



Using multilateral hedonic methods to capture product relaunches

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- Shrinkflation or product relaunches where the price changes proportionally less than the packing size are problematic in scanner data
- Products tend to get a new unique product identifier
- Typically matched-methods are used with scanner data, these product relaunches are missed

Cat-astrophe? Whiskas pet food criticised over shrinking servings

Cat food brand caught in shrinkflation storm as owners say new pouches and recipe leaves pets hungry



([The Guardian, 2023](#))

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- Statbel has been using scanner data from supermarkets to compile the CPI since 2015
- Method: GEKS-Törnqvist/CCDI multilateral method, with half splice on published indices with a 25-month window
- Stock keeping units as product identifiers
 - They don't capture shrinkflation
 - Solution: semi-automatic procedures using text mining and manual verification by price collectors
 - Problem: time consuming, possibility of human errors, problematic cases when time period between new and discontinued product is too large

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- Since 2022 we use scanner data for consumer electronics and household appliances in the CPI
- Specific characteristics of these segments:
 - High attrition rate of products (short life cycle)
 - Products have high entry and low exit price
 - Different features of products leaving and entering the market
- Need to take the difference between disappeared and new products into account
- Method: ITGEKS with bilateral time dummy hedonic indices
- Idea: what if we could use these multilateral hedonic methods to also capture shrinkflation or product relaunches?

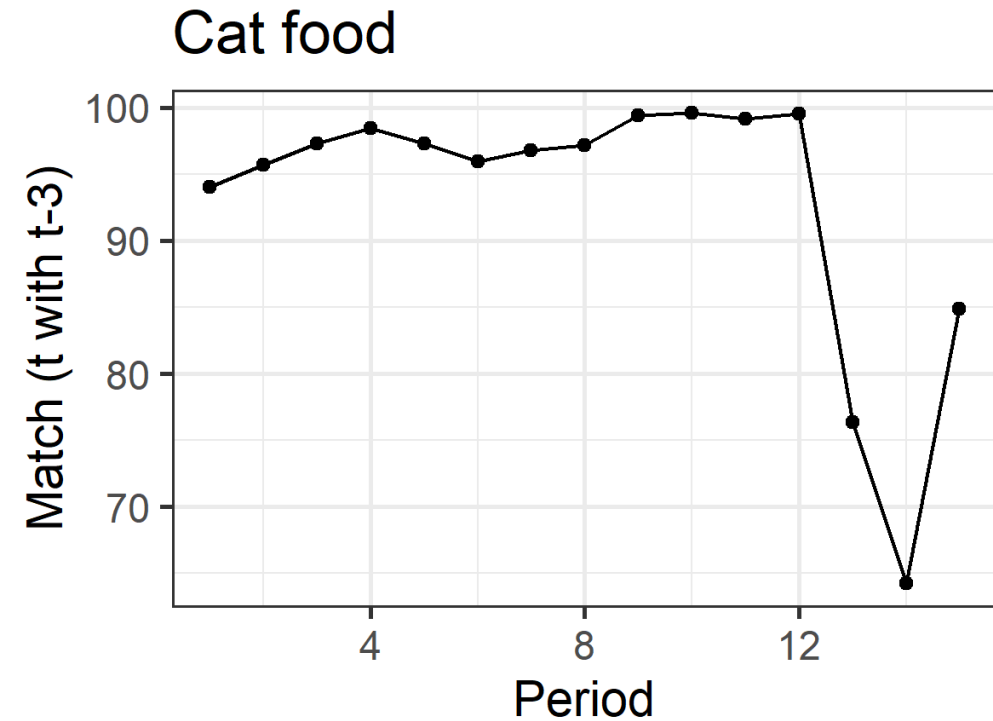
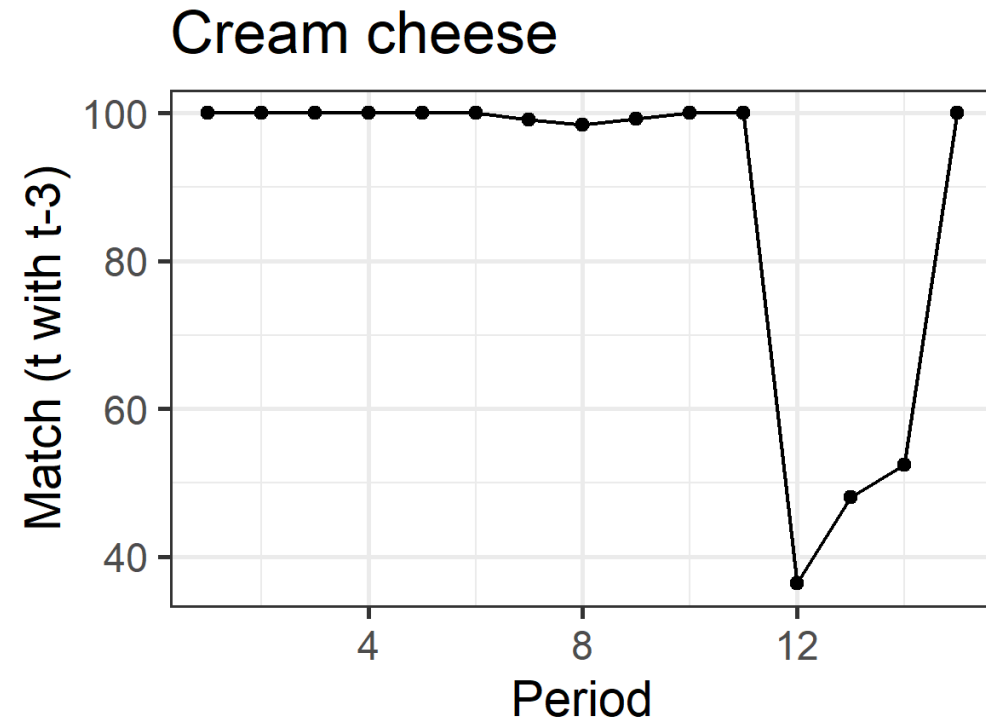
Which data to evaluate this idea?

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- Scanner data of supermarket chains used in production for the CPI/HICP:
 - A product identifier
 - Number of sales and turnover
 - Detailed product descriptions (i.e. separate variables for brand, variety and other information)
 - Separate variables for package size and the unit of measure (kilograms, litre, ...)
 - Our classification to ECOICOP
- Merged with the internal classification datasets of supermarket chains
- Product groups with problematic product relaunches (in this case shrinkflation) were examined as well as “normal” product groups

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Sales units from 3 months ago that can be matched in period t

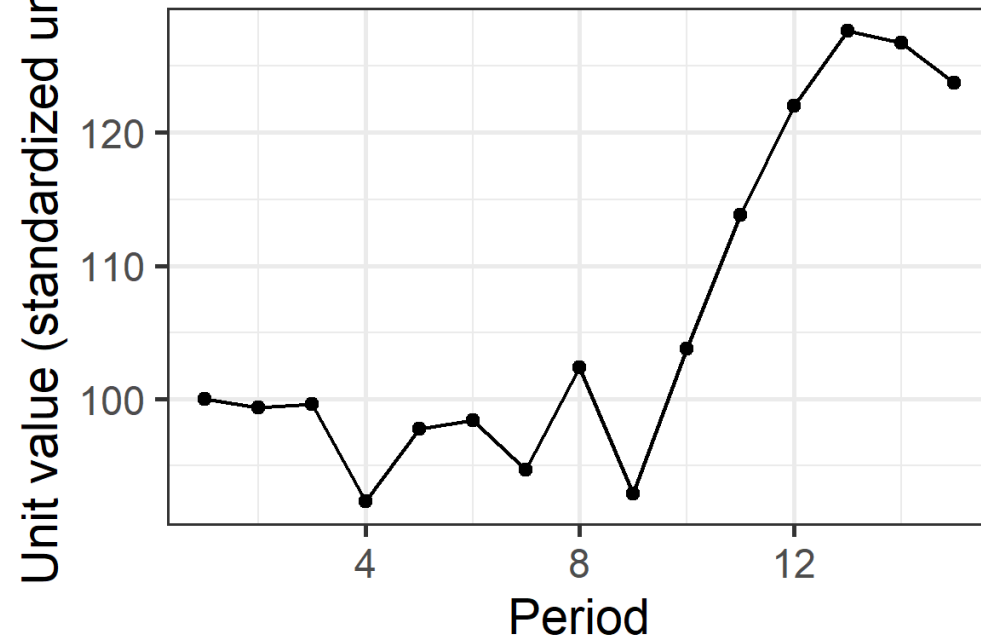


- Both segments experienced a significant product relaunch, e.g. for cream cheese in month 12, only $\pm 40\%$ of the number of sales of month 9 can be matched

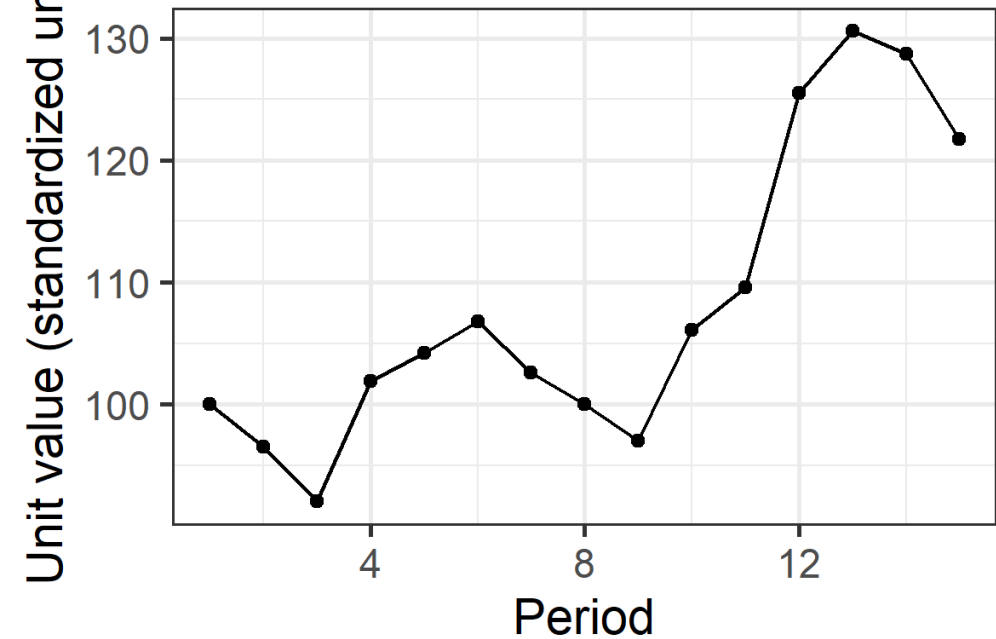
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Standardized unit value index

Cream cheese



Cat food



- Relaunches in both segments coincide with an increase in the standardised unit value price index (price/kg), indicating shrinkflation

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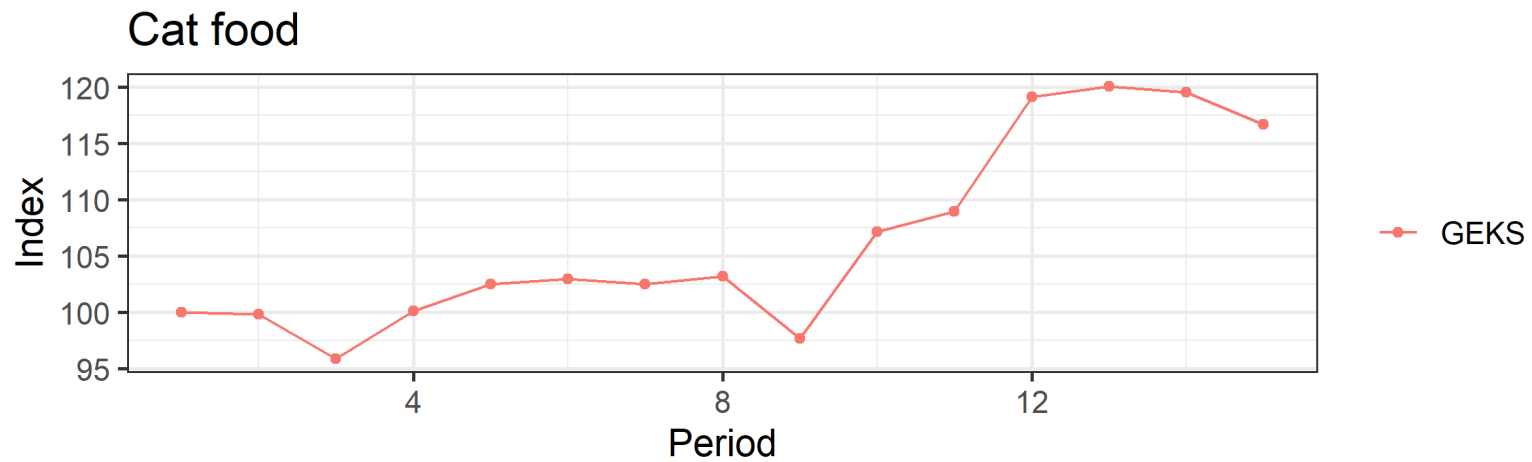
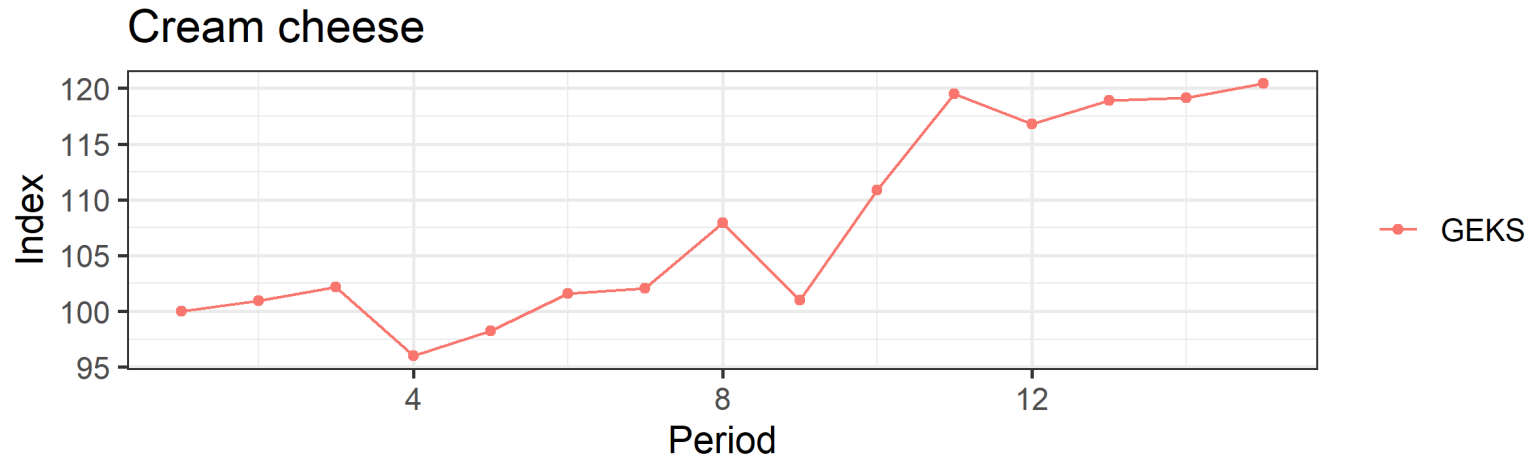
- Standard formula for the GEKS

$$P_{GEKS}^{0,t} = \prod_{l=0}^T (P^{0l} / P^{lt})^{(1/T+1)} = \prod_{l=0}^T (P^{0l} P^{lt})^{(1/T+1)}$$

- With P^{0l} and P^{lt} being Törnqvist-indices:

$$P_T^{0,t} = \prod_{i=1}^n \left(\frac{p_i^t}{p_i^0} \right)^{0.5 \left(\frac{p_i^0 q_i^0}{\sum_{j=1}^n p_j^0 q_j^0} + \frac{p_i^t q_i^t}{\sum_{j=1}^n p_j^t q_j^t} \right)} = \prod_{i=1}^n \left(\frac{p_i^t}{p_i^0} \right)^{0.5 (s_i^0 + s_i^t)}$$

- This is a matched items index using unique product identifiers



- GEKS index shows hardly any price increase at all from period 12, for cream cheese it even shows a small price decline at the moment of the product relaunch

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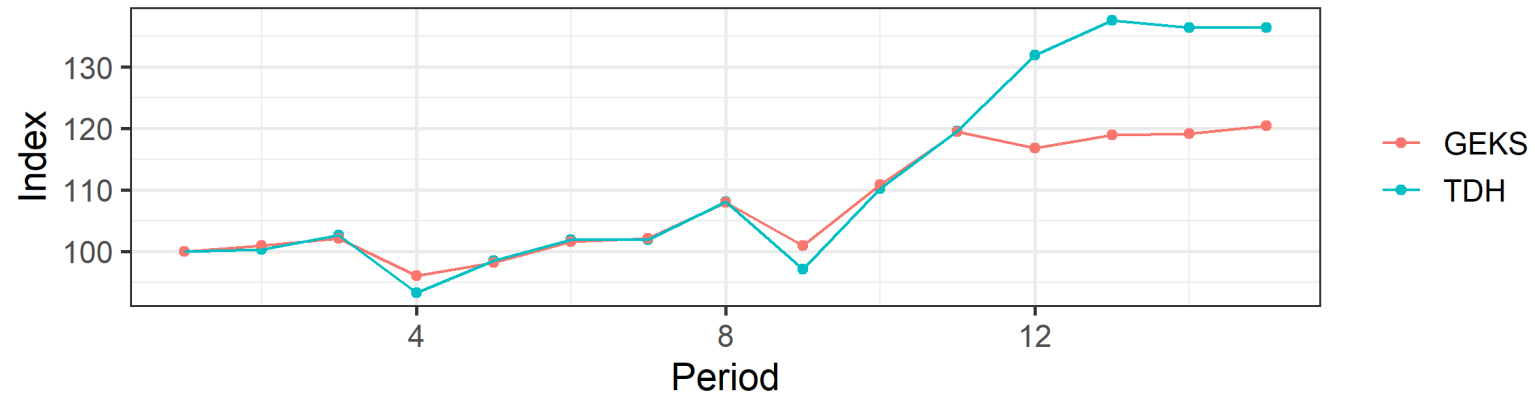
- Log-linear specification, with expenditure shares in each period serving as weights

$$\ln p_i^t = \alpha + \sum_{t=1}^T \delta^t D_i^t + \sum_{k=1}^K \beta_k z_{ik} + \varepsilon_i^t$$

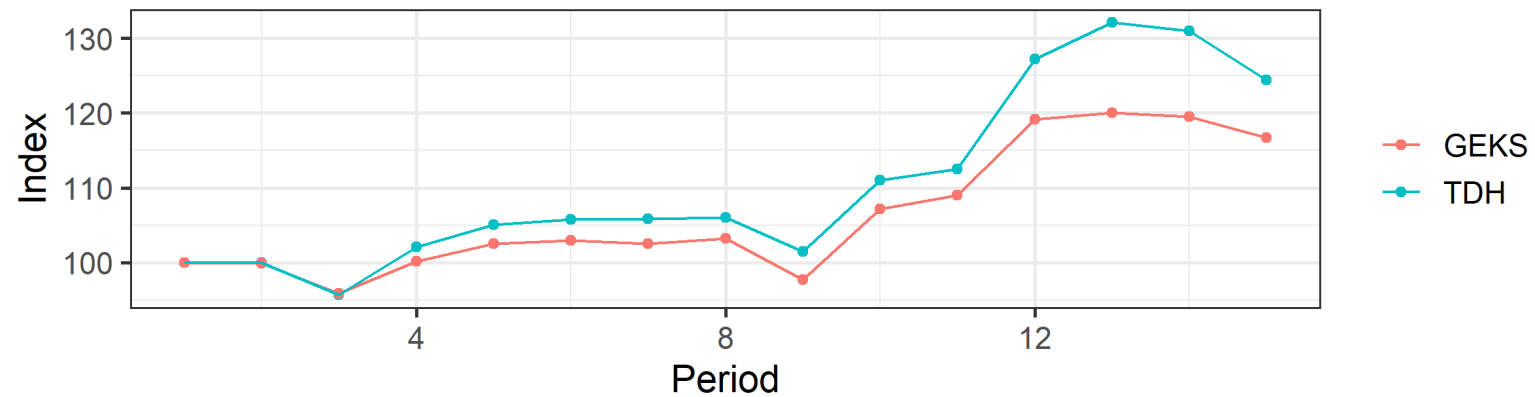
- Prices of all items (i) for several periods (t) are pooled in the same regression, on their characteristics (z_{ik}) and on dummy variables for the periods (D_i^t)
- Advantage: its simplicity, since the index follows directly from the estimated time dummy parameters
- Disadvantage: 1) it forces parameter fixity for the whole window and 2) without product churn the index does not equal the GEKS

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Cream cheese



Cat food



- TDH shows a correct price increase when the shrinkflation takes place in both product groups

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- Method we use for consumer electronics and household appliances scanner data
- ITGEKS with bilateral time dummy hedonic indices as inputs for the GEKS Törnqvist
- In the GEKS formula the Törnqvist-indices (P^{0l} and P^{lt}) are replaced with weighted bilateral time dummy hedonic indices

$$\ln p_i^t = \alpha + \delta^t D_i^t + \sum_{k=1}^K \beta_k z_{ik} + \varepsilon_i^t$$

- Advantage: fixity of the parameters only for the two periods being compared
- Disadvantage: many bilateral regressions must be run → window period of 25 months requires 300 bilateral time dummy hedonic regressions.

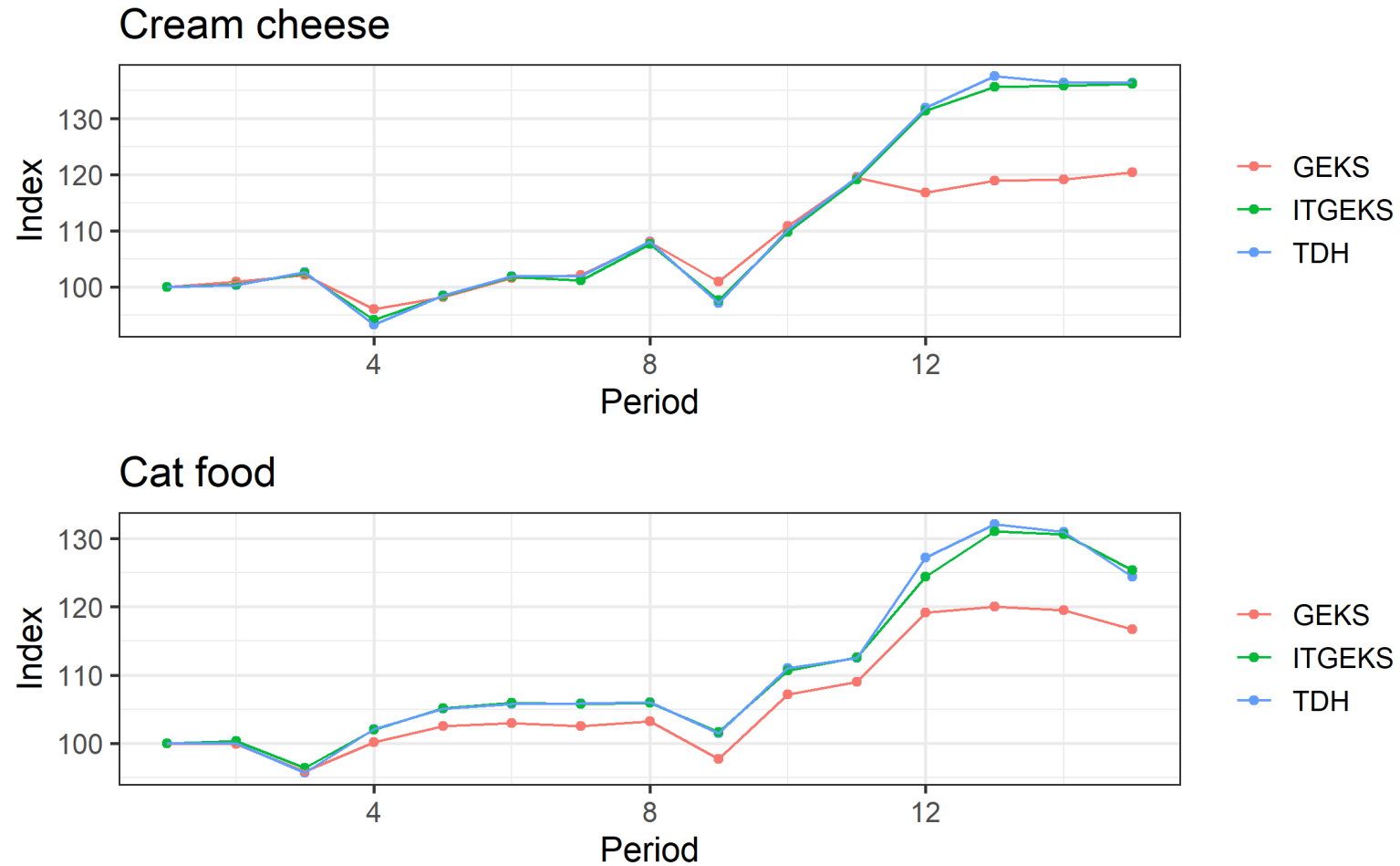
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- Which weights are used to estimate the bilateral time dummy hedonic indices?
- Mean expenditure shares for matched items (U_M^{0t}) and half expenditure shares for new (U_N^{0t}) and disappeared items (U_D^{0t}) → makes it algebraically equivalent to:

$$P_{ITGEKS}^{0,t} = \prod_{i \in U_M^{0t}} \left(\frac{p_i^t}{p_i^0} \right)^{0.5 (s_i^0 + s_i^t)} \prod_{i \in U_D^{0t}} \left(\frac{\hat{p}_i^t}{p_i^0} \right)^{0.5 (s_i^0)} \prod_{i \in U_N^{0t}} \left(\frac{p_i^t}{\hat{p}_i^0} \right)^{0.5 (s_i^t)}$$

- Advantage compared to TDH: without product churn the index equals the GEKS

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- Difference between the ITGEKS and the TDH is limited. Explained by the high R squared values (0.903 - 0.963), also indicates that parameter fixity is not that problematic.

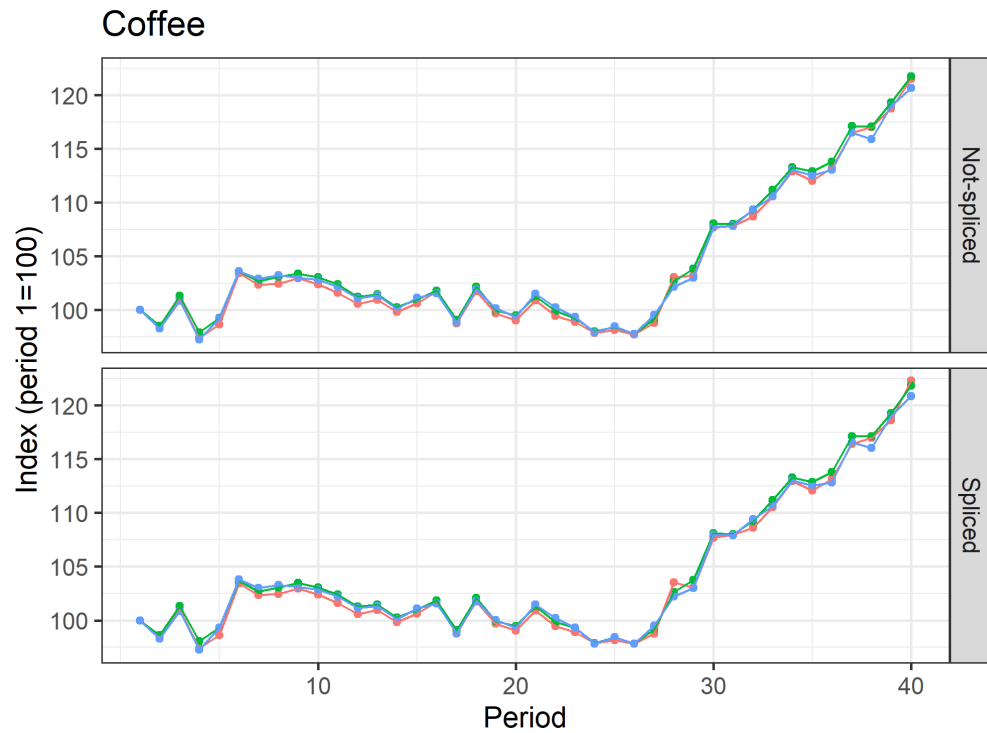
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- Extended the analysis to 4 randomly selected product groups where our “semi-automatic” had not found any problematic product relaunches: coffee, chocolate, soft drinks and breakfast cereals.
- These segments have product attrition, but no product relaunches that could result in a biased index
- Traditional GEKS Törnqvist could serve as a benchmark
- 40 months period, requires 780 bilateral time dummy hedonic regressions in the ITGEKS
- Compare index using entire window to an index with splicing (HASP-25)

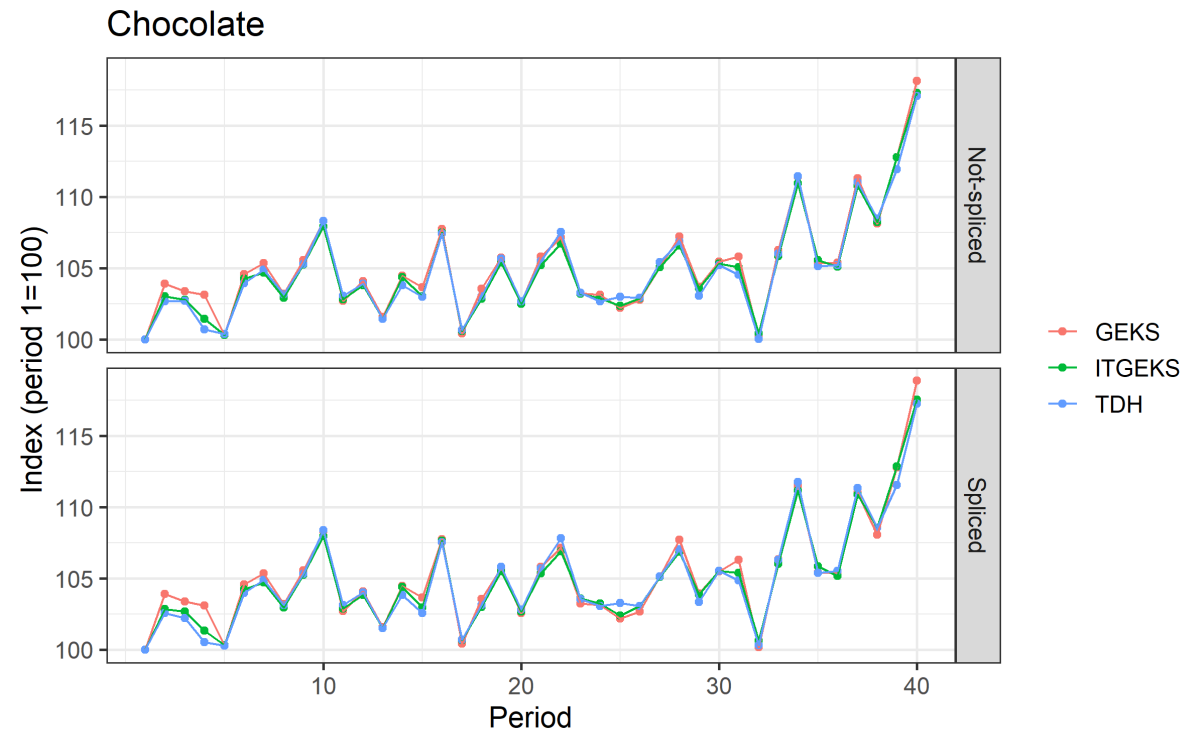
Product groups without shrinkflation

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- All 4 segments have similar conclusions and splicing doesn't make a difference



— GEKS
— ITGEKS
— TDH

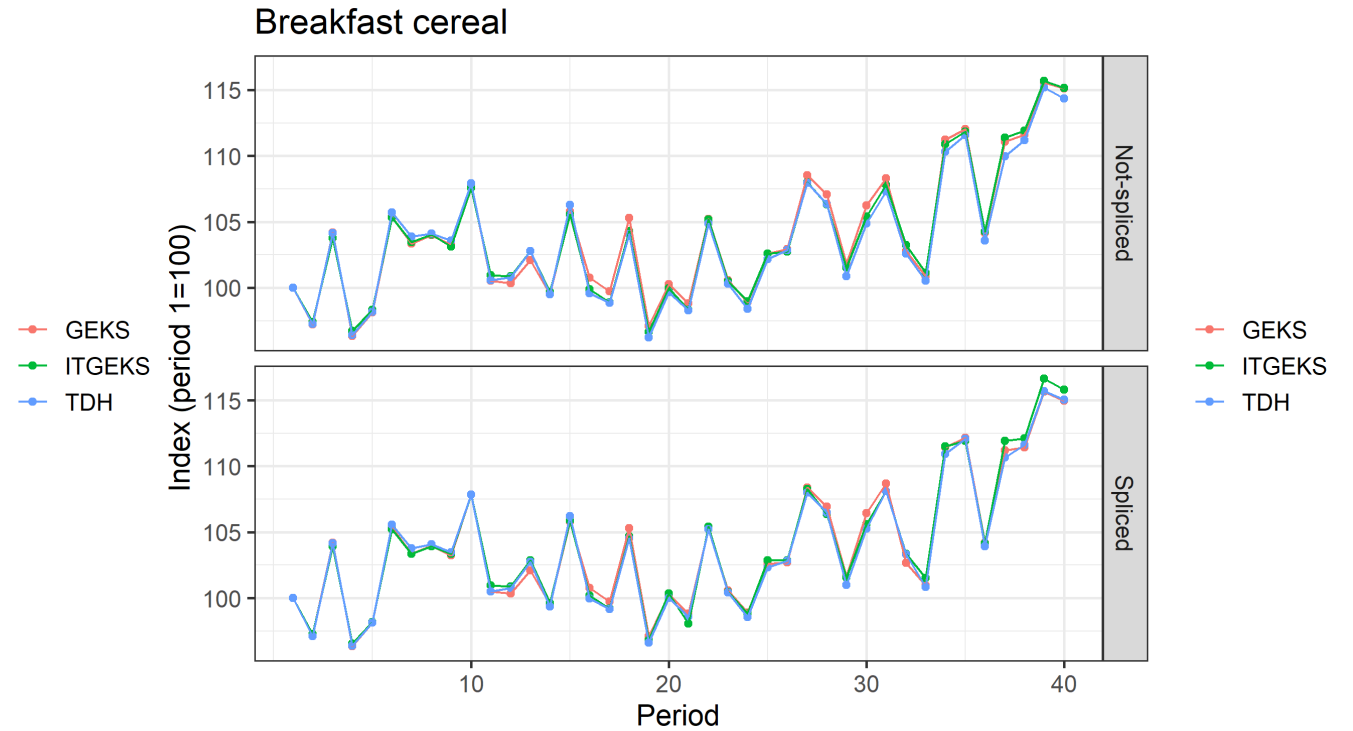
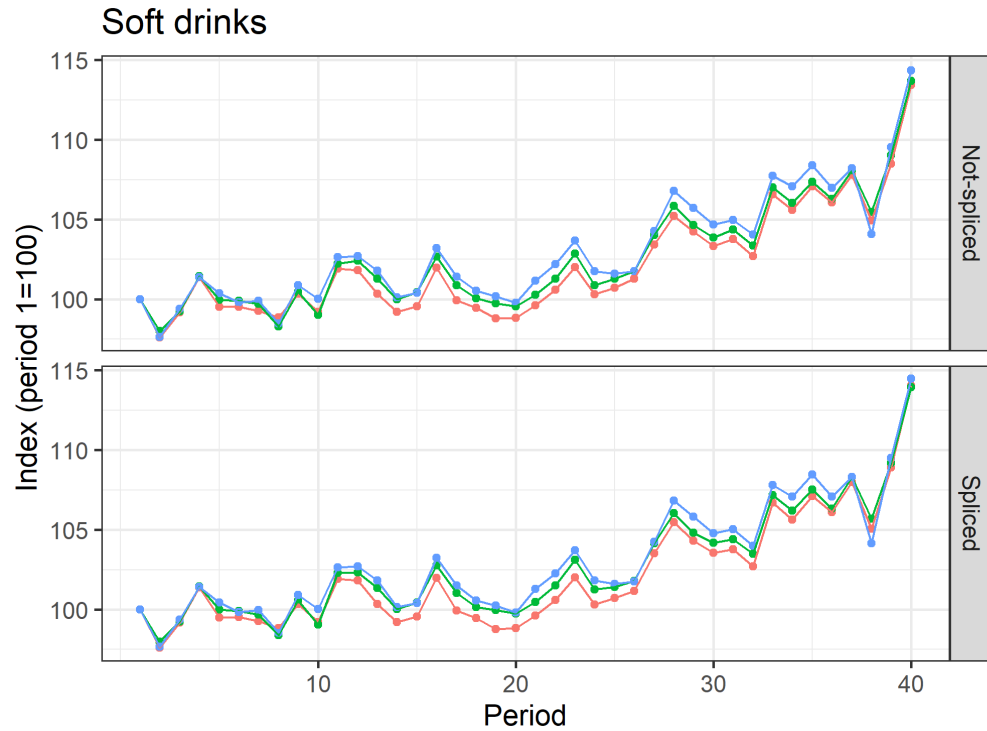


— GEKS
— ITGEKS
— TDH

Product groups without shrinkflation

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- Conclusion
 - using unique product identifiers to compile the index can lead to a (downward) bias
 - multilateral methods which use hedonics are applicable to supermarket scanner data
 - these methods can capture (certain) product relaunches or shrinkflation
 - difference between ITGEKS and the much easier to compile TDH was limited
 - splicing does not change the conclusions
 - even if countries lack sufficient metadata to do hedonics: still useful to compile a matching sales indicator and a standardised unit value price index to identify potential problematic groups
- Next steps:
 - extend our analysis to other product groups
 - compare with methods that use stratification or product clustering