# Finding the Goldilocks data collection frequency for the Consumer Price Index

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The views expressed in this presentation are those of the authors only and do not necessarily represent the views of the Bank of Italy or the Eurosystem.

### How often is "just right"?

#### **Assumptions:**

- •Timing of CPI price collection: prices often vary during the period
- Data collection, processing, and storage has a cost

#### **Objectives:**

Estimate variability of CPI m-o-m variations at different sampling-in-time frequencies

Propose and empirically validate a framework to determine the optimal data collection frequency



## Electricity and gas prices in Italy

 Mandatory publication of unregulated market offers for electricity and gas to guarantee transparency

- •Used by ISTAT for official CPI price data collection
- •Simplified data collection:
  - 9 cities
  - National consumption profile
    - Electricity consumption: 2700kWh/year
    - Gas consumption: 1400 m3/year
  - Selection of rates for electricity
  - Providers weighted by national market shares
  - Type of contracts (fixed or variable prices) weighted according to official reports



#### Daily Time-Product-Dummy index

•Weighted Time-Product Dummy index:

$$\ln p_{it} = \sum_{t=1}^{T} \delta_t D_t + \sum_{i=1}^{N} \beta_i D_i + \varepsilon_{it}$$

- $p_{it}$ : average price of operator *i* in month *t*
- $D_t$ : dummy equal to 1 if month is equal to t and zero otherwise
- $D_i$ : dummy equal to 1 for prices of operator *i* and zero otherwise.

•Weighted Least Squares using operator *i* market share as weight for each observation.

•Aggregate daily price level:

$$P_t = e^{\delta_t}$$

#### Month-on-month CPI variations

•Month-on-month rate of change:

$$\pi_{t} = \frac{\sum_{j=1}^{k} C P I_{j}^{t}}{\sum_{l=1}^{k} C P I_{l}^{t-1}} - 1$$

•Number of potential changes at each sampling frequency:

$$C(15,k)^{2} = {\binom{15}{k}}^{2} = {\binom{15!}{k! (15-k)!}}^{2}$$

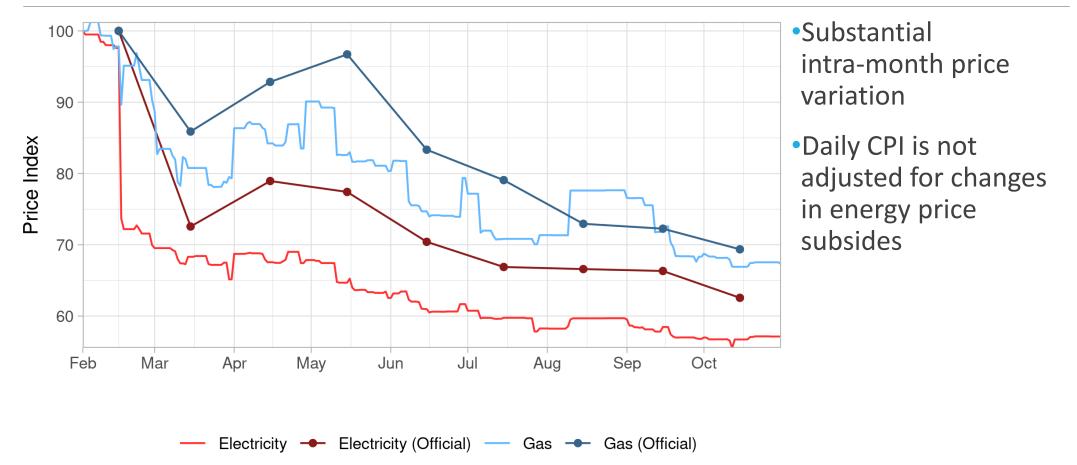
•Very large total number of combinations, over 155M each month for each utility. Almost 2.5B combinations for this research

#### Optimization framework

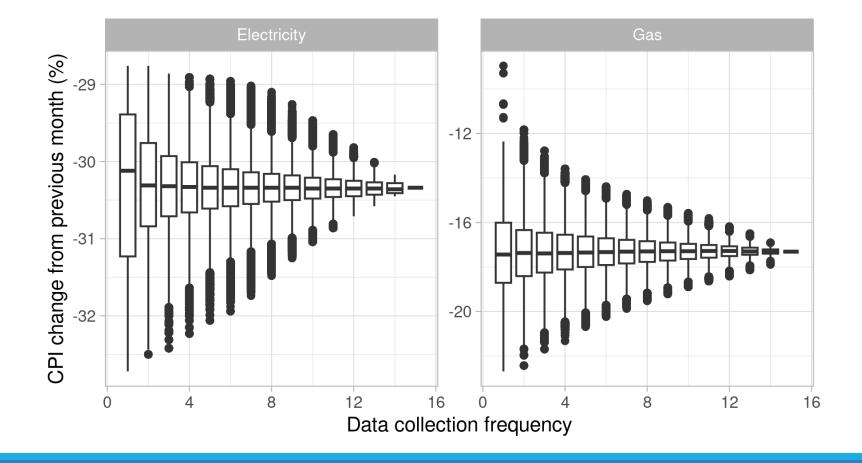
- •Uncertainty on CPI measurement can be represented by the StDev of potential m-o-m rates of change
- •Costs for data collection, processing, and storage are usually expressed in monetary terms
- •Need to convert one of the two metrics (or both) to have a common unit of measure for optimization
- •Our approach: express the collection cost in minimum reduction of StDev to be obtained through additional data collection efforts

 $min(UncertaintyCost(k) + CollectionCost(k)), k \in \mathbb{N}^+: \{1 \le k \le 15\}$ 

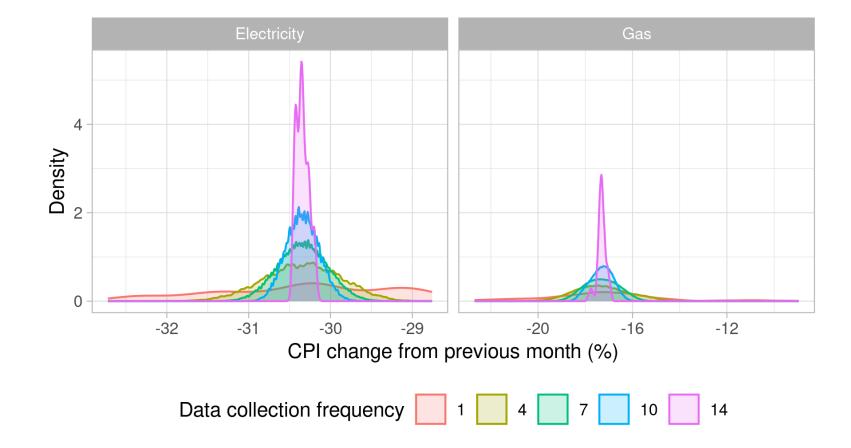




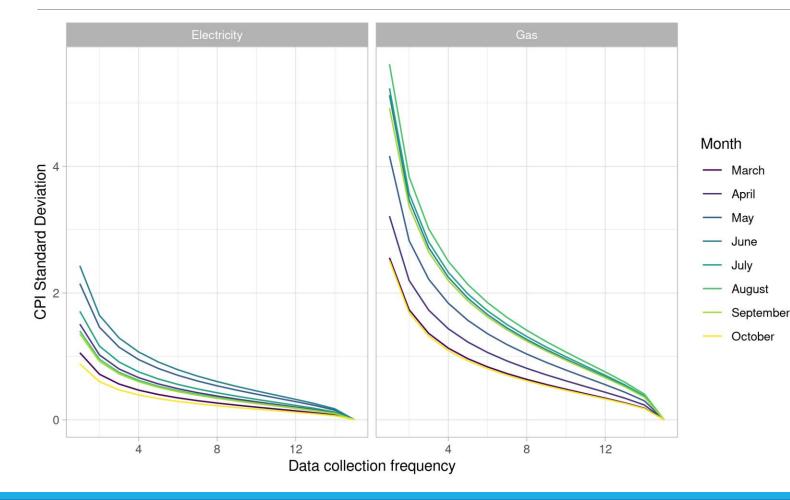
#### Boxplot of potential m-o-m CPI variations



#### Density of potential m-o-m CPI variations



#### StDev of potential m-o-m CPI variations



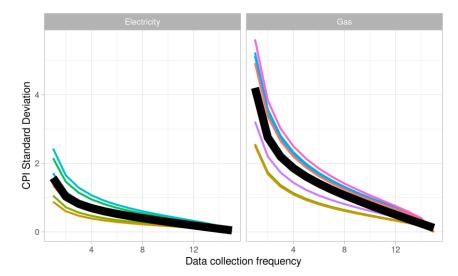
- •Gas prices were much more volatile than electricity ones
- Levels vary across months and utilities, but paths are
- consistent

#### Modeling uncertainty

$$UncertaintyCost(k) = \beta_0 + \beta_1 k + \frac{\beta_2}{k} + \eta$$

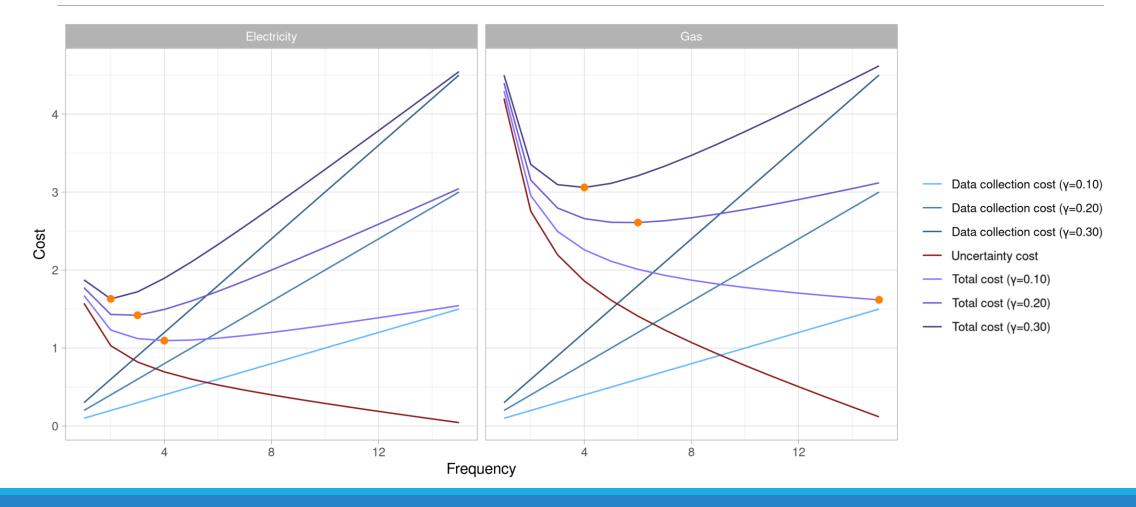
•Approximately hyperbolic cost function

 Parameter estimates consistent with expectations and across utilities



	Standard Deviation					
	Electricity			Gas		
	(1)	(2)	(3)	(4)	(5)	(6)
$\beta_0$	0.147*** (0.029)	1.167*** (0.049)	0.614*** (0.075)	0.394*** <i>(0.073)</i>	3.117*** (0.124)	1.649*** (0.182)
$\beta_1$		-0.084*** (0.005)	-0.042*** (0.006)		-0.224*** (0.014)	-0.114*** (0.016)
β2	1.581*** (0.090)		1.002*** <i>(0.117)</i>	4.218*** (0.225)		2.664*** (0.283)
Adjusted-R2	0.720	0.668	0.794	0.746	0.694	0.824
Note:	*p<0.1; **p<0.05; ***p<0.01					

#### Finding the Goldilocks frequency



#### Conclusions

•Variability of CPI m-o-m changes can be substantial at low data collection frequencies, affecting aggregated CPI figures

- Increasing data collection frequency yields diminishing returns in terms of CPI uncertainty reduction
- It is possible to determine an optimal data collection frequency leveraging historical price volatility and a cost function for data collection

•Optimality should be periodically reassessed, since both price variability and data collection costs may vary

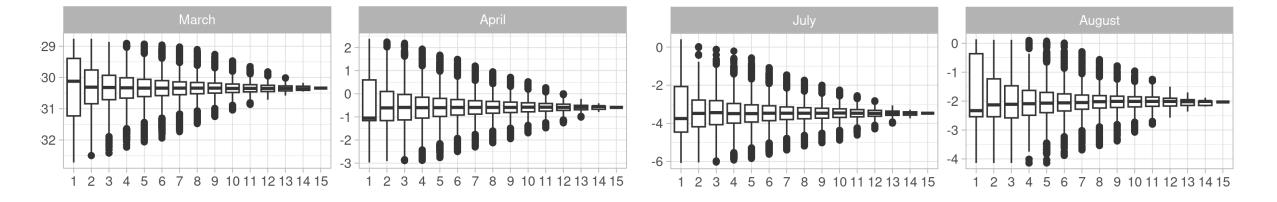
# Thank you!

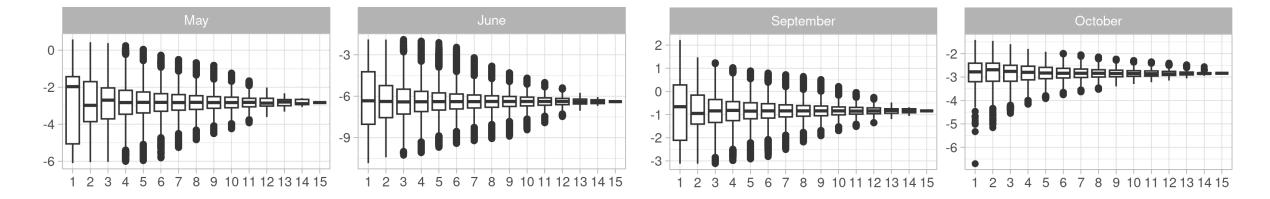
## Q&A

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#### Electricity: Boxplot of potential m-o-m CPI variations





#### Gas: Boxplot of potential m-o-m CPI variations

