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Studies of new data sources and techniques to improve CPI compilation in Brazil

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Case Study 1:

Automatic Collection of Airfares Using Web Scraping

Automatic Collection of Airfares: Experiment

Current price collection: once a week each of the 16 CPI local units collects prices manually for the selected routes from airline websites that represent typical consumer behavior.



Automatic Collection of Airfares: Experiment

- Flight arrivals on the most visited destinations for leisure purpose.
- 8 day trip with departure on Saturday and returns on Sunday.
- Airfares purchased 2 months in advance of the departure date.
- All ticket classes.

The objective is to replicate the manual procedure using web scraping collection.

How to replicate this procedure?

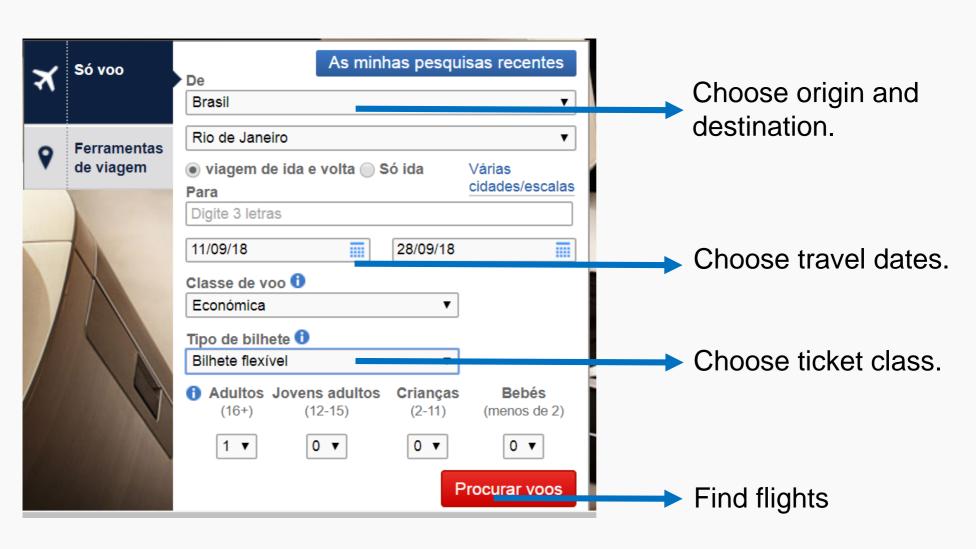
Web scraping concerns algorithms that convert data present in HTML to an easy and understandable structured format.

In general, airline websites require human-like action to extract information.

Development of in-house web scraper using R and Selenium with RSelenium package.

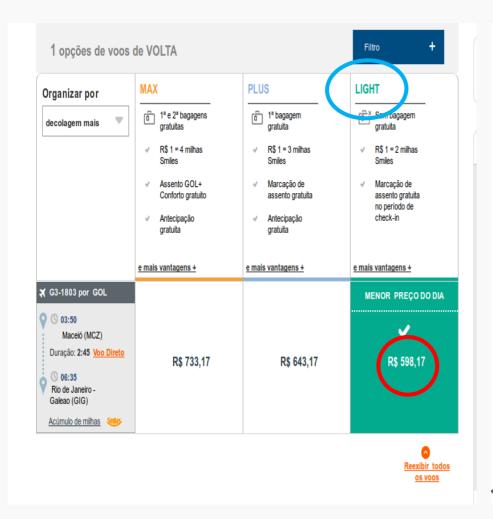


Robots emulate user navigation in the website.



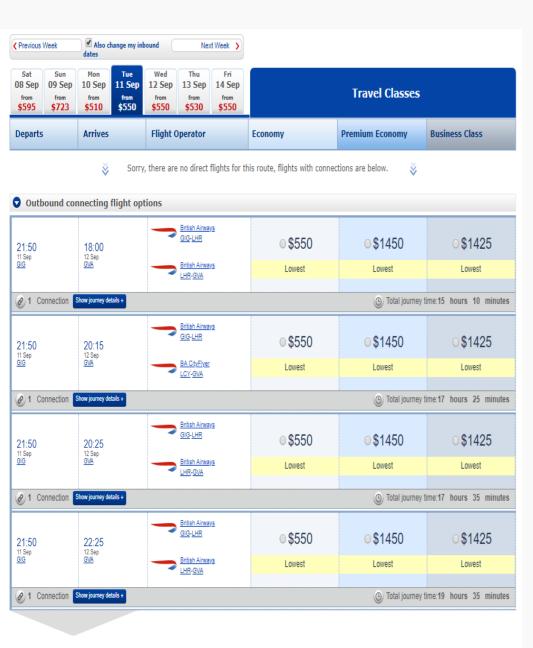
Example:

Airline website



Source code

```
<div>
      <div class="lessPriceBox"></div>
      <div class="taxaSelected">
       <div class="checkTaxaSelected"></div>
      </div><span class="smilesAndMoneyValue"></span><label class="textIdentFareValue" for="
                   R$ 643,17</span><span>tarifa Plus</span></label><input id="ControlGro"
    </div>
   <div class="lessPriceBox">
       <div class="lessPrice">Menor Preço do Dia</div>
      </div>
      <div class="taxaSelected">
       <div class="checkTaxaSelected"></div>
      </div><span class="smilesAndMoneyValue"></span><label class="textIdentFareValue" for="
                   R$ 598,17
    </div>
   <div id="market2 journey1" class="infoGrid bgGrid popupANAC"><span class="anacInformatio">
               </span></div>
```



Extracted information:

- Price.
- Airline.
- Depart and destination cities.
- Depart and return flight dates
- All ticket classes.

Scrapers have been running since January 2018.

And they have been following the same calendar as the manual collection.

Data analysis covers the period from January 2018 until September 2018, inclusive.

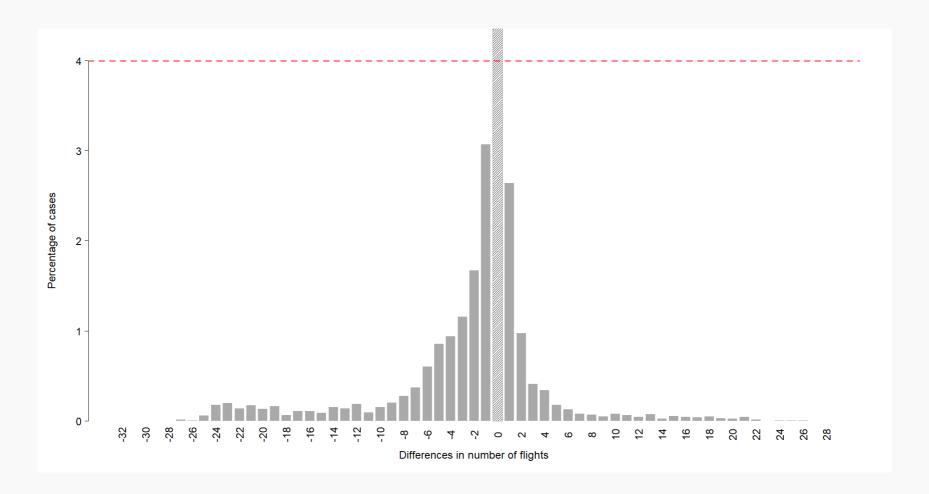
Product codifier: Company + Route + Depart and Return Date + Collection Date + ID for Depart or Return Flight.

Database: 25 weeks.

Manual Collection: 320,213 product prices.

Automatic Collection: 305,214 product prices.

Differences in the number of flights between automatic and manual collection for each product. Negative (positive) values represent cases in which manual processes found more (fewer) flights than the robots. Lighter bar (equals no difference) corresponds to 83% of cases.

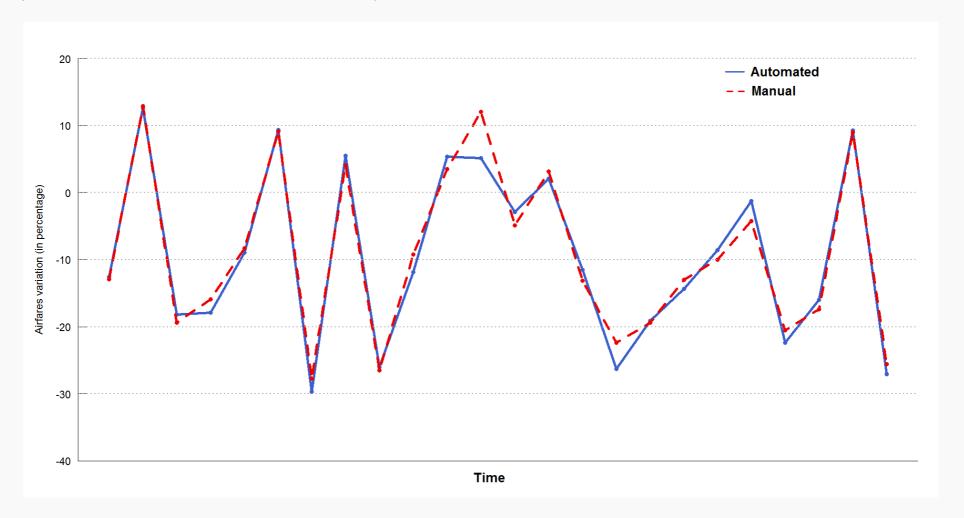


Why the discrepancies?

- Collection time.
- Data entry errors.

Sample database does not allow further investigation due to the absence of complementary flight information like flight code, depart and return time, etc.

Comparison between the variation for automatic and manual collection considering the differences in the number of flights in the interval [-5,5] (More than 95% of the cases).



Automatic Collection of Airfares: Final remarks

Pros:

Faster and cheaper than manual collection.

Screenshots and records of the prices collected.

Web scraping collection of airfares reproduces well the manual one.

Cons:

Changes in airline website designs may require the program code to be modified.

Technical issues: Internet connection instability and IP (Internet Protocol) blockage.

CPI compilation demands weekly collection.

Current stage: test phase → implementation

Case Study 2:

Use of Web Scraping to Support the Implementation of Hedonics at Brazilian CPI

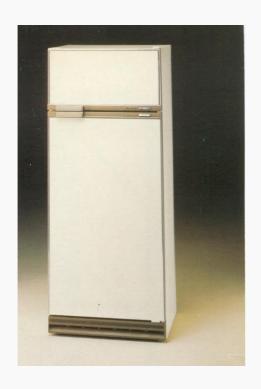
CPI pillars:

- Fixed basket.
- Matched Model Method.

Matched Model Method breakdowns when:

- New products are available.
- Disappearance of older ones.
- Evolution of technologies.

New or modified products may provide different degree of utility (quality) to the consumers respective to older ones.





Item/period	t	<i>t</i> +1	t+2	t+3	t+4
1	p_l^t	p_l^{t+1}	p_l^{t+2}	p_l^{t+3}	p_l^{t+4}
m	p_m^t	p_m^{t+1}	p_m^{t+2}		
n				p_n^{t+3}	p_n^{t+4}

From period t + 3, refrigerator m is no longer available for purchase and the refrigerator n is the replacement.

$$R_n^{t+3,t+2} = p_n^{t+3}/p_m^{t+2}$$

But if refrigerators have different attributes. Refrigerators m and n should not be directly matched. Pure price variation would not be measured.

How to deal with it?

Standard tool to minimize this problem relies on hedonic modeling techniques.

The hedonic approach states that each good is composed by a bundle of attributes and each of them has its marginal contribution for the final price.

Patching: for products with low rate of substitutions Hedonic indices: for products with high turnover or depreciation (used-cars for example).

Attribute database:

- Very resource intensive
- Important barrier for its implementation

Use of web scraping technique to overcome this costly process

Example of refrigerator attributes available at website:

Total Capacity	24.52 cubic feet	
Refrigerator Style	Side-by-Side	
Ice Maker	Yes	
Lighting Type	LED	
Color Finish	Stainless steel	

In-house web scraper using R to collect refrigerator prices and attributes. Run hedonic regression for quality adjustment.

Multiple regression between prices and attributes z.

$$p = \beta_0 + \beta_1 z_1 + \beta_2 z_2 + \dots + \beta_n z_n + \epsilon$$

Calculate the impact of each significant/relevant attribute z.

Patching approach: with the best fit hedonic model, it is possible to impute the estimated price for the new product n in the period t + 2.

Item/period	t	<i>t</i> +1	t+2	t+3	t+4
1	p_l^t	p_l^{t+1}	p_l^{t+2}	p_l^{t+3}	p_l^{t+4}
m	p_m^t	p_m^{t+1}	p_m^{t+2}		
n			\hat{p}_n^{t+2}	p_n^{t+3}	p_n^{t+4}

$$R_n^{t+3,t+2} = p_n^{t+3}/\hat{p}_n^{t+2}$$

May online prices be used for quality adjustment in the CPI sample (brick-and-mortar only)?

Collect online and offline data together and adjust hedonic regression for them.

Web Scraping for Hedonics in Brazilian CPI: Experiment Description and Data Collection

Online database:

In-house scraper using R software.

Extract prices (delivery fee is not included) and attributes for refrigerators that could have been purchased at the moment of the scraping.

One moment collection at February 2019.

Web Scraping for Hedonics in Brazilian CPI: Experiment Description and Data Collection

Offline database:

CPI refrigerator sample from January 15th until February 15th 2019.

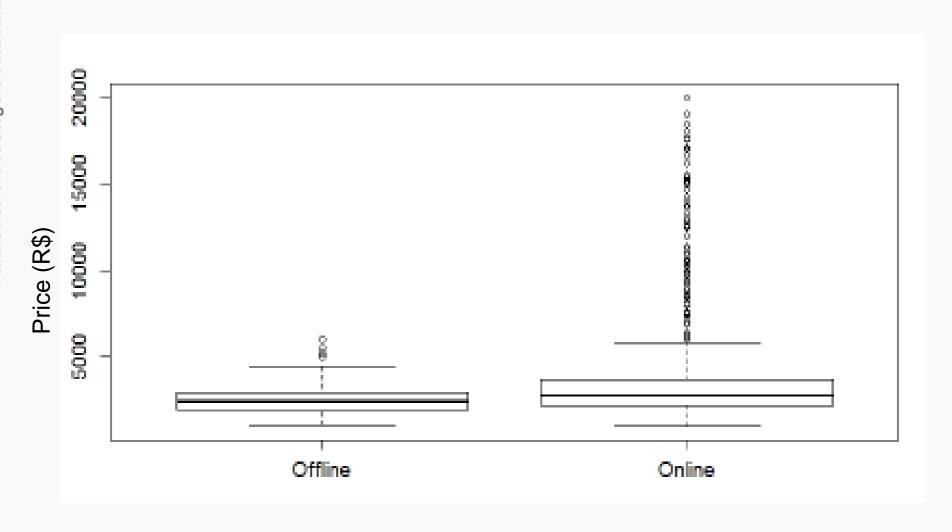
Only price data. How to obtain the attributes?



Reference: BRE57AK

Collectors were asked to get the references from refrigerator CPI sample. With that single information, we were able to get the attributes from database obtained by web scraping.

Box plot of offline and online prices



	Online	Offline
Prices	1663	1386
References	154	64
Stores	29	42

For a better comparison, the refrigerators were filtered as follow:

- Retailers that sell refrigerator online and offline.
- Reference refrigerators for selling online and offline for the retailers selected as the rule above.

Box plot of offline and online prices

Database: retailers and refrigerator references available online and offline



For 87% of refrigerator references, online mean price is smaller than offline mean price

Offline mean price is 11% greater than online mean price.

To run the hedonic regressions, the database with retailers and refrigerator references available both online and offline will be used.

Web Scraping for Hedonics in Brazilian CPI: Results

Run hedonic model to evaluate the significant attributes that explain the prices for refrigerator. Besides, test significance of dummy variable that identifies whether refrigerator prices are from online or offline shop

$$log(Pr) = \beta_0 + \beta_1 Br + \beta_2 Col + \beta_3 Sty + \beta_4 Defr + \beta_5 Cap + \beta_6 Shop$$

Final model is:

log(Price) = Brand + Color Finish + Style + Defrost + Total Capacity + Shop (Online or Offline)

Web Scraping for Hedonics in Brazilian CPI: Results

Model 1 – Attributes + Dummy (Online or Offline Shop)

```
Coefficients:
                Estimate Std. Error t value Pr(>|t|)
(Intercept)
                6.592e+00 2.905e-02 226.935 < 2e-16
BrConsul
               -1.619e-01 1.486e-02 -10.896 < 2e-16
BrElectrolux
              -4.476e-02 1.106e-02 -4.046 5.78e-05
Colinox
              1.003e-01 1.126e-02 8.909 < 2e-16
StyDuplex 1.166e-01 1.717e-02 6.791 2.35e-11
StyInverse 2.210e-01 2.212e-02 9.991 < 2e-16
DefrFrost Free 1.615e-01 1.045e-02 15.445 < 2e-16
              2.684e-03 6.284e-05 42.707 < 2e-16
Cap
ShopOnline
               -1.094e-01 8.593e-03 -12.736 < 2e-16
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.1001 on 713 degrees of freedom
Multiple R-squared: 0.8845, Adjusted R-squared: 0.8832
F-statistic: 682.5 on 8 and 713 DF, p-value: < 2.2e-16
```

There is difference between online and offline price level.

May online prices be used for quality adjustment in the CPI sample (brick-and-mortar only)?

There is difference in price level for online and offline refrigerators, but this result does not answer the question.

Does hedonic coefficients (price determining attribute estimates) rely on the kind of shop?

Web Scraping for Hedonics in Brazilian CPI: Results

Test interaction between significant attributes and dummy variable that identifies if the refrigerator is sold online or offline

Model 2 – Attributes + Dummy (Online or Offline Shop) + Interaction

```
Coefficients:
                         Estimate Std. Error t value Pr(>|t|)
(Intercept)
                        6.608e+00 3.035e-02 217.751
BrConsul
                       -1.901e-01 1.962e-02
                                             -9.692 < 2e-16
BrElectrolux
                       -2.154e-02 1.448e-02 -1.488 0.13726
colinox
                        1.099e-01 1.112e-02 9.878 < 2e-16
                        8.785e-02 1.757e-02 5.000 7.24e-07
StyDuplex
StyInverse
                        1.956e-01 2.214e-02 8.836 < 2e-16
                        1.539e-01 1.036e-02 14.848 < 2e-16
DefrFrost Free
Cap
                        2.692e-03 6.146e-05 43.804 < 2e-16
ShopOnline
                       -7.943e-02 1.892e-02 -4.198 3.04e-05 ***
BrConsul:ShopOnline
                       5.489e-02 2.661e-02 2.063 0.03948 *
BrElectrolux:ShopOnline -6.629e-02 2.141e-02
                                             -3.096 0.00204
                 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Signif. codes:
Residual standard error: 0.09787 on 711 degrees of freedom
Multiple R-squared: 0.8899, Adjusted R-squared: 0.8884
F-statistic: 574.9 on 10 and 711 DF, p-value: < 2.2e-16
```

Web Scraping for Hedonics in Brazilian CPI: Results

The only interaction that was significant was brand and the kind of shop

For model 2, adjusted R² is 0.8884, while for model 1 it is 0.8832

Explaining power of model 2 in comparison with model 1 is marginal.

Model 1 was chosen because it is more parsimonious

Web Scraping for Hedonics in Brazilian CPI: Final remarks

Rich attributes database obtained in cheap and efficient way with web scraping

Whether the products is sold online or offline does not impact in hedonic coefficients

So we can use web scraped data for quality adjustment on brick-andmortar CPI sample for refrigerator

Otherwise, offline prices + online attributes (via web scraping) to run hedonic regression

There are more refrigerator references online than offline

Web Scraping for Hedonics in Brazilian CPI: Final remarks

Estimated coefficients must be updated from time to time.

Using only online data makes it easier

Web scraping technique allows to identify products that is becoming more (less) representative based on the number of stores they are offered.

Next steps:

- Household Budget Survey (to be released this year)
- Implementation

Thank you for listening!

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