Common Ownership and Competition in the Ready-to-Eat Cereal Industry

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Today's Paper

Two Research Questions

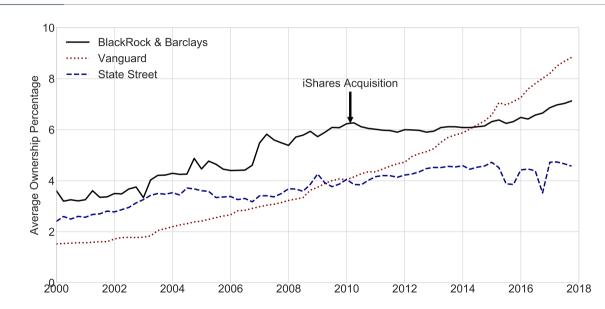
- 1. Applied: Do diversified investors with ownership stakes in multiple competitors cause firms to compete less aggressively? (aka the Common Ownership Hypothesis)
- 2. Methodological: How can we look at observational data on prices and quantities and use instruments to determine which model of competition generated our data? (Conduct Testing).

What is Common Ownership?

What is the Common Ownership Hypothesis?

- ▶ Economics starts from the premise that firms maximize profits.
 - Friedman (1953): natural selection of firms and billiards players.
 - Or, firms answer to investors, maximize shareholder value.
 - We generally think of these objectives as aligned (possibly subject to managers' agency frictions).
- ▶ So what do investors want?
 - Some (diversified) investors may hold stakes in you and your competitor. These are called "common owners."
 - Common owners are harmed when managers engage in business stealing.
- ► As a theory of firm behavior in joint ventures this is an old idea. The recent innovation is to extend this approach to passive or institutional investors

Rise of the Big Three (S&P 500 components, BCS AEJM 2021)



Common Ownership in the News

- ► The Atlantic: "Are Index Funds Evil?"
- ► The Economist: "Stealth Socialism"
- ▶ Bloomberg: "Index-Crazed Investors Turning S&P 500 Into One Gigantic Company"
- ► MoneyWeek: "Index Funds: Killing Capitalism?"
- ▶ Reuters: "When BlackRock Calls, CEOs Listen and do Deals"
 - "There is no CEO that doesn't return our call because typically we're up high on the shareholder register," Mark McCombe, global head of BlackRock's institutional client business, told Reuters reporters and editors attending the Reuters Global Wealth Management Summit on Friday.

"We are everybody's largest shareholder."

Some Reasonable (?) Assumptions (BCS 2020 P&P)

Most models of common ownership (implicitly) make the following assumptions:

Investor Portfolios Investors are entitled to a share of profits β_{fs} and hold portfolios $V_s = \sum_f \beta_{fs} \pi_f$

Manager Agency (Rotemberg 1984) Managers maximize a (γ) weighted average of investor payoffs $Q_f = \sum_s \gamma_{fs} V_s$.

These are pretty general since s could contain the manager herself.

Derivation of Profit Weight

Easy to show:

$$Q_f(x_f, x_{-f}) = \sum_{\forall s} \gamma_{fs} \cdot V_s(x_f, x_{-f})$$

$$\propto \pi_f + \sum_{g \neq f} \underbrace{\left(\sum_{\forall s} \gamma_{fs} \beta_{gs} \right)}_{\equiv \kappa_{fg}(\gamma_f \beta)} \pi_g$$

- $ightharpoonup \kappa_{fg}$ is interpreted as a profit weight, where one dollar of profits at firm g are valued as κ_{fg} dollars in the objective function of firm f.
- ▶ Depends on two primitives: β and γ .

Not appoint to a particular stratogic game

- β (cash-flow rights) is often directly observable.
 - γ embeds a model of corporate governance. Literature assumes $\gamma = \beta$ ("proportional control").

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A Simple Example (with $\gamma = \beta$)

- ► Firm 1 is controlled by an undiversified owner.
- ► Firms 2 and 3 have symmetric structures:
 - 60% undiversified, retail investors with no influence ($\gamma = 0$)
 - 20% two undiversified, institutional investors ($\gamma = 0.5$)
 - 20% commonly owned, institutional investor ($\gamma = 0.5$)

Then,

$$\mathcal{H}(\kappa) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1/2 \\ 0 & 1/2 & 1 \end{bmatrix}$$

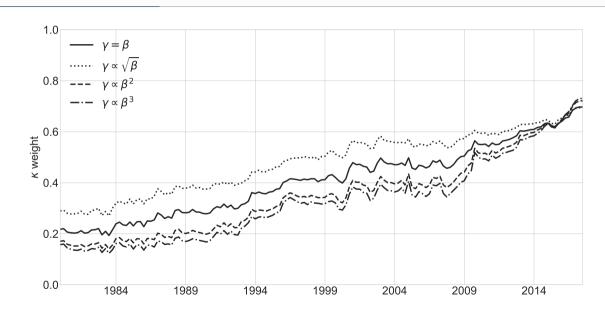
A Strange Example (still $\gamma = \beta$)

- ► Firm 1 has
 - \bullet N diversified, symmetric institutional investors with 1% each.
 - Undiversified retail investors $(\gamma = 0)$ own remainder.
- ► Firms 2 has
 - Same N institutional investors with x% each.
 - Undiversified retail investors $(\gamma = 0)$ own remainder.

Then,

$$\mathcal{H}(\kappa) = \begin{bmatrix} 1 & x \\ \frac{1}{x} & 1 \end{bmatrix}$$

From BCS in AEJ:Micro (2021)



Common Ownership and Pricing: Bertrand

Let κ represent the weight a firm places on competitors. Starting with the objective function,

$$\max_{p_j: j \in \mathcal{J}_f} \sum_{j \in \mathcal{J}_f} (p_j - mc_j) \cdot s_j(\mathbf{p}) + \underbrace{\sum_{g \neq f} \kappa_{fg} \sum_{k \in \mathcal{J}_g} (p_k - mc_k) \cdot s_k(\mathbf{p})}_{\text{unique to CO}}$$

We obtain first order conditions

$$s_{j}(p) + \frac{\partial s_{j}}{\partial p_{j}}(\mathbf{p}) \cdot (p_{j} - mc_{j}) + \sum_{k \in \mathcal{J}_{f}} \frac{\partial s_{k}}{\partial p_{j}}(\mathbf{p}) \cdot (p_{k} - mc_{k})$$

$$+ \underbrace{\sum_{g \neq f} \kappa_{fg} \sum_{k \in \mathcal{J}_{g}} \frac{\partial s_{k}}{\partial p_{j}}(\mathbf{p}) \cdot (p_{k} - mc_{k})}_{\text{unique to CO}} = 0.$$

From κ to Markups

A bit of notation, \mathcal{H} is the product-level version of κ :

$$\mathcal{H}_{jk}(\kappa) \equiv \begin{cases} 1 & \text{if } (j,k) \in \mathcal{J}_f \text{ for any } f, \\ \kappa_{fg} & \text{if } j \in \mathcal{J}_f \text{ and } k \in \mathcal{J}_g \text{ for any } (f,g), f \\ 0 & \text{otherwise.} \end{cases}$$

So that we can represent the system of first-order conditions where $\Omega_{ik}(\mathbf{p_t}) = -\frac{\partial q_k}{\partial \mathbf{p_t}}$

$$\mathbf{s_t} = (\mathcal{H}_t(\kappa) \odot \Omega(\mathbf{p_t})) \cdot (\mathbf{p_t} - \mathbf{mc_t}).$$

And therefore

$$\underbrace{) \odot \Omega(\mathbf{p_t}))^{-1} \mathbf{s}}_{\equiv \eta(\mathbf{p}, \mathbf{s}, \mathcal{H}(\kappa))}$$

Aside: MHHI and (Symmetric, Homogenenous Good) Cournot

If instead firms choose quantities

$$\max_{q_f} \left(p(Q) - mc_f \right) \cdot q_f + \sum_{g \neq f} \kappa_{fg} \cdot \left(p(Q) - mc_g \right) \cdot q_g$$

Gives share-weighted average markup:

$$\sum_{f} s_{f} \cdot \frac{p - mc_{f}}{p} = \frac{1}{\eta} \sum_{f} \sum_{g} \kappa_{fg} \cdot s_{f} \cdot s_{g}$$

$$MHHI = \mathbf{s}' \,\mathcal{H}(\kappa) \,\mathbf{s}$$

$$HHI = \mathbf{s} \cdot \mathbf{s}$$

$$MHHID = MHHI - HHI$$

Common Questions about Common Ownership

- 1. Isn't this just a partial-merger? (Yes).
- 2. In some cases (like double marginalization) don't partial-mergers better align incentives? (Yes).
- 3. Can we have $\kappa > 1$? (Yes). What does that mean? (Firms value own profits less than rivals, and have incentive to tunnel \rightarrow see BCS AEJ 2021).
- 4. What about an alternative for γ ? Banzhaf Power, Voting Models, etc. (Sure, but remember Arrow says microfounding social choice is hard).

Why Cereal?

Many Reasons

Long history in IO "model organism", but also non-Nash behavior isn't so crazy:

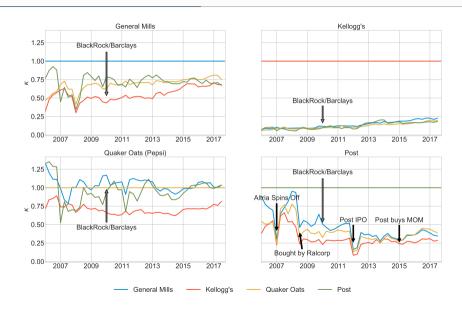
- ► Market structure: 4 big firms
- ► Nevo x 3
- ► FTC Case in the 1970's (Schmalensee 1978)
- ► Post/Nabisco Merger case (Rubinfeld 2000)
- ► Cartel/Price War in 1990s (Michel Weierergraeber 2022)

Cereal Data

Main Dataset is NielsenIQ (from Kilts) from 2007-2017

- ► Consolidate to dma-chain-week.
- ► Keep largest chains who price at chain level
 - Select based on # observations from panelist data.
- ► Consolidate upc → brand (Honey Nut Cheerios) from multiple package sizes and box designs.
 - Divide revenue by servings
 - Maintain the fiction that households purchase servings.

Cereal Data: Variation in κ



Foreshadowing

- ► Kellogg's should be a maverick who ignores cooperation from rivals and cares about own profit.
 - Instead they are the highest priced (margin) firm or are uniquely inefficient.
- ▶ Quaker Oats (Pepsi) should want to tunnel profits to rivals when $\kappa > 1$ and possibly price into inelastic part of demand curve.
 - They seem to be a low margin tough competitor. (or their mc is zero).
- \triangleright Post should cut prices sharply around large decreases in κ (de-listing events).
 - Actually they tend to raise prices around those times.

The "Forbidden" Slide

	(1)	(2)	(3)	(4)	(5)	(6)
	Log(Price)	Log(Price)	Log(Price)	Log(Price)	Log(Price)	Log(Price)
	b/se	b/se	b/se	b/se	b/se	b/se
hhi_servings	0.0495***	0.0857***	-0.0105	0.0181***	0.0578***	0.0559***
	(0.0068)	(0.0074)	(0.0066)	(0.0064)	(0.0057)	(0.0060)
mhhid_os_servings	-0.1216***	-0.1204***	-0.1523***	-0.1097***	-0.0833***	-0.0892***
	(0.0079)	(0.0081)	(0.0065)	(0.0064)	(0.0056)	(0.0059)
share_servings					-0.0116***	-0.0117***
					(0.0003)	(0.0003)
DMA FEs	No	Yes	No	Yes	Yes	Yes
Retailer FEs	No	No	Yes	Yes	Yes	Yes
Quadratic Time Trend	Yes	Yes	Yes	Yes	Yes	Yes
Parent FEs	Yes	Yes	Yes	Yes	Yes	Yes
Cubic HHI	No	No	No	No	No	Yes
r2_a	0.267	0.312	0.722	0.754	0.814	0.814
N	5173	5173	5173	5173	5173	5173

^{*} p<0.10, ** p<0.05, *** p<0.01

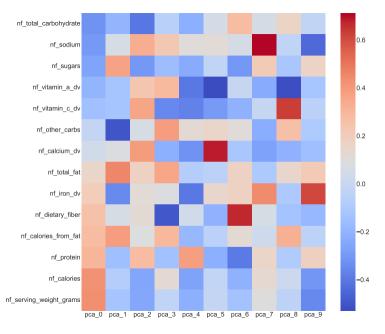
Both HHI and MHHI scaled for 1000 point increase.

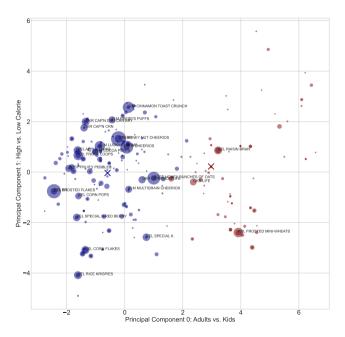
Empirical Application

Implementation: Demand Specification

$$u_{ijt} = h_d(\mathbf{x}_{jt}^{(1)}, \mathbf{v}_{jt}; \theta_1) - \alpha p_{jt} + \lambda \log(\mathrm{ad}_{jt}) + (\Sigma v_i + \Pi y_i) \cdot x_{jt}^{(2)} + \xi_{jt} + \varepsilon_{ijt}$$

- \triangleright y_i demographics: estimated at the dma-chain-year level (from panelists)
- y_i is joint distribution of (income, kids)
 - 1. Fit a lognormal for income to households w/ and w/o kids.
- $\triangleright \nu_i$ are random (normal) draws; price is lognormal.
- ▶ Lots of FE in $h_d(\cdot)$ (product, chain-dma, year, week of year)
- ▶ IV: Cost shifters, GH/Optimal IV $f(x_{-j})$, lagged advertising.





Demand Estimation

- ► We estimate demand system using PyBLP (Conlon Gortmaker RJE 2020)
- ► Highlights:
 - We estimate market size from milk and egg purchases.
 - Observable demographic preference shocks (income and children).
 - Random coefficients on: (constant, price, branded, servings per box, 3 PC's)
- ► Moments:
 - Own input costs and local demographic variables.
 - "Local" Gandhi-Houde differentiation instruments
 - We convert these into 21 "optimal instruments"
 - 520 micro-moments to get Π and Σ .

Implementation: Micro Moments

Also have 520 "micro-moments" grouped by DMA-Code/Retail Chain

$$\mathbb{E}\left[x_{jt} \times y_{it} \mid \text{purchase }\right] - \mathbb{E}\left[x_{jt} \times y_{it} \times \frac{s_{ijt}(\theta_1, \theta_2)}{1 - s_{i0t}(\theta_1, \theta_2)}\right] = 0.$$

- ► Match observed interactions of characteristics (constant, price, branded, servings per box, PC) & demographics from the model and the data.
- ► Conditional on purchase.
- ▶ We calculate these from Nielsen Panelist data by chain-dma-year.
- \blacktriangleright We carefully track # of observations to get variance calculations.
- ▶ We bootstrap the covariance from the sample (but not model).

See Conlon Gortmaker (2022) for details.

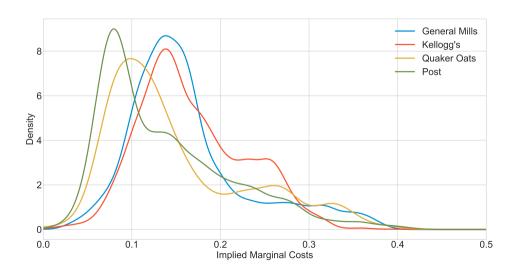
Diversion Ratios

$$D_{jk} = \frac{\partial q_j}{\partial p_k} / \left| \frac{\partial q_j}{\partial p_j} \right| = \frac{e_{jk}}{e_{jj}} \cdot \frac{q_j}{q_k}$$

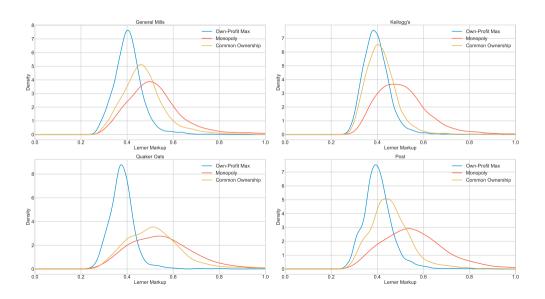
- ► Easier to interpret than cross elasticity
- ► Higher diversion implies closer competition
- \blacktriangleright See Conlon Mortimer (RJE 2021) for all kinds of tricks.

	Cheerios	Special K	Corn Flakes	Reese's Puffs	Capt Crunch	Froot Loops	Shares
HN Cheerios	5.07	4.27	3.75	5.33	3.58	3.48	2.69
Frosted Flakes	2.46	2.54	4.54	4.00	5.35	7.24	2.65
Cheerios	-	5.91	3.13	3.19	1.36	1.77	2.10
Honey Bunches	2.47	2.51	2.21	2.08	1.94	1.99	1.47
Cinn Toast Crunch	3.43	2.10	1.69	3.00	1.78	1.84	1.43
Froot Loops	1.26	1.19	1.64	1.69	1.82	-	1.18
Lucky Charms	2.18	1.64	1.57	2.99	1.59	1.58	1.14
Frosted Mini-Wheats	0.36	0.50	0.74	0.68	0.87	1.27	1.01
Corn Flakes	2.01	2.18	-	1.31	1.24	1.52	0.98
Rice Krispies	1.50	1.72	1.56	0.89	0.68	1.25	0.96
Apple Jacks	0.91	0.80	1.24	1.27	1.42	2.45	0.85
Raisin Bran (KEL)	0.46	0.47	0.63	0.78	0.82	1.24	0.79
Special K Red Berry	0.96	1.45	0.95	0.78	0.68	0.90	0.75
Special K	2.06	-	1.18	0.71	0.44	0.58	0.74
MG Cheerios	1.11	0.99	0.75	0.89	0.54	0.66	0.71
Reese's Puffs	1.36	0.86	0.87	-	1.08	1.01	0.69
Life	1.15	1.12	1.05	1.02	1.72	0.89	0.68
Cocoa Puffs	1.18	0.92	0.95	1.47	1.05	0.97	0.67
Capt Crunch	0.63	0.58	0.88	1.21	-	1.19	0.62
Capt Crunch Berry	0.68	0.61	0.83	1.15	3.29	1.00	0.58
Corn Pops	0.43	0.43	0.71	0.66	0.75	1.45	0.56
Cinn Life	0.76	0.75	0.83	0.84	1.59	0.78	0.54
Fruity Pebbles	0.61	0.59	0.71	0.71	0.75	0.77	0.44
Own Elas	-2.46	-2.66	-2.64	-2.70	-2.68	-2.71	-

Single Product: Implied Marginal Costs



Predicted Markups (Q4 2016)



Counterfactual Price Increases

	$GM ext{-}KEL$	$\operatorname{GM-QKR}$	GM-POST	KEL-QKR	KEL-POST	QKR-POST	Monopoly	κ^{CO}
GIS	6.94	1.53	3.30	-0.03	-0.09	-0.05	12.22	4.81
K	6.69	-0.03	-0.06	1.46	3.43	-0.03	12.07	6.87
PEP	-0.21	8.86	-0.22	8.72	-0.22	4.48	22.41	10.67
POST	-0.10	-0.08	7.43	-0.09	7.98	1.75	17.49	8.49
Overall	4.49	1.10	2.25	1.08	2.40	0.57	12.50	6.01

NB: Computed using marginal costs as predicted by own-profit maximization.

Greater than pairwise mergers, 48% of way to monopoly.

Main Results: These are N(0,1)

	Others' Cost	Demographics	BLP Inst.	Dmd.		
				Opt. Inst.		
Own Profit Max vs.	Panel 1: $A(\mathbf{z}_t) = \mathbf{z}_t$, linear $h_s(\cdot)$					
Single Product	1.6489	1.1116	0.9977	1.6432		
Common Ownership	-3.8928	-1.1957	0.5044	-1.0329		
Double Marginalization	1.4435	0.9892	-0.0427	5.5684		
${\bf Common\ Ownership + Double\ Marginalization}$	-0.1919	0.6815	0.1404	5.4688		
Perfect Competition	1.1730	0.4171	0.7364	3.9589		
Monopolist	-1.4097	-1.0680	-0.4523	-1.0908		
Own Profit Max vs.	Panel 2: $A(\mathbf{z}_t) = \mathbb{E}[\Delta \eta^{12} \mathbf{z}_t]$, linear $h_s(\cdot)$ and $g(\cdot)$					
Single Product	1.4264	0.5795	0.6662	0.7516		
Common Ownership	-2.3044	-0.5105	-0.0384	-1.4297		
Double Marginalization	0.8644	0.4421	-0.5311	2.9800		
${\bf Common\ Ownership + Double\ Marginalization}$	-0.9382	-0.2389	-0.3684	0.2460		
Perfect Competition	0.7164	0.6135	-0.1080	1.8776		
Monopolist	-0.8577	-0.4002	-0.3868	-1.1097		
Own Profit Max vs.	evs. Panel 3: $A(\mathbf{z}_t) = \mathbb{E}[\Delta \eta^{12} \mathbf{z}_t]$, random forest $h_s(\cdot)$ and $g(\cdot)$					
Single Product	5.0725	5.3347	5.2702	5.5758		
Common Ownership	-3.9823	-3.6775	-4.0786	-4.3977		
Double Marginalization	-6.3295	-9.9033	-6.5311	-7.6302		
Common Ownership + Double Marginalization	-6.6568	-6.8735	-6.6624	-7.5578		
Perfect Competition	-5.5005	-7.7749	-6.4083	-7.7653		
Monopolist	-3.7240	-4.4602	-3.6134	-3.8959		

An Internalization Parameter

Let κ represent the weight a firm places on competitors and τ the internalization of those weights.

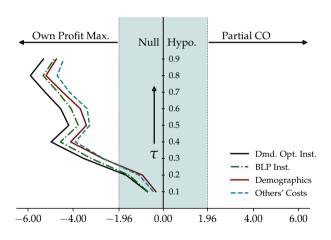
$$\arg \max_{p_j: j \in \mathcal{J}_f} \sum_{j \in \mathcal{J}_f} (p_j - mc_j) \cdot s_j(\mathbf{p}) + \sum_{g \neq f} \mathsf{T} \kappa_{fg} \sum_{j \in \mathcal{J}_g} (p_k - mc_k) \cdot s_k(\mathbf{p})$$

Now,

- ightharpoonup au = 0 implies own-profit maximization
- ightharpoonup $\tau = 1$ implies common ownership pricing
- $ightharpoonup \tau$ in between is..? Agency?

We test $\tau \in (0.1, ..., 0.9)$ against own-profit maximization.

Internalization Parameter Results



Stepping Back

- ▶ In order to evaluate the common ownership hypothesis, we developed a conduct testing procedure building on the identification results of Berry and Haile (2014)
- ▶ Have another specification of common ownership? We can test any pair of models such that
 - 1. Are fully specified, i.e. predict markups.
 - 2. Yield distinct markups $(\Delta \eta \neq 0)$.
 - 3. We have instruments that are relevant to $\Delta \eta$, excluded from the supply function $h_s(\cdot)$, and mean independent of ω^0 .
- ▶ Note that this is not specific to common ownership
 - Cartels: collusive versus oligopoly pricing?
 - Vertical contracts: DM or manufacturer pricing?
 - Labor: monopsony versus perfectly competitive labor markets?
 - Behavioral: suboptimal versus rational pricing rules?

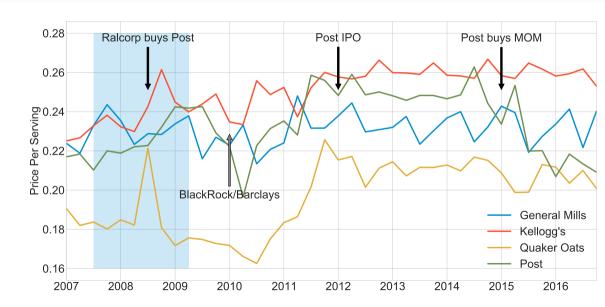
Wrapping Up

Takeaways:

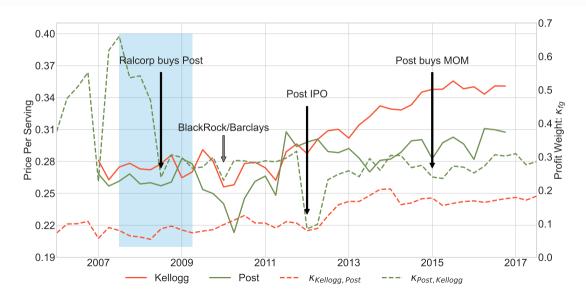
- ► Equilibrium markups are a nonlinear function of everything in the model. Using the model to get that nonlinearity right makes for a more powerful conduct test.
- ► In RTE cereal, we see strong evidence in favor of own-profit maximization rather than common ownership pricing.
- Discussion
 - We reject evidence of CO short-run price come petition in RTE Cereal.
 - Can't reject other mechanisms (CEO's living the quiet life)
 - Can't explain stock market pricing anomalies
 - In some sense CO is what we would see absent agency problems, so where are they?

Thanks!

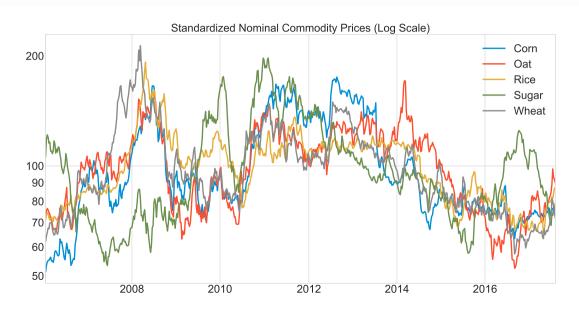
Cereal Data: Prices



Model Free Evidence?: Raisin Bran



Cereal Data: Input Prices



Cereal Data: Concentration

