

Discussion: Algorithmic Collusion of Pricing and Advertising on E-commerce Platforms (Zhao and Berman)

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Algorithmic Collusion

- ▶ Algorithmic price competition with no coordination
 - ▶ non-trivial: algorithms based on stationary environments
 - ▶ algorithmic competition: environment is endogenous/nonstationary

- ▶ Literature: Supra competitive prices with AI pricing
 - ▶ mechanisms in simulated markets: *facilitate repeated games* (Calvano et al 2020, Kline 2021), *correlated learning* (Hansen et al 2021), *sophistication* (Asker et al 2021), *hub and spoke* (Harrington 2021)
 - ▶ Limited/no theory: Results in the form “showing existence” and mechanisms are inferred

This paper

- ▶ Extends the literature to platforms
 - ▶ Sellers: set prices and bid for location

- ▶ Platform: runs auction to set location
- ▶ Buyers: some type does not search
- ▶ Main results (when search costs are high enough)
 - ▶ Q-learning results in lower advertising bid and lower prices
 - ▶ Does not hurt consumers or the platform

Main Thoughts

- ▶ Show algorithmic collusion results are knife-edge ▶

Does Q-learning represent seller behavior here?

- ▶ Q-learning's has slow convergence rate ($\sqrt{t} \vee \log(t)$)
- ▶ Extending Calvano et al to add bidding increases dimensionality of both the action and state space
- ▶ Q-learning would require ~ 100 millions time-periods
 - ▶ Unrealistic in real environments: requires consumer preferences to be stable over a long time-frame

Suggestions

- ▶ Setup
 - ▶ Search is exogenous (mental costs) – consider choice frictions

- ▶ Outcome of auctions bundled with profits – consider adding separately to the state
- ▶ Amazon data: very reliant on assumptions
 - ▶ estimating search is the key empirical result – consider click-through data
- ▶ Overall: An interesting addition to a growing literature