# Product Differentiation and Oligopoly: a Network Approach 

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# Autoridade da Concorrência 2023 Competition Policy Award Webinar 

## Research Question

- Motivation: large dispersion in markups across firms
- Rising level \& dispersion (De Loecker, Eeckhout \& Unger, 2020)
- Rising industry concentration (Kwon et al. 2022)
- Research Question: what's behind this heterogeneity? What's driving these trends? What are the welfare implications?
- Consumer surplus and deadweight loss due to oligopoly
- Challenge: IO question in a macroeconomic setting:
- Standard IO tools are not available (scalability, lack of data)
- No systematic, objective way to define product markets.


## This Paper

- Methodological contribution: bring IO in macroeconomics.
- Theory of oligopoly and markups in general equilibrium
- Forget about industries: in this model, oligopolistic firms compete in a network of product market rivalries.
- New demand system: Generalized Hedonic-Linear (GHL).
- Taken to the data (and validated) for universe of US public firms, using product similarity data by Hoberg \& Phillips (2016).
- Decompose markups into 2 forces: productivity and centrality.
- Welfare measurement: large, increasing oligopoly deadweight loss ( $12.7 \%$ of total surplus in 2019), major distributional effects.


## Literature

- Rising Markups and Industry Concentration: De Loecker, Eeckhout \& Unger (2020), Grullon, Larkin \& Michaely (2019); Kwon, Ma \& Zimmermann (2021), Eeckhout \& Veldkamp (2022).
- Distortions, Input/Output, Micro Origins of Aggregate TFP: Gabaix (2011); Acemoglu, Carvalho, Ozdaglar, Tahbaz-Salehi (2012); Baqaee \& Farhi (2020); Bigio \& La'O (2020); Edmond, Midrigan \& Xu (2019); Carvalho, Elliot \& Spray (2022);
- Hedonic Demand/Empirical IO: Lancaster (1968); Rosen (1974); Epple (1987) Berry, Levinsohn \& Pakes (1994); Nevo (2001)...
- Network Games: Ballester, Calvo-Armengol \& Zenou (2006); Galeotti, Golub, Goyal, Talamer \& Tamuz (2022).
- Text Analysis/Product Similarity: Hoberg \& Phillips (2016).


## Theory

## Generalized Hedonic-Linear Demand

- $i=1,2, \ldots, n$ firms that behave as oligopolists.
- Hedonic demand: each firm's product is a bundle of characteristics (Lancaster, 1968; Rosen, 1974; Epple, 1987, Berry, Levinsohn \& Pakes 1994; etc.)
- 1 unit of product $i$ provides:
- 1 unit of an idiosyncratic characteristic $i$
- a vector of $k$ common characteristics $\mathbf{a}_{i}$ (length 1 )


## A basic example: 2 firms, 2 characteristics



## Aggregating common characteristics

## Characteristics (Nutrient Intake)

Matrix of Coordinates (Nutrition Facts)

Product Bundle

$$
\left[\begin{array}{c}
x_{1} \\
x_{2} \\
\vdots \\
x_{k}
\end{array}\right]=\left[\begin{array}{cccc}
a_{11} & a_{12} & \cdots & a_{1 n} \\
a_{21} & a_{22} & \cdots & a_{2 n} \\
\vdots & \vdots & \ddots & \vdots \\
a_{k 1} & a_{k 2} & \cdots & a_{k n}
\end{array}\right]\left[\begin{array}{c}
q_{1} \\
q_{2} \\
\vdots \\
q_{n}
\end{array}\right]
$$

$$
\mathbf{x} \quad=
$$

Aq

## Defining Cosine Similarity



## Representative Consumer-Worker-Investor

- Quadratic utility $U(\mathbf{x}, \mathbf{y}, H)=$

$$
\alpha \cdot \sum_{k=1}^{m}\left(b_{k}^{x} x_{k}-\frac{1}{2} x_{k}^{2}\right)+(1-\alpha) \sum_{i=1}^{n}\left(b_{i}^{y} y_{i}-\frac{1}{2} y_{i}^{2}\right)-H
$$

- $H=$ hours worked - numeraire
- Consumer faces vector of prices $\mathbf{p}$ and chooses demand $\mathbf{q}$, subject to profits and labor income being $\geqslant \mathbf{p} \mathbf{\prime} \mathbf{q}$.


## Inverse Demand

$$
\mathbf{p}=\mathbf{b}-(\mathbf{I}+\mathbf{\Sigma}) \mathbf{q}
$$

where

$$
\boldsymbol{\Sigma} \stackrel{\text { def }}{=} \alpha\left(\mathbf{A}^{\prime} \mathbf{A}-\mathbf{I}\right)
$$

## Cost Function and Competition

- Cost function (can be relaxed): $h_{i}=f_{i}+c_{i}^{0} q_{i}+\frac{1}{2} \delta_{i} q_{i}{ }^{2}$
- Cournot Competition: firm $i$ chooses supply $q_{i}$ to maximize profits function $\pi_{i}$ (also quadratic)
- (Linear-quadratic) Network game
- Ballester, Calvó-Armengol \& Zenou, 2006
- Why? the matrix of cosine similarities $\mathbf{A}^{\prime} \mathbf{A}$ (proportional to $\boldsymbol{\Sigma}$ ) can be thought of as an adjacency matrix of a network



## Cournot-Nash Equilibrium

The expression above can be shown to be a measure of network centrality (Katz-Bonacich)

## Hedonic-Adjusted Productivity



- Accounts for product quality
- Volumetric-invariant
- Comparable across widely-different firms


## Decomposing Markups

## Monopolistic Markup <br> $$
=\left(1+\omega_{i}\right) / 2
$$

$$
\mu_{i}=\chi_{i}+\left(1-\chi_{i}\right) \bar{\mu}_{i}
$$

## Product Market Centrality

Depends on the entire matrix of cosine
similarities $\mathbf{A}^{\prime} \mathbf{A}$. The profit share of surplus
is a decreasing function of $\chi_{i}$ alone

## Data and Validation

## Hoberg \& Phillips (2016 JPE) Product Similarity

- By law, every public corporation in the US has to file SEC form 10-K on a yearly basis.
- First 6-10 pages contain the "Business Description".
- HP created time-varying measures of cosine similarity between firms by text-mining these business descriptions.
- Solve long-standing problems with NAICS/SIC: static, binary do not reflect product market competition, can be manipulated.
- Highly incentive compatible - standard in finance: use of NAICS and SIC is no longer considered acceptable to capture product market rivalries, at least for top finance journals.


## Construction

$$
\mathbf{v}_{i}=\left[\begin{array}{c}
v_{i, 1} \\
v_{i, 2} \\
\vdots
\end{array}\right] \quad \cos _{i j}^{\mathrm{HP}} \stackrel{\text { def }}{=} \frac{\mathbf{v}_{i}^{\prime} \mathbf{v}_{j}}{\sqrt{\left\|\mathbf{v}_{i}\right\|\left\|\mathbf{v}_{j}\right\|}}
$$

Identification: bijective mapping words $\rightleftarrows$ characteristics, $\mathbf{a}_{i}$ and $\mathbf{v}_{i}$ are collinear up to permutation $\Rightarrow \mathbf{a}_{i} \mathbf{a}_{j} \equiv \cos _{i j}{ }^{\mathrm{HP}}$

## Identification

- Compustat: Revenues $\left(p_{i} q_{i}\right)$, COGS $\left(\mathrm{TVC}_{i}\right)$, SG\&A $\left(f_{i}\right)$.
- Assume $\delta_{i}=0$ (later relaxed). Only one free parameter: $\alpha$.
- Proposition: $\partial \log p_{i} / \partial \log q_{j}$ is observed for firm pair (K,Q):

$$
\alpha=-\frac{\varepsilon_{\mathrm{KQ}} \cdot p_{\mathrm{K}} q_{\mathrm{K}}+\varepsilon_{\mathrm{QK}} \cdot p_{\mathrm{Q}} q_{\mathrm{Q}}}{2 \cdot \cos _{\mathrm{KQ}}^{\mathrm{HP}} \cdot \sqrt{p_{\mathrm{K}} q_{\mathrm{K}}-\mathrm{TVC}_{\mathrm{K}}} \cdot \sqrt{p_{\mathrm{Q}} q_{\mathrm{Q}}-\mathrm{TVC}_{\mathrm{Q}}}}
$$

- Every other object is identified in closed form (correct units).


## Identification

$$
\begin{gathered}
q_{i}=\sqrt{\pi_{i}} \\
c_{i}=\frac{\mathrm{TVC}_{i}}{q_{i}} \\
\mathbf{b}=(2 \mathbf{I}+\boldsymbol{\Sigma}) \mathbf{q}+\mathbf{c}
\end{gathered}
$$




|  |  |  | Demand Elasticity $\left(\frac{\partial q_{i}}{\partial p_{j}} \cdot \frac{p_{j}}{q_{i}}\right)$ |  |
| :---: | :---: | :---: | :---: | :---: |
| Market | Firm $i$ | Firm $j$ | Micro Estimate | GHL $($ text-based $)$ |
| Auto | Ford | Ford | -4.320 | -5.197 |
| Auto | Ford | General Motors | 0.034 | 0.056 |
| Auto | Ford | Toyota | 0.007 | 0.017 |
| Auto | General Motors | Ford | 0.065 | 0.052 |
| Auto | General Motors | General Motors | -6.433 | -4.685 |
| Auto | General Motors | Toyota | 0.008 | 0.005 |
| Auto | Toyota | Ford | 0.018 | 0.025 |
| Auto | Toyota | General Motors | 0.008 | 0.008 |
| Auto | Toyota | Toyota | -3.085 | -4.851 |
| Cereals | Kellogg's | Kellogg's | -3.231 | -1.770 |
| Cereals | Kellogg's | Quaker Oats | 0.033 | 0.023 |
| Cereals | Quaker Oats | Kellogg's | 0.046 | 0.031 |
| Cereals | Quaker Oats | Quaker Oats | -3.031 | -1.941 |
| Computers | Apple | Apple | -11.979 | -8.945 |
| Computers | Apple | Dell | 0.018 | 0.025 |
| Computers | Dell | Apple | 0.027 | 0.047 |
| Computers | Dell | Dell | -5.570 | -5.110 |

Variable: $\log \left|\frac{\partial q_{i}}{\partial p_{j}} \cdot \frac{p_{j}}{q_{i}}\right|$, residualized on $(i=j)$ dummy and Market Fixed Effects


## Markups: Time Series



## Markups: Cross-Section



## Empirics

## Distribution of Hedonic-Adjusted Productivity



## Distribution of Product Market Centrality



## Total Surplus and its Distribution



## Deadweight Loss from Oligopoly



## Robustness \& Extensions

- Private and foreign firms, entry and exit
- Aggregation result: add competitive fringes of atomistic firms in the form of a representative firms.
- Can be located using firm-sector similarity from FHP.
- Non-flat marginal cost
- Exclude "non-tradable" industries
- Bertrand
- Multi-product firms (using Compustat Segments)
- Input-Output Linkages (using Atalay et al. 2011 IO data)


## Diversion Ratio

To evaluate a merger's anti-competitive potential, the FTCDOJ merger guidelines recommend looking at Diversion Ratios:

Diversion Ratio $_{i j} \stackrel{\text { def }}{=} \frac{\partial q_{i}}{\partial p_{j}}\left(\frac{\partial q_{j}}{\partial p_{j}}\right)^{-1}=\frac{(\mathbf{I}+\boldsymbol{\Sigma})_{i j}^{-1}}{(\mathbf{I}+\boldsymbol{\Sigma})_{j j}^{-1}}$

## M\&A Activity: Diversion Ratios



## Take-aways

- A new GE theory of oligopoly with hedonic demand.
- Estimated for Compustat using 10-K product similarities.
- Distribution of markups is jointly determined by productivity and product market centrality.
- Both have undergone significant changes
- Rising Oligopoly Power
- increasing deadweight loss
- lower consumer surplus share.

I share the data! (elasticities, centrality, productivity...)

## The Great Startup Sellout and the Rise of Oligopoly

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## What is driving the increase in oligopoly?

## VC-backed startup exits (1985-2019)

$\square$ Initial Public Offerings $\square$ Acquisitions


## Entrant Productivity Premium

$$
\mathrm{EPP}_{i}=\frac{2 q_{i}-\sqrt{f_{i}}}{b_{i}-c_{i}-2 q_{i}+\sqrt{f_{i}}} .
$$

## Entrant Productivity Premium



## GAFAM Centrality



## thank you

