

# Dynamic Price Competition: Theory and Evidence from Airline Markets

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**Disclosures:** We thank the anonymous airline for giving us access to the data used in this study. Under the agreement with the authors, the airline had “the right to delete any trade secret, proprietary, or Confidential Information” supplied by the airline. We agreed to take comments in good faith regarding statements that would lead a reader to identify the airline and damage the airline’s reputation. All authors have no material financial relationships with entities related to this research.

## Dynamic pricing is commonly used in markets with fixed inventory and a sales deadline

- ▶ Examples: Airlines, trains, hotels, cruises, entertainment tickets, retailing, etc.
- ▶ Capacity drives price dynamics:
  - ▶ The opportunity cost of selling changes with scarcity
    - Value of a seat today depends on the ability to sell it in the future
    - Excess inventory → expect low prices
  - ▶ Demand may change over time
    - If high WTP consumers arrive in future, incentives to save seats

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  - ▶ E.g., if a firm has excess inventory, it might price *high* (not low) in order to get competitors to sell out early
  - ▶ E.g., fire sales by firm with less inventory to soften future competition
3. Open questions regarding dynamic price competition in perishable goods markets
  - a) Equilibrium prices and profits lack “nice” properties; can we characterize equilibrium outcomes?
  - b) Empirical welfare implications unknown

## Contributions of the research project

1. We introduce a tractable oligopoly framework for dynamic price competition
  - ▶ Provide a differential equation characterization of equilibrium dynamics e.g. Gallego & van Ryzin (1994)
  - ▶ Provide insights on existence, uniqueness, competitive dynamics, e.g., the role of “minimum capacity” see Martinez-de-Albeniz & Talluri (2011) for perfect substitutes
2. We estimate the welfare effects of dynamic pricing in the airline industry
  - ▶ We find the opposite results compared to studies in the single-firm setting: DP increases output and profits, decreases welfare  
single-firm setting: e.g., Hendel and Nevo (2013), Castillo (2021), Williams (2022)
  - ▶ Heuristics similar to airline practices increase surplus relative to DP heuristics differ from, e.g., Calvano et al (2020), Brown and MacKay (2021), Asker et al. (2021)

## Oligopoly model

- ▶ We consider a set  $\mathcal{J} = \{1, \dots, J\}$  of products and a set  $\mathcal{F} := \{1, \dots, F\}$  of firms
- ▶ Firm  $f$  owns products in  $\mathcal{J}_f \subset \mathcal{J}$
- ▶ Initial capacity of each product  $j$  is  $K_{j,0}$
- ▶ Firms must sell all units by time  $T$ , in periods  $t = \Delta, 2\Delta, \dots, T$
- ▶ In every period:
  - ▶ each firm  $f$  sets prices  $\mathbf{p}_t^f := (p_{j,t})_{j \in \mathcal{J}_f}$
  - ▶ a consumer arrives with probability  $\Delta\lambda_t \in (0, 1)$
  - ▶ consumer decides whether to buy a product or not and leave
- ▶ Firms observe history of all prices and inventories

## Demand model

- ▶ Consumers are passive/short-lived  $\rightarrow$  demand function  
(with forward-looking buyers, firm competing with its future self e.g., Board & Skrzypacz (2016); Dilme & Li, (2019); Gershkov, Moldovanu, Strack (2017))

- ▶ A consumer who arrives at time  $t$  chooses product  $j$  with probability:

$$s_{j,t}(\mathbf{p}) = s_j(\mathbf{p}; \boldsymbol{\theta}_t, \mathcal{A}_t) \in [0, 1]$$

where  $\boldsymbol{\theta}_t$  are demand parameters and  $\mathcal{A}_t$  is the set of available products

- ▶ We impose some regularity assumptions that guarantee that the profit-maximizing price vector of a single firm is unique and satisfies the system of FOCs (see paper).
- ▶ Demand assumptions hold beyond IIA, e.g., nested logit



## Solution concept: Markov-perfect equilibrium

- ▶ We analyze Markov-perfect equilibria
- ▶ Payoff-relevant state: vector of inventory  $\mathbf{K} := (K_j)_{j \in \mathcal{J}}$  and time  $t$
- ▶ Denote the Markov pricing strategy for product  $j$  by  $p_{j,t}(\mathbf{K})$

## Let's start with the single-firm case

- ▶ The single-firm case has “nice” properties that mostly do not extend to oligopoly
- ▶ Useful to establish notation and motivates solution strategy
- ▶ Assume that a single firm “M” owns all products
- ▶ “Opportunity cost” or value of a seat  $j$  in state  $(\mathbf{K}, t)$ :

$$\omega_{j,t}(\mathbf{K}) := \underbrace{\Pi_{M,t+\Delta}(\mathbf{K})}_{\text{Continuation Profit}} - \underbrace{\Pi_{M,t+\Delta}(\mathbf{K} - \mathbf{e}_j)}_{\text{Continuation Profit if M sells 1 unit of } j}$$

- ▶ When we study oligopolies we will call these differences in value functions *scarcity effects*

## Optimal control problem

- ▶ The continuation profit of a single firm with capacities  $K_j > 0$  for  $j \in \mathcal{J}$  at  $t$ :

$$\begin{aligned} \Pi_{M,t}(\mathbf{K}; \Delta) = \max_{\mathbf{p}} \quad & \Delta \lambda_t \sum_{j \in \mathcal{J}} \underbrace{s_{j,t}(\mathbf{p})}_{\text{Probability of sale}} \underbrace{(p_j + \Pi_{M,t+\Delta}(\mathbf{K} - \mathbf{e}_j; \Delta))}_{\text{Continuation value after sale}} + \\ & \underbrace{(1 - \Delta \lambda_t \sum_{j \in \mathcal{J}} s_{j,t}(\mathbf{p}))}_{\text{Probability of no sale}} \underbrace{\Pi_{M,t+\Delta}(\mathbf{K}; \Delta)}_{\text{Continuation value after no sale}} \end{aligned}$$

- ▶  $\Pi_{M,t}(K; \Delta)$  converges as  $\Delta \rightarrow 0$  uniformly in  $t$  and  $\mathbf{K}$  to  $\Pi_{M,t}(\mathbf{K})$  which solves the differential equation

$$\begin{aligned} \dot{\Pi}_{M,t}(\mathbf{K}) = -\lambda_t \max_{\mathbf{p}} \sum_{j \in \mathcal{J}} s_{j,t}(\mathbf{p}) \underbrace{(p_j - (\Pi_{M,t}(\mathbf{K}) - \Pi_{M,t}(\mathbf{K} - \mathbf{e}_j)))}_{=:\omega_{j,t}(\mathbf{K})} \end{aligned}$$

## Properties of the single-firm case

### Proposition 1

1. Value function  $\Pi_{M,t}(\mathbf{K})$  is decreasing in time  $t$  and increasing in capacity
  2. Opportunity costs  $\omega_{j,t}(\mathbf{K})$  are decreasing in time  $t$  and capacity
  3. The stochastic process  $\omega_{j,t \wedge \tau}(\mathbf{K}_t)$ ,  $\tau := \inf\{t \geq 0 | K_{j,t} \leq 1\}$  is a submartingale
- ▶ Insight: Observed average price (conditional on  $K_j > 1$ ) is increasing  $\rightarrow$  demand uncertainty leads to increasing average prices
  - ▶ None of these properties carry over to the oligopoly case!

Now, we consider the duopoly game. A new scarcity force.

- ▶ Back to a duopoly where each firm owns one product:  $\mathcal{J} = \mathcal{F} = \{1, 2\}$
- ▶ Each firm  $f$  has its own continuation profit in state  $(\mathbf{K}, t)$ :  $\Pi_{f,t}(\mathbf{K}; \Delta)$

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- ▶ Now, there are two scarcity effects for each firm  $f$ :

- ▶ “Own-scarcity effect”

$$\omega_{f,t}^f(\mathbf{K}) := \Pi_{f,t+\Delta}(\mathbf{K}) - \Pi_{f,t+\Delta}(\mathbf{K} - \mathbf{e}_f)$$

- ▶ “Competitor-scarcity effect”

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- ▶ This defines a matrix of scarcity effects:

$$\Omega_t(\mathbf{K}) = \begin{pmatrix} \omega_{1,t}^1 & \omega_{2,t}^1 \\ \omega_{1,t}^2 & \omega_{2,t}^2 \end{pmatrix}.$$

# Differential equation characterization of equilibrium

## Proposition 2 (Continuous-time limit)

Assume sufficient conditions on demand system. For every  $\mathbf{K}$ , there exists a  $T_0(\mathbf{K}) > 0$ , non-increasing in  $\mathbf{K}$ , so that the value function  $\Pi_{f,t}(\mathbf{K}; \Delta)$  converges to a limit  $\Pi_{f,t}(\mathbf{K})$  as  $\Delta \rightarrow 0$  that solves the differential equation

$$\dot{\Pi}_{f,t}(\mathbf{K}) = -\lambda_t \left( s_f(\mathbf{p}^*(\Omega_t(\mathbf{K}); \theta_t)) (p_f^*(\Omega_t(\mathbf{K}); \theta_t) - \underbrace{(\Pi_{f,t}(\mathbf{K}) - \Pi_{f,t}(\mathbf{K} - \mathbf{e}_j))}_{\text{own-scarcity effect}}) \right. \\ \left. - s_{-f}(\mathbf{p}^*(\Omega_t(\mathbf{K}); \theta_t)) \underbrace{(\Pi_{f,t}(\mathbf{K}) - \Pi_{f,t}(\mathbf{K} - \mathbf{e}_{-f}))}_{\text{competitor-scarcity effect}} \right)$$

where  $f' \neq f$ , with natural boundary conditions and  $p^*(\Omega, \theta)$  is an equilibrium of a stage game parameterized by  $(\Omega; \theta)$ .

Allows us to empirically investigate DPs in oligopoly with large state spaces



## The value function

- ▶ The Markov structure allows us to summarize the impact of today's price on future revenues into "scarcity effects."
- ▶ Given a pricing strategy  $\mathbf{p}_t(\mathbf{K}) := (p_{1,t}(\mathbf{K}), p_{2,t}(\mathbf{K}))$ , firm  $f$ 's value function is

$$\Pi_{f,t}(\mathbf{K}; \Delta) = \Delta \lambda_t \left( \underbrace{s_{f,t}(\mathbf{p}_t(\mathbf{K})) \left( p_{f,t}(\mathbf{K}) + \Pi_{f,t+\Delta}(\mathbf{K} - \mathbf{e}_f; \Delta) \right)}_{\text{revenue of own sale}} \right) +$$

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where  $f \neq f'$ .

## The stage game with equilibrium prices $\mathbf{p}^*(\Omega_t(\mathbf{k}); \theta_t)$

- ▶ We can write for each firm  $f \neq f'$

$$\begin{aligned} & \Pi_{f,t+\Delta}(\mathbf{K}; \Delta) - \Pi_{f,t}(\mathbf{K}; \Delta) = \\ & -\Delta\lambda_t \left( \underbrace{s_{f,t}(\mathbf{p}_t(\mathbf{K})) \left( p_{f,t}(\mathbf{K}) - \omega_{f,t}^f(\mathbf{K}) \right) - s_{f',t}(\mathbf{p}_t(\mathbf{K})) \omega_{f',t}^f(\mathbf{K})}_{\text{stage game payoff of firm } f} \right). \end{aligned}$$

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- ▶ This stage game can have multiple equilibria if e.g.  $\omega_{f',t}^f$  is very negative
- ▶ Typically  $\omega_{f',t}^f$  is negative because sale of competitor creates future scarcity

## Findings from simulations of this system of differential equations

- ▶ Profits are non-monotonic in the own capacity
- ▶ Profits are non-monotonic in competitor capacity
- ▶ Profits are neither concave nor convex in capacity: Both scarcity effects can be positive or negative
- ▶ But the dynamics of scarcity effects close to the deadline depends on which firm has the minimum capacity:
  - ▶ competition fiercest when firms have symmetric inventory (independent of symmetry in other dimensions)
  - ▶ largest price effects when the firm with min cap sells
- ▶ see paper for new markup rule

## Data Overview

- ▶ Use third-party data provided to us by a large US airline:
  - ▶ Daily prices and quantities of *competing airlines*
  - ▶ Observe all bookings (counts); including connecting traffic, tickets purchased via travel agents, etc.
  - ▶ Daily prices for each flight—we'll use the lowest available economy ticket
  - ▶ Data identifies firms, flight numbers, departure dates, etc.
- ▶ Think of this as the Nielsen of airline data
- ▶ We observe search data, the pricing technology, and all output from the firm's pricing system for one airline
  - ▶ Use search data to estimate arrivals, which are then scaled up (robust to scaling factor)

## Facts on Routes Studied

- ▶ Analysis concentrates on duopoly markets with a large % of local, nonstop traffic
- ▶ Distribution of fares similar to all duopoly markets
- ▶ Many markets are from large cities to medium-size cities; 58 directional pairs total

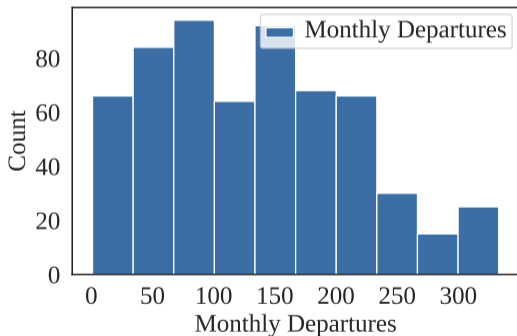


Figure: A few frequencies a day

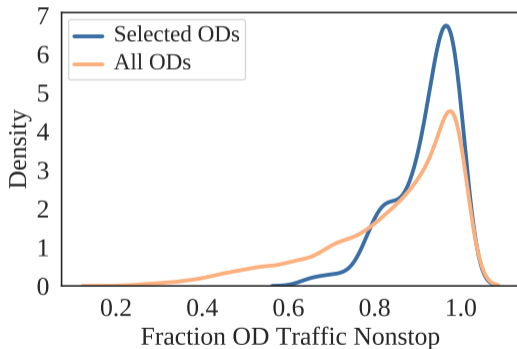


Figure: Non-stop traffic distribution

# Summary Statistics—Dynamics

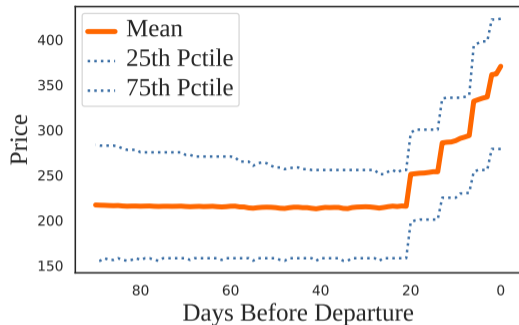


Figure: Prices over Time

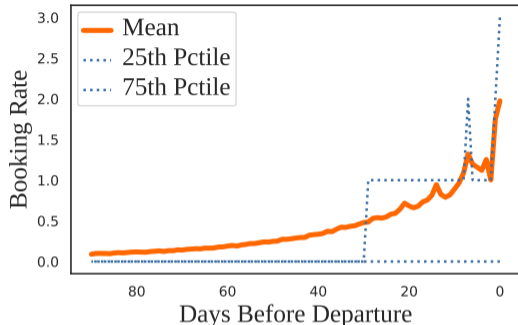


Figure: Booking Rate over Time

- ▶ Distribution of fares follow step pattern—AP discounts substantially increase fares
- ▶ Booking rates increase, due to both more arrivals (partially observed) and (we will find) higher WTP

## Average outcomes across competitors

- ▶ No competitor sells consistently a larger fraction of its seats
- ▶ Price differences across carriers are small, but one carrier charges relatively lower prices earlier on and higher prices later on (on average)

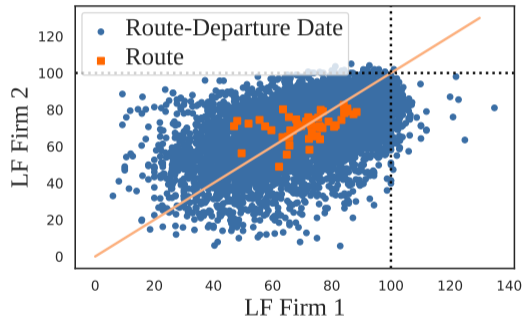


Figure: Average load factors for carriers in duopoly markets

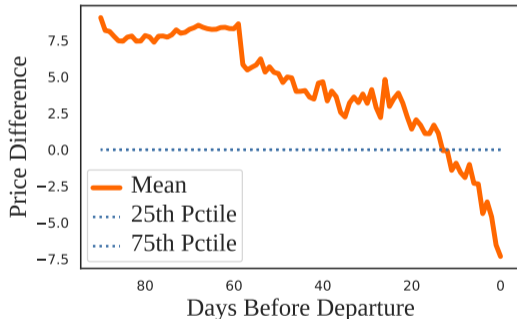


Figure: Difference in prices for markets in which each firm offers exactly one flight



## Empirical Model of Demand—Nested Logit with 2 Nests

- ▶ Let  $j$  be a carrier-flight,  $d$  a departure date,  $t$  is day before departure,  $r$  a route
- ▶ Conditional on arrival, we specify consumer utilities as

$$u_{i,j,t,d,r} = \mathbf{x}_{j,t,d,r}\boldsymbol{\beta} - \alpha_t p_{j,t,d,r} + \zeta_{i,J} + (1 - \sigma)\varepsilon_{i,j,t,d,r},$$

where

- ▶  $\zeta_{i,J} + (1 - \sigma)\varepsilon_{i,j,t,d,r}$  follows a type-1 extreme value distribution, and  $\zeta_{i,J}$  is an idiosyncratic preference for the inside goods;
- ▶ We allow price sensitivity parameters  $\alpha_t$  to vary with time
- ▶ Nesting parameter  $\sigma$  captures flight substitutability
- ▶ Each arriving consumer solves their utility maximization problem such that consumer  $i$  chooses flight  $j$  if and only if  $u_{i,j,t,d,r} \geq u_{i,j',d,t,r}, \forall j' \in \mathcal{J}_{t,d,r} \cup \{0\}$ .
- ▶ Estimates robust to adding an unobservable  $\xi$ , estimated with control function

## Empirical Model of Demand—Poisson Arrival

- ▶ We assume daily arrivals are distributed Poisson, with rates  $\lambda_{t,d,r}$  equal to

$$\lambda_{t,d,r} = \exp \left( \tau_r^{\text{OD}} + \tau_d^{\text{DD}} + \tau_{t,d}^{\text{SD}} + f(\text{DFD})_t \right),$$

where  $f(\cdot)$  is a polynomial expansion

- ▶ Therefore,  $q_{j,t,d,r} = \min\{C_{j,t,d,r}, \lambda_{t,d,r} \cdot s_{j,t,d,r}(p; \theta)\}$ , which is censored Poisson
- ▶ We scale up arrivals using a factor (1-3.5) to account for unobserved searches, after accounting for the percentage of direct bookings/searches for a single carrier

## Demand Estimates Over Time

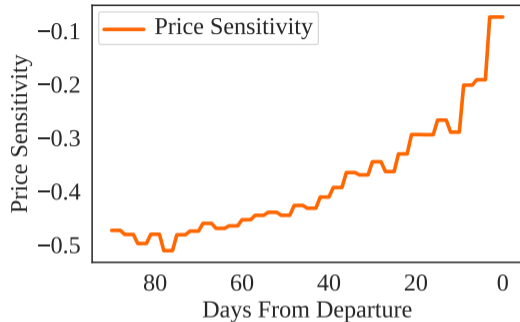


Figure: Price Sensitivity Parameters

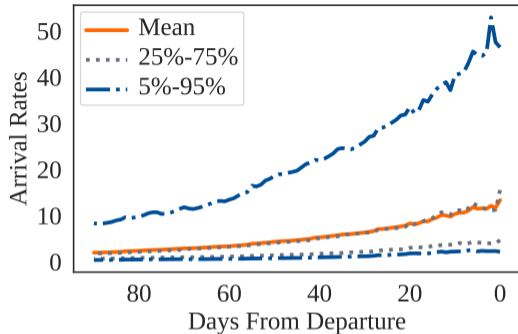


Figure: Arrival Rates

- ▶ Estimate nesting parameter = 0.5; avg. elasticity of -1.438
- ▶ Both the number of arriving customers and the average price sensitivity are increasing towards the deadline

## Counterfactual Analysis: Dynamic Pricing

	Price	Firm 1 Rev.	Firm 2 Rev.	CS	Welfare	Q	LF	Sellouts
Benchmark	226.3	5571.5	5759.4	16698.2	28029.0	20.0	70.6	9.2
Uniform	250.8	4629.6	4925.7	19042.4	28597.6	19.2	69.7	7.9
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4. Fewer units are sold with uniform pricing, and there are fewer sell outs



## Counterfactual Results: Dynamic Pricing

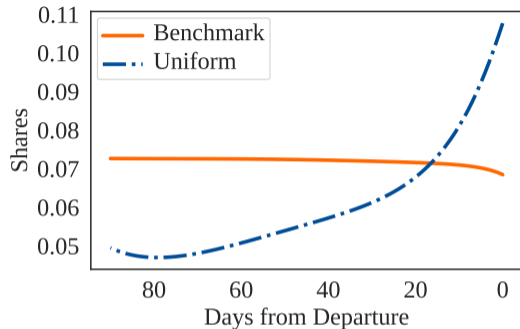


Figure: Shares over Time

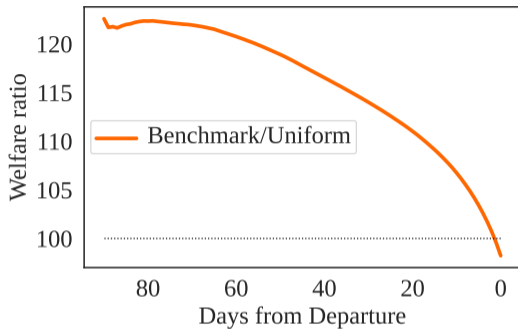


Figure: Cumulative Welfare Comparison

- ▶ Early-arriving customers pay a higher price with uniform pricing, late-arriving customers a higher price.
- ▶ The firm keeps inefficiently few seats for late-arriving customers under DP to soften competition close to the deadline (**competitor-scarcity effect**)

## Counterfactual Results: Heuristics

### 1. Lagged-price model

- ▶ Firm assumes last observed price will continue until deadline

### 2. Deterministic model

- ▶ Firms believe competitors will follow a fixed price path according to the minimum filed price

	Price	Firm 1 Rev.	Firm 2 Rev.	CS	Welfare	Q	LF	Sellouts
Benchmark	100.0	100.0	100.0	100.0	100.0	100.0	100.0	100.0
Lagged	104.6	104.1	105.3	103.3	103.9	100.0	100.1	101.0
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## Conclusion

- ▶ We introduce a framework to study dynamic price competition in perishable goods markets
- ▶ We show that competitor scarcity is a key driver of price dynamics and captures the incentive to soften competition in the future
- ▶ We apply our framework to airlines and find that DP expands output but decreases welfare in the routes studied
- ▶ Open questions remain regarding the use of dynamic versioning, loyalty, and the influence of forward-looking buyers