The Scope of Sequential Screening with Ex-Post Participation Constraints

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Joint work with D. Bergemann (Yale) and G. Weintraub (Stanford)

Microsoft, March 2019

Problem: Sequential Screening

- When and how to sell when a buyer learns her valuation over time?
- Classic example: Airline tickets
- ▶ Initial purchase is based on an imperfect estimate: buyer's type could be leisure/business travelers (**Period 1**)
- Buyer knows true willingness-to-pay only at date of travel(Period 2)

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What is the revenue maximizing menu of contracts?

- ► Classic paper of Courty and Li (2000); also Akan et.al. (2015)
- Menu of upfront fees/refund contracts

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- ► Ex.1: in online shopping buyers can return purchases at low or no cost (Krähmer and Strausz 2015).
- ► Ex. 2: online display advertising markets: auction based and typical business constraint.

Online Display Advertising Motivation





· White House officials helped provide Devin Nunes, the Republican chairman of the House Intelligence Committee, with reports that showed incidental surveillance of the Trump team.

· The revelation is likely to fuel criticism that Mr. Nunes has been too eager to do the bidding of the Trump administration.

#1127 Comments

'Zoot Suit' Revival



Desperate, on a Road to Nowhere

Times journalists spent weeks documenting the stories of people living along a desert highway in Niger, interviewing more than 100 residents scattered by Boko Haram. The Daily 360: A View of the Highway

California Today: Theater Company Has a Hit With a Plus, the governor unveils an



The Opinion Pages

Ambassador for Working Foreign Policy Women The president is Her policies are little insensitive to the promore than lip service. United States political climate And they could make real in Latin America. reform impossible.

Ivanka Trump Is a Bad

Editorial: Ignoring History and the Promise of Diplomacy Collins: Trump Remembers the Indies Kristof: President Trump vs. Big Bird Join us on Facebook »

TIMES INSIDER . Foreign Correspondents as They Live and Breathe

Church Cruelty in Ireland THE CROSSWORD ... Play Today's Puzzle

The Blind Spots in Trump's

Edsall: Trump is ignorant of

The Empty Supreme Court

Confirmation Hearing

Whatever Hannened to

France's Famed 'Liberté'?

Suffer the Little Children:

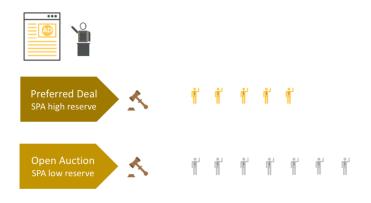
His Own Ignorance

One in three breast cancer patients under 45 removed the healthy breast along with the breast affected by cancer, a new study found.

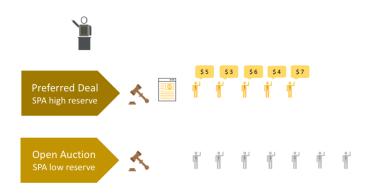


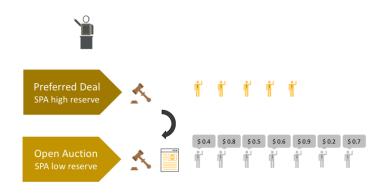






Think of period 1





Think of period 2

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 - ► Cai et. al (2016) and Devanur & Weinberg (2017) dual approach also applies
- ► (Partially) Shed light on practical mechanisms as effective price discrimination devices such as Waterfall Auctions

Seller: single item Single Buyer

Period 1 Period 2

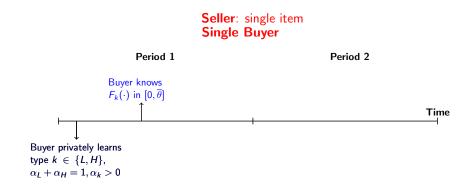
Time

Seller: single item Single Buyer

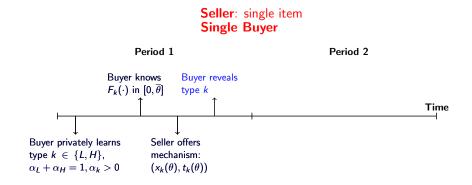
Period 1 Period 2



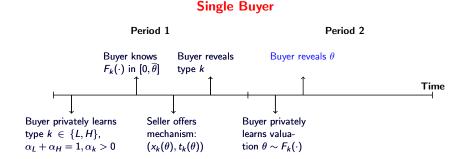
 $\begin{aligned} & \text{type } k \ \in \ \{L,H\}, \\ & \alpha_L + \alpha_H = 1, \alpha_k > 0 \end{aligned}$



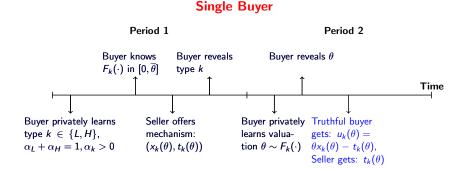








Seller: single item



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Seller: single item

Model primitives are common knowledge

mechanism:

 $(x_k(\theta), t_k(\theta))$

Parties are risk-neutral

type $k \in \{L, H\}$,

 $\alpha_L + \alpha_H = 1, \alpha_k > 0$

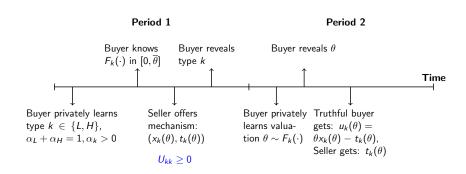
lacktriangle Non-increasing hazard rates. WLOG $\hat{ heta}_L \leq \hat{ heta}_H$

learns valua- gets: $u_k(\theta) =$

tion $\theta \sim F_k(\cdot)$ $\theta x_k(\theta) - t_k(\theta)$,

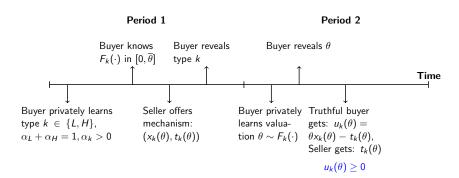
Seller gets: $t_k(\theta)$

Revenue Maximizing Mechanisms



► Courty and Li: What is the revenue maximizing sequential screening mechanism under interim participation constraints?

Revenue Maximizing Mechanisms



Our Question: What is the revenue maximizing sequential screening mechanism under ex-post participation constraints?

[Ex-post PC]
$$u_k(\theta) \ge 0, \ \forall k \in \{L, H\}, \ \forall \theta$$



Seller's Problem

The seller's problem is

$$(\mathcal{P}^d) \max_{0 \leq \mathbf{x} \leq 1, t} \qquad \sum_{k \in \{L, H\}} \alpha_k \cdot \int_0^{\bar{\theta}} t_k(z) \cdot f_k(z) dz$$
 s.t. $u_k(\theta) \geq \theta \cdot x_k(\theta') - t_k(\theta') \quad \forall k, \theta \text{ [Ex-post IC]}$
$$\int_0^{\bar{\theta}} u_k(z) f_k(z) dz \geq \int_0^{\bar{\theta}} u_{k'}(z) f_k(z) dz, \forall k, k \text{ [Interim IC]}$$

$$u_k(\theta) \geq 0, \quad \forall k, \theta \text{ [Ex-post PC]}$$

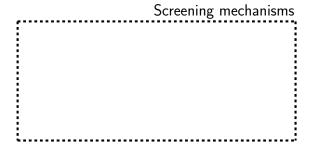
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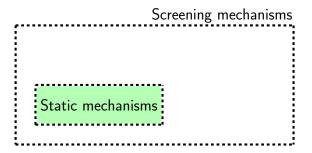
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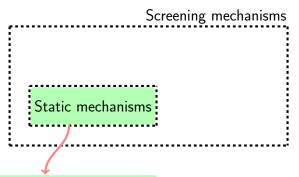
$$\begin{array}{ll} (\mathcal{P}^d) & \max_{0 \leq \mathsf{x} \leq 1, t} & \sum_{k \in \{L, H\}} \alpha_k \cdot \int_0^{\bar{\theta}} t_k(z) \cdot f_k(z) dz \\ \\ & \text{s.t.} \quad u_k(\theta) & \geq \quad \theta \cdot x_k(\theta') - t_k(\theta') \quad \forall k, \theta \text{ [Ex-post IC]} \\ \\ \int_0^{\bar{\theta}} u_k(z) f_k(z) dz & \geq \quad \int_0^{\bar{\theta}} u_{k'}(z) f_k(z) dz, \forall k, k \text{ [Interim IC]} \\ \\ u_k(\theta) & \geq \quad 0, \quad \forall k, \theta \text{ [Ex-post PC]} \\ \end{array}$$

- ▶ Ex-post IC: By the envelope theorem it is enough to solve for non-decreasing allocations $x_k(\cdot)$ and the utility of the lowest ex-post types $u_k(0)$
- ▶ Interim IC: More challenging (together with ex-post PC)

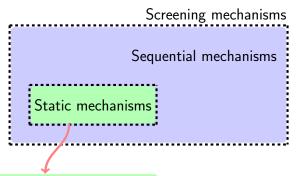




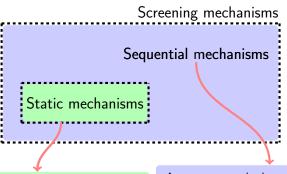




- A contract such that $x_k(\cdot) \equiv x(\cdot)$ and $t_k(\cdot) \equiv t(\cdot)$ for all k in $\{L, H\}$
- Pooling of interim types
- Myerson for the mixture distribution: **posted price** θ^s



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- –A contract such that $x_L(\cdot) \neq x_H(\cdot)$ and $t_L(\cdot) \neq t_H(\cdot)$
- Separate interim types
- Contract can be arbitrarily complex

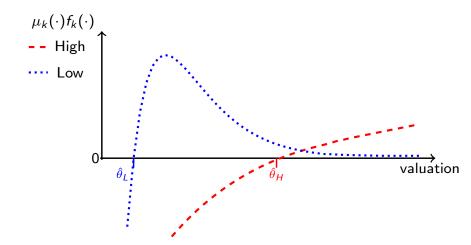
Research Questions/Contributions

- 1. When is a static contract optimal? When it is not?
 - ▶ Krähmer and Strausz 2015: Sufficient condition
 - Us: Necessary and sufficient condition

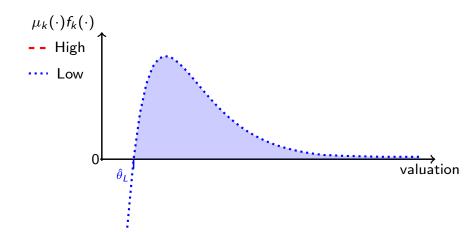
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- 1. When is a static contract optimal? When it is not?
 - Krähmer and Strausz 2015: Sufficient condition
 - Us: Necessary and sufficient condition
- 2. If a sequential contract is optimal, what does the optimal mechanism look like?
 - Us: Full characterization
 - Very different to Courty and Li
 - Significant revenue improvement over static contract

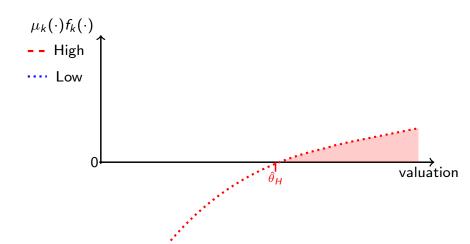
$$\mu_k(\theta) = \theta - \frac{1 - F_k(\theta)}{f_k(\theta)}$$



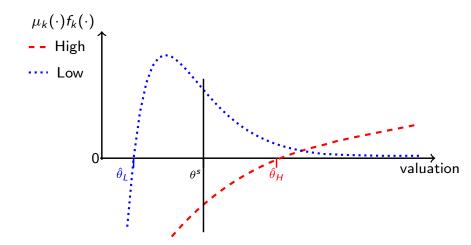
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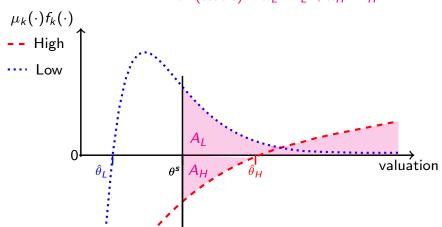


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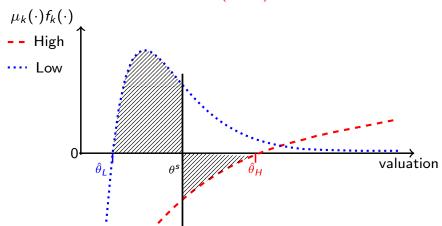
Rev (static)=
$$\alpha_L \cdot A_L + \alpha_H \cdot A_H$$



Let's look at weighted virtual values ("marginal revenues");

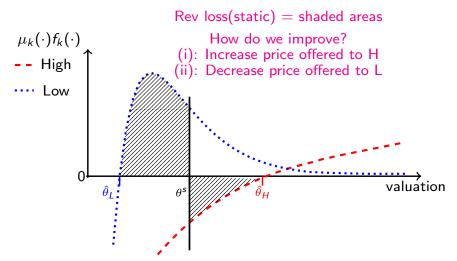
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Rev loss(static) = shaded areas



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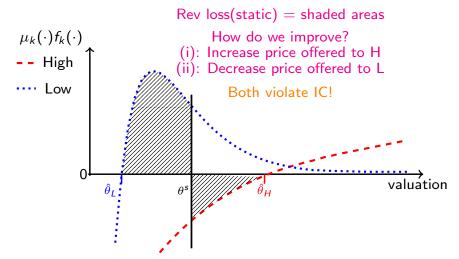
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13/23

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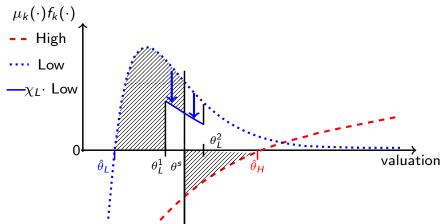


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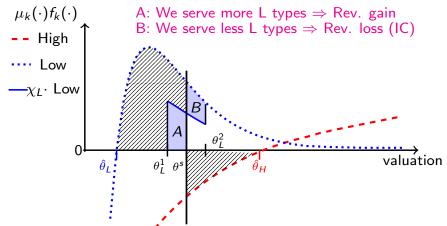
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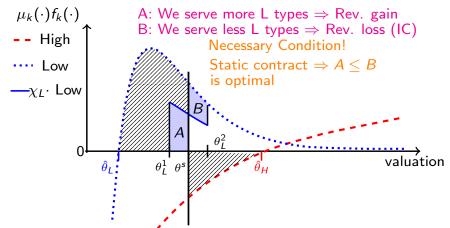
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General Necessity and Sufficiency

Theorem

The static contract is optimal if and only if

$$\max \left\{ \underbrace{\textit{Region A}}_{\text{revenue gain}} \right\} \leq \min \left\{ \underbrace{\textit{Region B}}_{\text{revenue loss}} \right\}$$

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1. Condition can be rigorously stated in terms of primitives:

$$\max_{\theta \leq \theta^{s}} \frac{\int_{\theta}^{\theta^{s}} \mu_{L}(\theta) f_{L}(\theta) d\theta}{\int_{\theta}^{\theta^{s}} (1 - F_{H}(\theta)) d\theta} \leq \min_{\theta^{s} \leq \theta} \frac{\int_{\theta^{s}}^{\theta} \mu_{L}(\theta) f_{L}(\theta) d\theta}{\int_{\theta^{s}}^{\theta} (1 - F_{H}(\theta)) d\theta}$$

- 2. Sharp intuitive characterization for optimality of static contract!
- Necessity formalizes picture above; sufficiency relaxes L to H IC and applies Lagrangian duality.

$$f_k(\theta) = \lambda_k e^{-\lambda_k \theta}, \quad k = \{L, H\} \quad \theta \ge 0, \quad \lambda_L > \lambda_H.$$

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Proposition

The static contract is optimal if and only if

$$\lambda_L - \lambda_H \le \frac{1}{\theta^s}$$

- \triangleright θ^s : optimal Myerson price for mixture distribution
- \triangleright λ_L and λ_H close then screening is not optimal
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Corollary

Assume $\lambda_L \in (\lambda_H, 2\lambda_H]$, then the static contract is optimal.

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Assume $\lambda_L > 2\lambda_H$, then there exists $\bar{\alpha} \in (0,1)$ such that the sequential contract is optimal iff $\alpha_L \in (0,\bar{\alpha})$.

General Necessity and Sufficiency

Krähmer and Strausz 2015 sufficient condition:

$$\frac{\mu_{\ell}(\theta)f_{\ell}(\theta)}{\bar{F}_{k}(\theta)}$$

is increasing for all ℓ, k

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- Stronger pointwise condition
- It implies our condition
- Ex.: does not necc. hold for exponential valuations.

General Characterization

Full characterization for optimal sequential contract!

Theorem

Consider problem (\mathcal{P}^d) and assume profit-to-rent cond. does not hold, the optimal solution has allocations

$$x_L^{\star}(\theta) = \begin{cases} 0 & \text{if } \theta < \theta_L^1 \\ \chi_L \in [0,1] & \text{if } \theta_L^1 \leq \theta \leq \theta_L^2 \\ 1 & \text{if } \theta_L^2 < \theta, \end{cases} \qquad x_H^{\star}(\theta) = \begin{cases} 0 & \text{if } \theta < \theta_H \\ 1 & \text{if } \theta_H \leq \theta, \end{cases}$$

for some values θ_L^1 , θ_H , θ_L^2 with $\theta_L^1 \le \theta_H \le \theta_L^2$, and $u_L(0) = u_H(0) = 0$



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- Optimal contract is simple
- Optimal contract departs from bang-bang contract with one buyer and one item
- $\hat{\theta}_L \le \theta_L^1, \ \theta_H \le \hat{\theta}_H$



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 - H type allocation has one or more intermediate steps

The Value of Sequential Screening: Optimal Revenues

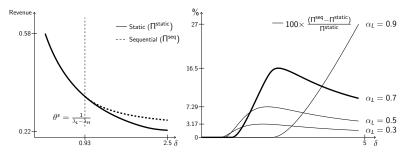


Figure: Left: Optimal expected revenue for static and sequential. Right: Percentage improvement of the sequential over the static contract. In both figures we set set $\lambda_L = \lambda_H + \delta$ with $\lambda_H = 0.5$.

Back to Waterfall Auctions

- In Waterfall Auctions low type buyers are randomized: can only bid when high-reserve auction does not clear
- "high-reserve auction does not clear" \Leftrightarrow high type value $\leq \theta_H$
- ▶ Seller revenue:

$$\max_{\theta_H \ge \theta_L \ge 0, (IC)} \alpha_L \underbrace{F_H(\theta_H)}_{\text{randomization}} \bar{F}_L(\theta_L)\theta_L + \alpha_H \bar{F}_H(\theta_H)\theta_H$$

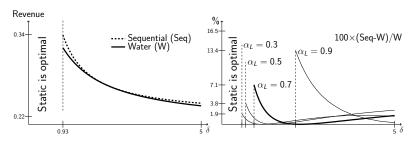


Figure: Left: Optimal expected revenue for Waterfall and Sequential. Right: Percentage improvement of the Sequential over the Waterfall contract. In both figures we set set $\lambda_L = \lambda_H + \delta$ with $\lambda_H = 0.5$.

Multiple Interim Types

▶ We partially extend result of necessary and sufficient condition for optimality of static contract.

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Multiple Interim Types

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- We prove that for exponential valuations there is at most one randomization step in optimal sequential contract. We partially extend this result.
- Multiple-type analysis is more complex because there is not an obvious relaxation of the math program.

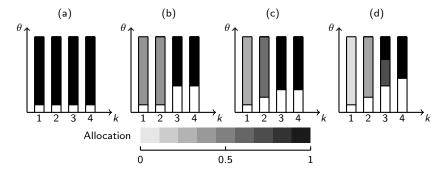


Figure: Optimal allocations for 4 interim types with exponential valuations. In each panel the vertical axis corresponds to buyers' valuations and the horizontal axis corresponds to the type. Each bar represents the allocation for each type, lighter grey indicates lower probability of allocation while darker grey indicates higher probability of allocation. Different distributions of interim types across instances.

Conclusions and Future Work

Summary

- Complete characterization for the optimal mechanism for two interim types and one buyer
- Both static and sequential contracts can be optimal
- When the sequential contract is optimal, the seller has to randomize the low-type and give a deterministic allocation to the high-type
- Some extensions to multiple types

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Current and Future work

- Study multiple buyers: may need ironing
- Connections with practical real-world mechanisms, such as waterfall auctions (randomization)
- Analyze performance guarantees of "simple mechanisms"