

The Billion Prices Project

Research and Inflation Measurement Applications

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Micro Price Data in Macroeconomics

- Data Sources
 - Statistical Offices (CPI, PPI, IPI)
 - Scanner Data (eg. Nielsen)
 - Online Data (eg. Billion Prices Project)
- Uses
 - Research in Macroeconomics
 - Price Dynamics (Price Stickiness, Real Rigidities)
 - Market Segmentation
 - Research in International Economics
 - Pass-through and Border Effects
 - Law of One Price and PPP
 - Real Exchange Rates
 - Measuring inflation and other economic indicators

Each Data Source has Advantages and Disadvantages

CPI Data

- Purpose: measure inflation

Advantages	Disadvantages
<ul style="list-style-type: none">• Representative sample<ul style="list-style-type: none">• carefully-chosen goods• many retailers and locations• Long Time Series• Collection of 'posted prices in stores	<ul style="list-style-type: none">• Very costly to collect and access• Low frequency (monthly)• Limited number of goods and varieties• Some unit values and imputed prices• Difficult international comparisons

Each Data Source has Advantages and Disadvantages

Scanner Data

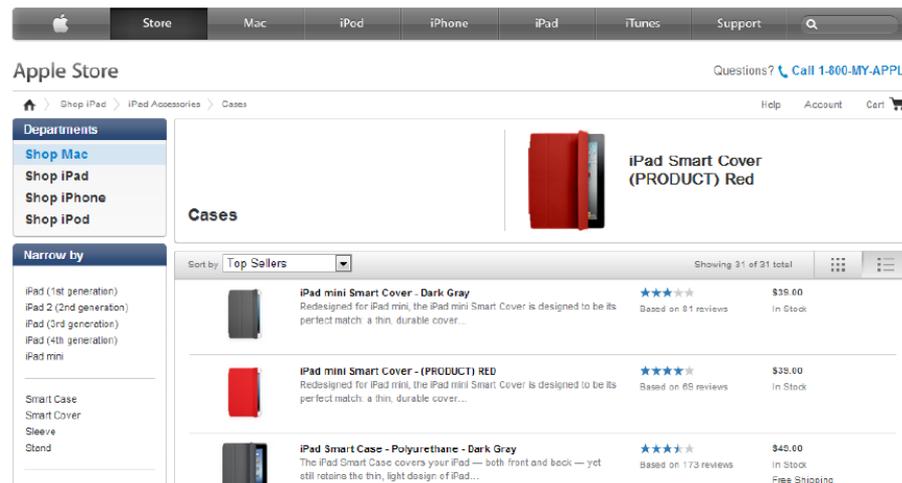
- Purpose: marketing analytics (eg. market shares)

Advantages	Disadvantages
<ul style="list-style-type: none">• Granularity<ul style="list-style-type: none">• Some product details for all goods <i>sold</i>• Transaction data<ul style="list-style-type: none">• Contains quantities and sometimes costs• Frequency (weekly)	<ul style="list-style-type: none">• High cost to collect/acquire• Limited coverage (supermarkets, department stores)• Data characteristics vary greatly depending on provider, location, time period, etc.• Extremely difficult to compare internationally• Unit values and time-averages (eg: prices are often calculated as sales/quantity in a week)

Each Data Source has Advantages and Disadvantages

Online Data

- Can be collected using automated web-scraping software
- Every day, a *robot* downloads a public webpage, analyses its HTML code, extract price data, and stores it in a database



The screenshot shows the Apple Store website's 'Cases' section for iPads. The navigation bar at the top includes 'Store', 'Mac', 'iPod', 'iPhone', 'iPad', 'iTunes', and 'Support'. The main content area displays a list of iPad Smart Covers. The first product is the 'iPad Smart Cover (PRODUCT) Red' with a red cover. Below it, a list of products is shown with their names, descriptions, ratings, and prices. The products listed are:

Product Name	Description	Rating	Price	Availability
iPad mini Smart Cover - Dark Gray	Redesigned for iPad mini, the iPad mini Smart Cover is designed to be its perfect match: a thin, durable cover...	★★★★☆ (Based on 81 reviews)	\$39.00	In Stock
iPad mini Smart Cover - (PRODUCT) RED	Redesigned for iPad mini, the iPad mini Smart Cover is designed to be its perfect match: a thin, durable cover...	★★★★☆ (Based on 69 reviews)	\$39.00	In Stock
iPad Smart Case - Polyurethane - Dark Gray	The iPad Smart Case covers your iPad — both front and back — yet still retains the thin, light design of iPad...	★★★★☆ (Based on 173 reviews)	\$40.00	In Stock Free Shipping

```
<html>
<!-- START product -->
<ahref="productId=MD963LL"></a>
<p class="productname">Ipad Mini Smart Cover – Dark Grey</p>
<td class="Price">$39.00</td>
<!-- END product -->
....
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Each Data Source has Advantages and Disadvantages

Online Data

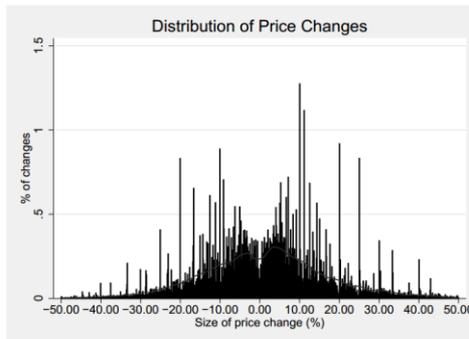
Advantages	Disadvantages
<ul style="list-style-type: none">• Frequency (daily)• Cheap to collect (but complicated)• Granularity<ul style="list-style-type: none">• All product details (brands, size, anything shown online)• All goods and varieties available for sale (census)• New goods automatically sampled• Easier to compare internationally	<ul style="list-style-type: none">• Fewer retailer and locations than CPI• Short time series• Not all categories of goods and services are online (not yet)• Online and Offline prices may behave differently

Online Data in Macro and International Research

- Billion Prices Project at MIT
 - Daily data from 2008 to the present
 - Sample of retailers and countries has grown over time (hundreds of retailers in 70 countries)
- We have written papers on:
 - Price Stickiness
 - Border Effects
 - Law of One Price
 - Inflation Measurement

Research Examples: Price Stickiness

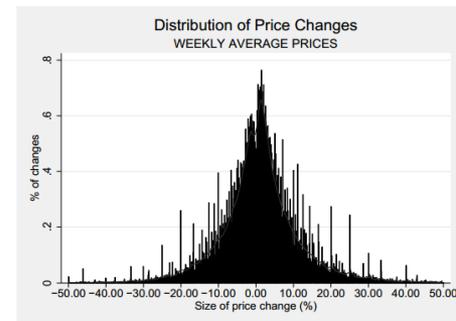
- We use online data to re-evaluate puzzling stylized facts coming from CPI or scanner data
- For example, in contrast to the literature, we find that in most retailers and countries price-change distributions are bi-modal, with little mass close to zero percent.
 - Consistent with state-dependent models of price adjustment where small price changes are not optimal (given the existence of an adjustment or menu'cost).



(a) Chile



With unit values or weekly averages



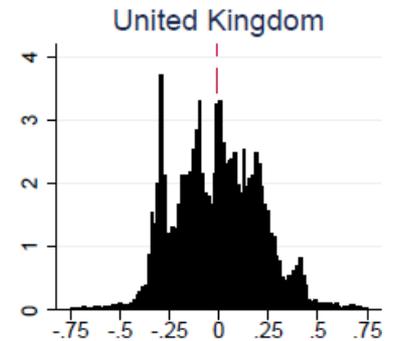
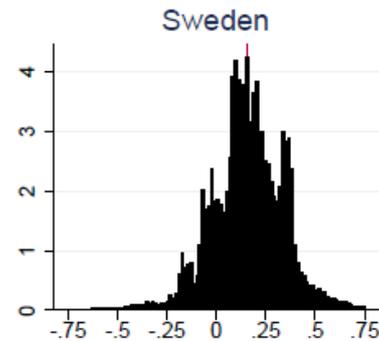
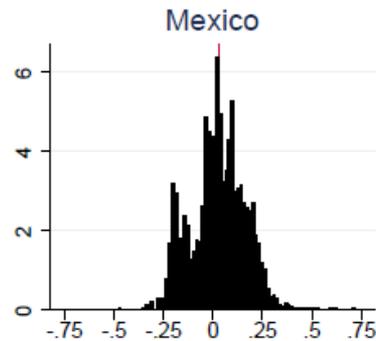
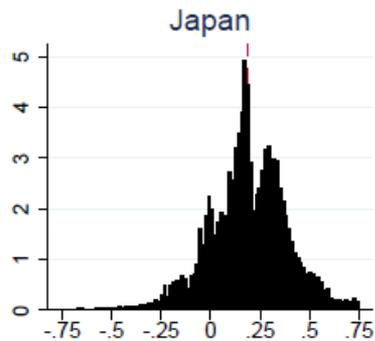
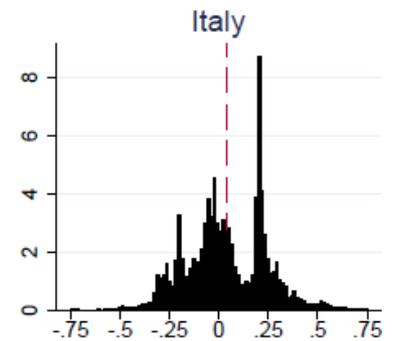
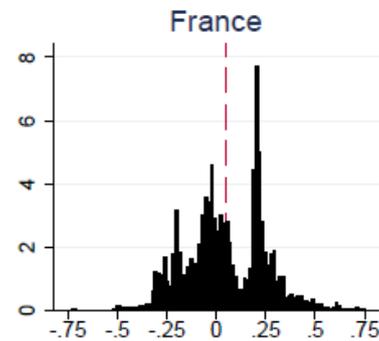
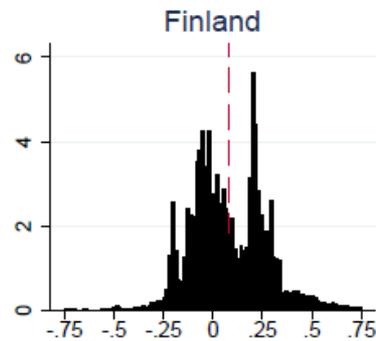
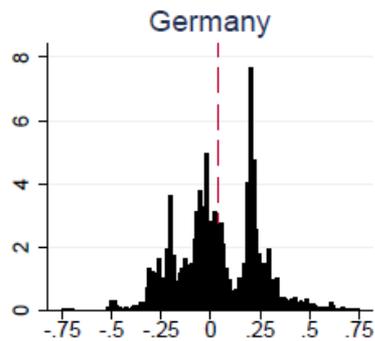
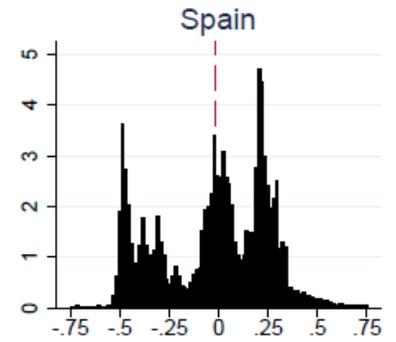
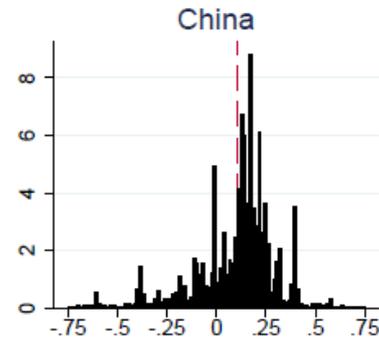
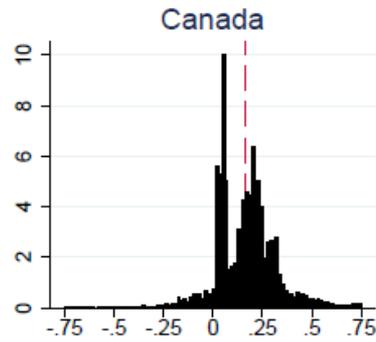
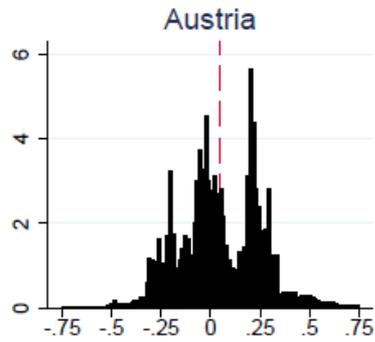
(c) Chile

See Cavallo (2012) "Scraped Data and Sticky Prices", Cavallo & Rigobon (2012) "The Distribution of the Size of Price Changes"

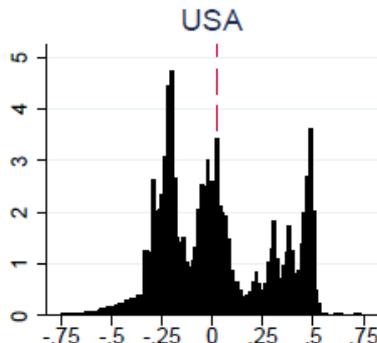
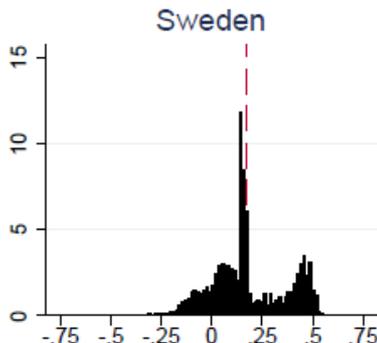
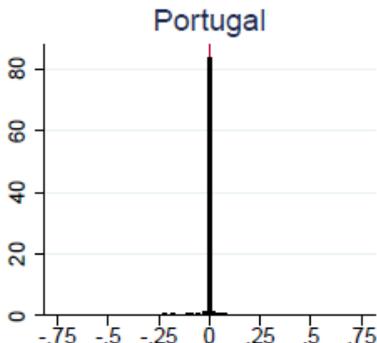
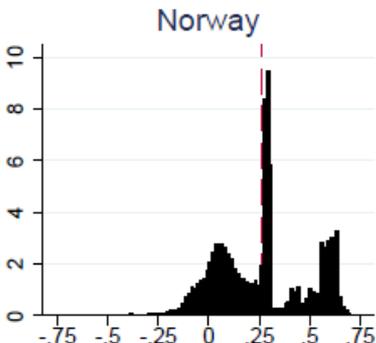
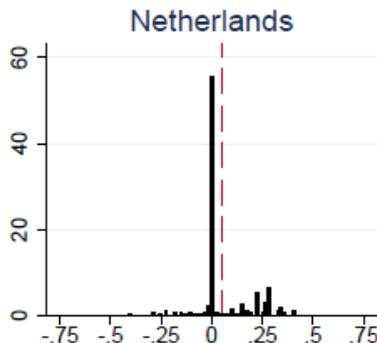
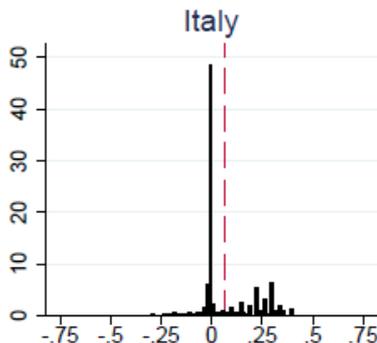
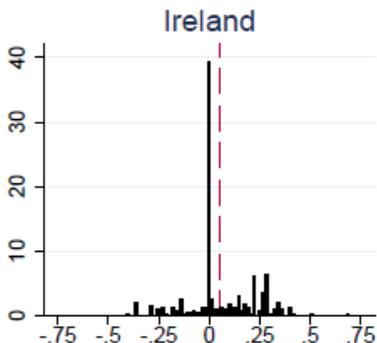
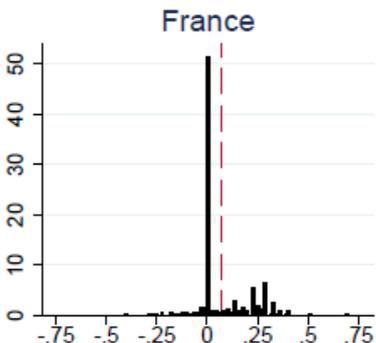
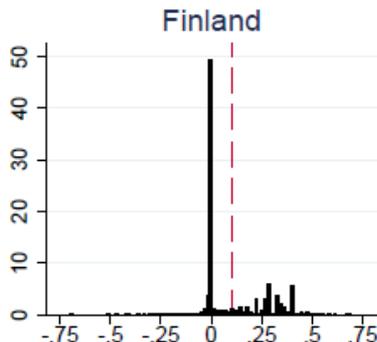
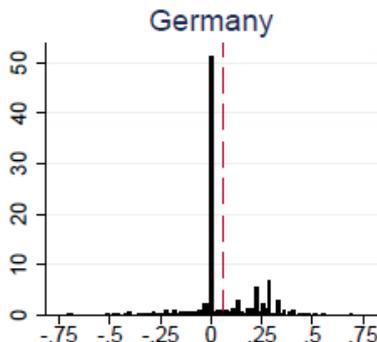
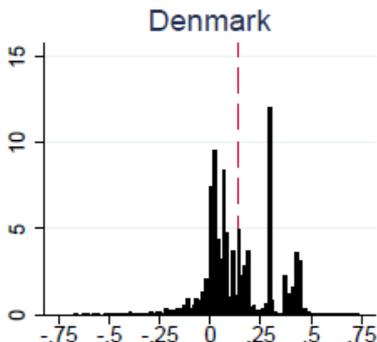
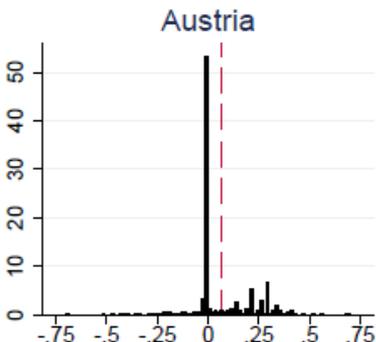
Research Examples : Law of One Price (LOP)

- Cavallo, Neiman, and Rigobon (2014) QJE. “Currency Unions, Product Introductions, and the Real Exchange Rate”
- We evaluate LOP deviations using a large dataset of identical tradeable goods, sold by global retailers in three industries and dozens of countries (Apple, Ikea, Zara, H&M, and others)

Good-level RERs q_{ij} for $j = \text{United States}$



Good-level RERs q_{ij} for $j = \text{Spain}$



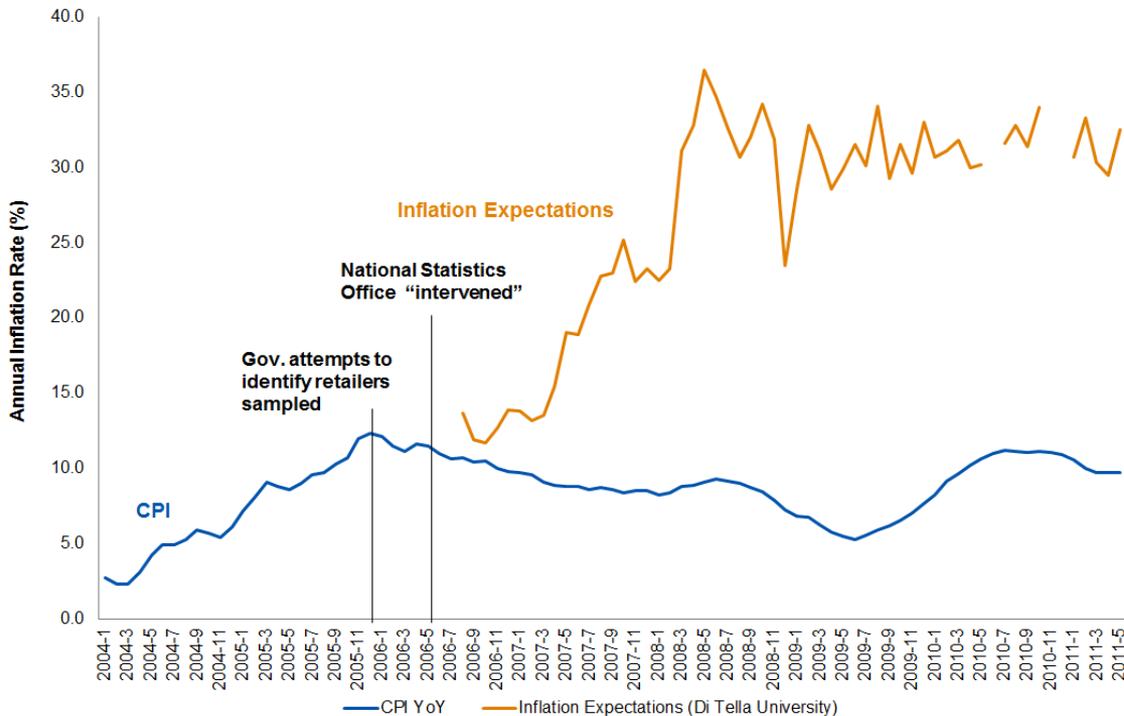
Research Examples : Law of One Price (LOP)

Average Absolute Value of Good-Level Log RER		All Stores	Apple	IKEA	H&M	Zara
All Data	Currency Unions	0.062	0.005	0.117	0.021	0.087
All Data	NER Pegs	0.149	0.047	0.164	0.141	0.142
All Data	Floats	0.182	0.139	0.185	0.152	0.192

- Two main findings:
 - LOP holds within Currency Unions, fails otherwise (even in pegged regimes)
 - A new decomposition shows that the RER at the time of product introduction is most important (yet not reflected in traditional CPI-based RERs), and moves closely with the Nominal Exchange Rate.

Research Examples: Inflation Measurement

- Argentina's inflation data is widely questioned
 - Statistical Office "intervened" in 2007.
 - Since then CPI inflation has been stable around 10%
 - Inflation expectations have been consistently above 25%



Research Examples: Inflation Measurement

- Using scraped data, I showed that online price indices could closely match CPIs in four other Latin American countries

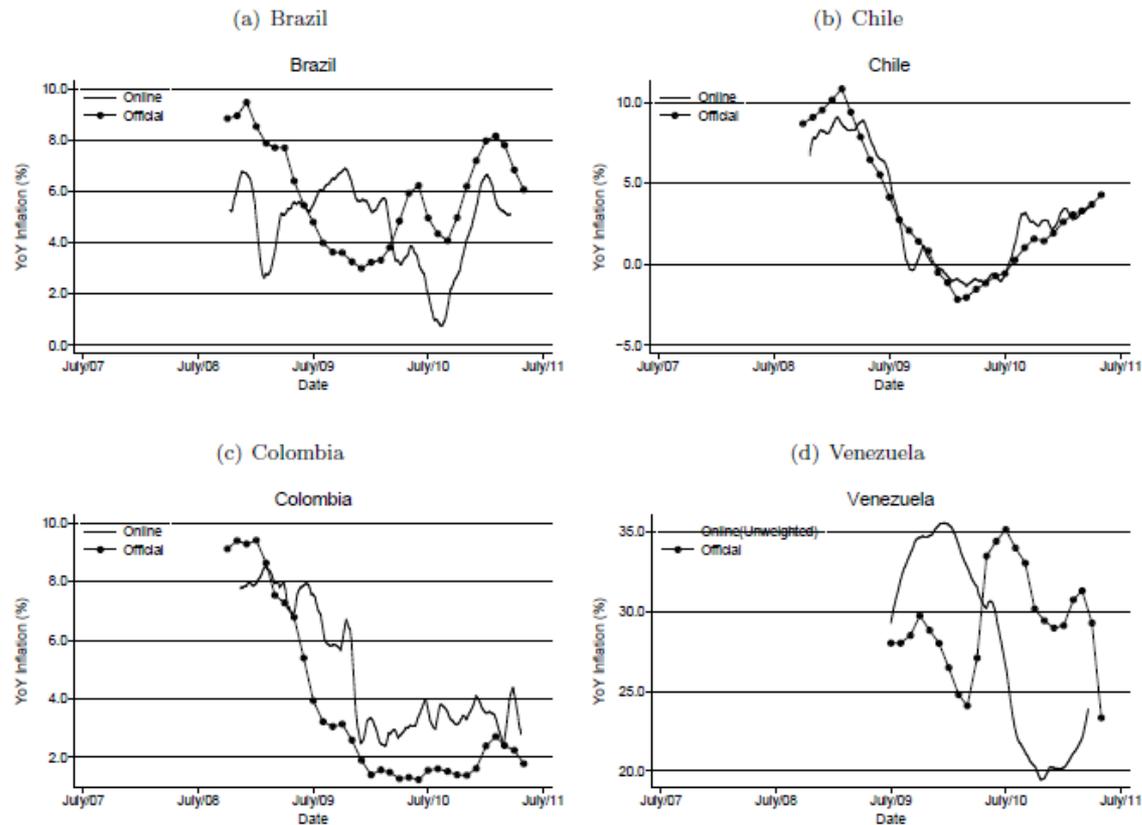


Figure 2: Online and Official Indexes - Annual Inflation Rate

Brazil, Chile, Colombia, and Venezuela

Table 3: Online vs Official Series

	Argentina	Brazil	Chile	Colombia	Venezuela
Mean Annual Inflation (%)					
Online Index		4.72	3	4.88	27.43
Official Index		5.91	3.19	3.73	29.38

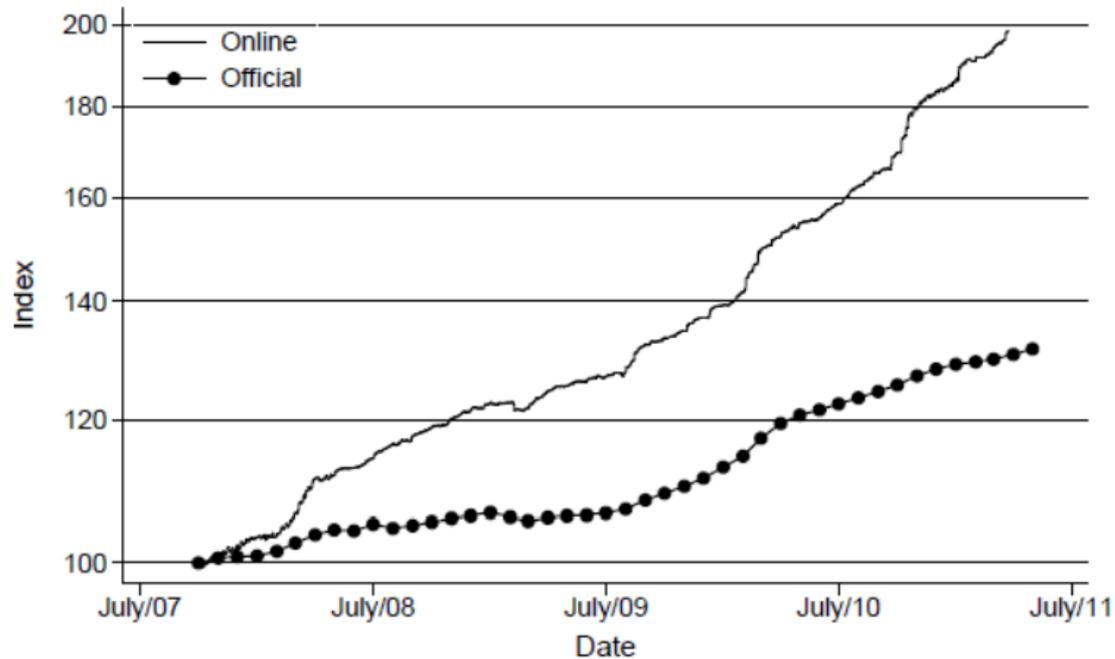
- Matching is best on Chile and Colombia, where:
 - Supermarkets have larger market shares (27% and 30%, vs. only 15% in Brazil)
 - City where online data is collected accounts for most of the CPI (55% in Santiago)
- Good news for Argentina! → the supermarket I used had 28% market share, Buenos Aires is 100% of CPI data

Research Examples: Inflation Measurement

- Argentina`s inflation rate

Argentina

(a) Daily Index



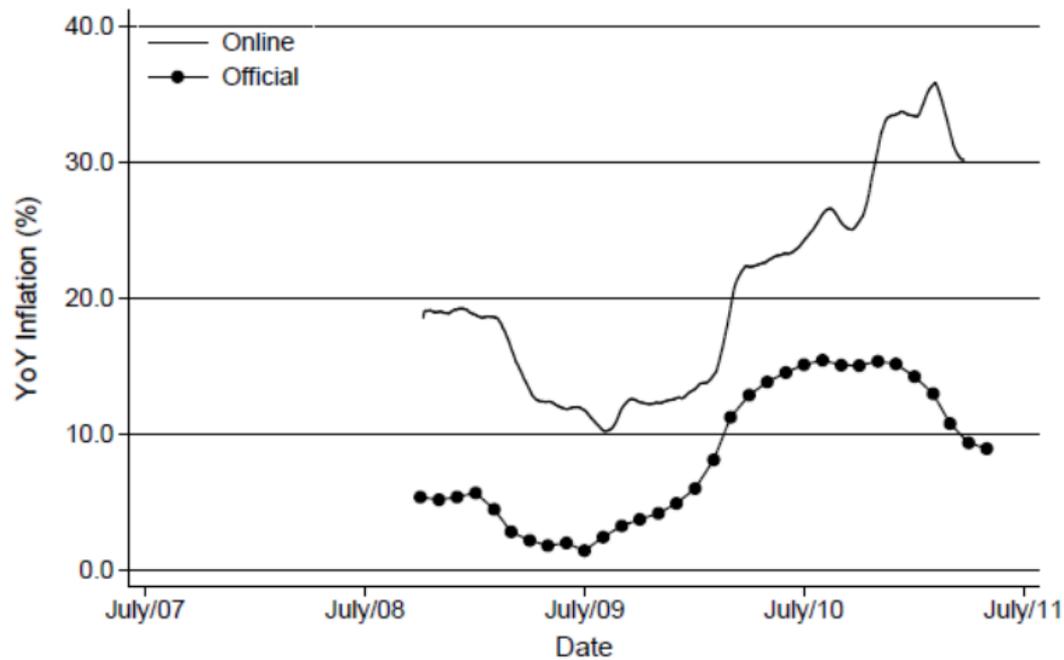
Note: Log scale

Research Examples: Inflation Measurement

- Argentina`s inflation rate

Argentina

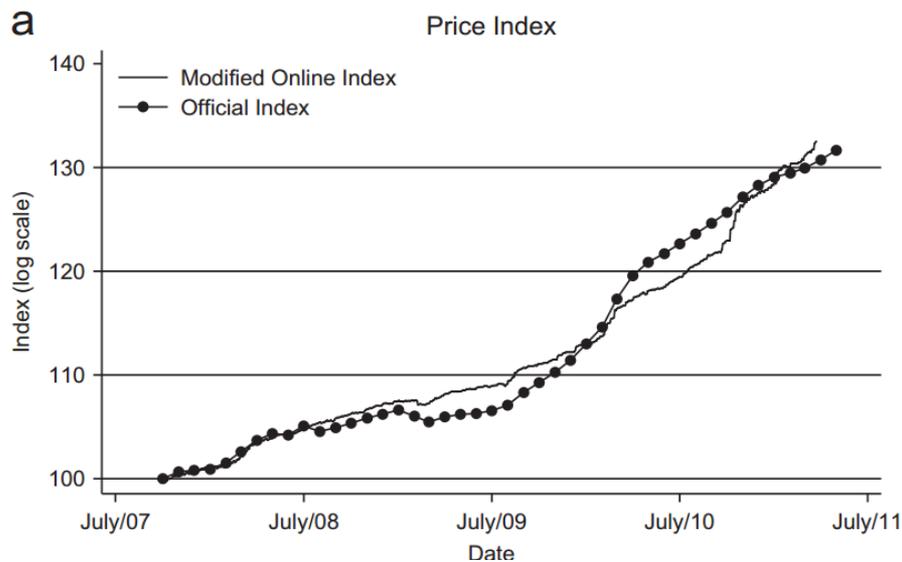
(b) Annual Inflation



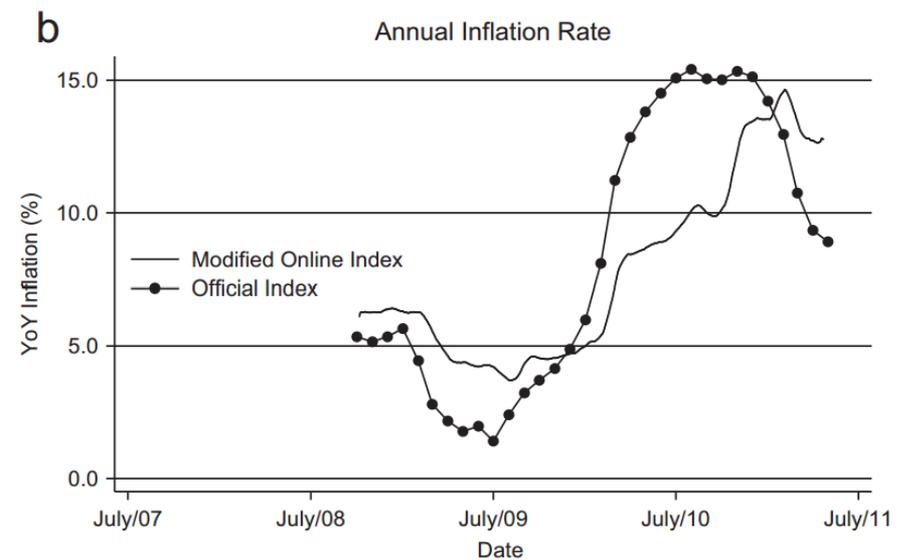
Research Examples: Inflation Measurement

- How to best approximate the official inflation rate?
- Answer: Take the true inflation rate (online), and divide by 3.

Argentina

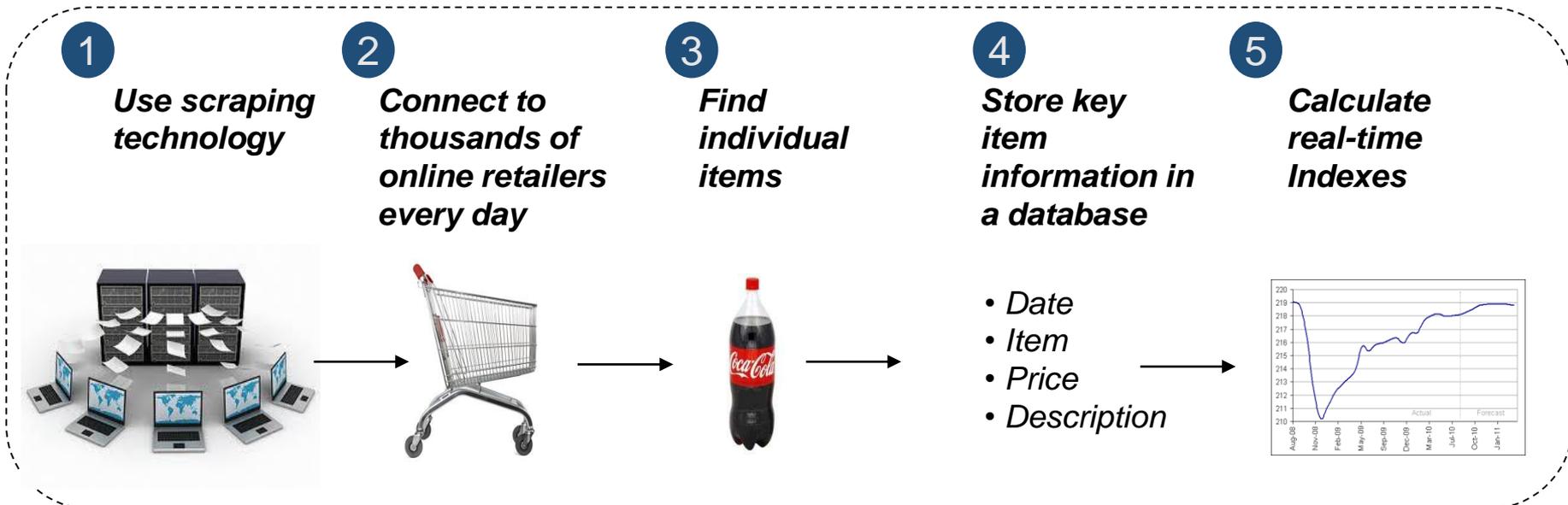


Argentina



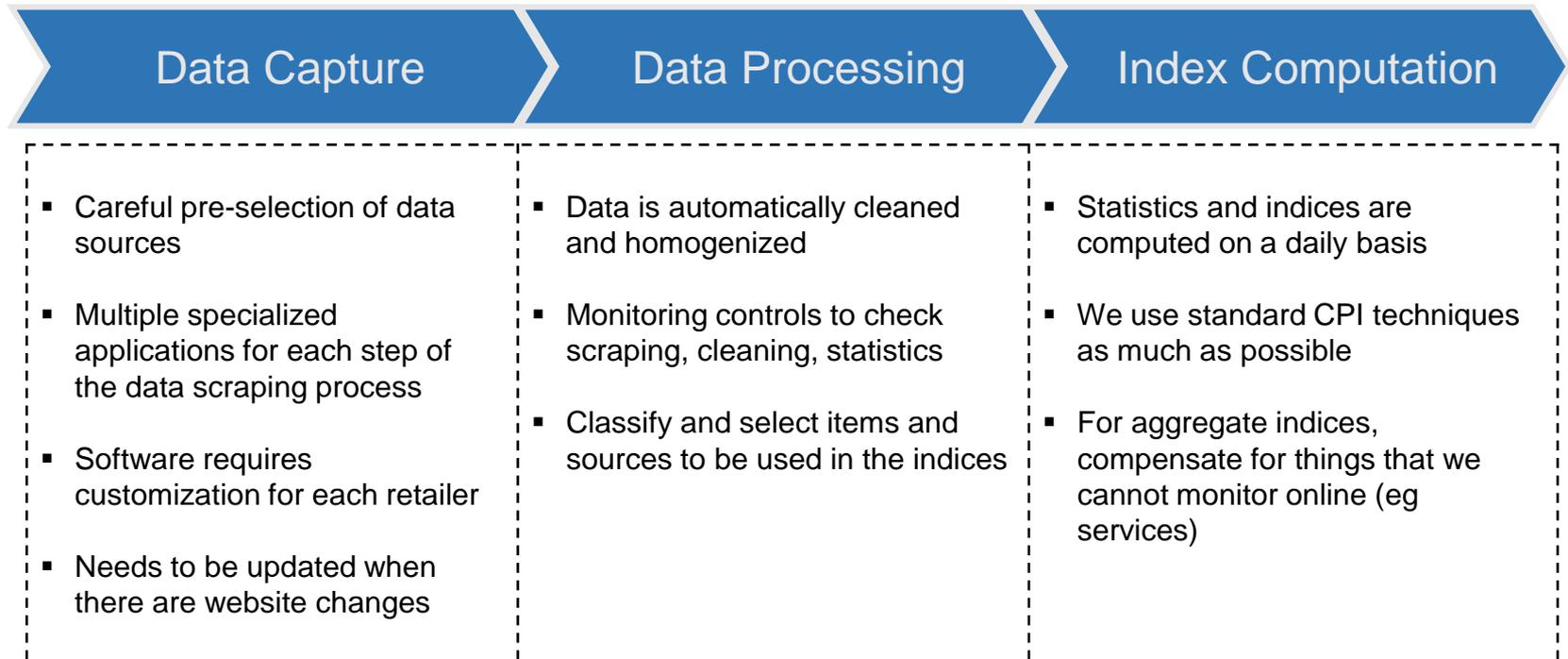
Online Data and Daily Inflation Measurement

- In 2008 we started publishing a daily online index for Argentina
- In 2010, we started publishing a daily index for the US on the BPP website
- Since 2011, PriceStats has been publishing daily inflation indices in 22 countries in real-time (3-day lag).



The Process Requires Three Stages

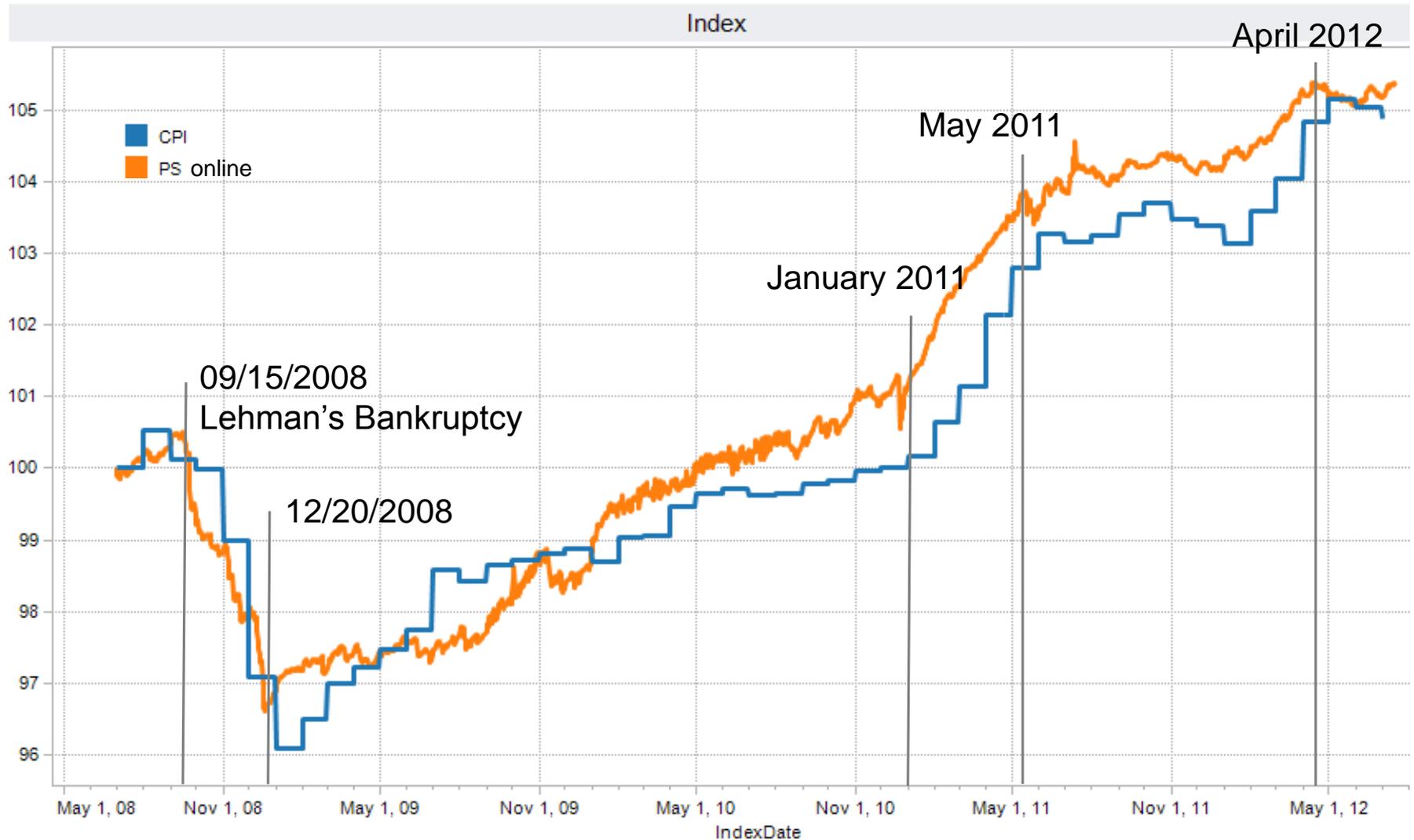
Technology & Processes



Online Price Indices vs Official CPIs

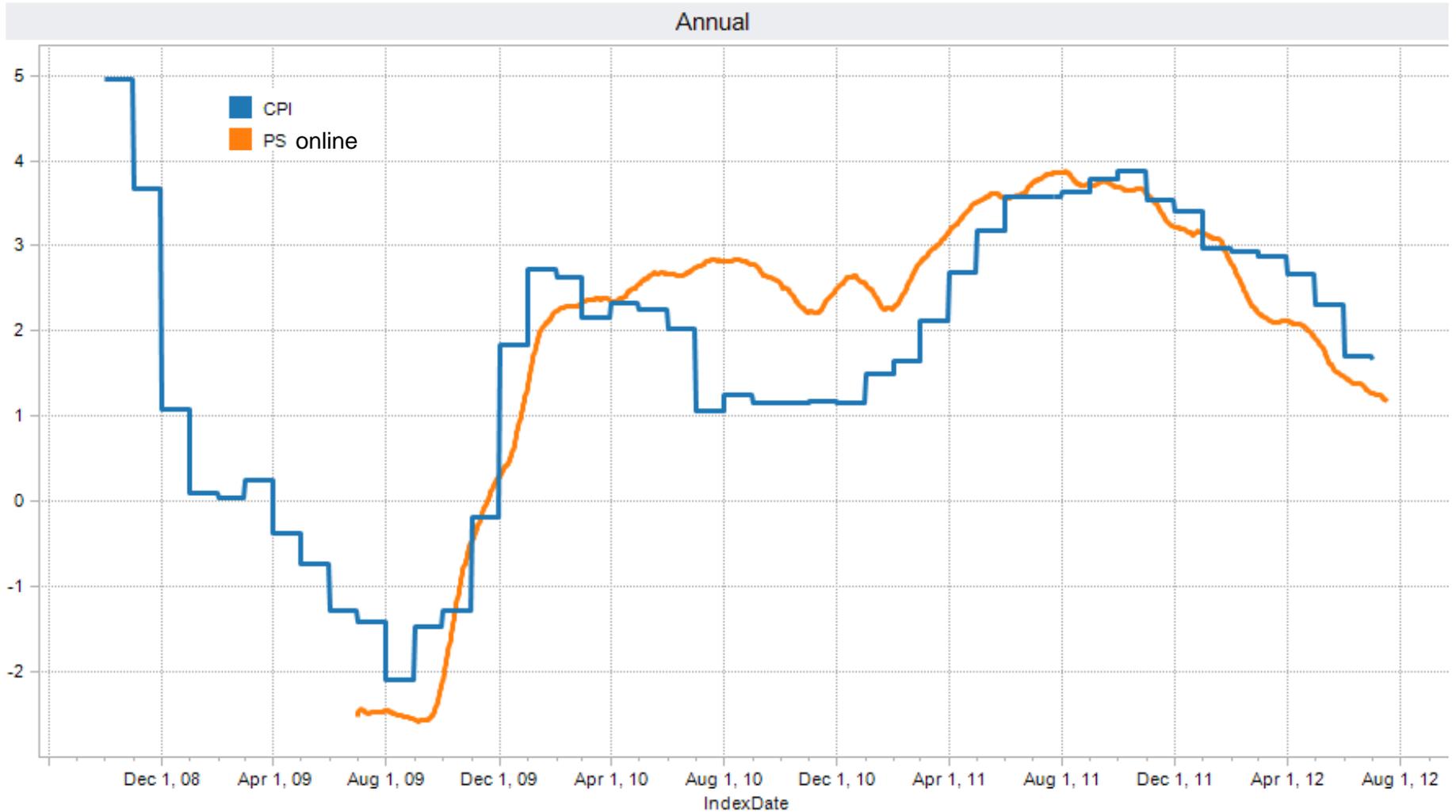
- We focus on measurement of the same phenomenon (inflation) with an alternative source of data
- Compared to CPIs, three main characteristics
 - Congruence
 - Differences in the short run
 - Anticipation of major changes in inflation trends

US Daily Price Index



Source: BPP – PriceStats – BLS (CPI-U, US city-average, all items, NSA). Updated until 7/17/2012.

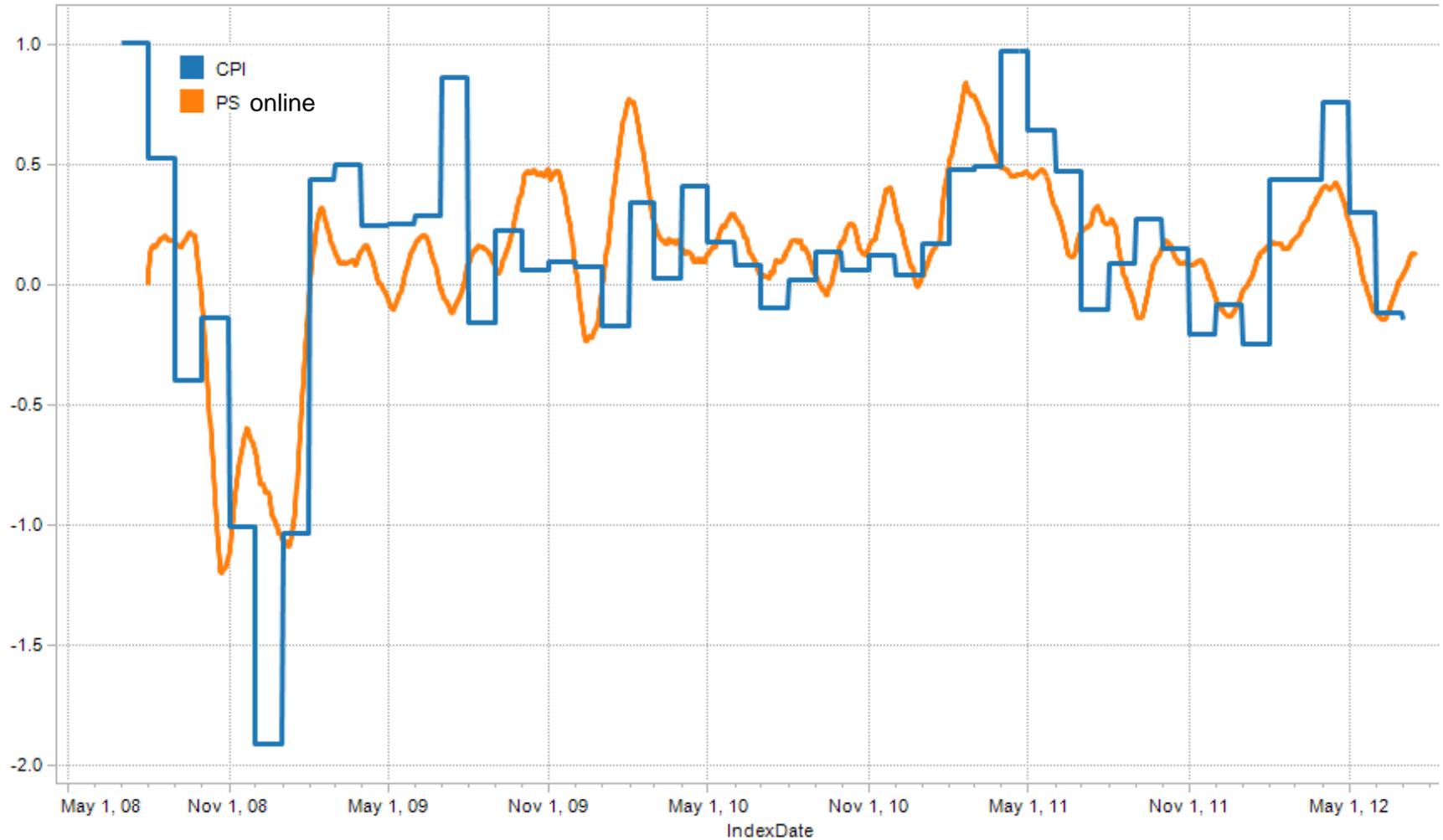
US Annual Inflation



Source: BPP – PriceStats – BLS (CPI-U, US city-average, all items, NSA)

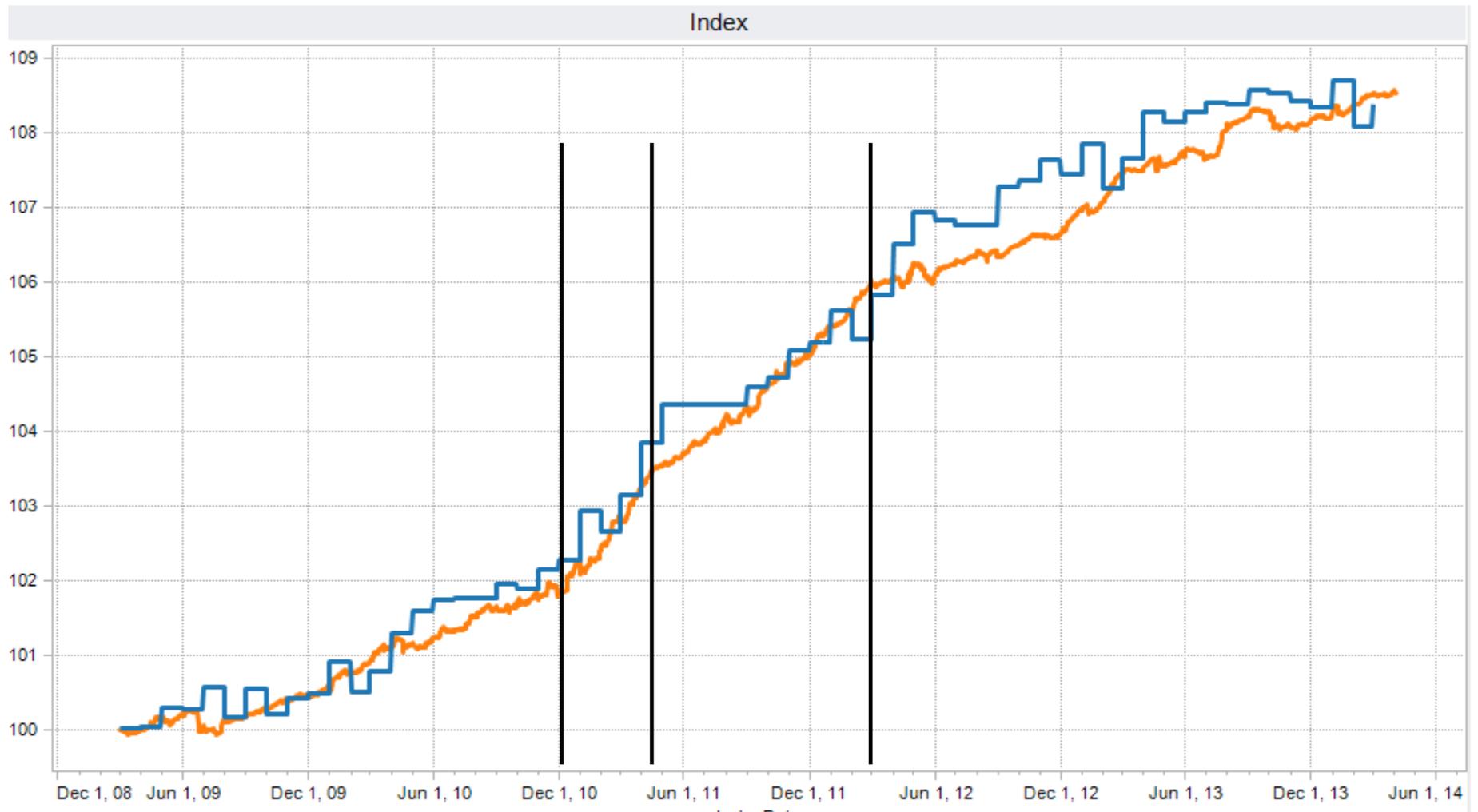
US Monthly Inflation

Monthly

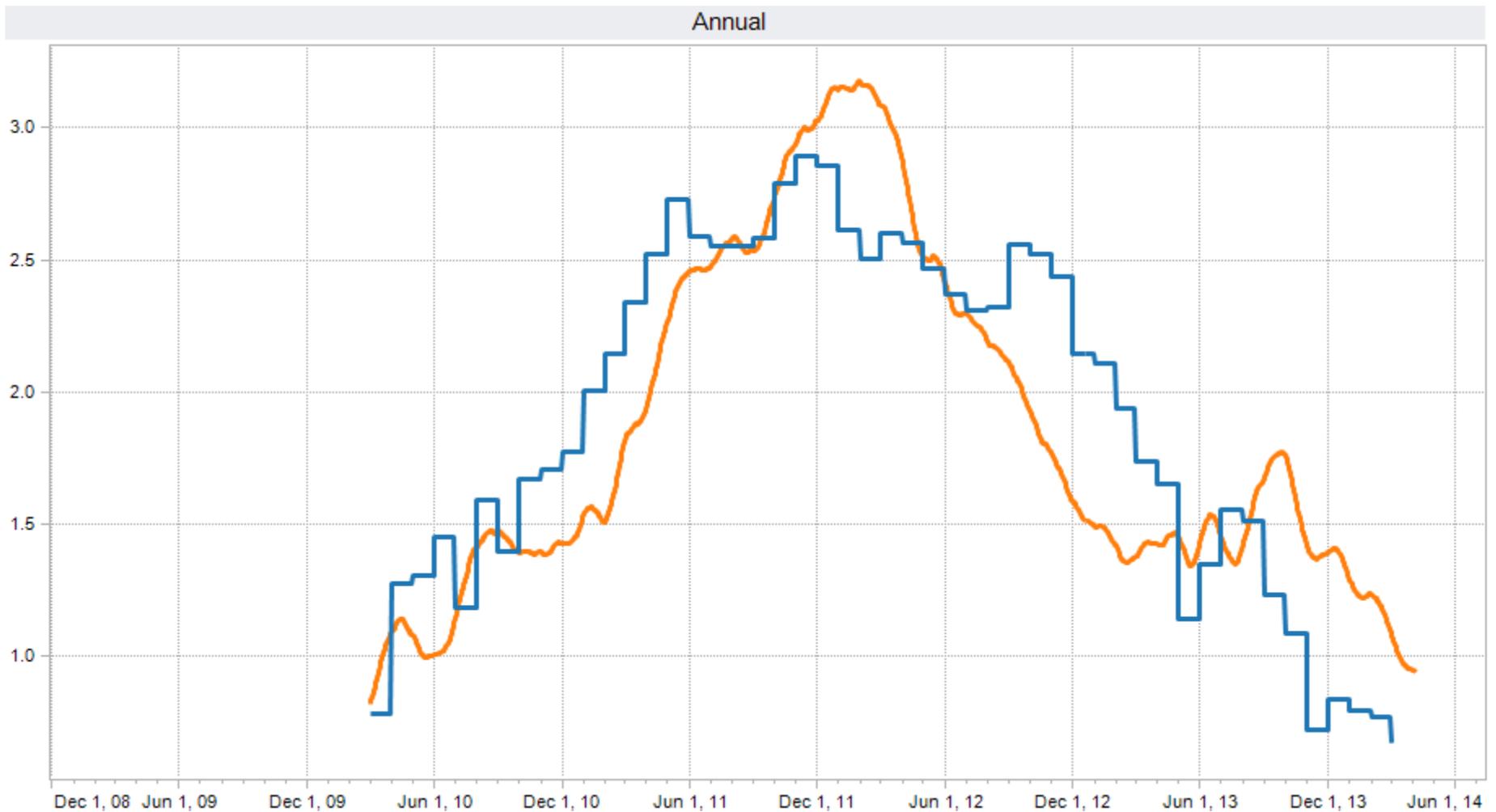


Source: BPP – PriceStats – BLS (CPI-U, US city-average, all items, NSA)

Eurozone Index

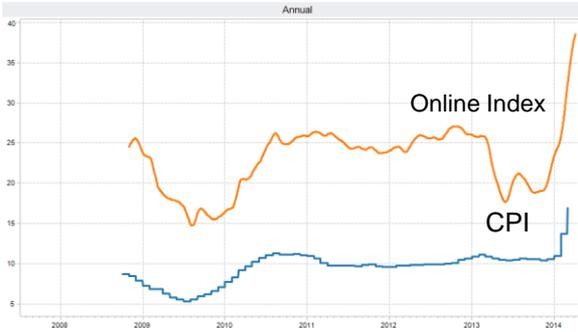


Eurozone Annual Inflation



Annual Inflation Rates in Other Countries

Argentina



Colombia



China



Germany



France



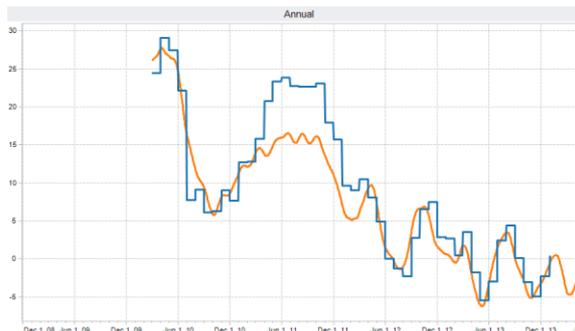
Ireland



Developed Food



Developed Fuel



Global - Aggregate

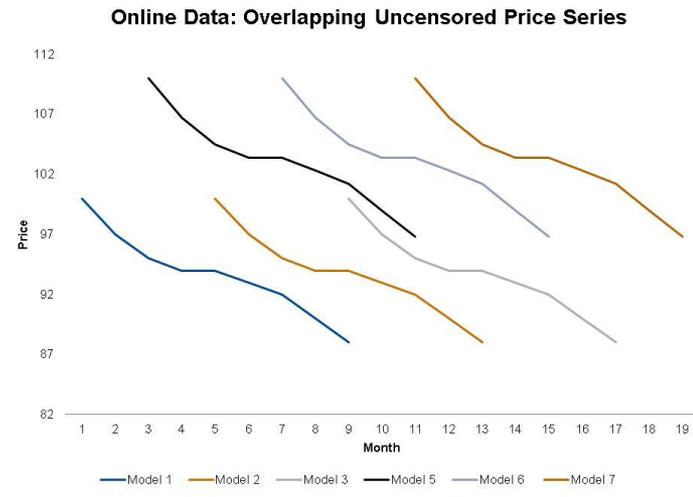
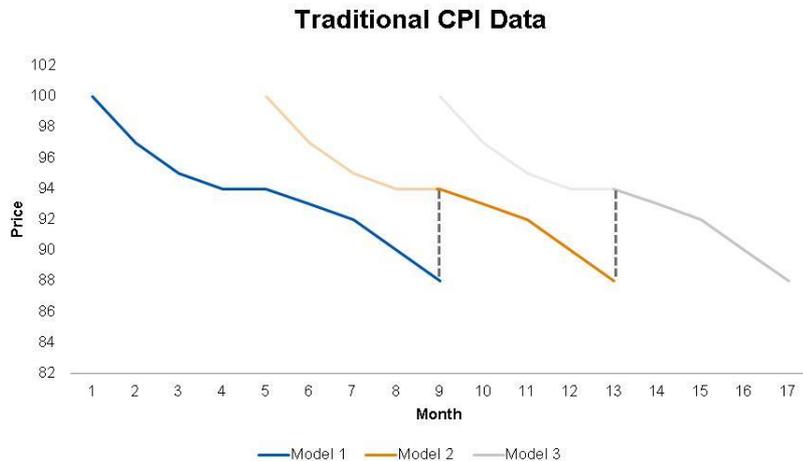


Differences: Online vs Offline Markets

- Are online prices representative of the economy?
 - Online sales are still only about 10% of retail sales in developed countries
- However:....
 - Online and Offline market are tightly integrated in many countries (people search online even when they end up buying offline)
 - Studies with simultaneous sampling show retailers tend to have either identical online and offline prices (eg: Apple, Ikea, Zara, H&M, Cavallo et al 2014) or stable online 'markups' (see Cavallo (2012))
 - The `online store` is effectively the *largest* store for most retailers
 - Eg: Walmart has 4759 stores in the US. The median store has 0.02% of sales. The `online store` has 8% of sales
- Still, as always with macro, the right answer is "it depends" (on the country, on the particular conditions, etc).

Differences: Quality Adjustments

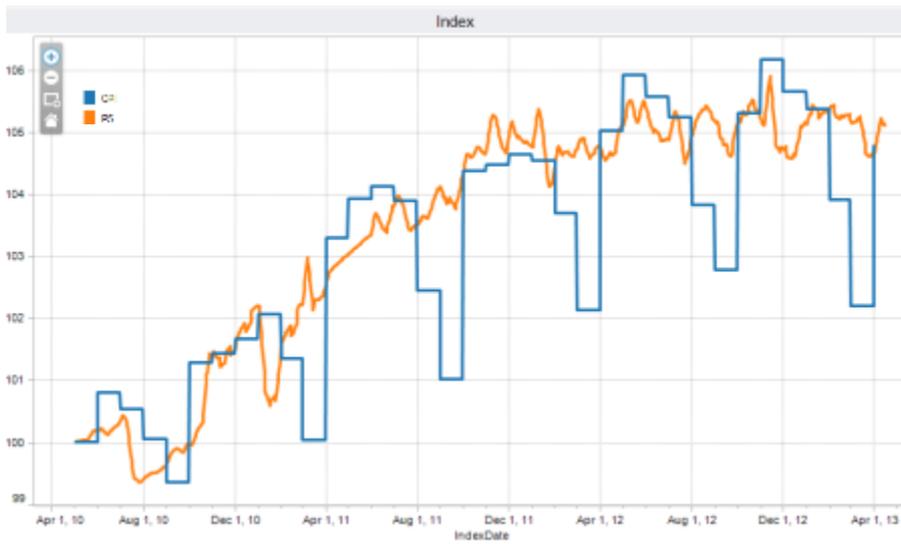
- Many complex techniques applied in CPI methods, such as hedonic quality adjustments, are needed because the data has inherent limitations
- Online data has “big data” advantages:
 - uncensored spells (automatically included at introduction)
 - all varieties/models on display



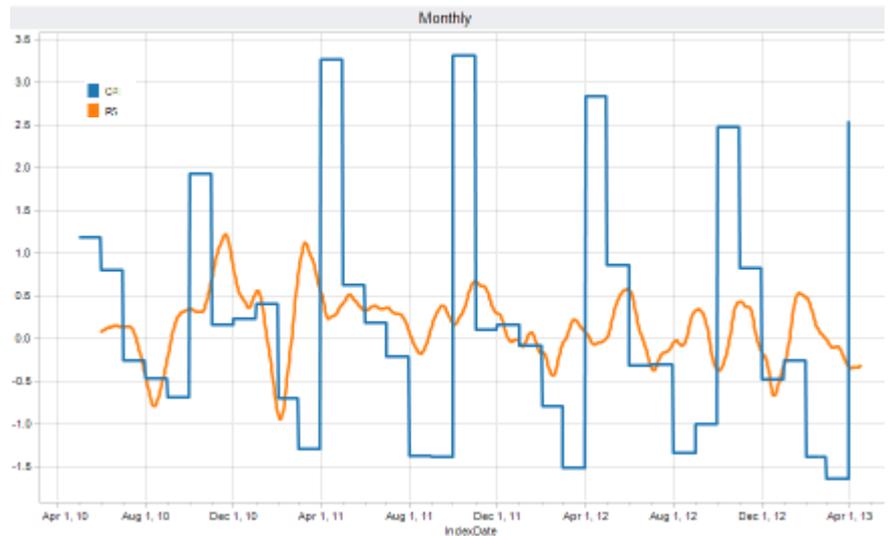
Differences: Quality Adjustments

- Simple indices can approximate the level and trend of CPI inflation in hedonic-adjusted categories (as suggested in Silver & Heravi (99), Aizcorbe, Corrado & Doms (2003))
- Implication: online series tend to be smoother than CPIs

Greece Price Index



Greece Monthly Inflation

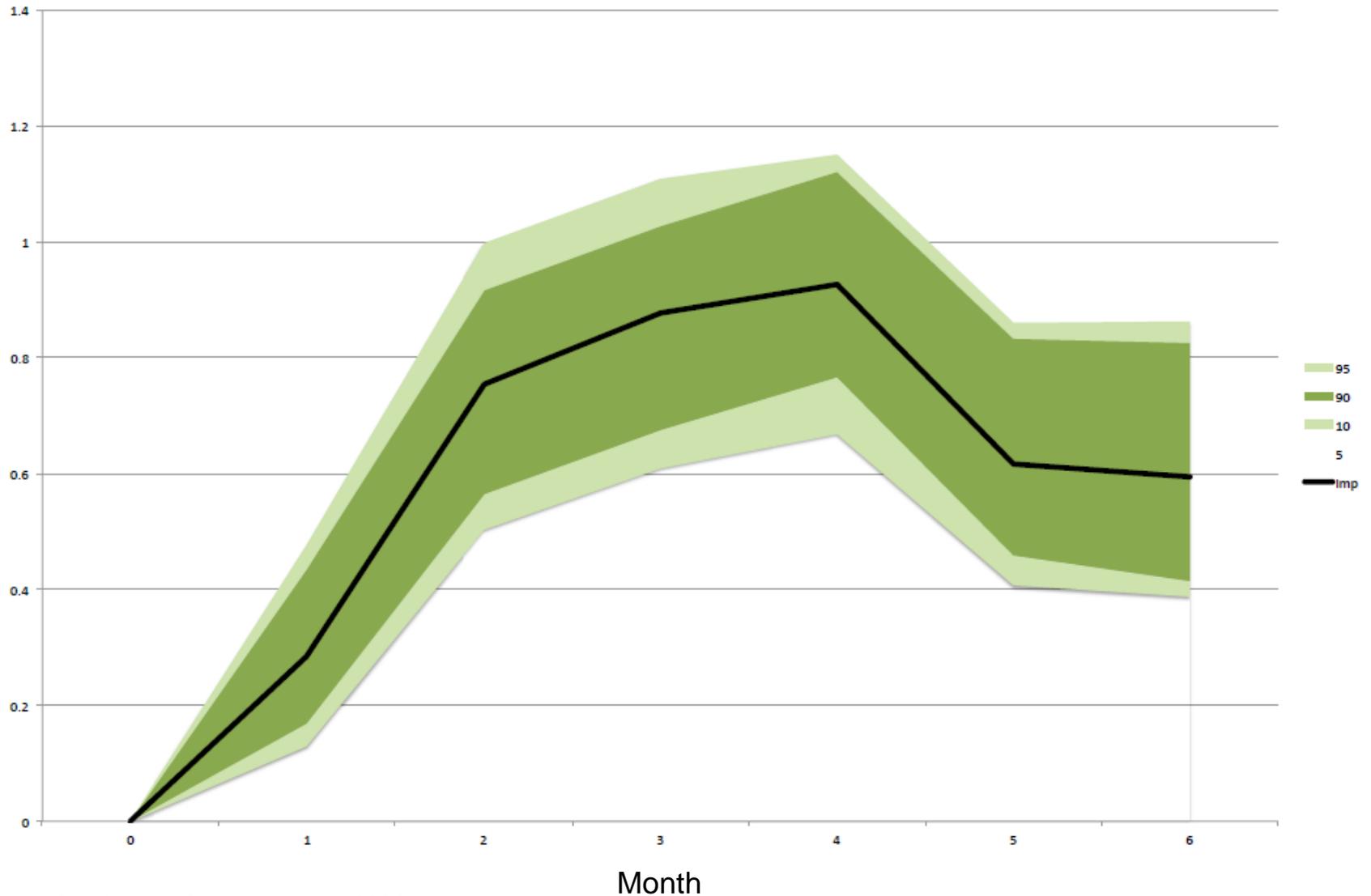


Anticipation

- We consistently find that online prices tend to anticipate changes in inflation trends.
 - This goes beyond the ability to collect and publish the data faster
 - Online prices tend to *react faster* to shocks.
 - Why?
 - Lower adjustment (or menu) costs
 - Online shoppers may be less sensitive to price changes
 - More intense and transparent competition
- We can study the link between online data and CPIs using simple VARs.
 - VAR regressions with Δ CPI on the LHS and lags of Δ CPI and Δ PS on the RHS (monthly data)
 - Impulse responses show the impact of a 1% shock in PriceStats on future CPI (reflecting additional information not contained in lagged CPI)

Impulse Response USA

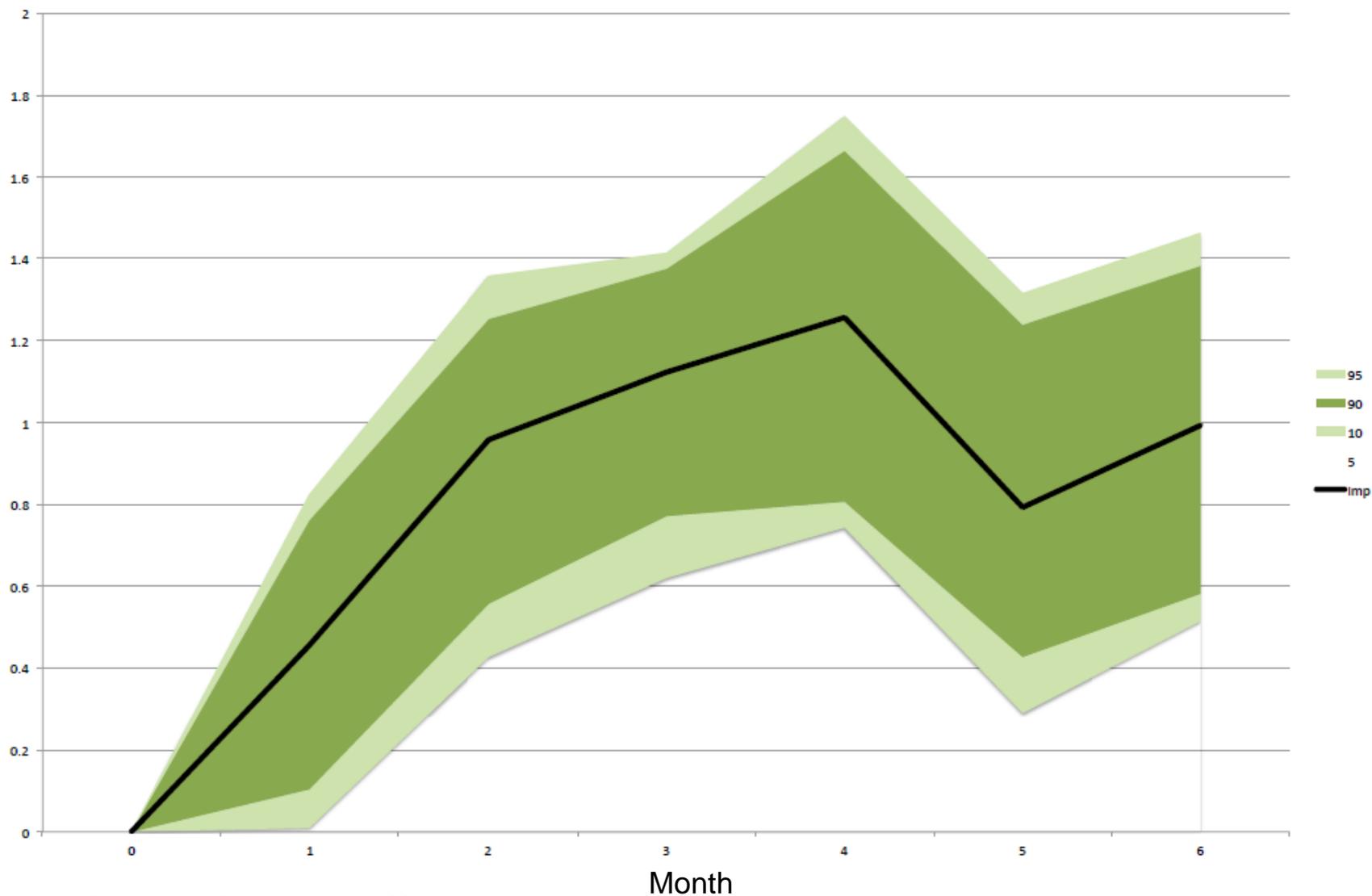
Cummulative IRF - 1% Shock to PriceStats Aggregate Inflation



Source: PriceStats – Data until March 2014

Impulse Response Eurozone

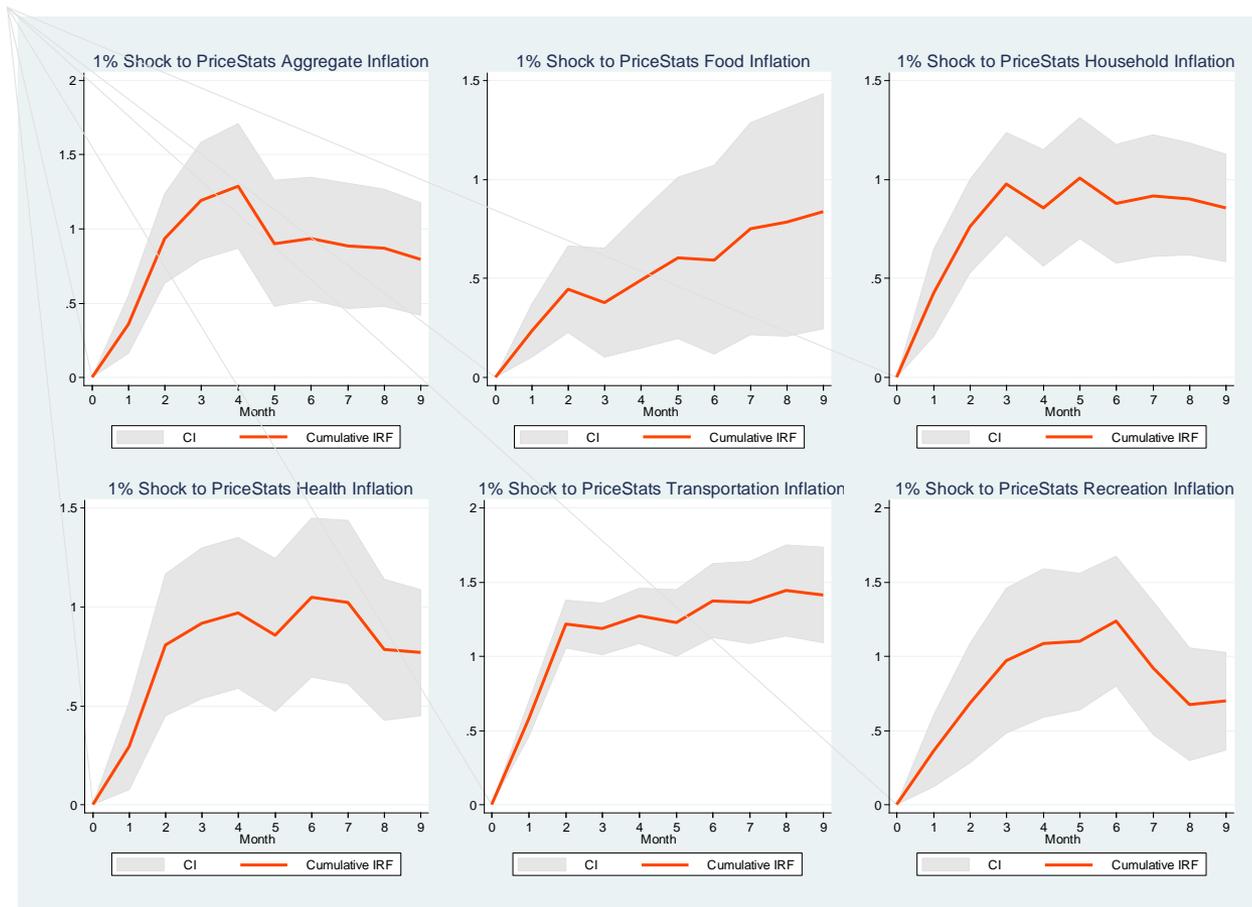
Cummulative IRF - 1% Shock to PriceStats Aggregate Inflation



Source: PriceStats – Data until March 2014

Impulse Responses for US Sectors

- There are differences in anticipation across US sectors
 - Shortest in Fuel (1 month)
 - Longest in Recreation (5 months)



Source: PriceStats – Data until June 2013

Other variables?

- Do online prices add more information than Gas Prices?
- We can test it with a simple extension:
 1. Run a regression of ΔCPI on lags of ΔCPI .
 2. Regress the residuals of 1) on Δ Gas Prices.
 3. Regress the residuals of 2) on Δ PS Aggregate and Δ PS Transportation (US)

Results	R2
1)	38.915%
2)	60.596%
3)	22.325%

- PriceStats series explain 22% of the variability of CPI inflation that is not explained by Gas Prices and lagged CPI
- This is all within-sample

Forecasting Example

- We forecast the US CPI inflation rate for 24 months, from January 2012 to January 2014.
- Each month t we first estimate the following model:

$$\Delta \log \text{CPI}_{t-1,t-2} = c_0 + \alpha_0 \Delta \log \text{HFD}_{t-1,t-2} + \sum_{i=1}^l \alpha_i \Delta \log \text{HFD}_{t-1-i,t-2-i} + \sum_{i=1}^l \beta_i \Delta \log \text{CPI}_{t-1-i,t-2-i}$$

- Each variable is first seasonally adjusted by regressing it on monthly dummies and using the residuals.
- The high frequency data (HFD) is either Gas Prices (values from last week of month) or any of the PriceStats (PS) series (values from the last day of month).
- We try models with different number of lags $l \in \{1, 2, 3\}$, and using different estimation windows $W \in \{24, 36, 48, 60 \text{ months}\}$
- We compute the fitted value for $\widehat{\Delta \log \text{CPI}}_{t,t-1}$ using the coefficients from the previous regression and the high frequency data available on month t . This forecast is repeated for 24 months.

Forecasting Example

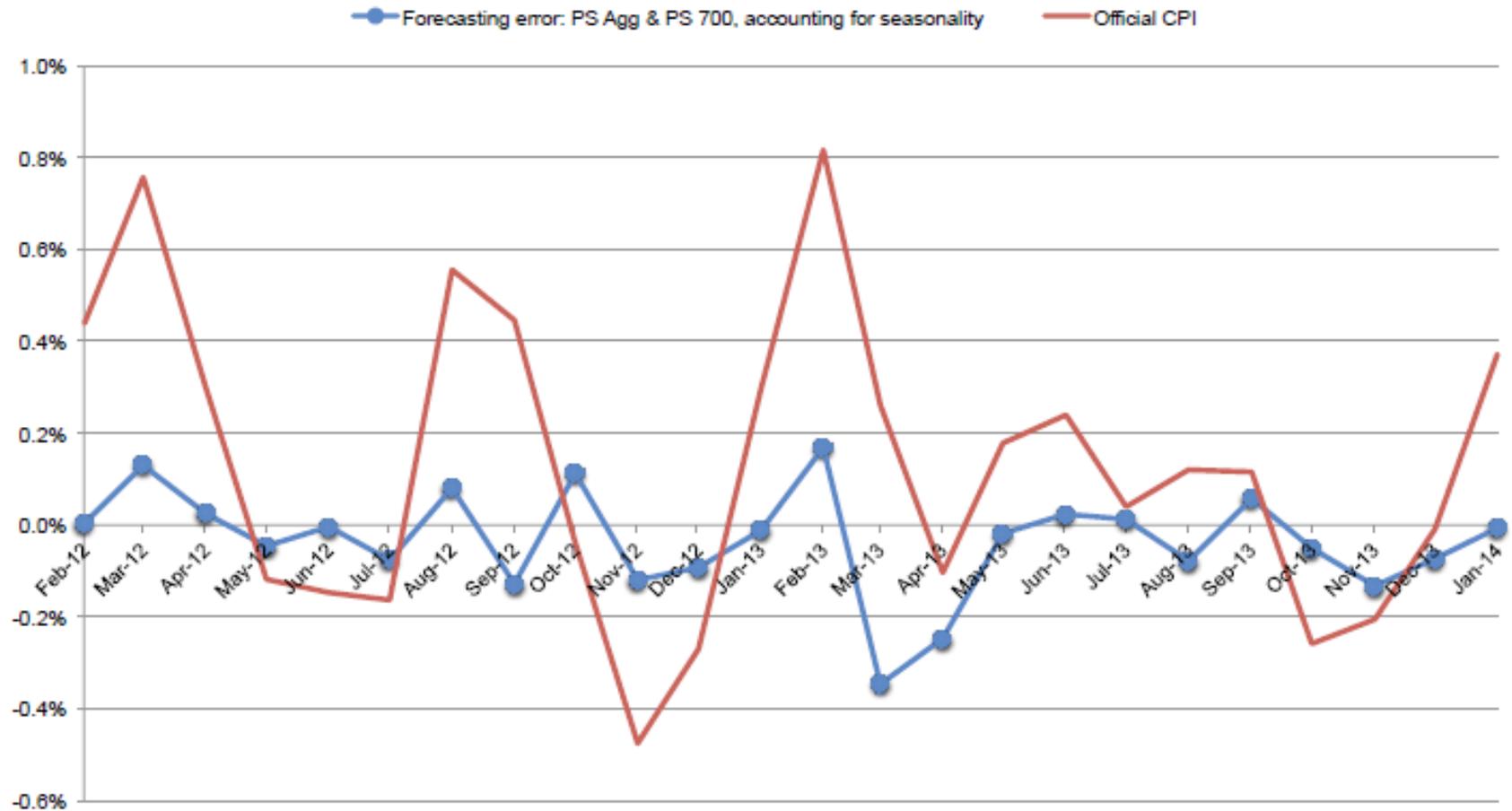
- We tried alternative models using
 - Only CPI
 - CPI and Gas Prices
 - CPI and PriceStats Series

Regressors	MAE %	RMSE %
CPI	0.164	0.221
CPI + Gas Prices	0.101	0.149
CPI + PS Aggregate	0.143	0.184
CPI + PS Food + PS Transportation	0.110	0.134
CPI + PS Aggregate + PS Transportation	0.085	0.117

Note: Averages of models with 1, 2, and 3 lags

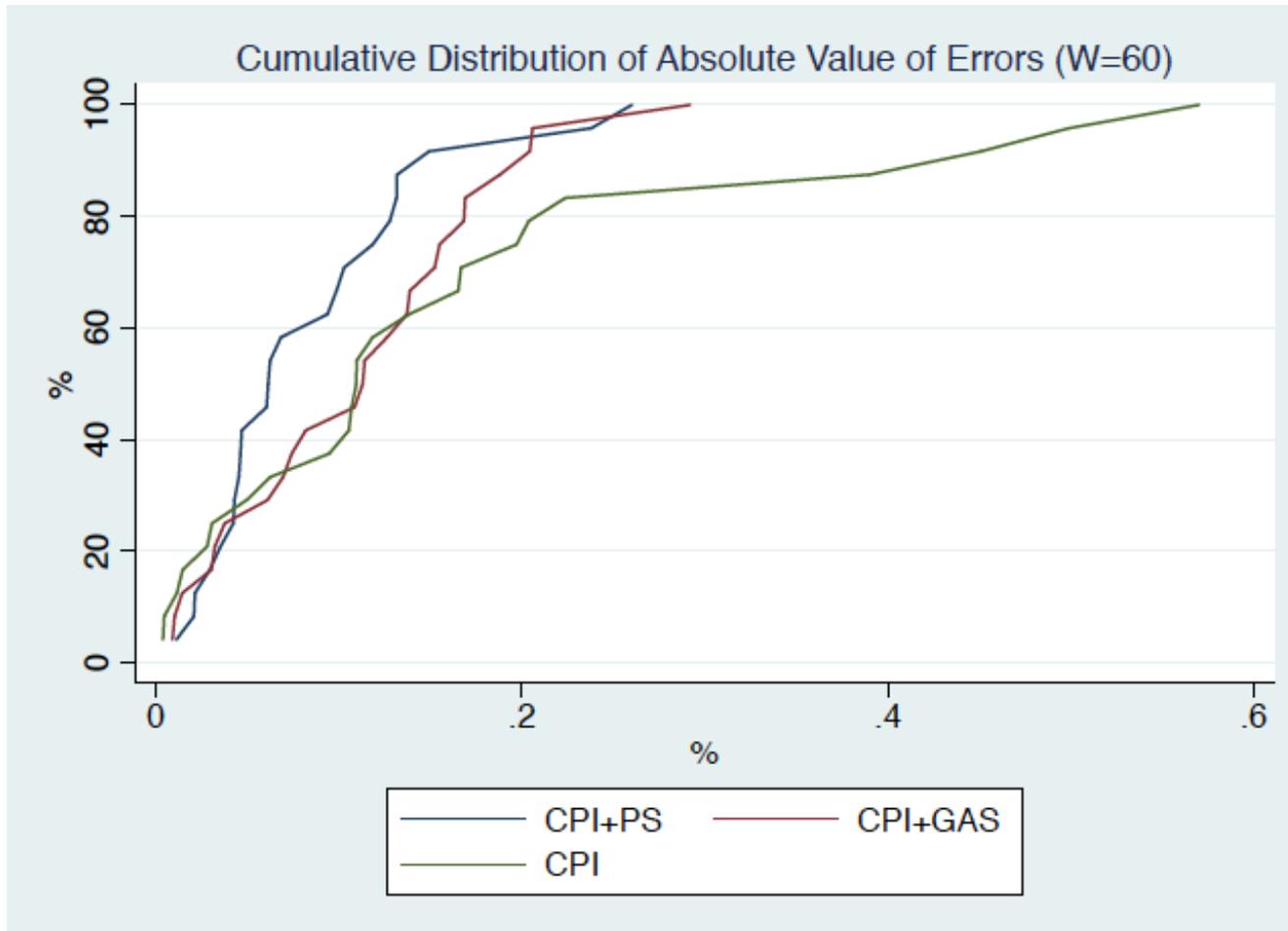
Forecasting Example

- Errors for model with $W=24$, seasonally adjusted, average of lags



Forecasting Example

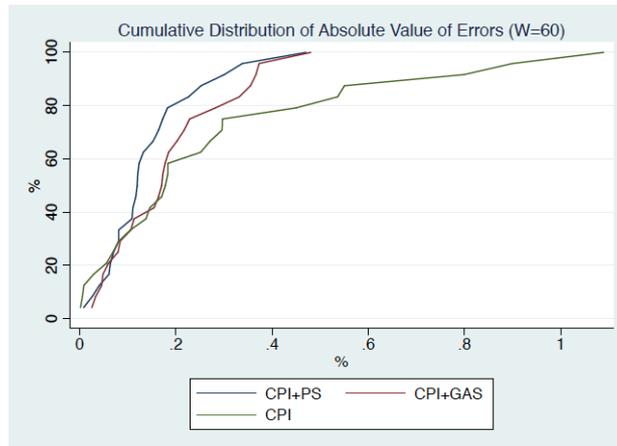
- Cumulative Distribution of Abs Value of Errors



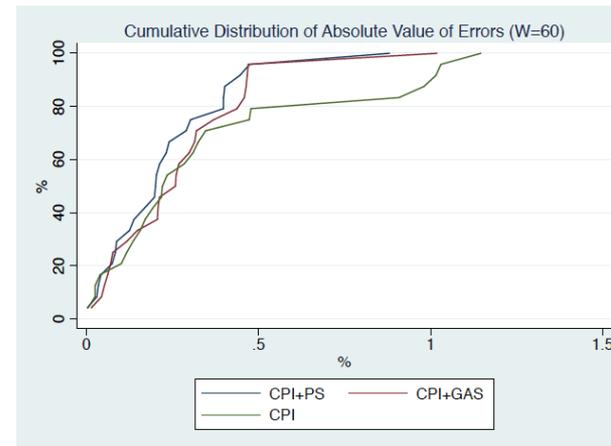
Forecasting Example

- We repeated the same exercise, but this time forecasting the inflation from t to $t+i$, where $i \in \{2, 3, 4, 5 \text{ months}\}$

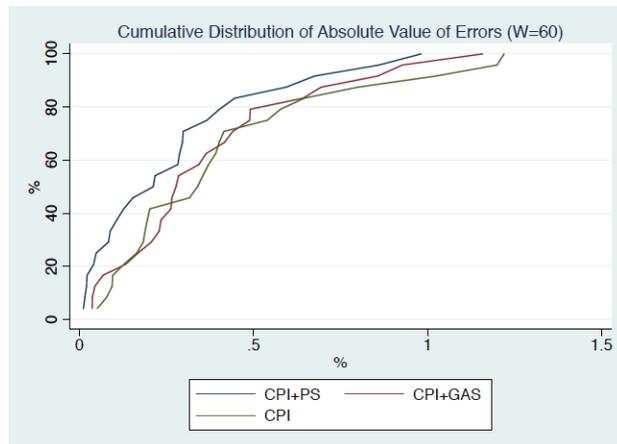
$i = 2$



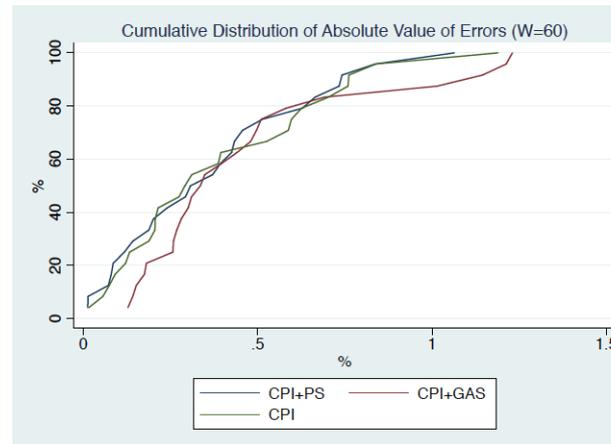
$i = 3$



$i = 4$



$i = 5$



Conclusions

- Online prices have the potential to dramatically increase the amount and quality of micro price data available for academic research.
 - Data characteristics help re-evaluate old empirical puzzles and answer questions that could not be tackled before
- Online data can also be a reliable source of information for inflation measurement
 - Congruence, differences, and heterogeneity
 - ``Big data`` characteristics can greatly *simplify* measurement
 - Best when used as an alternative source of data, not as a separate sector that needs special treatment
- Online prices tend to anticipate changes in inflation trends
 - Typical anticipation is 2-3 months
 - Provides unique information in real-time, different from other sources such as gas prices,
 - Particularly useful to forecast times of inflation ``surprises``
- Access to online data will become widespread in the following years.

