

Experimental clothing indexes using Australian web scraped data



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Australian Bureau of Statistics
Informing Australia's important decisions



- ▶ ABS in a transformation environment – seeking ways to utilise ‘big data’ for compilation of economic statistics
- ▶ March quarter 2014 – Transactions (scanner) data introduced into the Australian CPI
- ▶ December quarter 2017 – Expansion of transactions data and introduction of multilateral index methods

- ▶ What alternative big data sources are available to obtain price information?
- ▶ Web scraping – the extraction and transformation of unstructured data from the web into structured data
- ▶ The ABS is currently expanding its use of web scraped data in the CPI
 - Progressively incorporated since March 2017

Background

- ▶ Clothing and footwear – high priority for ABS
- ▶ Challenges with clothing and footwear:
 - High collection and data editing costs
 - Competitive market environment
 - Strong seasonality

Table 1: Typical data structure

Date	Retailer	Category	Item Name	Price	Item Count
10-Jul-16	Retailer ABC	Women's Tops	Short Sleeve Regular Shirt "Brand XYZ"	\$55.00	1
13-Jul-16	Retailer ABC	Women's Tops	S/S Regular Shirt Brand XYZ	\$55.00	1
13-Jul-16	Retailer ABC	Women's Tops	Short Sleeved Oversized Shirt "Brand XYZ"	\$55.00	1
13-Jul-16	Retailer ABC	Women's Tops	Long Sleeve Shirt "Brand XYZ"	\$65.00	1
28-Jul-16	Retailer ABC	Women's Tops	L.S. Shirt "Brand XYZ"	\$65.00	1
28-Jul-16	Retailer ABC	Women's Tops	Short-Sleeve Reg Shirt "Brand XYZ"	\$55.00	1
07-Jul-16	Retailer ABC	Women's Tops	Short Sleeved O/S Shirt "Brand XYZ"	\$55.00	1

Product Definition

- ▶ Matched model indexes (e.g. Jevons, Törnqvist) rely on the ability of price analysts to identify which items are identical (i.e. homogenous) from the consumer's perspective
- ▶ Broader product definitions improve product matching over time but increase the risk of average price bias
- ▶ 3 alternative product definitions considered:
 - Item Name
 - Brand + Product Type
 - Brand + Product Type + Product Characteristics

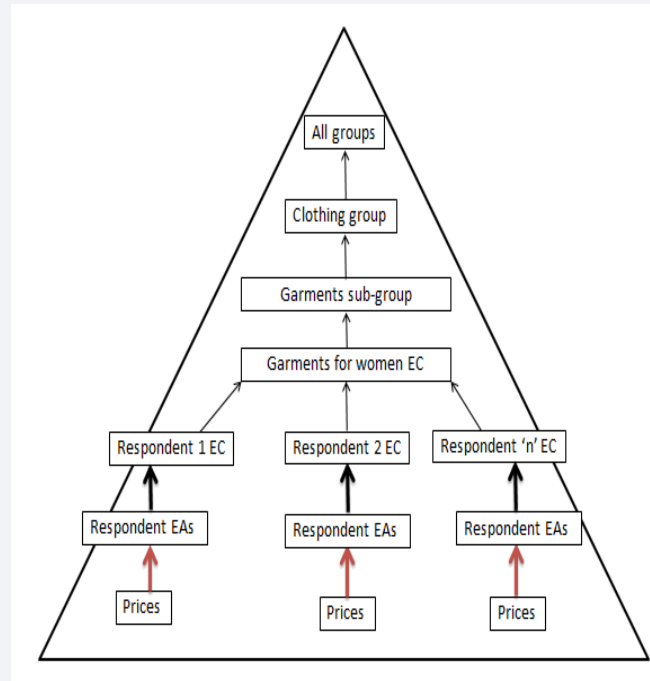
Product Extraction

- ▶ A keyword approach was used to extract potentially important product information from item name strings
- ▶ Product information extracted included:
 - Brand
 - Product Type (e.g. t-shirt, dress, shorts)
 - Product characteristics (e.g. sleeve length, material, length)

Brand	Type	Characteristics	Item Name
Brand XYZ	Shirt	Short_Sleeve~Regular	Short Sleeve Regular Shirt "Brand XYZ"
Brand XYZ	Shirt	Short_Sleeve~Regular	S/S Regular Shirt Brand XYZ
Brand XYZ	Shirt	Short_Sleeve~Oversized	Short Sleeved Oversized Shirt "Brand XYZ"
Brand XYZ	Shirt	Long_Sleeve	Long Sleeve Shirt "Brand XYZ"
Brand XYZ	Shirt	Long_Sleeve	L.S. Shirt "Brand XYZ"
Brand XYZ	Shirt	Short_Sleeve~Regular	Short-Sleeve Reg Shirt "Brand XYZ"
Brand XYZ	Shirt	Short_Sleeve~Oversized	Short Sleeved O/S Shirt "Brand XYZ"

Aggregation Structure

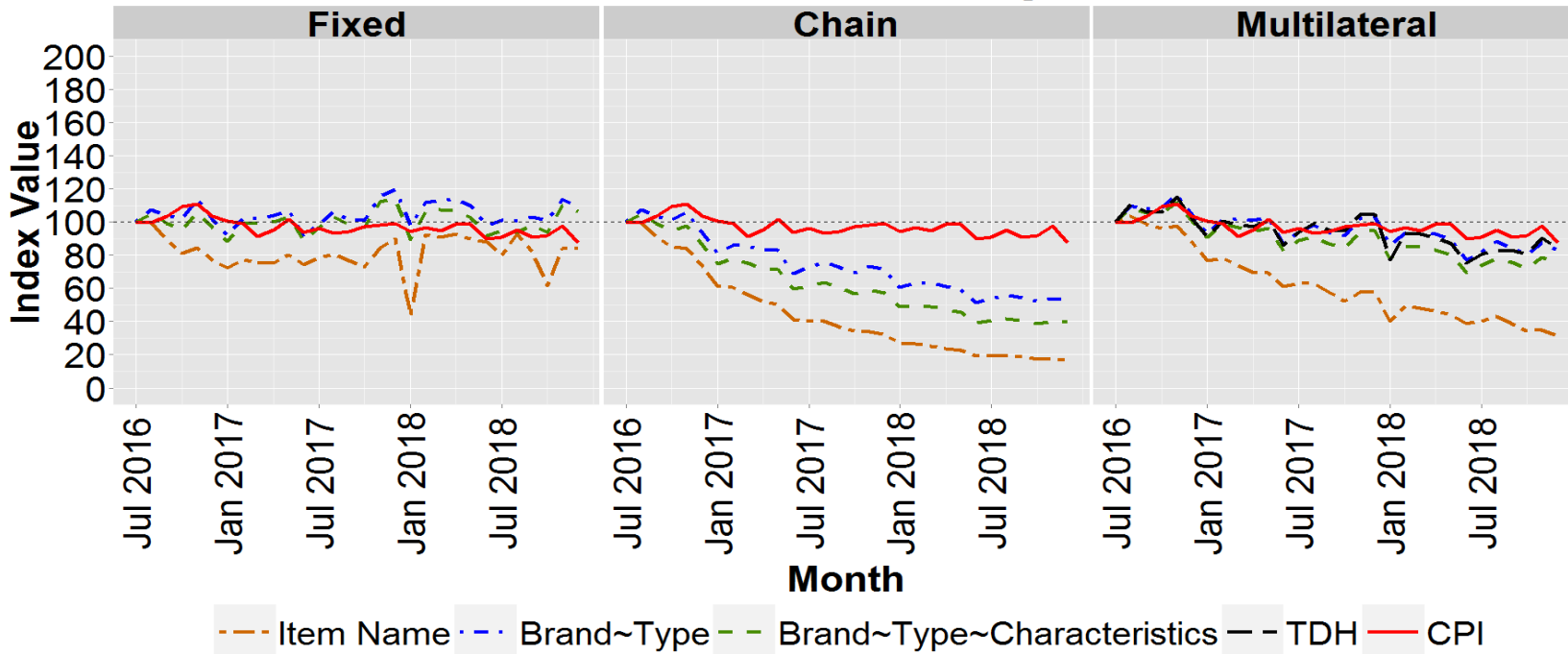
- ▶ ABS currently aggregates clothing and footwear products across retailers to derive elementary aggregates (EAs)
- ▶ This presentation instead aggregates products to EAs within each retailer
- ▶ Aggregation across retailers is carried out at the Expenditure Class (EC) level



- ▶ Unweighted index methods are required since web scraped data does not contain expenditure or quantity information
- ▶ Multilateral index methods can be used to match products across multiple time periods and resolve the “chain drift” problem with chained indexes
- ▶ 2 unweighted multilateral index methods are considered:
 - GEKS-Jevons (GEKS-J)
 - Time dummy hedonic (TDH) model with OLS weights

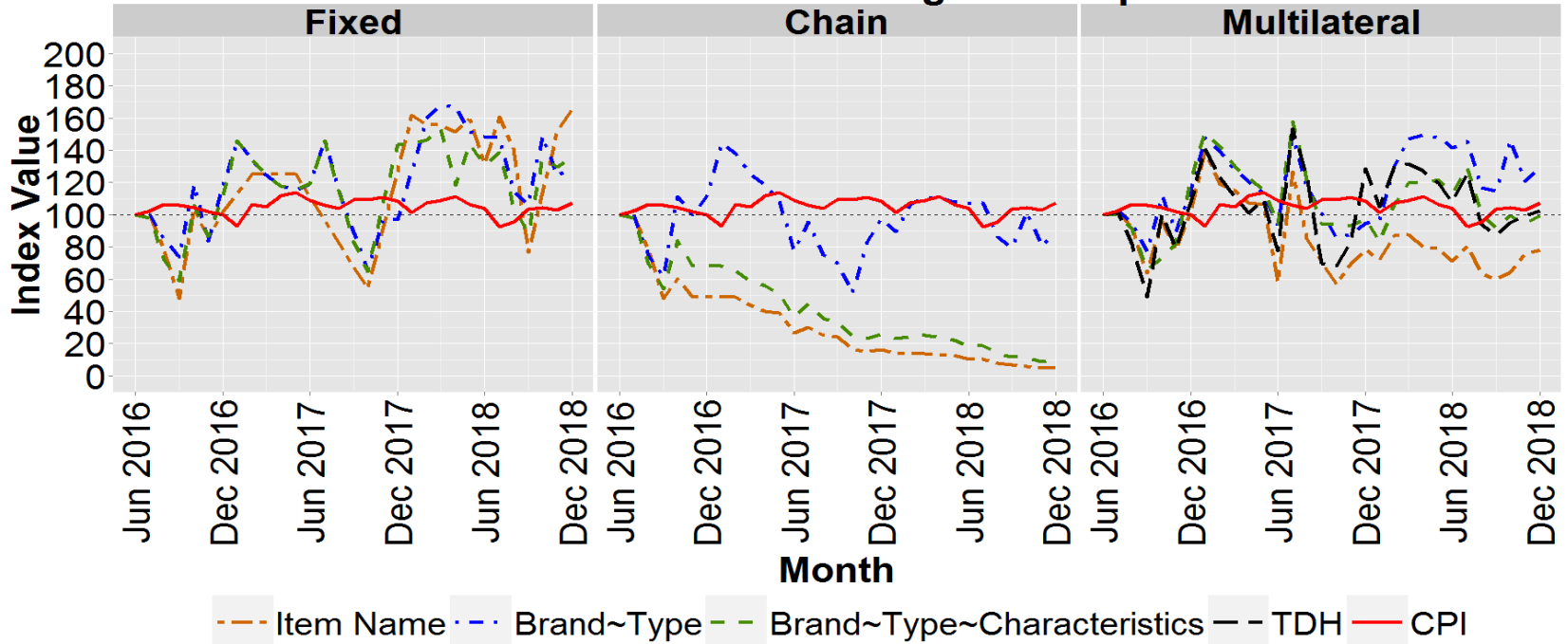
Elementary Aggregate Results

Retailer 5 - Womens T Shirts



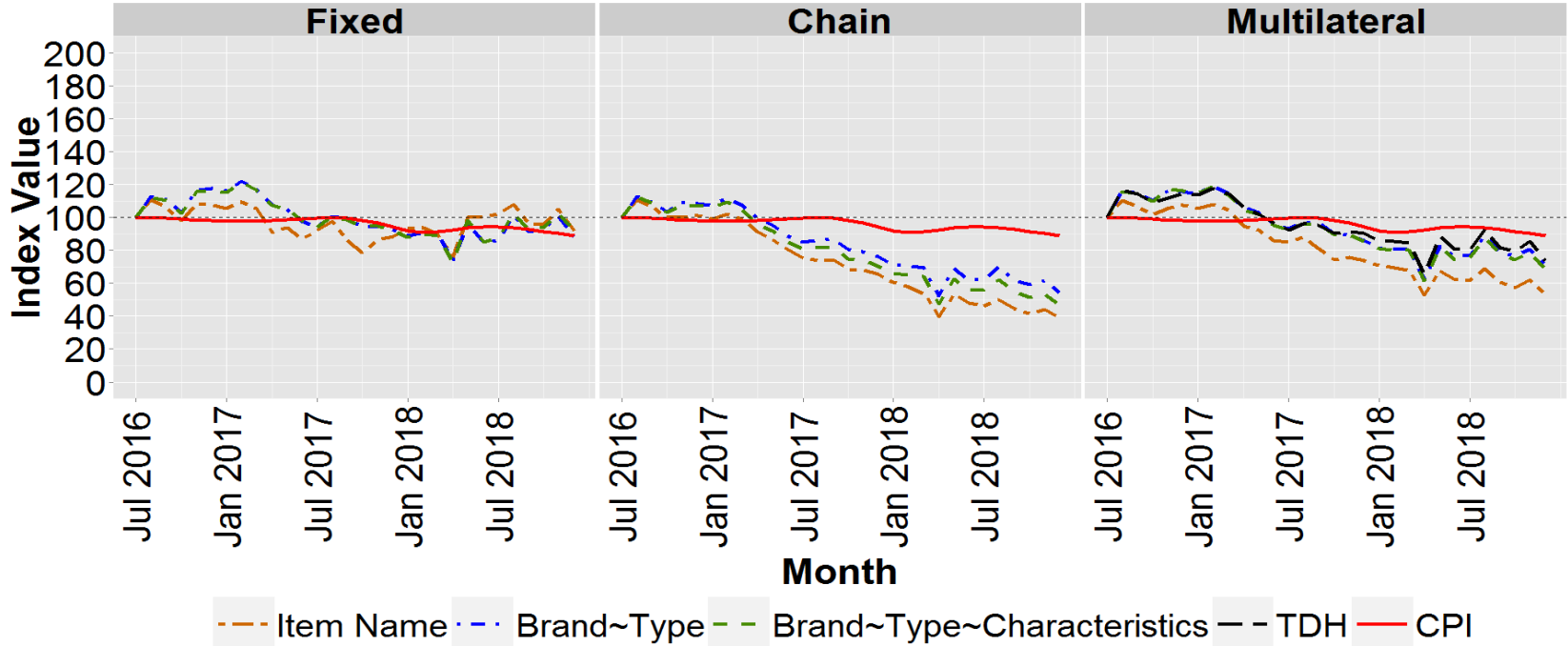
Elementary Aggregate Results

Retailer 2 - Mens Cardigans Jumpers



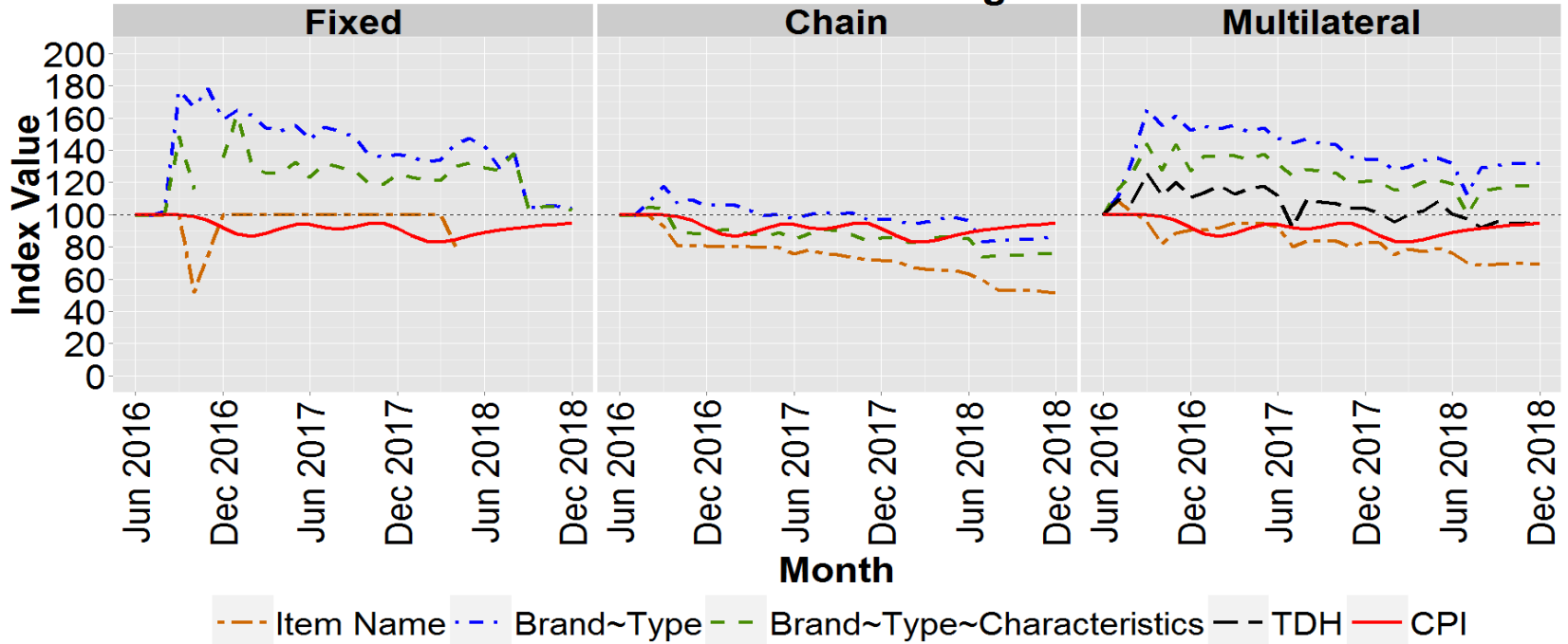
Elementary Aggregate Results

Retailer 12 - Womens Casual Footwear



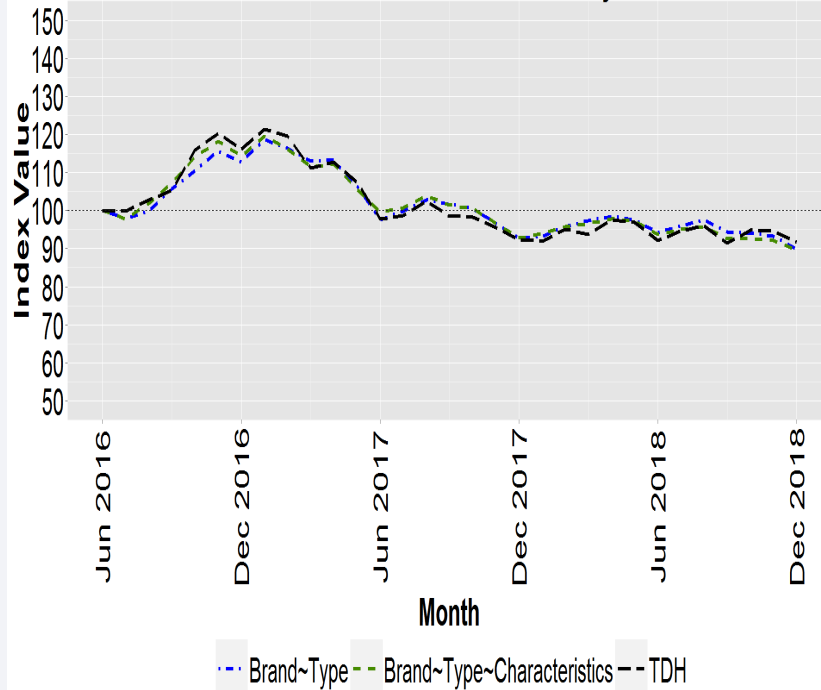
Elementary Aggregate Results

Retailer 17 - Accessories Bags Briefcases



Expenditure Class Results

Garments for Women - Monthly

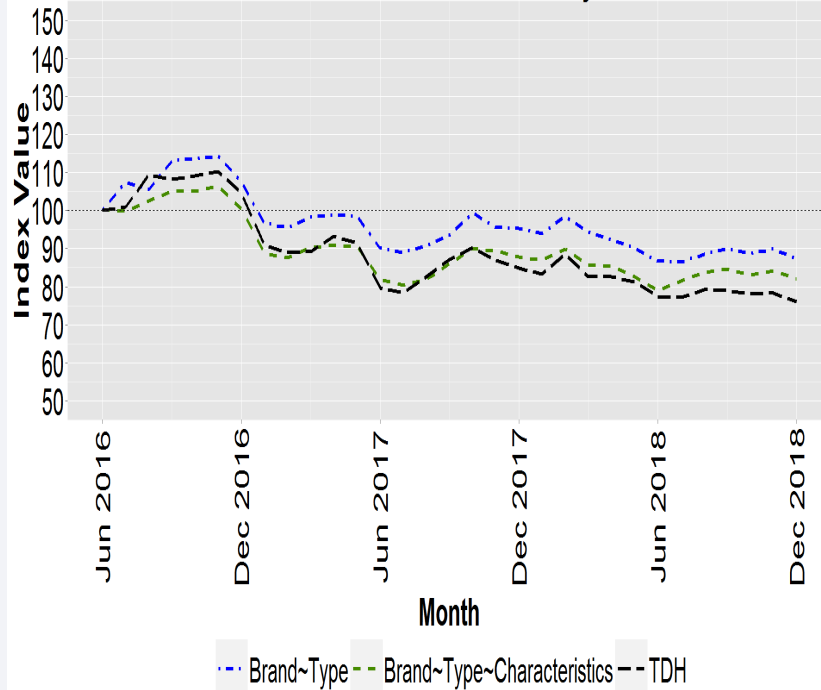


Garments for Women - Quarterly

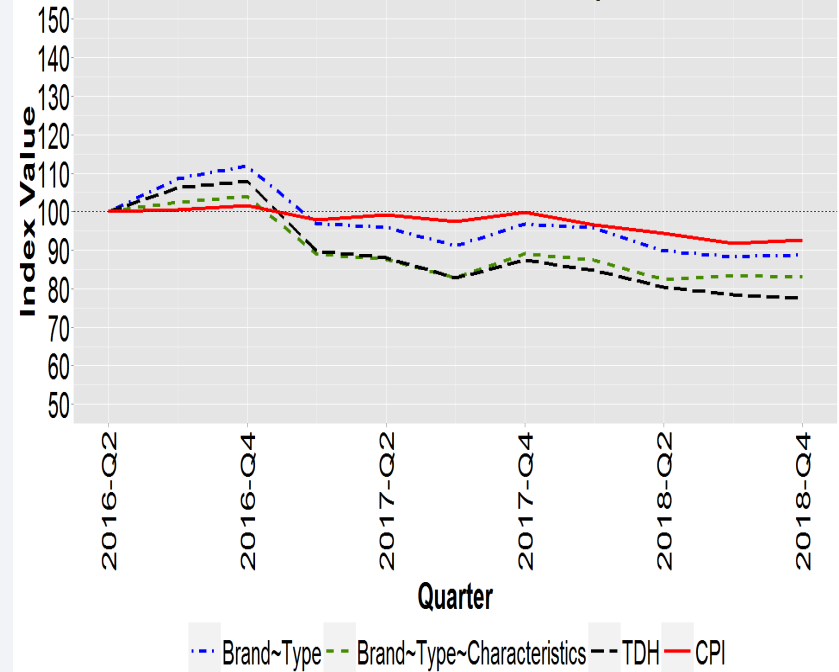


Expenditure Class Results

Footwear for Men - Monthly



Footwear for Men - Quarterly

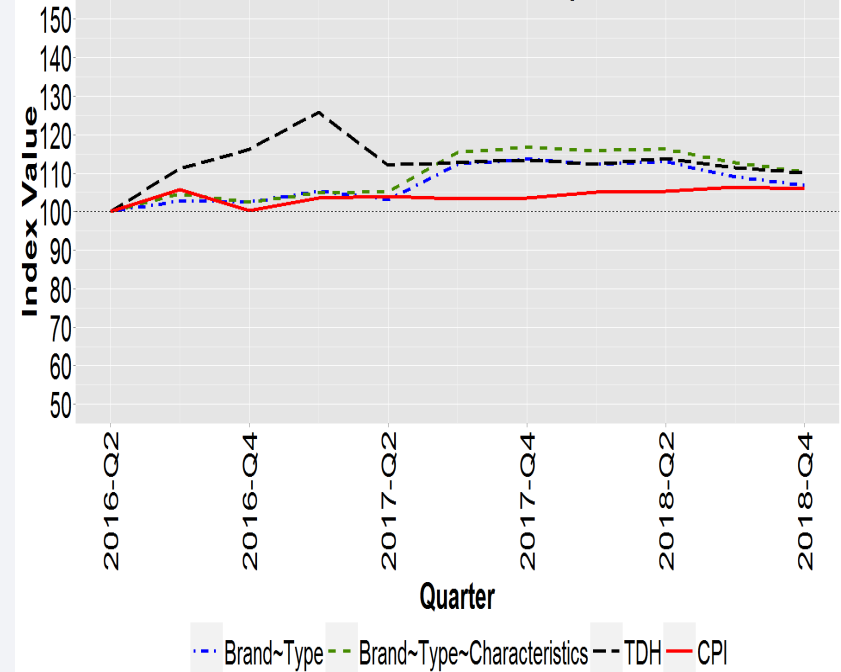


Expenditure Class Results

Accessories - Monthly



Accessories - Quarterly



Conclusions

- ▶ Pre-processing to form ‘clustered’ homogenous products is one viable strategy for NSOs to consider for ‘dynamic’ basket categories
- ▶ At the elementary level, our clothing results exhibit downward drift for chained indexes
- ▶ Fixed and multilateral indexes produced the ‘most plausible’ results with broader clothing product definitions

Conclusions

- ▶ Characteristic extraction more difficult with some footwear and accessory indexes – sparse text data means some heterogeneity still exists in our broader product definitions
- ▶ At the published level, experimental multilateral results broadly comparable with CPI equivalent

Further development

- ▶ Web scrapers maintained by ABS Prices Branch – funding attempts to expand across organisation
- ▶ Alternative strategies for forming clustered homogenous products
- ▶ Alternative strategies for respondent aggregation
- ▶ Alternative strategies for weighting individual products within clustered homogenous definitions

A decorative graphic on the left side of the slide, consisting of a series of overlapping triangles in various colors (blue, yellow, red, green, purple) and a background of binary code (0s and 1s) and a line graph.

Questions?