D I S C O N T I N U O U S DEMAND FUNCTIONS: ESTIMATION AND PRICING

Arnoud V. den Boer University of Amsterdam N. Bora Keskin Duke University

Rotterdam May 24, 2018 Dynamic pricing and learning:

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- Cont. armed MAB, observing demand d(p) and reward $r(p) = p \cdot d(p)$

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- Learning optimal selling price from accumulating sales data
- Cont. armed MAB, observing demand d(p) and reward $r(p) = p \cdot d(p)$
- Standard assumption: $d(\cdot)$ is continuous

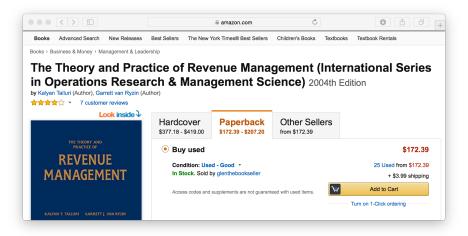




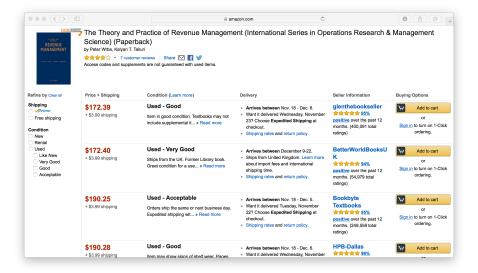
Nous admettons que la fonction F(p) qui exprime la loi de la demande ou du débit est une fonction **continue**...

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- Rankings in online marketplaces (e.g. Amazon's BuyBox)
- Product search with price thresholds

Motivation

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\$700 - \$1,500 Over \$1,500 \$		Amkov 24MP 1080P 3.0 Hd Screen Digital Camera With Shooting 4X Zoom W/ Camcord 54.88 from eBay - melfff Amkov 24MP (2009 30*HD Screen Digital Camera With Shooting 4X Zoom w/ Camcord Store category Sign Up Now I You may also like 24 MP - CMOS - With Video - With Image Stabilization				

Motivation

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Cameras & Optics	Set Price Alert Similar Products					
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 Digital Point & Shoot Cameras 	Set Price Alert Similar Products					
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Camera & Optic Accessories	Set Price Alert Similar Products	122 Seller Reviews				
Electronics Accessories	Nikon DK-5 Eyepiece Shield DK-5 Eyepiece Shield	Go To Adorama	Free shipping \$3.5 See Details			
Under \$600	Set Price Alert Similar Products					
\$600 - \$1,300	No Image Lantern Guide for Canon Eos 7D Magic Lantern Guide for Canon Eos 7D	Go To Adorama	Free shipping \$3.6			
\$1,300 - \$2,500	Available Set Price Alert Similar Products	727 Seller Reviews	See Details			
\$2,500 - \$4,500	EA-SPC5D3 Screen Protector for Canon 5D Mark III Cameras					
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- Not treated in dynamic pricing or MAB literature

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- If yes, how to implement estimation and pricing in the presence of demand discontinuities?

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- Consumer demand: Poisson random variable with mean $d(p_t)$

$$d(p) = \begin{cases} e^{\alpha_0 + \beta_0 p} & \text{if } \kappa_0 \le p \le \kappa_1 \\ e^{\alpha_n + \beta_n p} & \text{if } \kappa_n$$

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• Model uncertainty:

unknown demand parameters $\theta_n = (\alpha_n, \beta_n)$ (n = 0, 1, ..., N)unknown discontinuity points κ_n (n = 1, ..., N) $\boldsymbol{\theta} = (\theta_0, \theta_1, ..., \theta_N) \in \Theta$ $\boldsymbol{\kappa} = (\kappa_1, ..., \kappa_N) \in \mathcal{K}$

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• Pricing policy: $\pi = (p_1, p_2, ...)$ non-anticipating

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 - Regret or "revenue loss due to demand model uncertainty"

$$\Delta_{\pi}(T, \boldsymbol{\kappa}, \boldsymbol{\theta}) = \sum_{t=1}^{T} \mathbb{E}_{\pi} \left\{ \sup_{p \in [\underline{p}, \overline{p}]} \{ R(p, \boldsymbol{\kappa}, \boldsymbol{\theta}) \} - R(p_t, \boldsymbol{\kappa}, \boldsymbol{\theta}) \right\}$$

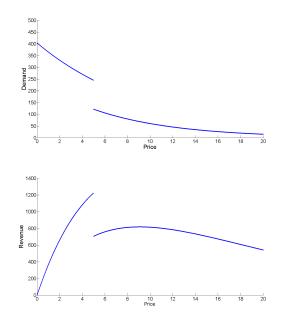
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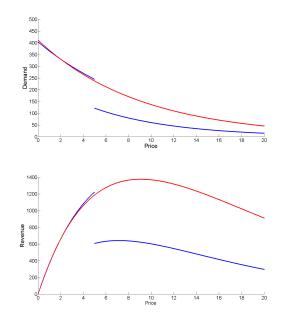
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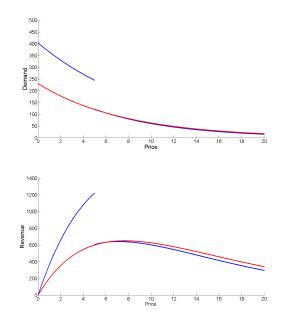
• **Objective:** choose π to minimize

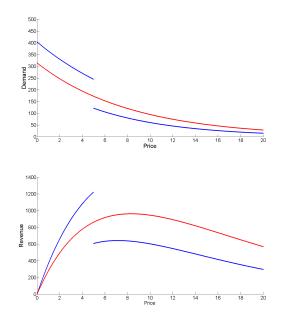
$$\mathcal{R}_{\pi}(T) = \sup \left\{ \Delta_{\pi}(T, \boldsymbol{\kappa}, \boldsymbol{\theta}) : \boldsymbol{\kappa} \in \mathcal{K}, \, \boldsymbol{\theta} \in \Theta \right\}$$

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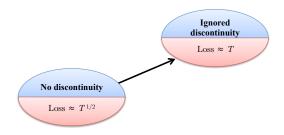












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Estimating a discontinuous demand function

• Two-step maximum likelihood estimation:

• Log-likelihood function

$$\mathcal{L}_t : (\varkappa, \vartheta) \mapsto \sum_{s=1}^t \sum_{n=0}^N \left(d_s \vartheta_n \cdot (1, p_s) - e^{\vartheta_n \cdot (1, p_s)} \right) \mathbb{I}\{\kappa_n < p_s \le \kappa_{n+1}\}$$
$$\hat{\theta}_t(\varkappa) = \arg \max_{\vartheta} \{ \mathcal{L}_t(\varkappa, \vartheta) \}$$

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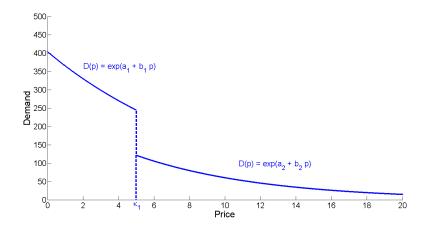
• <u>Step 1 (discontinuity estimation)</u> $\hat{\kappa}_t = \arg \max_{\varkappa} \left\{ \mathcal{L}_t(\varkappa, \hat{\theta}_t(\varkappa)) \right\}$

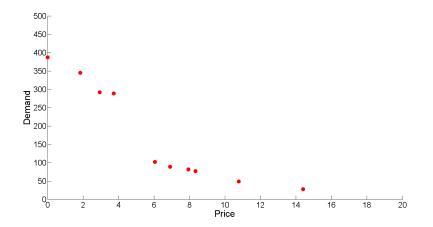
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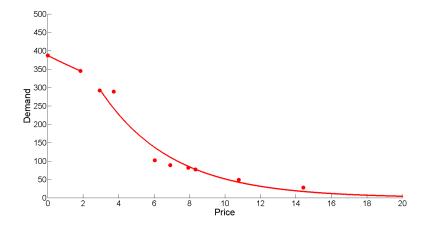
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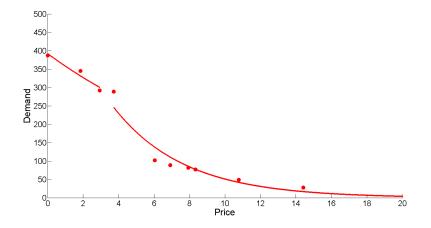
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- Step 2 (demand parameter estimation) $\hat{\boldsymbol{\theta}}_t = \hat{\boldsymbol{\theta}}_t(\hat{\boldsymbol{\kappa}}_t)$



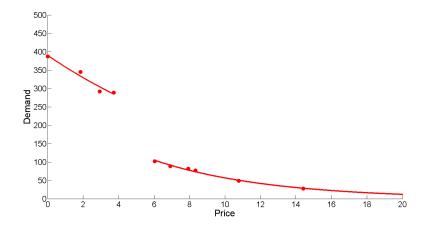




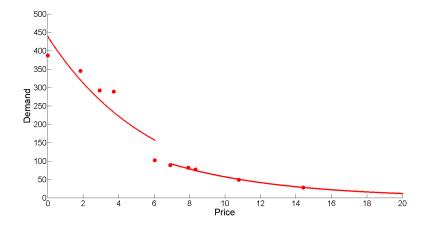
 $p_{(2)} \le \hat{\kappa}_1 < p_{(3)}$



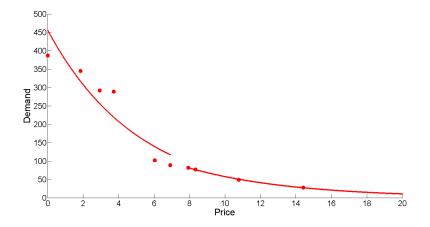
 $p_{(3)} \le \hat{\kappa}_1 < p_{(4)}$



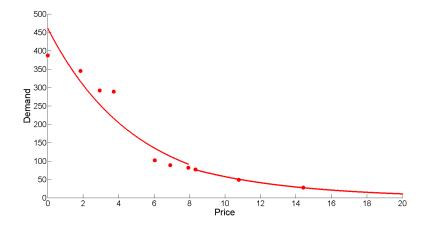
 $p_{(4)} \le \hat{\kappa}_1 < p_{(5)}$



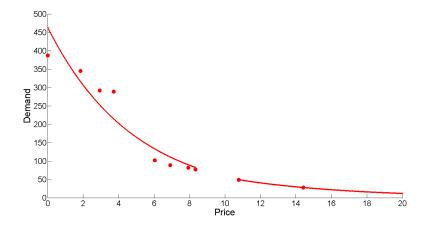
 $p_{(5)} \le \hat{\kappa}_1 < p_{(6)}$



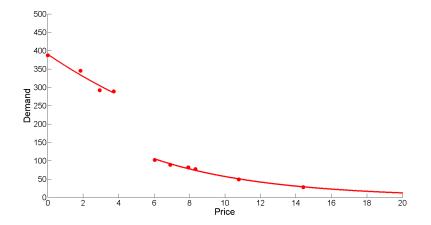
 $p_{(6)} \le \hat{\kappa}_1 < p_{(7)}$



 $p_{(7)} \le \hat{\kappa}_1 < p_{(8)}$



 $p_{(8)} \le \hat{\kappa}_1 < p_{(9)}$



Highest likelihood if $p_{(4)} \leq \hat{\kappa}_1 < p_{(5)}$.

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- (3) [Exploit] Based on $\hat{\kappa}$ and $\hat{\theta}$, use the estimated optimal price in the remaining T M periods, but a factor $\log(M)/M$ away from the estimated discontinuities.

Theorem (discontinuity estimation error)

There exist constants $M_1, z_1, \gamma_1 > 0$ such that, if $M \ge M_1$, then

$$\mathbb{P}_{\pi}\left\{\left|\hat{\kappa}_{n}-\kappa_{n}\right| > \frac{z_{1}\log M}{M} \text{ for some } n=1,\ldots,N\right\} \leq \frac{\gamma_{1}}{M}.$$

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Theorem (parameter estimation error)

There exist constants $M_2, z_2, \gamma_2 > 0$ such that, if $M \ge M_2$, then

$$\mathbb{P}_{\pi}\left\{\|\hat{\theta}_n - \theta_n\|^2 > \frac{z_2 \log M}{M} \text{ for some } n = 0, 1, \dots, N\right\} \leq \frac{\gamma_2}{M}.$$

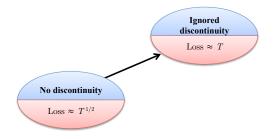
Theorem (upper bound on regret)

There exists a constant C > 0 such that, if $M = \lceil \sqrt{T} \rceil$, then

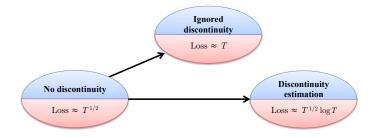
 $\mathcal{R}_{\pi}(T) \leq C\sqrt{T}\log T$

for all $T \ge 4(N+1)^2$.

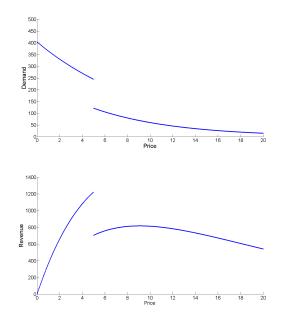
Summary of results



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Some intuition



Include change-point detection module in policy Retains $O(\sqrt{T}\log T)$ regret

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 - Nonparametric approach to discontinuous MABs.

THE END