

Dynamic pricing as a challenge for Consumer Price Statistics

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Overview

- What is dynamic pricing?
- Goals of the analysis
- Price collection via internet for the German CPI/HICP
- Set-up of the study
- Results
- **■** Implications for Consumer Price Statistics
- Outlook



What is dynamic pricing?

- Application of automatic algorithms to change prices in short intervals due to market conditions and parameters indicating consumers willingness to pay
- Possible parameters used to set prices
 - Calendar effects (holidays, time of the day)
 - Weather
 - Prices of competitors
 - Devices as indicator for the individual willingness to pay?
 - _ ...
- Dynamic pricing: price changes in time
- Individualized pricing: different prices for different consumers



Goals of the analysis

- Dynamic pricing is analysed, individualized pricing not
 - To investigate the existence and the extent of individualized pricing would require more complex study designs (different consumer profiles)
- To which extent is dynamic pricing applied by online shops?
 - Identification of online shops applying dynamic pricing
 - Frequency and level of price changes for products relevant for CPI/HICP



Price collection via internet for CPI/HICP

- Centralized price collection for goods in online shops
 - Approx. 10.000 single prices per month
 - Share in whole basket: approx. 5%, increasing...
- Centralized price collection for services
 - Internet as data source
- Way of price collection
 - To a large extent done manually at one time per month
 - Automatized for certain fields (via web scraping)
 - passenger transport by railway, rental cars, long-distant coaches, online pharmacies



Price collection via internet for CPI/HICP

- Sample design
 - 2,680 products out of the sample of CPI/HICP
 - Different product groups
 - 14 online shops
- Automatized price collection via web scraping
- Hourly collection of prices at constant times
- Observed period: 3 months (09.12.2016 06.03.2017)
- Overall: 2097 prices per product, 5,559,933 single prices



Set-up of the study

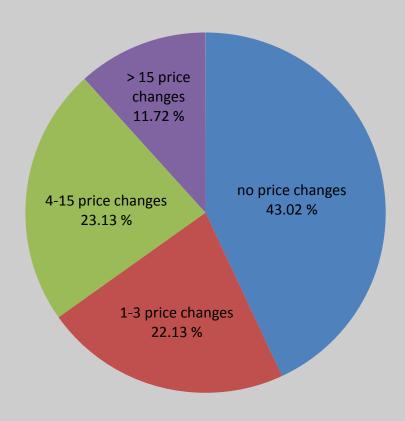
Input Java Output

- Java
 - Realization of program logic
 - Data cleaning
 - Read/Store data from database
- Selenium
 - Extraction of information
 - XPaths are used to find information on respective website
 - Plug-in for common internet browsers
- MySQL database is used for input data and to store extracted data
- Windows Scheduled Tasks is used to start the automation at a certain time



Results (1)

Price changes per product (observed period of 3 months)





Results (2)

| Shop | Share of | number | | | |
|---------|----------|--------|--------|--------|-------------------------|
| | 0 | 1 - 3 | 4 - 15 | >15 | price series overall |
| Shop 1 | 7.91% | 9.30% | 35.35% | 47.44% | 215 |
| Shop 2 | 47.94% | 19.50% | 20.39% | 12.16% | 559 |
| Shop 3 | 50.72% | 26.09% | 21.26% | 1.93% | 207 |
| Shop 7 | 54.39% | 19.59% | 16.33% | 9.68% | 888 |
| Shop 9 | 4.88% | 29.27% | 48.78% | 17.07% | 123 |
| Shop 13 | 68.81% | 30.28% | 0.92% | 0.00% | 109 |
| Shop 14 | 17.86% | 30.71% | 47.86% | 3.57% | 140 |
| other | 39.64% | 28.25% | 25.74% | 6.38% | 439 |
| all | 43.02% | 22.13% | 23.13% | 11.72% | 2680 |



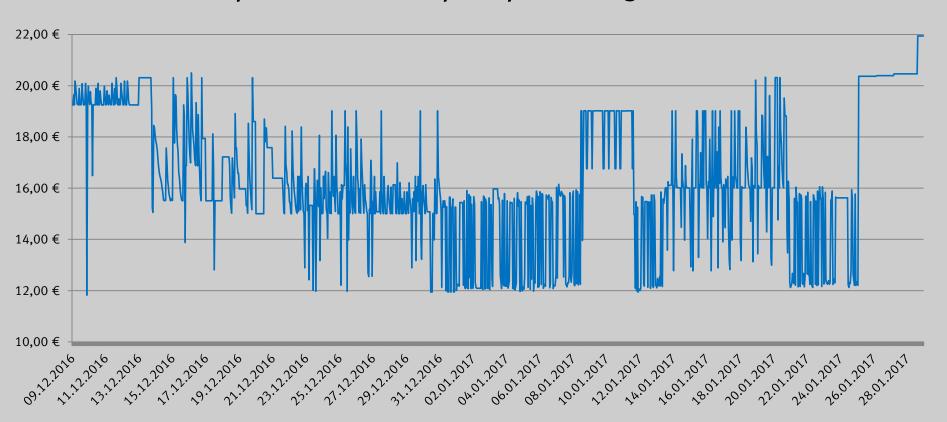
Results (3)

| Shop | (p | share on price series | | | |
|---------|-----------------|-----------------------|-------------|-------|---------|
| | < 0 , 05 | 0,05 - <0,1 | 0,1 - <0,25 | >0,25 | overall |
| Shop 1 | 99 | 46 | 27 | 6 | 82.8% |
| Shop 2 | 67 | 44 | 52 | 19 | 32.6% |
| Shop 3 | 26 | 19 | 2 | 1 | 23.2% |
| Shop 7 | 75 | 40 | 67 | 49 | 26.0% |
| Shop 9 | 47 | 24 | 9 | 1 | 65.9% |
| Shop 13 | 0 | 0 | 1 | 0 | 0.9% |
| Shop 14 | 5 | 6 | 57 | 4 | 51.4% |
| other | 75 | 44 | 18 | 4 | 32.1% |
| all | 394 | 223 | 233 | 84 | 34.9% |



Results (4)

Example of extreme frequent price changes: aftershave





Results (5)

- Summarized results
 - Differences in price setting behavior among shops, not among product categories
 - Dynamic pricing is applied by few online shops in a remarkable extent
 - Volatility of prices is critical in some of these shops
 - Time of price changes: more in the first third of the day probably for technical reasons and to hide price changes



Implications for Consumer Price Statistics

- Information about online shops and their price setting behavior can be used to manage resources for price collection
 - Concentrate on online shops with high frequency of price changes and high volatility of prices
 - More frequent dates of price collection, additional checks
 - Use of new tools for price collection (web scraping) as soon as possible



Outlook – web scraping

- Automatized price collection (web scraping)
 - Is applied in some fields and will be applied on a broader basis
 - Will be further developed technically, project with IT
- Methodological challenges
 - How to deal with replacements and quality adjustments?
 - How to calculate average prices?
- Legal basis for the access to websites does not exist currently



Outlook - transaction data

- Dynamic pricing in physical shops?
 - Introduction of electronic price signs
- Traditional price collection will no longer be sufficient
- Use of transaction data (scanner data) necessary
 - Project on scanner data has just started



Digitale Preise: Die Zukunft des Bezahlens © dpa



THANK YOU FOR YOUR **ATTENTION!**























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