

OLIGOPOLY DYNAMIC PRICING: A REPEATED GAME WITH INCOMPLETE INFORMATION

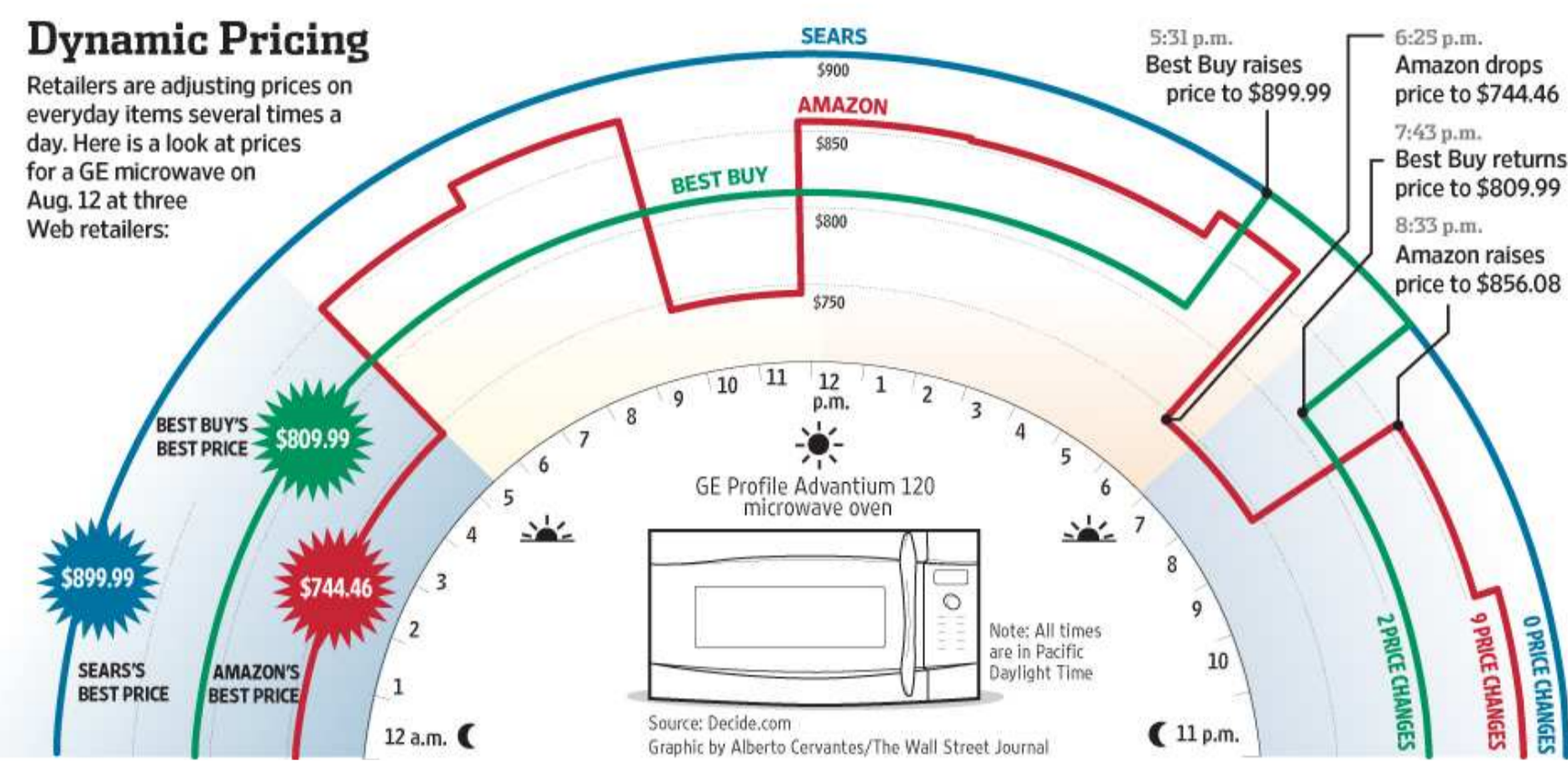
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Oligopoly Dynamic Pricing

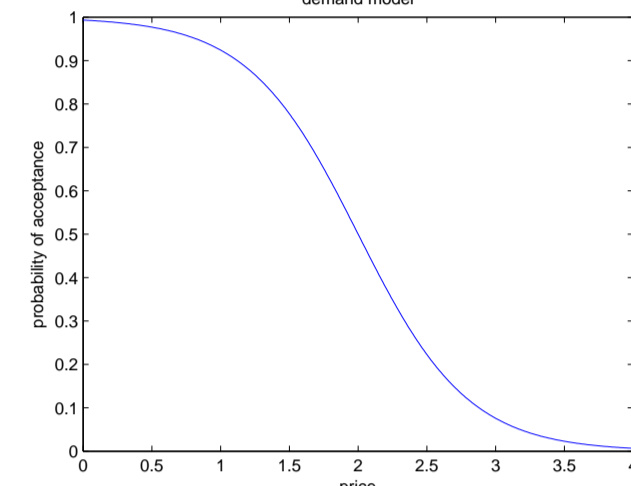
Dynamic Pricing

Retailers are adjusting prices on everyday items several times a day. Here is a look at prices for a GE microwave on Aug. 12 at three Web retailers:



Single Seller Dynamic Pricing

- The *profit-maximization* problem for a seller with an *unlimited* supply of identical goods.
- The seller offers prices sequentially to a stream of potential customers.
- The marginal cost is c .
- For the t -th customer, the seller chooses a price $p(t) \in [c, p_u]$.
- The seller experiences either success or failure.
- The probability of success at price p at any given time is $\rho(p)$.



Single Seller Dynamic Pricing

Unknown Demand Model with Finite Space Uncertainty

- The unknown underlying demand model is $\rho(p)$.
- $\rho(p)$ is unknown but belongs to a known finite set $\{\rho^{(m)}(p)\}_{m=1}^M$.
- $\rho(p)$ is strictly decreasing and satisfies the increasing generalized failure rate (IGFR).
- Intuitively, IGFR means that given a seller can sell a product at price p , the probability of sale at $p + \Delta p$ is decreasing in p .

Profit

- The expected profit: $r^{(\omega)}(p, c) = (p - c)\rho^{(\omega)}(p)$.
- The profit-maximizing price $p^{(\omega)}(c) = \arg \max_{p \in [c, p_u]} r^{(\omega)}(p, c)$.

Lemma 1. Suppose that $\rho(p)$ is strictly decreasing and satisfies IGFR. Then $p^{(\omega)}(c)$ is unique, and $r^{(\omega)}(p, c)$ is continuous and strictly increasing with p over $[c, p^{(\omega)}(c)]$.

Oligopoly Dynamic Pricing

- N sellers.
- The marginal cost for seller i is c_i .
- Without loss of generality, we assume $c_1 < c_2 < \dots < c_N$.
- Unknown demand model $\{\rho^{(m)}(p)\}_{m=1}^M$.
- Sellers propose their prices simultaneously.
- The customer accepts the lowest price p with probability $\rho(p)$.

An Infinitely Repeated Game

Public and private history

- Sellers' price offers are public.
- Each seller's sale history is private.

Strategy of Seller i : A mapping σ from public information and seller i 's private information to a price offer at current time.

Seller i 's one-shot payoff $u_i(p_i)$

- If a single seller i offers the lowest price p_i :
 - $u_i(p_i) = (p_i - c_i)\rho(p_i)$.
 - $u_j(p_j) = 0 \forall j \neq i$.
- If K sellers offer the same lowest price p :
 - $u_i(p) = \frac{1}{K}(p - c_i)\rho(p)$.

Seller i 's payoff for the infinitely repeated game:

$$U_i^{(\omega)}(\sigma) = \liminf_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T u_i^{(\omega)}(a^t(\sigma)).$$

Equilibria and Efficiency

Equilibria

- Nash Equilibrium (NE) σ
 - $U_i(\sigma) \geq U_i(\sigma'_i, \sigma_{-i})$ for every seller i , every demand model $\rho^{(m)}$ and all strategies σ' .
- Subgame perfect equilibrium
 - NE may not be sequentially rational.
 - For every subgame of the original game, the induced continuation strategy is a NE of the subgame.

Efficiency

- Pareto Efficiency:
 - The payoff of any player cannot be increased without reducing the payoff of at least one other player.
- Learning efficiency: Regret
 - Regret is defined as the accumulated profit loss in the unknown demand case as compared to the known demand case.

Dynamic Pricing under Known Demand Model

The colluding strategy σ_C under known demand model:

- Seller 1 forms the optimal collusion of K sellers to maximize its own profit.

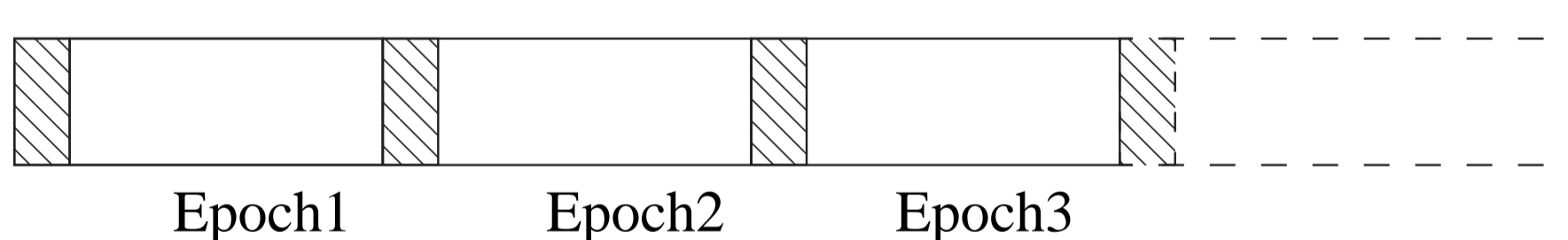
$$K = \arg \max_k \frac{1}{k} r^{(\omega)}(c_{k+1}, c_1)$$

- Sellers with $c_i < c_{K+1}$ offer the collusive price c_{K+1} .
- Sellers with $c_i \geq c_{K+1}$ will not participate.
- Any deviations will trigger a punishment that seller 1 offers $c_2 - \epsilon$ and seller $i \neq 1$ offers c_i forever.

Theorem 1. The colluding strategy σ_C is a *subgame-perfect* and *Pareto-efficient* Nash equilibrium.

Demand Learning under Collusion

- The time horizon is partitioned into *fixed length epochs*.
- Each epoch starts with a *declaration* time slot followed by *cooperation* time slots.



□ Declaration time slots □ Cooperation time slots

Demand Learning under Collusion

Demand Learning under Collusion

In the declaration slot of epoch t :

- Seller 1 carries out a *maximum likelihood estimate* $\hat{\omega}(t)$ of the underlying demand model using its private history.
- Seller 1 then offers the profit-maximizing colluding price $\hat{p}(\hat{\omega}(t))$.
- All other sellers offer the same price they offered in the cooperation slots in the previous epoch.

In each cooperation time slot of epoch t

- Seller i with $c_i < \hat{p}(\hat{\omega}(t))$ offers $\hat{p}(\hat{\omega}(t))$.
- Not participate if $c_i \geq \hat{p}(\hat{\omega}(t))$.

Trigger strategy for punishing any deviations

- Any deviations in declaration slots from seller 2, \dots , N and any deviations in cooperation time slots will trigger a everlasting punishment.
- Punishment is that seller 1 offers $c_2 - \epsilon$ and seller $i \neq 1$ offers c_i forever.

Properties of DLC

Theorem 2.

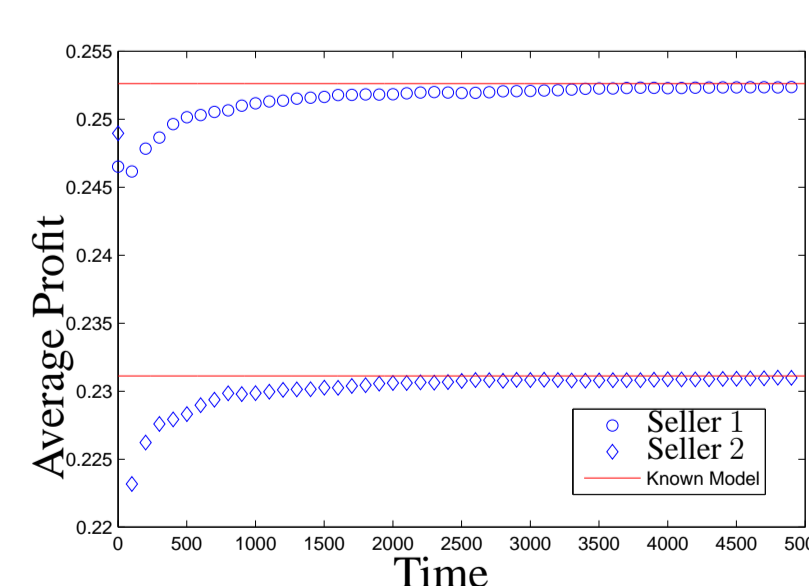
- DLC is a *subgame-perfect* Nash equilibrium.
- DLC is a *Pareto-efficient* Nash equilibrium.
- DLC achieves a *bounded regret*, i.e., under any demand model $\rho^{(\omega)} \in \{\rho^{(\omega)}\}_{\omega=1}^M$, there exists a positive constant C such that

$$R_{DLC} \leq C.$$

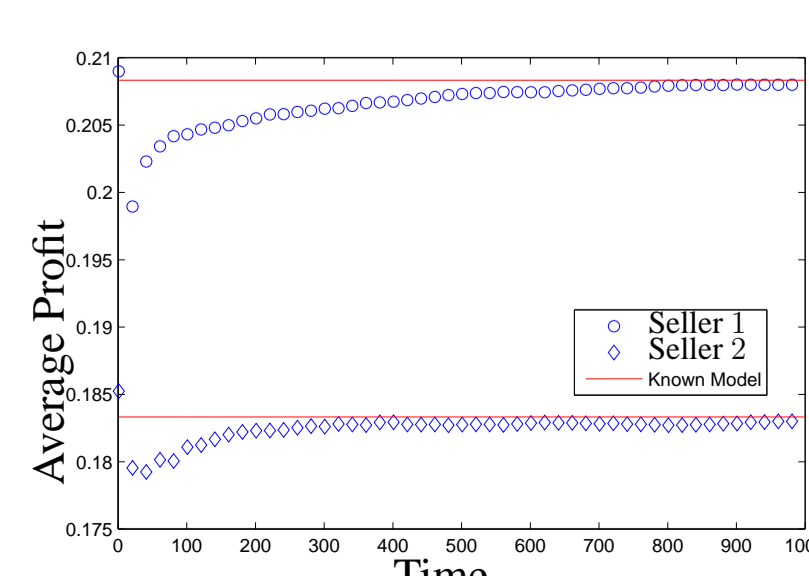
Implications of bounded regret

- The action profiles converges to the optimal action profile played as under the known demand model.

Simulation



Average Profit when demand model is $\rho^{(2)}$ ($c_1 = 0.5, c_2 = 0.6$)



Average Profit when demand model is $\rho^{(1)}$ ($c_1 = 0.5, c_2 = 0.6$)

Conclusion

Oligopoly dynamic pricing

- Infinitely repeated game with private observations.
- Incomplete information: payoff determined by an unknown demand model.
- The optimal collusion with a subset of sellers.

Demand Learning under Collusion (DLC):

- Subgame perfect Nash equilibrium.
- Pareto Efficient.
- Efficient online learning with bounded regret.