



Centre for Competition Law and Economics

Analysing cartel episodes: A Markov-switching application



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RATIONALE

- In many cartelized markets, periods of collusion are periodically interrupted by periods of competition
 - **Price wars**
- Studying empirical modelling of recurrent collusion speaks to core questions of collusive overcharge
 - **Empirical models of damage estimation commonly accept period of harm as a given**
 - **Studying transition to/from collusion**

RATIONALE

- Features of empirical model of recurrent collusion
 - **Establish ‘collusive’ episodes**
 - **Distinct data-generating processes during collusive and non-collusive periods**
 - **Estimate overcharge across collusive episodes**
 - **Account for transitions between collusive and non-collusive periods**

LITERATURE

- Collusion is state-dependent, often related to demand
 - Rotemberg & Saloner (1986); Haltiwanger & Harrington (1991); Bagwell & Staiger (1997), Fabra (2006)
 - State dependence implies recurrent nature
- Empirical studies
 - More important work on structural break tests
 - Boswijk *et al.* (2017), Crede (2015)

METHODOLOGY: RS

- Reduced-form, regime-switching model:

$$p_t = \begin{cases} c_0 + \omega + \sum_{l=1}^m a_l p_{t-l} + \sum_{l=0}^n \gamma_l \mathbf{x}_{t-l} + \varepsilon_t, & S_t = 1 \\ c_0 + \sum_{l=1}^m a_l p_{t-l} + \sum_{l=0}^n \gamma_l \mathbf{x}_{t-l} + \varepsilon_t, & S_t = 2 \end{cases}$$

with $\varepsilon_t \sim N(0, \sigma^2)$, p_t price, \mathbf{x}_t vector of demand and cost drivers

- S_t denotes regime in operation: $S_t = 1$ for collusive regime and $S_t = 2$ for non-collusive regime
- Alternative specifications also possible

METHODOLOGY: OVERCHARGE

- Replace intercept in ARDL with smoothed probabilities (α_{it})

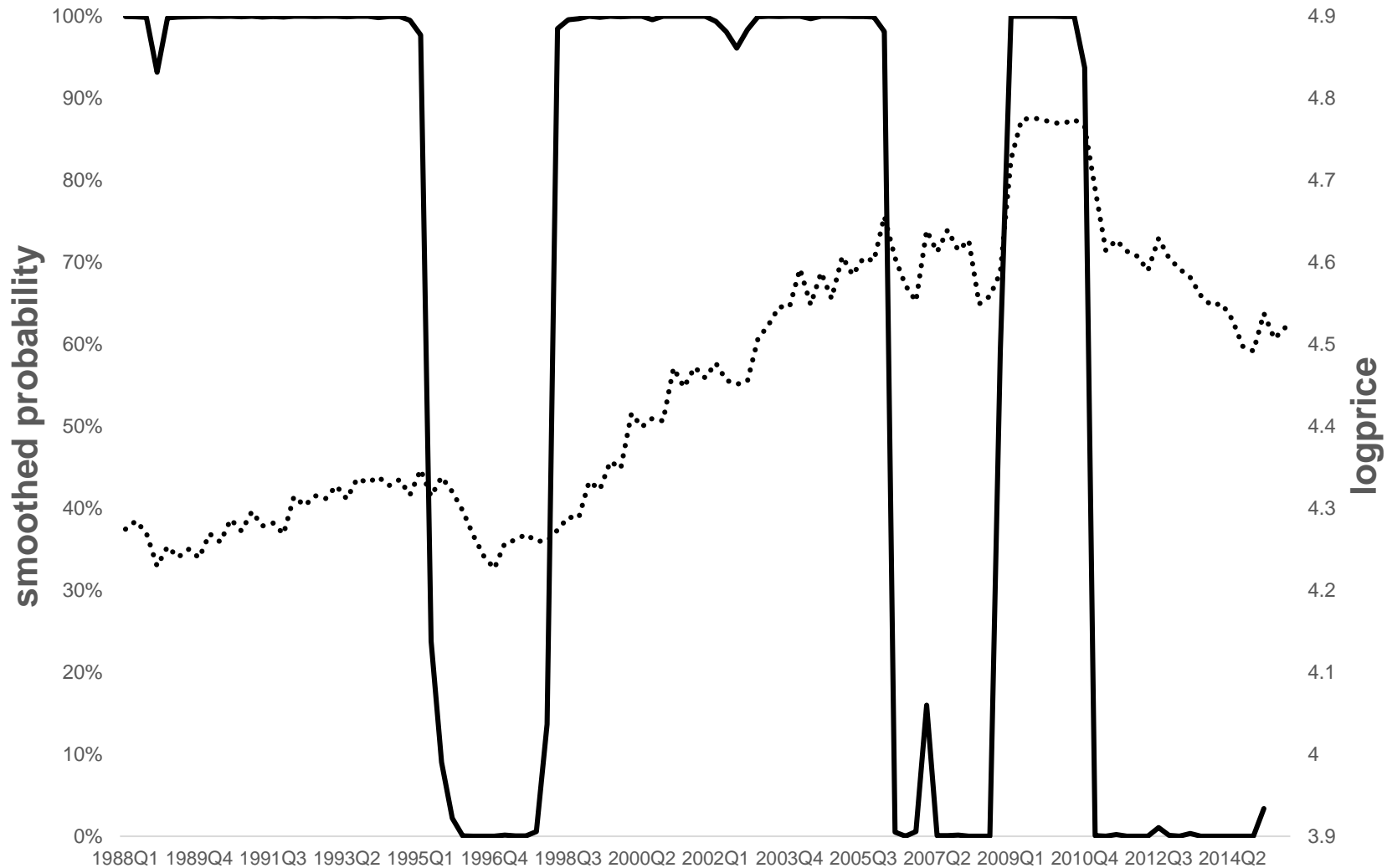
$$p_t = \beta \alpha_{it} + \sum_{l=0}^m a_l p_{t-l} + \sum_{l=0}^n \gamma_l x_{l-t} + \varepsilon_t$$

- $\alpha_{it} = \xi(S_t = i | \Omega_T; \theta)$ is the **smoothed probability from the RS model**
 - p_t is the **cement price**
 - x_t is a **vector of demand and cost drivers**
- Dynamic overcharge is taken as
 - $100 \times (e^\beta - 1) \times \alpha_{1,t}$

CASE STUDY: SA CEMENT MARKET

- History
 - Legal cement cartel 1940s until 1986
 - Exemption from competition law until 1996
 - New agreement in 1998, starting 1999
 - Inconclusive competition law investigation in 2000
 - Investigation in 2008 and subsequent leniency and settlement agreements
 - Court established illegal collusion from 1999 to 2009
- Sample period 1988 – 2015
- Drivers in model
 - Electricity, lime and limestone
 - House prices, sales volumes

RESULTS: REGIME PROBABILITIES

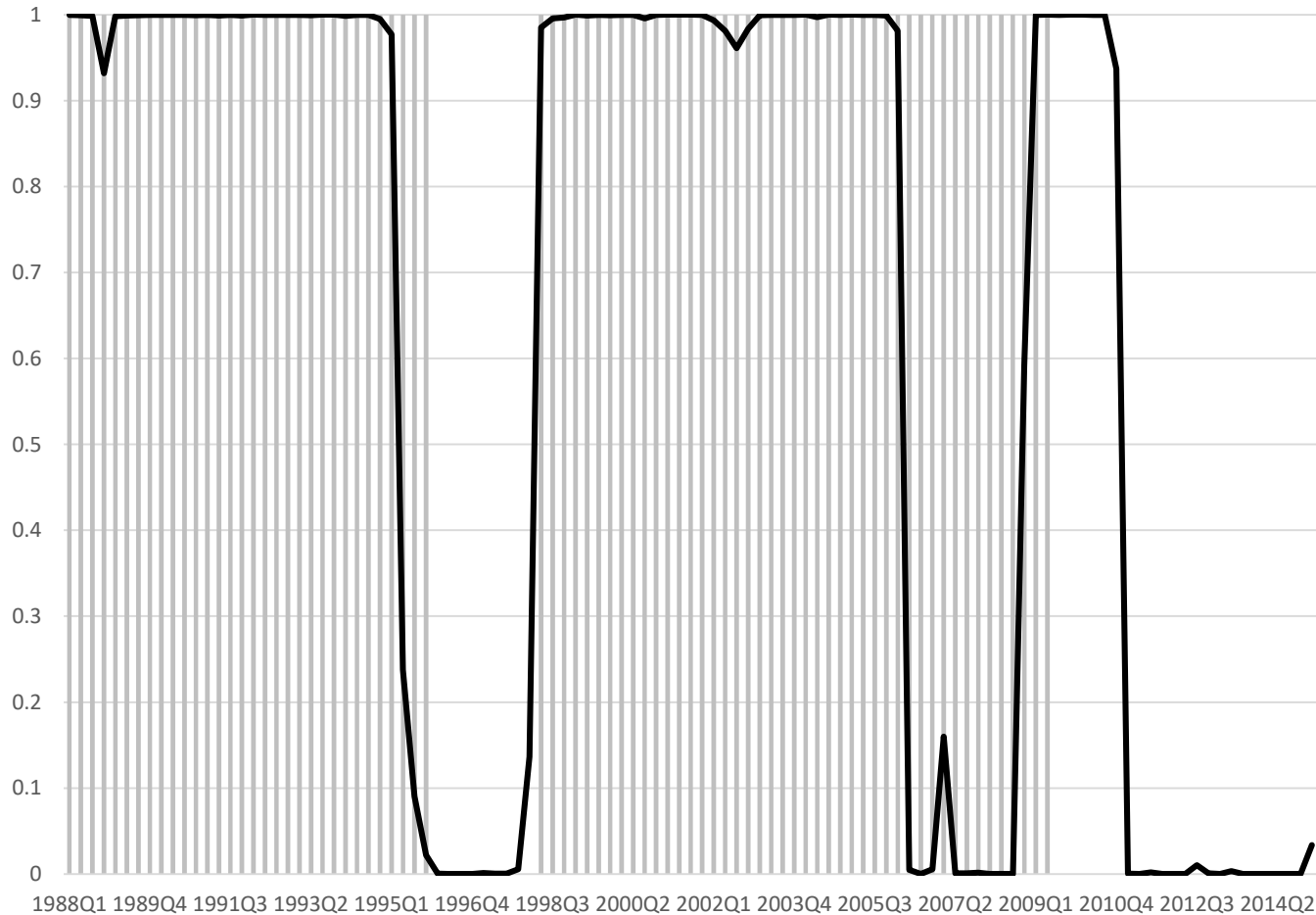


RESULTS: STATIC ESTIMATES FOR OVERCHARGE

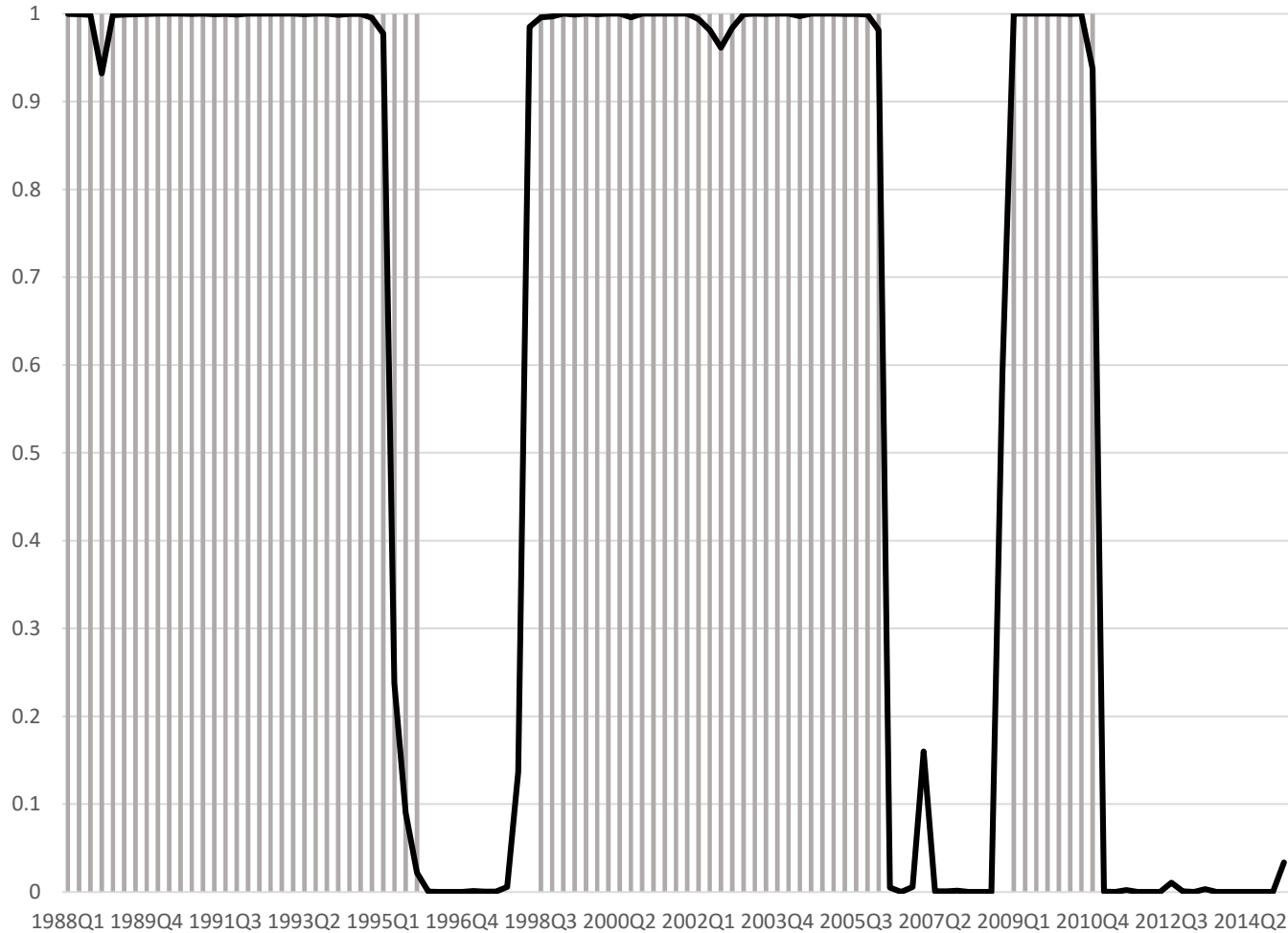
$$p_t = \beta \alpha_{it} + \sum_{l=0}^m a_l p_{t-l} + \sum_{l=0}^n \gamma_l x_{l-t} + \varepsilon_t$$

| Variable | Coefficient | Std. Error | t-Statistic | p-value |
|--------------------|-------------|------------|-------------|---------|
| Lime and limestone | 0.22 | 0.09 | 2.33 | 0.02 |
| House price | 0.18 | 0.02 | 9.59 | 0.00 |
| Sales | 0.54 | 0.09 | 5.74 | 0.00 |
| Electricity prices | 0.05 | 0.02 | 0.63 | 0.53 |
| Overcharge | 0.18 | 0.09 | 1.94 | 0.05 |

RESULTS: COMPARISON TO COURT DETERMINED DUMMY



RESULTS: DUMMY WITHOUT TRANSITION



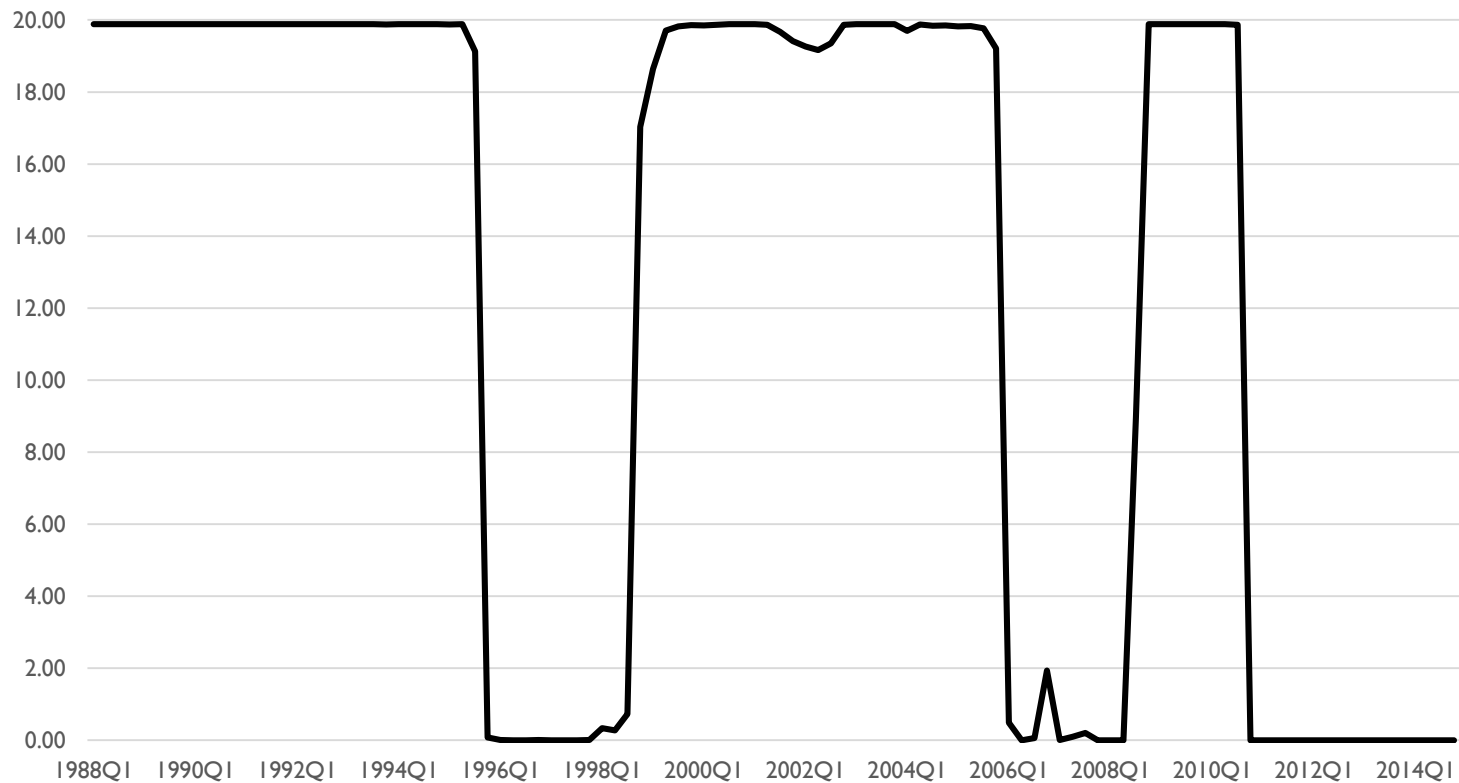
RESULTS: AVERAGE OVERCHARGE COMPARISON

| Smoothed probabilities | 'Official' dummy | Dummy based on Bai-Perron structural break | Smoothed probabilities but excluding transitions |
|-------------------------|------------------|--|--|
| OUR MODEL | | | |
| 18% | 1% | 4% | 13% |
| STATIC OLS MODEL | | | |
| 12% | 2% | 2% | 11% |

Average overcharge (18%) higher than standard dummy variable approaches

RESULTS: DYNAMIC OVERCHARGE

$$100 \times (e^{\beta} - 1) \times \alpha_{1,t}$$



RESULTS: DIAGNOSTICS OF ARDL WITH COURT DUMMY

| TEST | Test statistic | p-value |
|--|-----------------------------|---------|
| Jarque-Berra | $\chi^2(2) = 15.38$ | 0.26 |
| Breusch-Godfrey Serial correlation LM | $(n - 2) \times R^2 = 8.66$ | 0.01 |
| Breusch-Pagan-Godfrey Heteroskedasticity | $n \times R^2 = 42.03$ | 0.01 |
| ARCH-LM | $n \times R^2 = 1.18$ | 0.28 |

RESULTS: DIAGNOSTICS OF ARDL WITH SMOOTHED PROBABILITY

| TEST | Test statistic | p-value |
|--|------------------------------|---------|
| Jarque-Berra | $\chi^2(2) = 15.38$ | 0.73 |
| Breusch-Godfrey Serial correlation LM | $(n - 2) \times R^2 = 41.87$ | 0.23 |
| Breusch-Pagan-Godfrey Heteroskedasticity | $n \times R^2 = 29.96$ | 0.62 |
| ARCH-LM | $n \times R^2 = 13.17$ | 0.11 |

RESULTS: RS DIAGNOSTIC TESTS

| TEST | Test statistic | p-value |
|--------------|------------------------|---------|
| Jarque-Berra | $\chi^2(2) = 4.28$ | 0.978 |
| Ljung-Box | $\chi^2(8) = 9.48$ | 0.3 |
| ARCH-LM | $n \times R^2 = 13.17$ | 0.12 |

CONCLUSIONS

- This paper suggests an empirical model of recurrent collusion using a RS methodology
 - **Allows explicit testing for presence of multiple regimes**
 - **Allows simultaneous detection of periods of collusive harm and estimation of overcharge**
 - **Allows for smooth transitions between collusive and non-collusive episodes**



Thank you