

Multilateral indices and the relaunch problem

Product clustering and alternative solutions

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Relaunch problem

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Relaunches

- Product replaced by (almost) identical one
- Price change often larger than quality change
- Matched-model methods miss price changes due to relaunches (if nothing is done)
- Here focus on: Geary-Khamis, TPD and GEKS-Törnqvist



Product clustering



Method

- Items combined into product clusters
- Cluster prices computed as unit values
- Unit value bias for heterogeneous items
- Crucial: Suitable cluster definition



Example: Unit value bias

	Price (0)	Price (1)		Quantity (0)	Quantity (1)
Product 1	10		12	100	300
Product 2	18		20	300	100
Unit value price	16		14		



Clustering: pros and cons

Pros

- Broad applicability
- Well established
- Easy

Cons

- Unit value bias (heterogeneous strata)
- Analysis below cluster level impossible (loss of details)



Imputation

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Method

- Price estimated for non-sold items
- Aim: solve relaunch problem rather than to complete data -> might give interpretation problems
 e.g. imputed price for a nonseasonal item
- Non-trivial choice between imputation methods
- Methods available for some indices (e.g. GEKS-Törnqvist) but less well known for others (e.g. TPD)



Relation imputation and clustering

Imputation methods exist that give the same results as product clustering

- GEKS- Törnqvist:
 - clustering same as imputing each price with unit value (unobserved **and** observed prices)
- TPD and Geary-Khamis:

Same as for GEKS- Törnqvist, but also quantities need to be imputed



Relation imputation and clustering

- Unit value bias in clustering corresponds to a rigorous imputation approach
- In paper: new imputation methods that:
 - mimic product clustering
 - less rigorous replacement of prices

If all prices are observed -> No adjustment (contrary to clustering) If many prices are missing -> Similar results as clustering



Product matching



Method

- Each new product matched with a disappeared product
- Replacement and replaced products should be similar
- Semi-automatic procedures proposed: text mining and manual analysis
- Automatable methods needed for transaction data
- Solution needed if number of new and disappeared products are unequal
- In paper: a simple procedure with arbitrary choices

Matching

<u>Pros</u>

No adjustment of prices and quantities (contrary to clustering and imputation)

<u>Cons</u>

Arbitrary choices (matching procedure)

Interpretation of the matched products



Simulation



Setup

- 12 month data TV's, chocolate and potato products
- Fixed population: products that have been sold each month
- Simulated relaunches (change of product ID)
- Monte carlo simulation (100 replicates)
- Comparison:

no correction, clustering, imputation, matching versus 'true' index without simulated relaunches

- Criterion: median abs difference of index values



Scenario 1: random relaunches

TV's : Median distance from true index

	GEKS-Törnqvist	Geary Khamis	TPD
No correction	1.18	<u>1.32</u>	<u>0.74</u>
Imputation	0.65	4.74	4.83
Matching	1.38	1.48	2.33
Clustering	4.56	5.52	5.63

Chocolates: Median distance from true index

	GEKS-Törnqvist	Geary Khamis	TPD
No correction	<u>0.37</u>	<u>0.16</u>	<u>0.20</u>
Imputation	0.47	0.94	1.47
Matching	0.46	0.58	0.98
Clustering	0.78	1.07	1.03



Scenario 1: random relaunches

Potatoes: Median distance from true index			
	GEKS-Törnqvist	Geary Khamis	TPD
No correction	1.96	1.40	1.22
Imputation	1.78	1.31	1.15
Matching	<u>0.48</u>	<u>0.53</u>	<u>0.51</u>
Clustering	1.27	1.37	1.37

- Best correction method:
 - For TV's and Chocolates: 'No correction'
 - For Potatoes: 'Matching'
- Matching always better than clustering



Scenario 2: Non-random relaunches

- Relaunches occur at one time period for randomly selected
 75% of all items.
- Each relaunch goes along with a simulated, permanent price increase by 20%.



Scenario 2: Non-random relaunches

TV's : *Median distance from true index*

	GEKS-Törnqvist	Geary Khamis	TPD
No correction	11.14	7.39	10.54
Imputation	<u>1.24</u>	3.81	5.11
Matching	2.10	<u>1.44</u>	<u>3.62</u>
Clustering	5.63	6.66	6.70

Chocolates: Median distance from true index

	GEKS-Törnqvist	Geary Khamis	TPD
No correction	14.77	11.90	14.61
Imputation	1.83	<u>0.34</u>	3.04
Matching	<u>0.20</u>	4.09	<u>0.54</u>
Clustering	0.74	0.85	0.81



Scenario 2: Non-random relaunches

Potatoes: Median distance from true index			
	GEKS-Törnqvist	Geary Khamis	TPD
No correctio	n 13.33	10.42	13.40
Imputation	<u>0.83</u>	2.06	1.28
Matching	0.95	1.58	<u>1.05</u>
Clustering	1.10	<u>1.31</u>	1.20

- Correction method necessary ("no correction" gives large errors)
- Best correction method depends on data set and index method:
 - * Matching (5 cases)
 - * Imputation (3 cases)
 - * Clustering (1 case)
- Matching better than clustering for 7 out of 9 cases



Conclusion

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- Correction for relaunches needed (given a price increase)
- Clustering easy, well-understood, broad applicable, but unit value bias for heterogeneous strata
- Imputation and matching mostly give better results in a simulation study (especially matching)
- Drawbacks: arbitrary choices, unnatural to 'add' values to transaction data, interpretation difficulties.

Thank you!

