

# Real-Time Pricing in the Spanish Residential Electricity Market

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## Assessing the impacts of real-time pricing

- ▶ **Efficiency impacts:**

- ▶ Shift away consumption from high-demand expensive hours
- ▶ Long-run savings in the generation mix, e.g., less need for peakers or batteries

- ▶ **Equity impacts:**

- ▶ Distributional effects based on load profiles: peak consumers lose under RTP
- ▶ Bill shocks due to inability to respond in high price events

**We study these issues in the context of the Spanish electricity market, which rolled out real-time pricing for a large share of residential customers.**

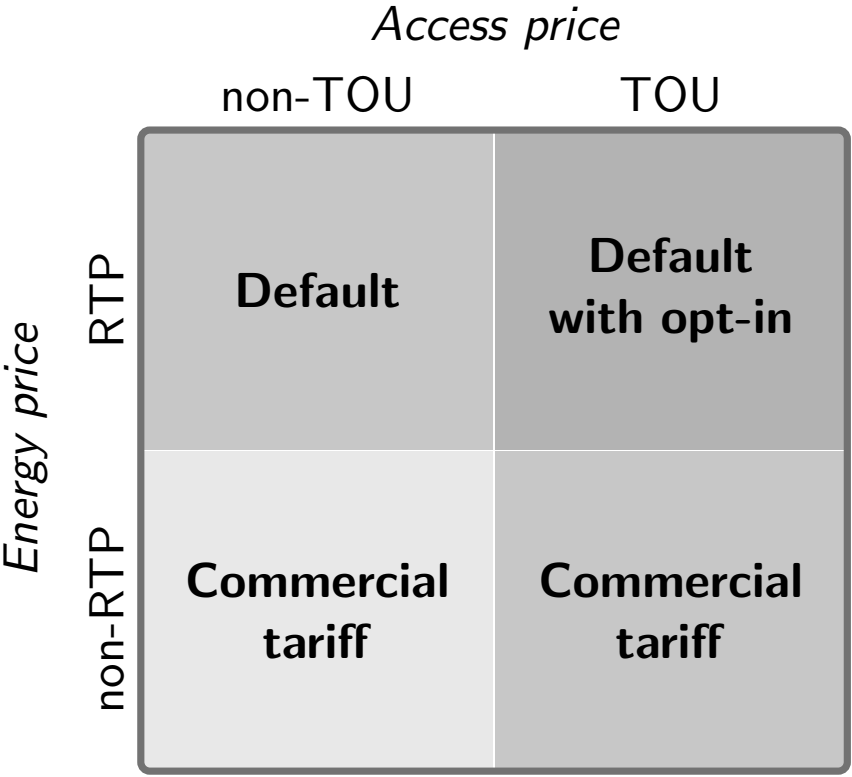
## Two projects

- ▶ **Efficiency impacts:**
  - ▶ **Estimating the Elasticity to Real-Time Prices**, joint work with David Rapson (UC Davis), Natalia Fabra (UC3M) and Jingyuan Wang (Northwestern), in preparation for AER P&P
- ▶ **Equity impacts:**
  - ▶ **The Distributional Implications of Real-time pricing**, joint work with Michael Cahana (UChicago), Natalia Fabra (UC3M) and Jingyuan Wang (Northwestern)

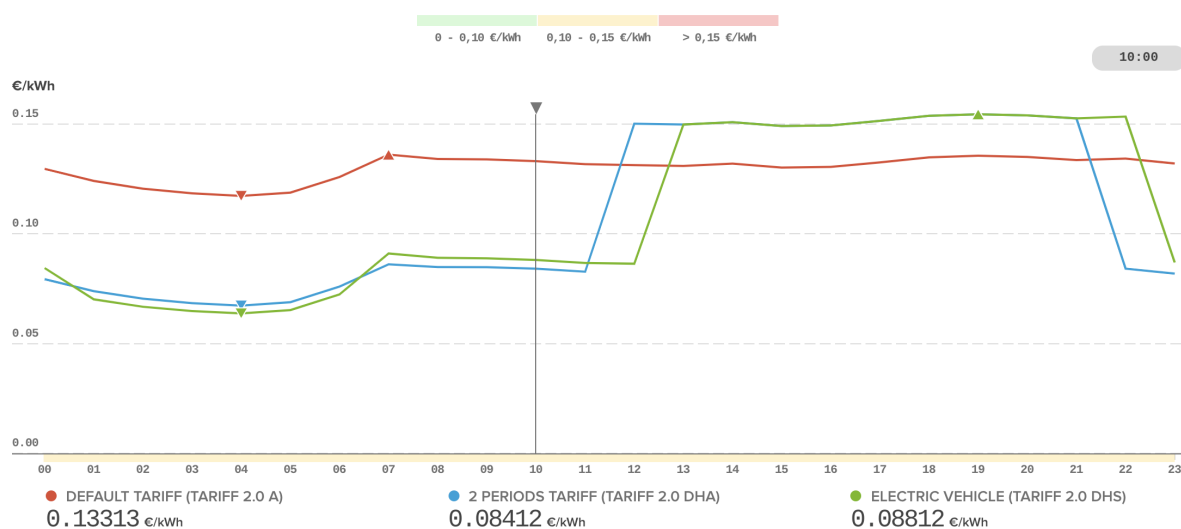
## Dynamic electricity pricing in Spain

- ▶ April 2014: Spain becomes the only country in which RTP is the **default option for all households**
  - ▶ *The case of Spain with a regulated default dynamic price contract is unique (EC, 2019)*
- ▶ Electricity **marginal price** composed of two parts:
  - ▶ **Energy price**: determined hourly as a function of the wholesale electricity market (**RTP**), or time-invariant
  - ▶ **Access price**: regulated costs charged as a function of consumption; peak/off-peak prices (**TOU**) or time-invariant
- ▶ Customers defaulted into RTP and non-TOU

# Tariff taxonomy



## Tariff taxonomy: prices over the day



**Figure:** Electricity prices over day: RTP and Non-TOU (red) and RTP and TOU (blue)

## Data

- ▶ We obtain smart-meter data for over 4M households, from two large Spanish utilities.
- ▶ For each household (January 2016-July 2017):
  - ▶ hourly electricity consumption
  - ▶ plan characteristics (pricing, maximum capacity)
  - ▶ postal code
- ▶ We link the postal code with detailed Census data:
  - ▶ education, income and age distribution, avg number of rooms...

Data: electricity consumption area

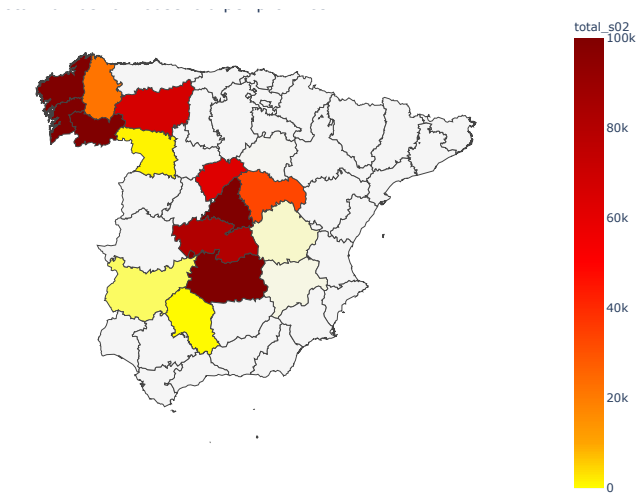


Figure: Naturgy area

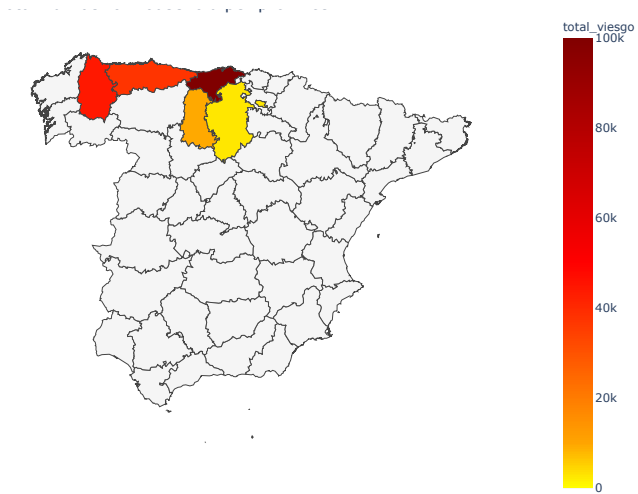


Figure: Viesgo area



## A first look at the data: prices

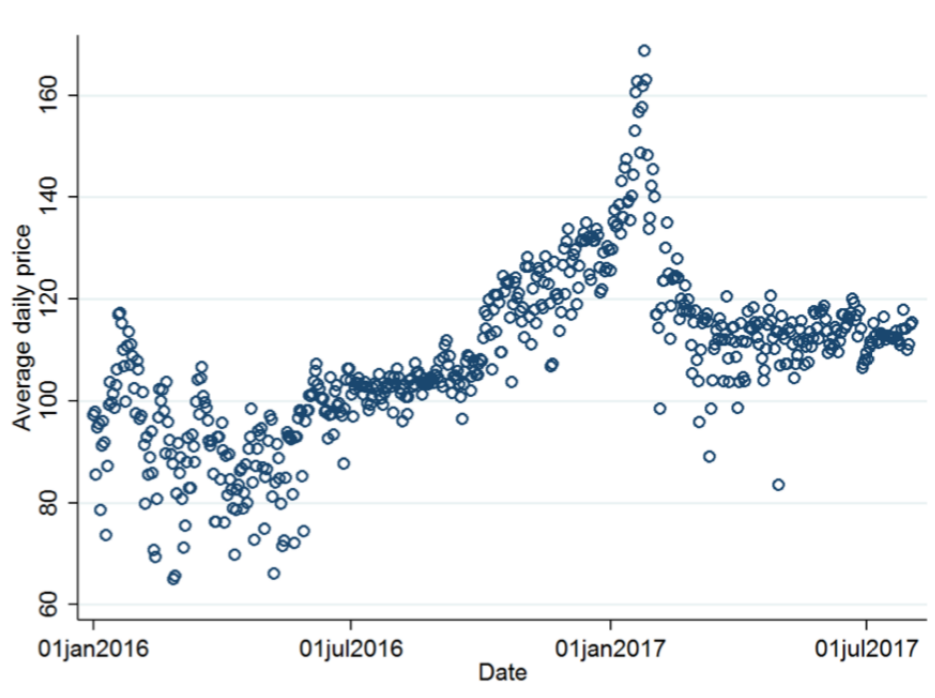


Figure: Average prices over the sample period

## A first look at the data: price variation

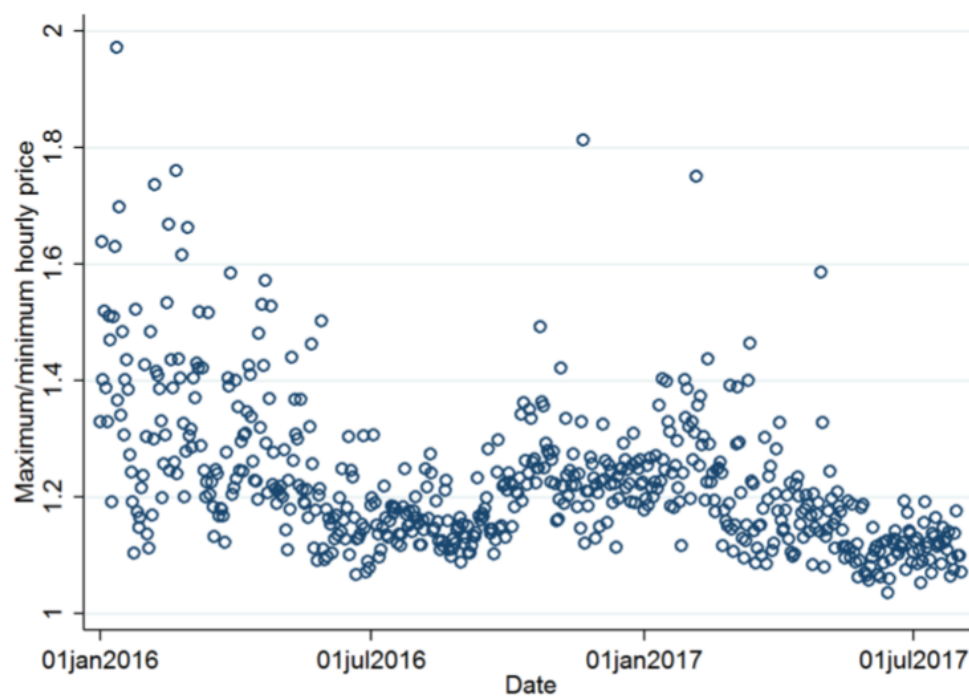


Figure: Ratio of the highest to lowest price every day

## Project 1: Objectives and findings

### Objectives:

- ▶ Measure households' demand response to RTP.
- ▶ Compare households' price response to RTP and TOU.

### Main findings:

- ▶ **RTP vs non-RTP** consumers appear to mostly not respond.
  - No impact of short run variation of RTP on demand
- ▶ **TOU vs non-TOU** consumers appear to behave differently.
  - Selection or actual response?
  - Important to disentangle for policy implications

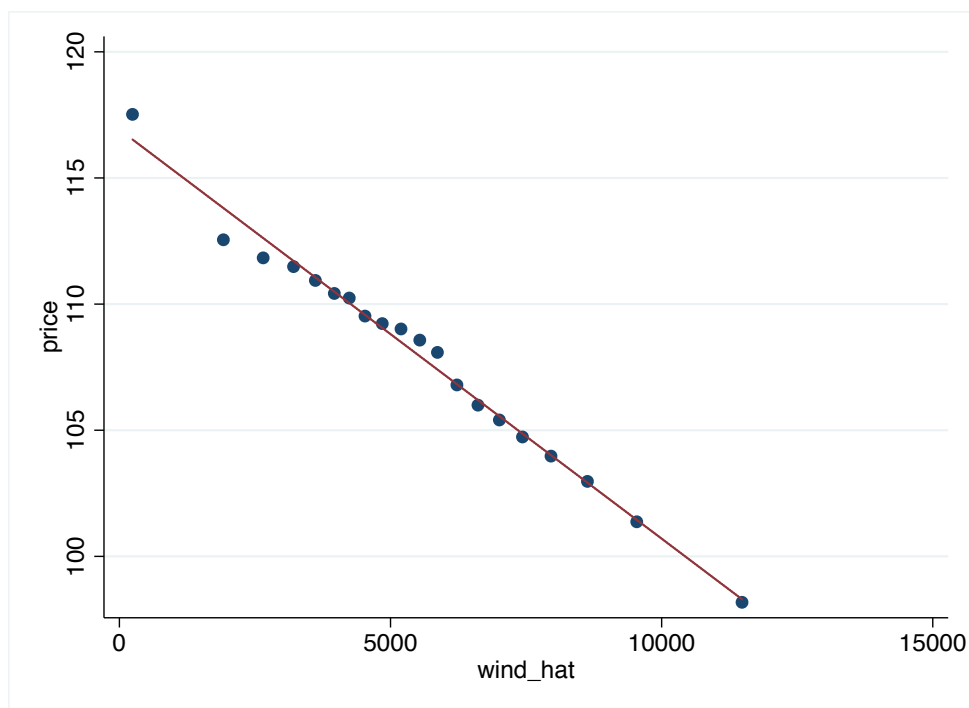
## Empirical strategy for RTP response

- ▶ We estimate the short-run price elasticity of households.
- ▶ Main regression (individual by individual):

$$\ln q_{ith} = \beta \ln p_{ith} + \phi X_{ith} + \gamma_{ith} + \epsilon_{ith}$$

- ▶ In baseline specifications, we control for:
  - ▶ Temperature bins by hour
  - ▶ Fixed effects: hour x month, year x month, day of week
  - ▶ Wind power forecasts as an IV for short-run price changes

## Instrumental Variable strategy



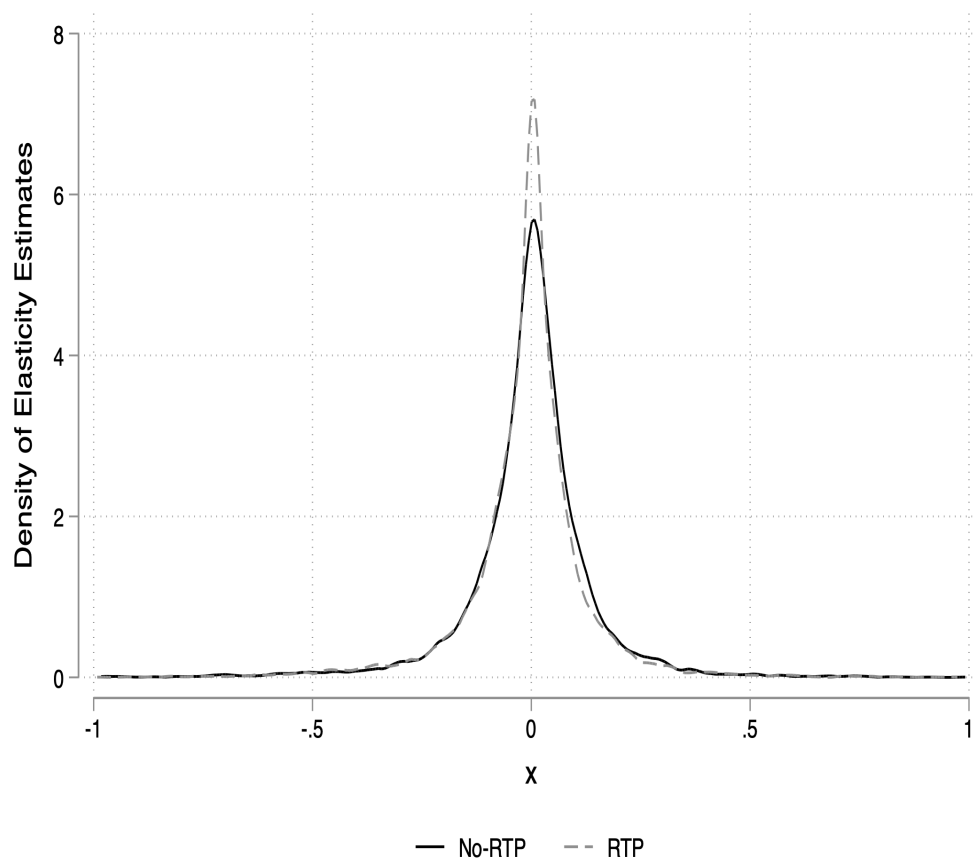
- ▶ Instrument shows strong first stage, also after conditioning
- ▶ Plausibly exogenous after controlling for local weather conditions

## Instrumental Variable challenges

- ▶ Most consumers do not consume electricity explicitly based on wind patterns, so exclusion restriction plausibly valid.
- ▶ Yet, wind patterns are intertwined with weather.
- ▶ Weather can affect electricity consumption in many ways: temperature control, sunset/sunrise, type of activities, time at home, etc.
- ▶ Difficult to control for potentially all confounders.
- ▶ High-frequency data can easily lead to significant spurious patterns due to omitted variable bias.

**We consider an array of fixed-effect individual specifications together with a lasso estimator.**

We find similar distributions of price elasticities



- Distribution centered around zero, median of no response.

Average elasticities by group are close to zero

	(1) p_iv11	(2) p_iv21	(3) p_iv31	(4) p_lasso
rtp	-0.00513 (0.00238)	-0.00430 (0.00237)	-0.00374 (0.00220)	-0.00468 (0.00217)
Constant	-0.00473 (0.00244)	-0.00883 (0.00252)	-0.0117 (0.00182)	-0.0237 (0.00274)
Observations	14598	14598	14598	14598

Standard errors in parentheses

► Not much of an effect from RTP.



## Heterogeneity in responses

- ▶ We link the individual estimates to Census demographic information at the zip code level.
  - ▶ We do **not** find covariates that could pick up potential heterogeneity in responses.
- ▶ For one of the utilities, we also observe whether consumers use an app (weekly).
  - ▶ We do **not** find any systematic differences in behavior for consumers that use the app frequently.

## TOU vs non-TOU

- ▶ Compare **TOU vs non-TOU customers**:
  - ▶ Are TOU consumers more responsive?
- ▶ Evidence suggests at the very least they are aware of pricing scheme.

TOU	Off-Peak	Peak
0	0.53	0.47
1	0.58	0.42

**Caution:** Comparison is endogenous due to selection, but suggests consumers either self-select or react to prices.

## Policy implications: RTP vs TOU

- ▶ Currently, RTP does not appear to induce large responses.
  - ▶ Demand response with automatic devices?
  - ▶ Demand response with larger price differences?
  - ▶ Demand response in the medium-run?
- ▶ TOU potentially **more effective** (certainty, salience?)...
- ▶ ...but theoretical literature emphasizes the **limits of TOU** to delivering all benefits from demand response
- ▶ **Key challenge:** intermittency really not addressed with TOU, at the very least it requires general patterns with seasonal adjustments (e.g., solar), but it doesn't work for wind