanner data implementation team



How do we complete this puzzle?



New data

New UK data



Rail fares transaction data
Daily data feeds covering all transactions
One supplier for all of GB
~100% market coverage

Go live in 2023



Second-hand cars web data
Daily data feeds covering all advertisements
One supplier for all of UK
~70% market coverage?

Go live in 2023



Groceries scanner data
Weekly/monthly feeds, some data daily some weekly
3 (+) major supermarket chains
~50% market coverage

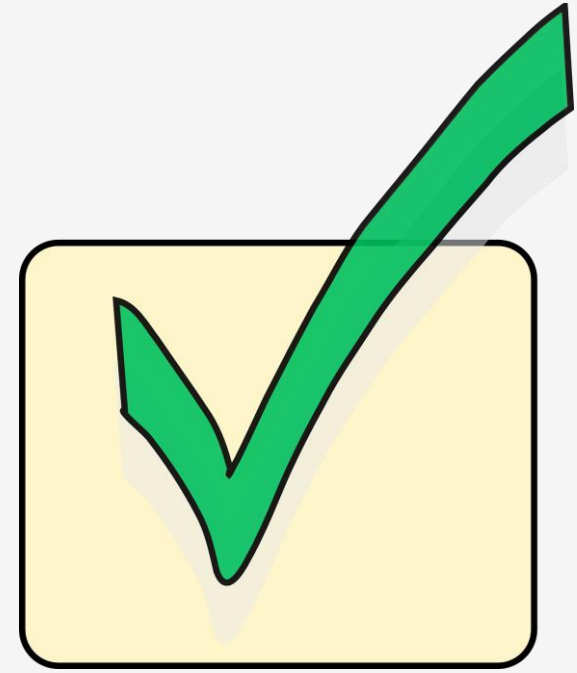
Go live in 2024

Quality assurance and classification

Quality assurance

Series of initial quality checks:

- Have the expected files arrived?
- Are variables correct data type?
- Are values within expected range?
- Are the data size/shape as expected?
- Are there a consistent (low) number of null values?



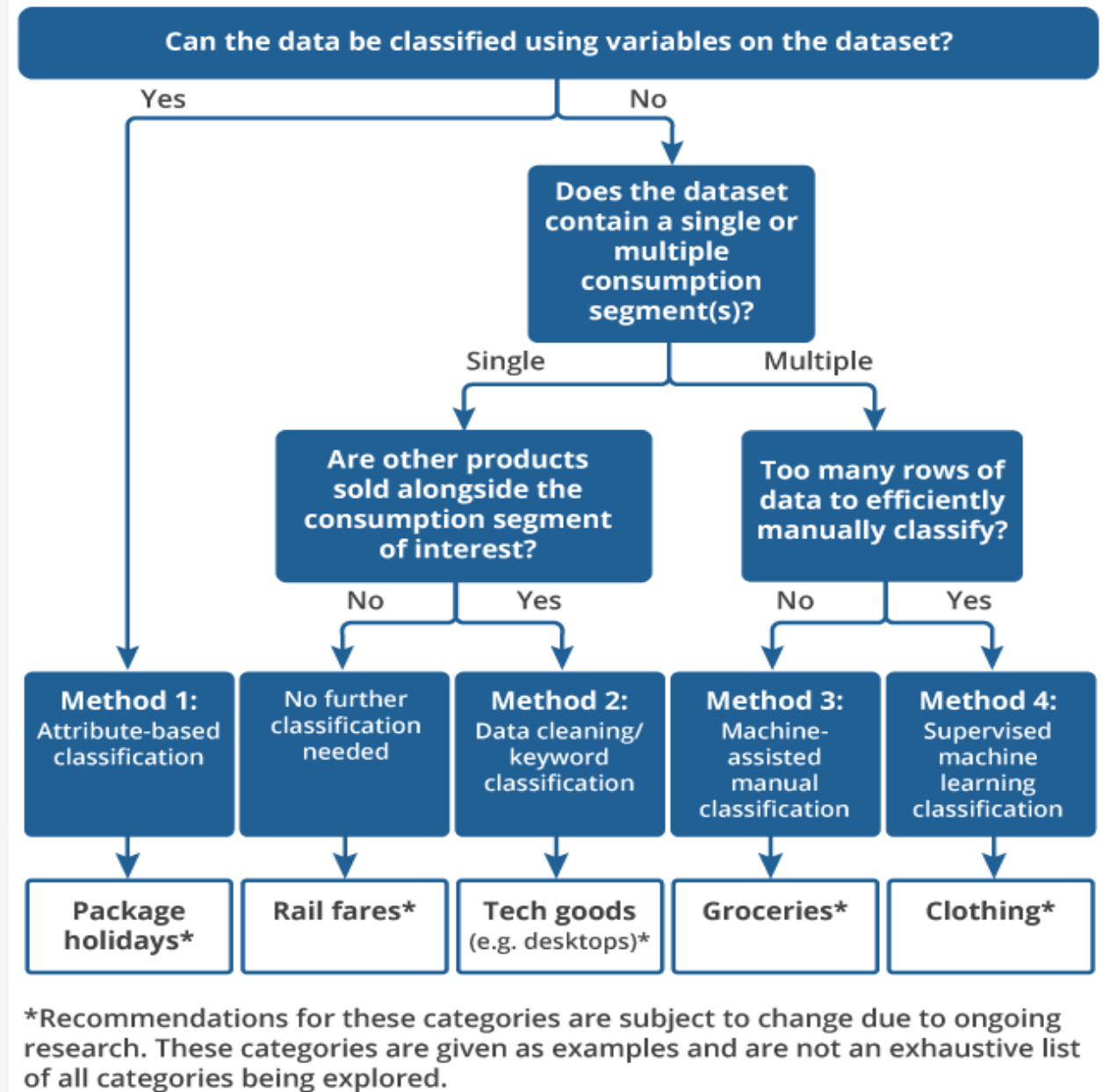
If any failures, issues raised with retailer and redelivery arranged if necessary

Further “curiosity” checks to investigate sensibility of data

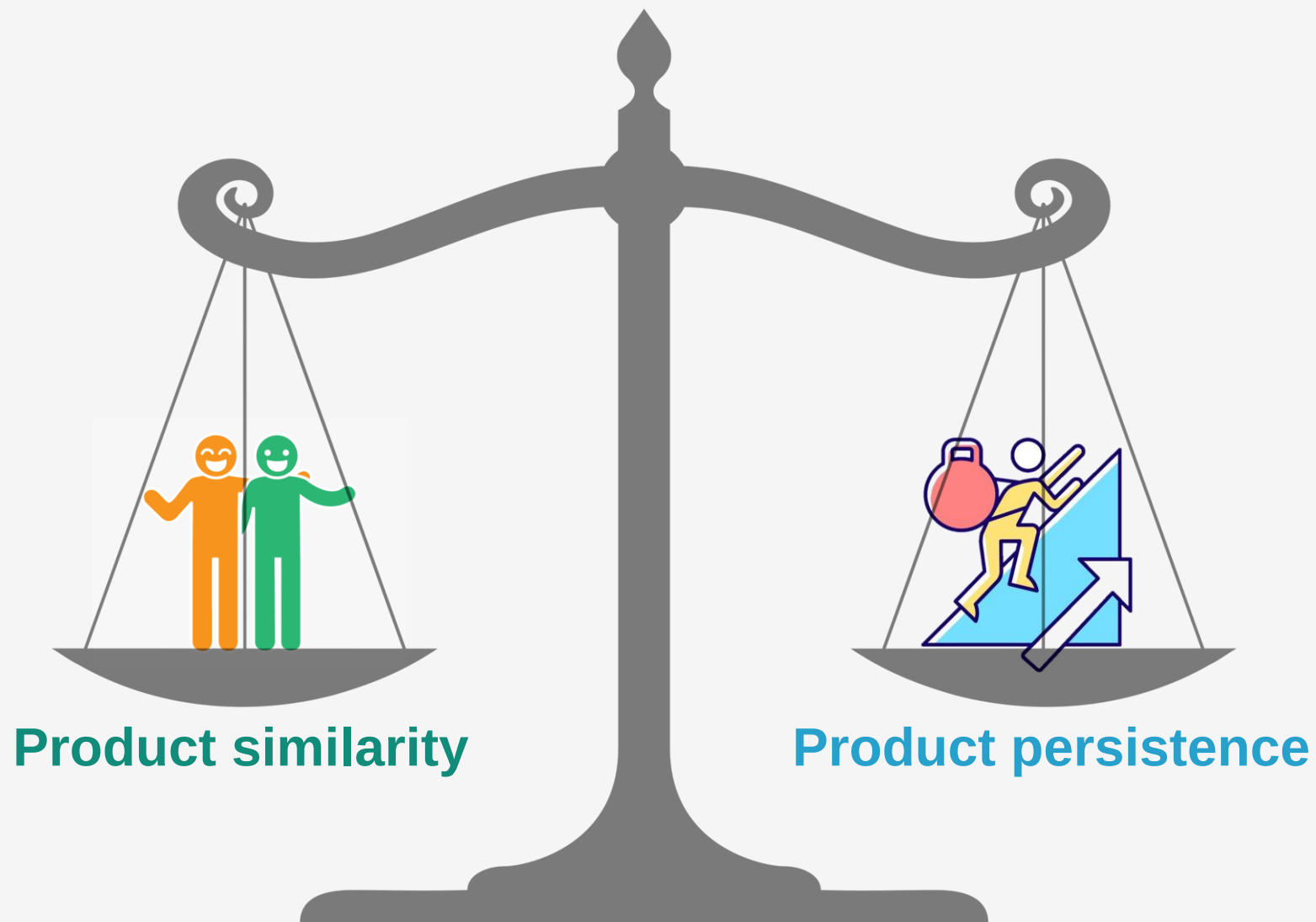
Classification/filtering

For **rail fares & second hand cars**, other products are sold alongside products of interest (e.g. station parking and second-hand motorbikes), therefore a simple data cleaning classification method is applied

Groceries cannot be classified using variables on the dataset and contain multiple consumption segments (COICOP6). For now manual classification has been achievable – in future we look to improve the efficiency of this using machine learning to assist the classification process



Defining a product



Groceries

(has unique identifiers)

GTIN

SKU

Relaunch linking

Relaunch linked

SKU + unit of
measurement

Rail fares

(no unique identifiers)

Origin/destination
stations

+ Ticket type

+ Ticket class

+ Journey type

+ Region?

+ Traveller
demographic?

Second-hand cars

(can't use unique identifier!)

Make

+Model

+Mark?

+Trim?

+Age?

+Mileage?

+Body type?

+Fuel type?

MARS ([Chessa, 2019](#))

Helps to identify
which variables
should be used to
define a unique
product to ensure
balance between
product similarity and
persistence through
time

Aggregation

New UK aggregation structure (2024)

To enable the introduction of new data sources we will move to a new aggregation structure.

This has been developed under four key considerations:

1. We can realise more potential from alternative data sources, while continuing existing practices for our traditional collection to ensure good market coverage
2. We have the flexibility to use alternative data in combination with, or in place of, traditionally collected data, weighted according to our best information on retailer market share
3. We can more readily calculate regional/sub-national consumer price statistics, for which there has been growing demand in the UK
4. We enable a smooth transition towards the latest iteration of Classification of Individual Consumption According to Purpose (COICOP 2018), while also realigning our numerical system for our detailed (COICOP 6) level of the hierarchy coding with higher COICOP levels

Realising more potential?

Traditional: Sample item selected to represent broader group, e.g. Basmati rice chosen to represent all rice. Reduces collection cost/burden & maintains homogeneity within elementary aggregate (EA) indexes

Scanner: Now receive data for all rice varieties, so can include and weight according to economic importance. BUT only from some retailers, how do we then use all these rice data without implicitly exaggerating the weight for the retailers who provide scanner data?

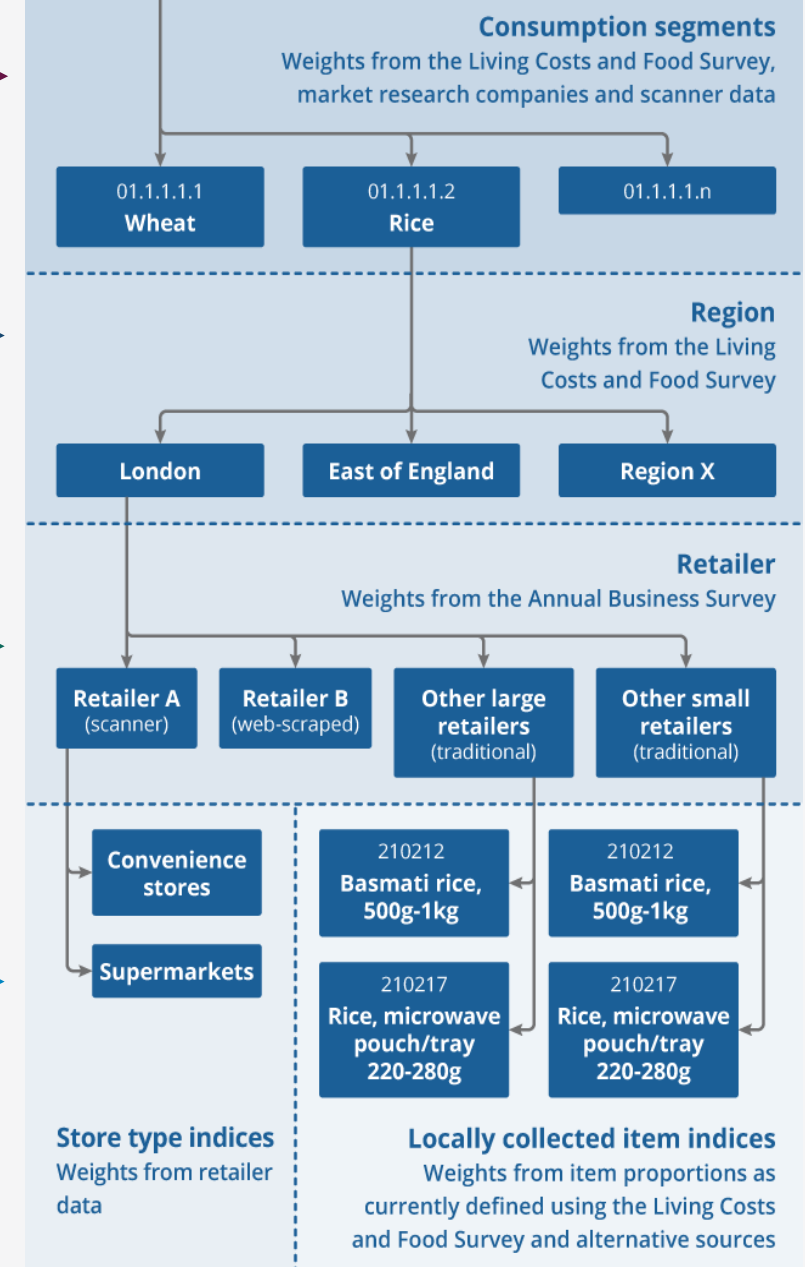


New COICOP 6 level, “consumption segments”, replacing old COICOP6 level “items”. These are typically equivalent or broader, allowing us to use more of the scanner data

New regional level, allowing us to more readily produce sub-national statistics

Retailer level allows us to weight retailers’ according to their market share, independently of how many data are provided – consumption segments are only introduced when we have >70% market coverage through either scanner data, traditional data or a combination

Stratum level indices allow us to maintain item definitions for traditional collections and produce more granular indices for some retailers helping us understand the households experiencing higher inflation. EAs can use either multilateral or bilateral methods



Index methods

Traditional data EA indices:

- Jevons; Dutot; Lowe/Laspeyres
- Fixed base
- $\text{Jan}(y)=100$

Scanner data EA indices:

- GEKS-T (otherwise known as CCDI)
- Mean splice
- 25 month window, re-referenced to $\text{Jan}(y)=100$

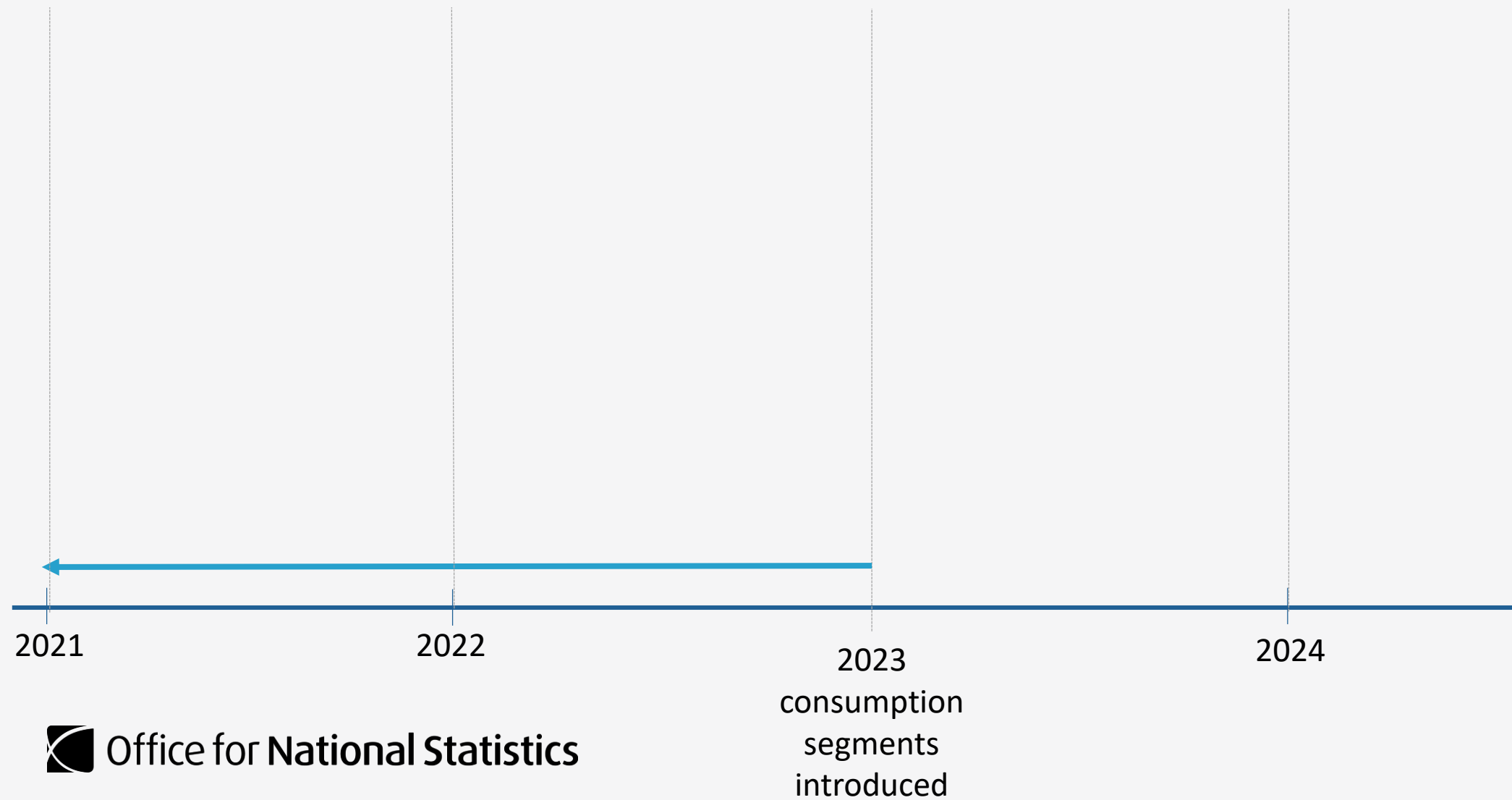
Aggregation of EA indices:

- Lowe
- Fixed-base
- $\text{Dec}(y)=100, \text{Jan}(y)=100$

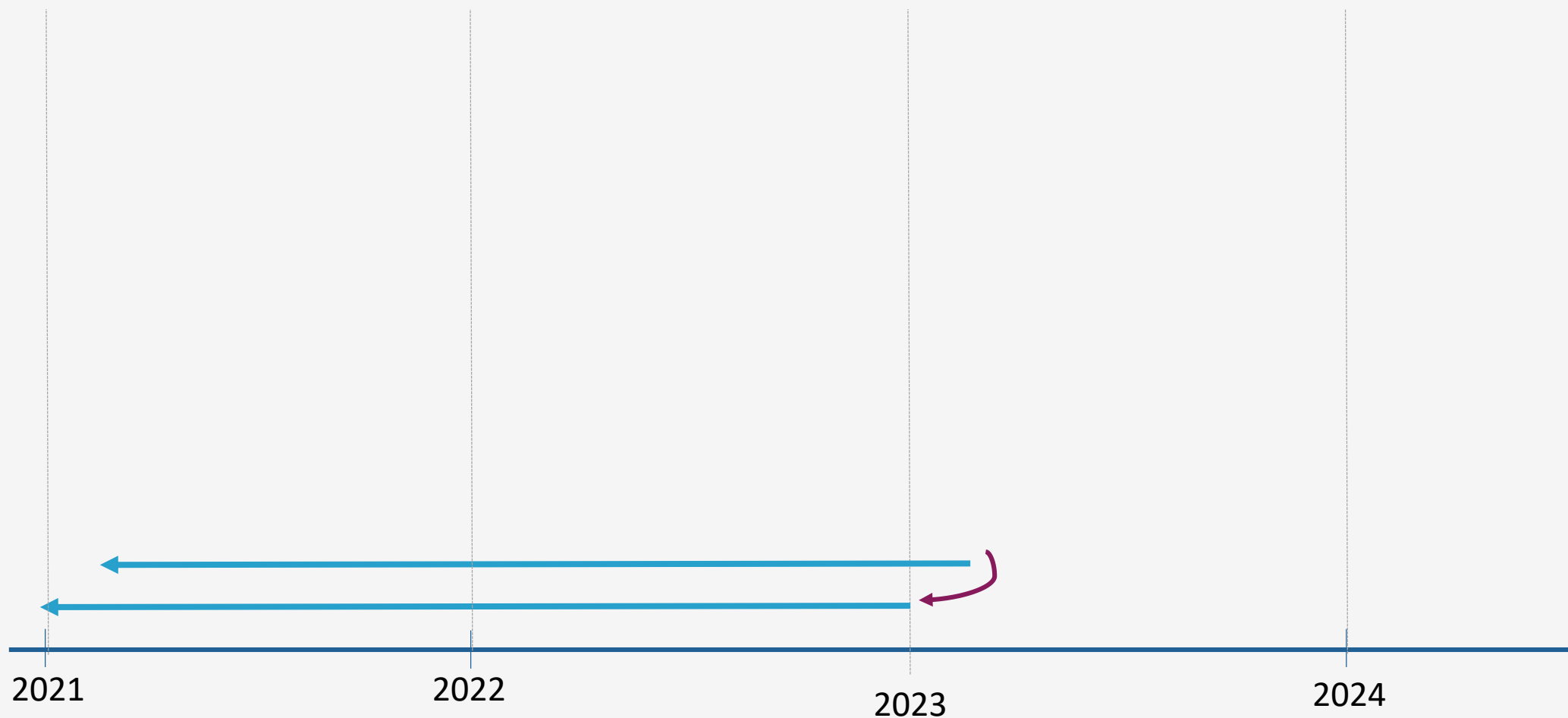
Annual process

How does a traditional basket update work within a 25-month window multilateral method framework?

1. Clean and classify 25 months historic data and calculate GEKS across full window

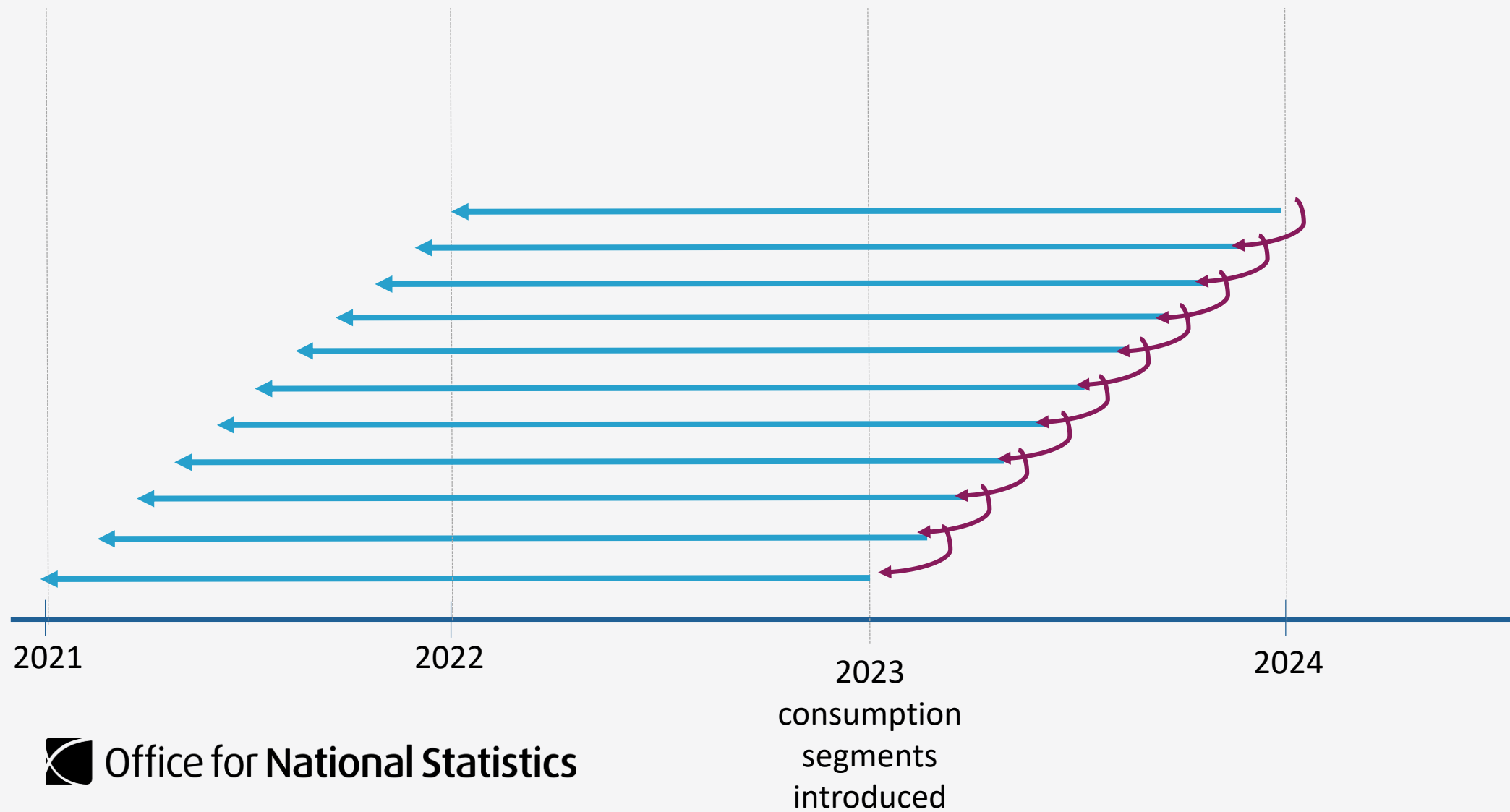


2. Append month and calculate new 25 month window, **mean splice** onto existing series, rereference to $\text{base_month}(y)=100$

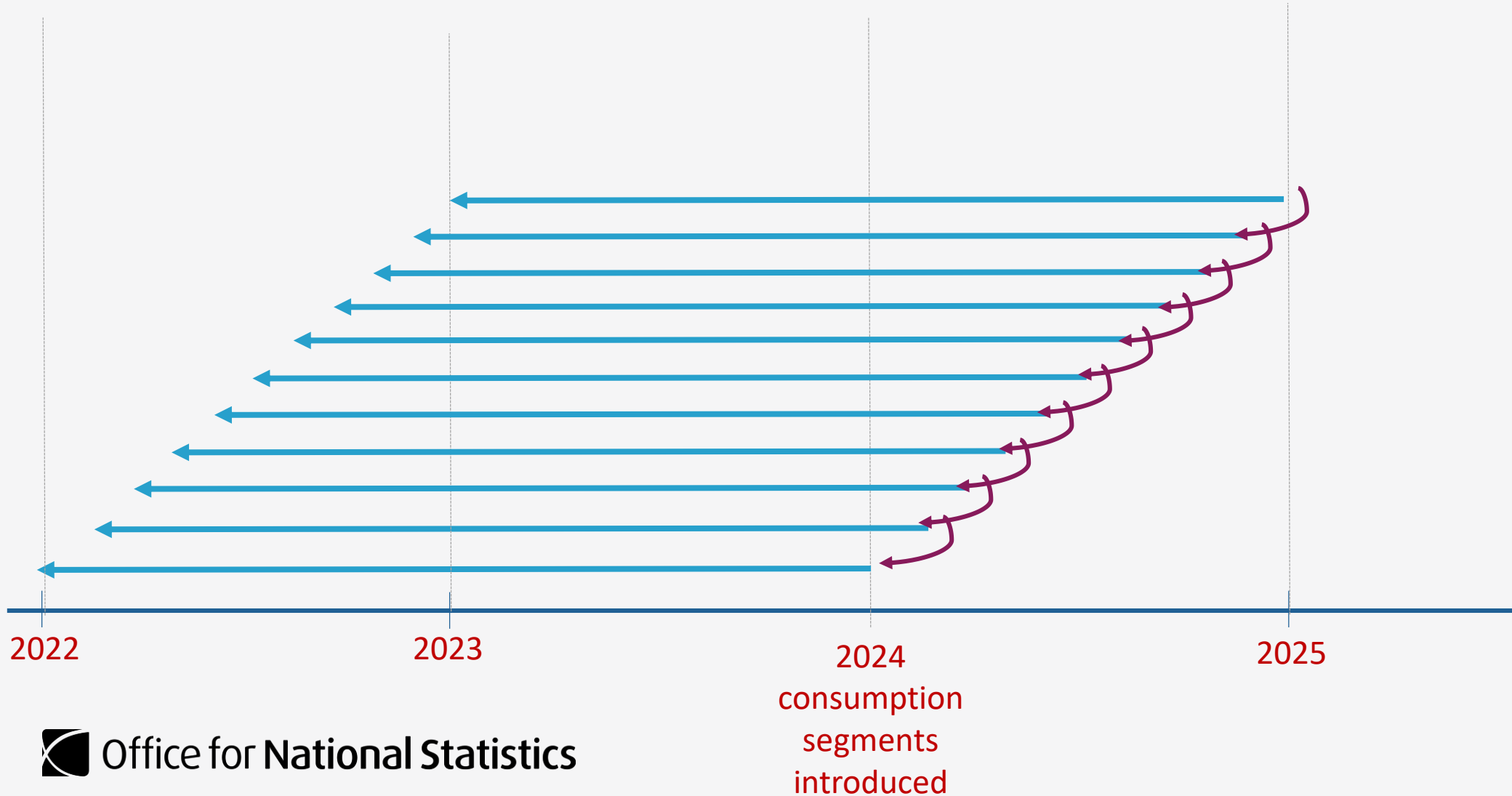


consumption
segments
introduced

3. Repeat monthly until new base month



4. Repeat process for each **consecutive year** with updated consumption segments



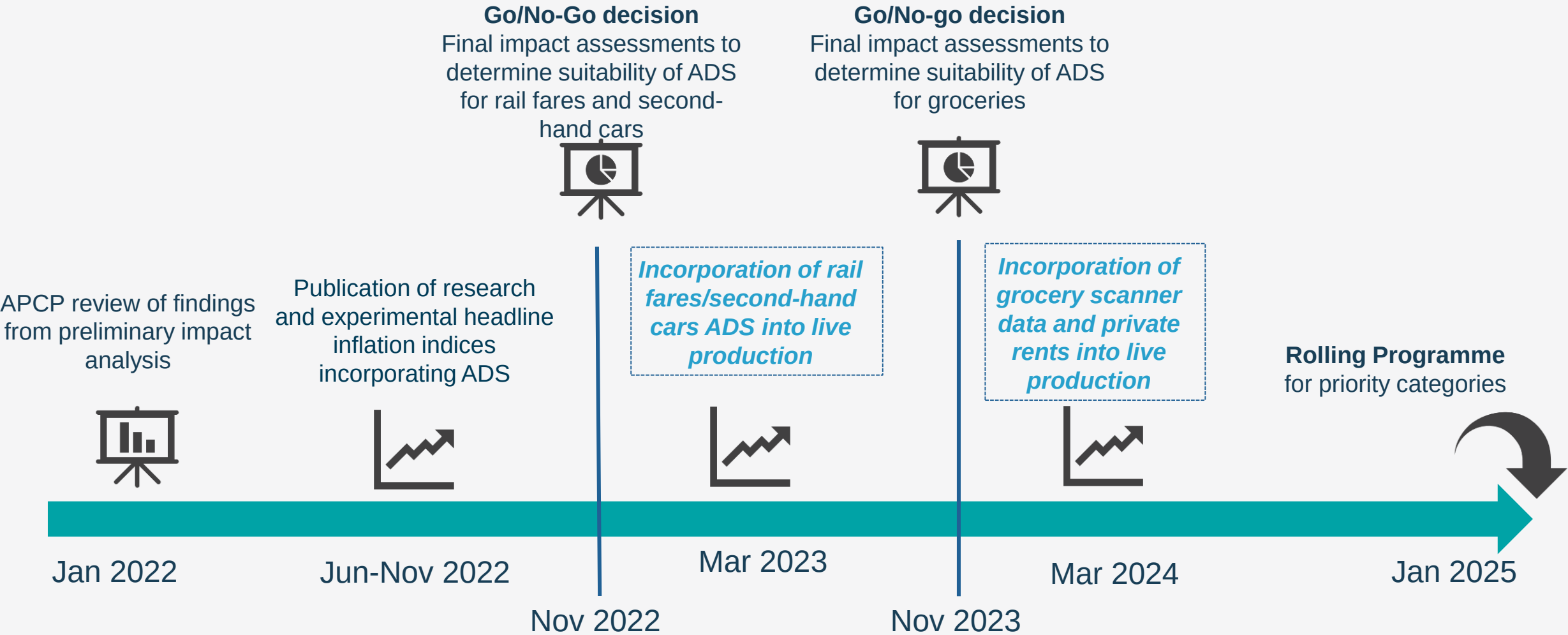
Ongoing work and next steps

The puzzle still isn't complete!

Ongoing work

- Data cleaning methods
- Continued exploration of multilateral index methods
- Interaction between multilateral methods and higher level chaining
- Exploration into use of web-scraped data, including continued work on ML-based classification, product grouping and expenditure proxies
- Building end-end system

Timeline



Questions?

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