Consumer Data and Consumer Welfare: Evidence from the Hotel Booking Market.

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Introduction			
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Motivation.

- Platforms like Amazon, Google, Expedia, etc., accumulate consumers' data.
- They claim to use this data to provide better services via better product targeting.
- It raises economic questions.
 - How does it change competition between firms that advertise on platforms?
 - How does consumer welfare change?
- Lack of empirical study on firms' response to better targeting and its effect on consumer welfare.

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(a) Logged-in to a personal account



(b) Logged-out + anonymized

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E'RE OPEN	Ink 48 Hotel Manhates, New Yok, Severan see Plana Genera Manhath Thro Genera Bed	Very Good 468 reviews 1 night, 1 adult \$179 Select your room
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Introduction			
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Main Idea			

- Compare market outcomes in two situations.
 - Common ranking.

The platform uses market-aggregated data only. Same ranking for all consumers.

Personal ranking.

The platform uses consumer's personal data. Consumer-specific ranking of products.

- The key component is consumers incur the search costs to learn product quality.
- With search, consumer's consideration set and, hence, demand, depend on ranking.
- Demand change affects firms' pricing.

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Trade-off and Contribution

Trade-off:

- Targeting reduces search expenditures and provides better product match.
- Targeting segments the market thereby softens price competition.

Contribution:

- First empirical paper studying the equilibrium model.
 - Analyze firms' price response to better targeting.
- The personalized ranking is harmful to consumers due to price increase.

Data			
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Data structure.

- Expedia hotel search and booking data. 200,000 consumers. Nov 2012 June 2013.
- Data for each consumer query:
 - An ordered list of hotels shown to the consumer with their characteristics.
 - What hotels clicked and what, if any, booked.
 - Query details (stay length, # of rooms, # of adults, etc.).

	Approach •		

Approach

- Sequential search model for consumer demand. Estimated with a click and purchase data.
- Using demand estimation, calculate hotels' costs.
- Permutation of hotels' positions \rightarrow demand change \rightarrow prices change \rightarrow consumer welfare change.

	Demand ●○○		

Optimal consumer search.

- Weitzman (1979) optimal sequential search model. Utility *u*_{ij}, Reservation Utility *r*_{ij}.
- · Search and purchase decisions are determined by

$$u_{ij} = \delta_{ij} + \epsilon_{ij}, \quad r_{ij} = \delta_{ij} + H_0^{-1}(s_{n_{ij}})$$

• Utility observed prior to search is modeled as:

$$\delta_{ij} = \alpha_i \boldsymbol{p}_j + \boldsymbol{\beta_i}^T \boldsymbol{x_j} + \boldsymbol{\xi_j}$$

- Ursu (2018) showed that position affects search costs only. Ursu (2018) details
- Search cost $s_{n_{ij}} \sim F^{s_{n_{ij}}}(s)$ with mean $\mu_{n_{ij}} = log(1 + exp(\gamma \cdot n_{ij}))$
- Consumer problem is dynamic \rightarrow complicated demand system.

	Demand O●O		

Consumer choice.

Choi, Dai, Kim (2018, ECMA) showed that sequential consumer search model can be translated into a familiar discrete-choice model.

 $w_{ij} \stackrel{\text{def}}{=} \min\{u_{ij}, r_{ij}\}$

Proposition

The consumer i purchases product j if and only if $w_{ij} \ge \max_{\substack{k \neq j \\ k \neq j}} \{w_k\}$

Intuituion

If u_{ij} or r_{ij} are low, the hotel is never booked.

- If r_{ij}, the hotel is not explored.
- If u_{ij}, the hotel is not booked even if explored.

			Demand			
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Consumer demand.

	Mean Utility Parameters	Utility Parameters Standard Deviations
Utility		
Constant	4.1167***	6.2731***
	(0.1742)	(1.152)
Price (\$100)	-2.4881***	0.6499***
	(0.1203)	(0.1259)
Star rating	1.2369***	0.0166
	(0.0705)	(0.1703)
Review score	0.1118	0.0916
	(0.0895)	(0.171)
Location score	0.0737	0.9509***
	(0.1006)	(0.0726)
Chain dummy	0.7346***	1.6514***
	(0.2656)	(0.4475)
Search cost		
Position (γ)	0.0625***	_
	(0.0032)	-

Note: Stars indicate estimates significant at the 99% level.

Monte Carlo simulation

Consumer Data and Consumer Welfare: Evidence from the Hotel Booking Market.

		Supply •	

Supply side.

Knowing the distribution of consumers' demand elasticity and search costs, find hotels' opportunity costs

$$egin{aligned} p_j^* &= rgmax_{p_j} \left[\left(p_j - c_j
ight) D_j(p_j)
ight] \ c_j &= p_j^* + rac{D_j(p_j^*)}{rac{\partial D_j(p_j)}{\partial p_j}}
ight|_{p_j = p_j^*} \end{aligned}$$

The effect of price on demand

- Effect on consideration set
 - Position in ranking
 - Reservation utility *r_{ij}*

• Effect on purchase probability conditional on consideration set

• Consumer utility *u*_{ij}

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		Counterfactuals	
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Counterfactuals.

- Compare consumer welfare under common and personal rankings.
- Algorithm:
 - Oddel hotels prices under common and personal rankings, using consumer demand estimation and hotels' cost.
 - ② Draw consumers from consumer distribution and simulate their decisions under common and personal rankings.
 - Ompare results.

					Counterfactuals	
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Effect on consumer welfare.

Consumers in average experience 3.1% (\$4.2) direct utility reduction.

- Gain 0.5% (\$0.7) due to finding a better-fitted hotel.
- Loss 3.6% (\$4.9) due to increased transaction prices.

(a) Consumers utility distribution. Personal vs Common rankings





		Counterfactuals	
		000	

Effect on consumer welfare.

Consumers in average experience .8% (\$1.1) gain due to a reduction in search intensity.

Figure: Positions of booked hotels. Personal vs Common rankings



			Conclusion
			•

Concluding remark

- First empirical paper studying the effect of consumer-personalized products ranking in the equilibrium search model.
- The average consumer welfare reduces under the personal ranking in contrast with previous researches. The effect is heterogeneous over consumers. Some of the consumers experience welfare gain.

Data.

	Observations	Mean	Median	SD	Min	Max
Query level						
Number of hotels displayed	206,657	27.12	31.00	8.10	5	38
Trip length (days)	206,657	2.42	2.00	1.98	1	40
Booking window (days)	206,657	39.26	18.00	53.89	0	498
Saturday night (percentage)	206,657	0.50	1.00	0.50	0	1
Adults	206,657	2.00	2.00	0.90	1	9
Children	206,657	0.39	0.00	0.79	0	9
Rooms	206,657	1.12	1.00	0.44	1	8
Total clicks	206,657	1.12	1.00	0.61	1	25
Two or more clicks (percentage)	206,657	0.07	0.00	0.25	0	1
Transaction	206,657	0.66	1.00	0.48	0	1
Random ranking (percentage)	206,657	0.31	0.00	0.46	0	1
Hotel level						
Price	5,511,851	156.49	129.00	101.28	10	1000
Stars	5,383,647	3.31	3.00	0.88	1	5
Review Score	5,505,786	3.86	4.00	0.91	0	5
Chain	5,511,851	0.65	1.00	0.48	0	1
Location Score	5,511,851	3.09	3.00	1.52	0	7
Promotion	5,511,851	0.24	0.00	0.43	0	1

Appendix: Monte Carlo simulations.

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1,000 consumers, each searching among 30 hotels. 10,000 draws from $F^{\theta}(\theta)$.

	True values	Estimated values
Utility		
Price	-1	-0.9608
		(0.0525)
Price heterogeneity	0.3	0.2849
		(0.0430)
Quality	2	1.8941
		(0.0883)
Quality heterogeneity	0.6	0.5335
		(0.0542)
Search cost		
Position	0.1	0.0856
		(0.0065)



Effect of ranking on clicks and sales.



(a) Expedia ranking

(b) Random ranking



Appendix: Ursu (2018) details.

$$u_{ij} = \delta_{ij} + \epsilon_{ij}, \quad r_{ij} = \delta_{ij} + H_0^{-1}(s_{ij})$$

Position affects	Prediction	Result
δ	Purchase conditional on search is affected by position	Ruled out
ϵ	Reservation utility is increasing in position	Ruled out
S	Opaque offers increase search costs	Ruled in
δ,ϵ	Purchase conditional on search is affected by position	Ruled out
$\delta, {\it s}$	Purchase conditional on search is affected by position	Ruled out
$\epsilon, {m s}$	Reservation utility is increasing in position	Ruled out
$\delta,\epsilon,{m s}$	Purchase conditional on search is affected by position	Ruled out

Back

Supply side.

Knowing the distribution of consumers' demand elasticity and search costs, find hotels' opportunity costs

$$\Pi_{jt}(p_{jt}) = \left((1-f_j)p_{jt}-c_{jt}\right)D_j(p_{jt})$$

$$\frac{c_{jt}}{1-f_j} = p_{jt} - \frac{D_j(p_{jt})}{\frac{\partial D_j(p_{jt})}{\partial p_{jt}}}$$

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Hotels' opportunity costs histogram.

- The opportunity cost is different from the marginal cost of serving the room.
- The opportunity cost might be negative if, for example, the hotel expects an increase in competition in the future.
- The opportunity cost captures the dynamic aspect of optimal pricing decision.



Effect on prices is heterogeneous.

• The average transaction price increased.



Appendix 0000000000

Effect on prices is heterogeneous by position.

• Less prominent hotels have a lower incentive to decrease the price.



Effect on prices is heterogeneous by position.

(a) Positions of hotels that charges lower prices under the personal ranking



(b) Positions of hotels that charges higher prices under the personal ranking



Appendix 0000000000

Price change by the measure of market vertical differentiation.

 As the level of products vertical differentiation increases, the advertising effect becomes less important since consumers have stronger preferences toward one of the products, which leads to higher price difference.

(a) Position #1

(b) Position #15

