

Consumer-Optimal Segmentation in Multi-Product Markets

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January 5, 2024

- Second- and third-degree price discrimination (quality differentiation and market segmentation, respectively) may either benefit or harm consumers
- Most research analyzes these practices in isolation
- Increasingly, these tools are deployed simultaneously, particularly in the data-rich digital economy
 - Student versions of digital products are typically inferior, while enterprise versions have more features
 - Streaming services offer differentiated subscription tiers
- Open question of how these two forms of price discrimination interact and impact consumers

- Comprehensive analysis of consumer-optimal segmentation with a monopolist that offers quality-differentiated products

Main results:

- ① Completely characterize the value of segmentation
- ② Consumption is monotone in valuation across segments (and within segments, by incentive compatibility)
- ③ Seller offers “essentially” same quality to identical consumers
- ④ Every segmentation hurts consumer surplus when aggregate demand elasticity is sufficiently high

- Present our model of second- and third-degree price discrimination
- Work through simple example (binary value setting) to introduce concepts used in analysis
- Formally state our main result: value of segmentation

① Model

② A Simple Example

③ General Setting

- There is a monopolist and a continuum of consumers
- Produces a vertically differentiated good with quality $q \in \mathbb{R}_+$
- Cost of producing good of quality q is given by increasing and convex function $c : \mathbb{R}_+ \rightarrow \mathbb{R}_+$
- Each consumer has a privately known willingness to pay v drawn from a finite set $V = \{v_1, \dots, v_K\}$
- Consumer's gross utility is $v \cdot q$

- A *market* $x \in \Delta V$ is a distribution over V , where $x_k \triangleq x(v_k)$ is the probability of value v_k
- Given a *menu* of prices $p(q)$, consumers purchase quality which maximizes their utility:

$$q(v, p) \triangleq \arg \max_q [v \cdot q - p(q)], \quad U(v, p) \triangleq \max_q [v \cdot q - p(q)]$$

- Profits are

$$\Pi(v, p) \triangleq p(q(v)) - c(q(v))$$

- The profit-maximizing menu is denoted p^x :

$$p^x \triangleq \arg \max_{p(q)} \sum_{k=1}^K x_k \Pi(v_k, p)$$

- This gives us the profit and consumer surplus in a market:

$$U(x) \triangleq \sum_{k=1}^K x_k U(v_k, p^x), \quad \Pi(x) = \sum_{k=1}^K x_k \Pi(v_k, p^x)$$

Segmentations

- Fix some *aggregate market* $x^* \in \Delta V$
- A *segmentation* σ of x^* is a distribution over ΔV such that

$$\sum_{x \in \Delta V} \sigma(x)x = x^*$$

- Given a segmentation σ of x^* , the seller optimally sets prices p^x for every market x in the support of σ
- We focus on the *consumer-optimal segmentation*, which solves

$$\max_{\sigma \in \Delta(\Delta V)} \left[\sum_x \sigma(x)U(x) \right] \text{ s.t. } \sum_x \sigma(x)x = x^*$$

- This is a Bayesian Persuasion (BP) problem over K states

① Model

② A Simple Example

③ General Setting

- Assume $V = \{v_L, v_H\}$ with $x^* = (x_L^*, x_H^*)$
- Define the *hazard rate*

$$h^x \triangleq (v_H - v_L) \frac{1 - x_L}{x_L}, \quad h^* \triangleq (v_H - v_L) \frac{1 - x_L^*}{x_L^*}$$

- Denote by Q the “supply function”:

$$Q(v) = (c')^{-1}(v)$$

- Applying standard results (Mussa and Rosen (1978)), total consumer surplus is equal to

$$U(x) = x_H(v_H - v_L)Q(v_L - h^x)$$

- Given a segmentation σ , define the following weights:

$$\lambda^x \triangleq \frac{\sigma(x)x_L}{x_L^*}$$

- λ^x defines a distribution over hazard rates

$$\sum_x \lambda^x = 1$$

which average to the *aggregate hazard rate*

$$\sum_x \lambda^x h^x = h^*$$

- Next, define the *local information rent*

$$u_L(h) \triangleq h \cdot Q(v_L - h)$$

- We can rewrite

$$U(x) = x_L u_L(h^x) \implies \sum_x \sigma(x) U(x) = x_L^* \sum_x \lambda^x u_L(h^x)$$

- Consumer surplus is the expected value of the local information rent over the distribution λ^x

- Consider the Bayesian persuasion problem where we directly choose a distribution over hazard rates:

$$\max_{\lambda \in \Delta \mathbb{R}_+} \left[x_L^* \sum_h \lambda(h) u_L(h) \right] \quad \text{s.t.} \quad \sum_h \lambda(h) h = h^* \quad (1)$$

- Kamenica and Gentzkow (2011): let \bar{u}_L be the concave upper envelope of u , then the solution to (1) is $x_L^* \bar{u}_L(h^*)$
- Already showed: every segmentation σ is feasible in (1)
- Can every feasible λ be generated by a segmentation?

Proposition

The consumer-optimal segmentation achieves the concavification bound (1) with equality:

$$\max_{\sigma} \left[\sum_x \sigma(x) U(x) \right] = x_L^* \bar{u}_L(h^*).$$

- Segmentation increases consumer surplus if and only if

$$\bar{u}_L(h^*) > u_L(h^*)$$

- We can read off the (unique) consumer-optimal segmentation from the solution to (1)

Example: Isoelastic Costs I

- Suppose that

$$c(q) = \frac{q^\gamma}{\gamma}, \quad \gamma > 1$$

- u_L and \bar{u}_L plotted below

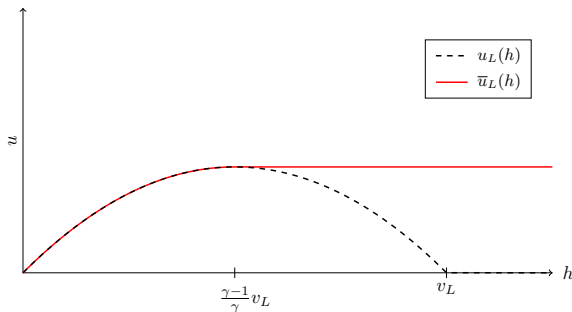


Figure 1

- Segmentation improves consumer surplus if and only if

$$h^* > \frac{\gamma - 1}{\gamma} v_L \iff \eta^* < \frac{\gamma}{\gamma - 1}$$

- If the optimal segmentation is non-trivial, v_L consumes quality

$$q(v_L) = \left(\frac{v_L}{\gamma} \right)^{\frac{1}{\gamma-1}}$$

① Model

② A Simple Example

③ General Setting

- Denote by D^x the demand associated with market x :

$$D_k^x \triangleq D^x(v_k) = \sum_{i \geq k} x_i$$

- Generalize objects from before to higher dimensions:

$$h_k^x \triangleq (v_{k+1} - v_k) \frac{D_{k+1}^x}{x_k}, \quad u_k(h) = h \cdot Q(v_k - h)$$

- In a regular market,¹

$$U(x) = \sum_{k=1}^{K-1} x_k u_k(h_k^x) \tag{2}$$

- Repeat the analysis of the two-value case $K - 1$ times, but now we need an aggregate constraint

¹It is without loss to restrict to distributions supported on regular markets.

- Let $D \prec D^*$ denote the majorization constraint ► Definition
- Define the hazard rate associated with D :

$$h_k^D \triangleq (v_{k+1} - v_k) \frac{D_{k+1}}{x_k^*}$$

Theorem

The consumer-optimal σ attains the concavification bound:

$$\max_{\sigma} \left[\sum_x \sigma(x) U(x) \right] = \max_{D \prec D^*} \sum_{k=1}^{K-1} x_k^* \bar{u}_k(h_k^D) \quad (3)$$

Theorem

The consumer-optimal σ attains the concavification bound:

$$\max_{\sigma} \left[\sum_x \sigma(x) U(x) \right] = \max_{D \prec D^*} \sum_{k=1}^{K-1} x^*(v_k) \bar{u}_k(h_k^D)$$

- Original problem is a maximization over $\Delta(\Delta V)$; this is constrained maximization over \mathbb{R}_+^K (c.f. Kleiner et al. (2021))
- In paper: frequently, $D = D^*$ is optimal

► Proof Idea

- Across segments, quality consumed across types is monotone, and we can bound quality dispersion within types
- Qualitative description of optimal segmentation under a broad class of cost functions and aggregate markets
 - Low types consume higher quality than under no segmentation
 - Segmentation has no benefit when demand elasticities in aggregate market are sufficiently high
- Extend methodology to Pareto frontier of surplus divisions

Market segmentation and consumer surplus: Pigou (1920); Robinson (1969); Schmalensee (1981); Varian (1985); Aguirre et al. (2010); Cowan (2016)

Second degree price discrimination: Mussa and Rosen (1978); Maskin and Riley (1984); Johnson and Myatt (2003)

Single good benchmark: Bergemann et al. (2015)

Segmentation with multi-product monopolist: Haghpannah and Siegel (2022, 2023); today's results are orthogonal

- Characterize how market segmentation affects consumer surplus when monopolist can engage in quality differentiation
- Suggests that, for competition authorities, segmentation should be evaluated based on market characteristics
- Further work may be able to develop empirical tests

Thank you!

References I

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- We say that $D \in \mathbb{R}_+^K$ is *majorized* by D^* , $D \prec D^*$, if

$$\sum_{i=k}^{K-1} (v_{i+1} - v_i) D_{i+1} \leq \sum_{i=k}^{K-1} (v_{i+1} - v_i) D_{i+1}^*$$

for all $k \in \{1, \dots, K-1\}$

- D is a “quasi-market”: may be increasing at points
- If D were a market, equivalent to MPS of D^*

- Majorization constraint arises from analyzing the way support gaps affect the average hazard rate across markets
- From this, upper bound is straightforward as a double relaxation: u to \bar{u} , and D^* to $D \prec D^*$
- Showing that the upper bound is tight is convoluted; proof explicitly constructs a segmentation achieving it