# How Competition Shapes Information in Auctions

Agathe PERNOUD

Simon GLEYZE

**Connections Workshop** 

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- E.g., takeover auctions, broadband licences auctions, procurement auctions....
- All involve large **due diligence costs**.

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Investment is only worth it if buyers have a fair chance to win:

- Incentivize buyers to inquire about the valuations of their competitors.
- A high-value bidder can discourage others from learning their own.



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Local landline company was participating and presumed to win... ...whereas local incumbent was disqualified in Chicago. In 1994-95, FCC ran an ascending auction for mobile phone broadband licenses. Bidding stopped at low price of \$26 per capita for L.A. license... ...whereas price was \$31 for less profitable city of Chicago.

Local landline company was participating and presumed to win... ...whereas local incumbent was disqualified in Chicago.

Other participants bid cautiously...

... whereas they aggressively competed in Chicago.

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We study second-price auctions in which buyers can flexibly acquire information at some cost.

- ▷ Buyers' valuations are independently drawn.
- ▷ Valuations are **unknown** to buyers ex ante.
- ▷ Can acquire a signal about their own valuations as well as those of **others**.

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Cheaper to first assess competition, and only learn own value if worth it.

#### How does buyers' information shape competition in return?

These learning incentives **hurt** the performance of the auction.

Intuition: Losing buyers fail to learn their values  $\Rightarrow$  regression to the mean of bids.

▷ **Revenue loss** compared to standard model.

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Market design implications:

- ▷ An additional bidder is **less** valuable than optimizing the auction's design.
- ▷ Seller gains from maintaining **uncertainty** over competition (e.g., via NDAs).

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THIS PAPER: focuses on second-price auctions to study interplay between *competition and learning incentives*.

*Learning and entry costs in auctions:* Milgrom (1981), Haush and Li (1993), Levin and Smith (1994), Persico (2000), Bergemann and Välimaki (2002), Compte and Jehiel (2007), Shi (2012) Lu and Ye (2013), Quint and Hendricks (2018), Lu, Ye and Feng (2021), Bobkova (2021), Marquez (2021), ...

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*The value of competition—"Auctions vs. Negotiations":* Bulow and Klemperer (1996, 2009), Aktas et al (2010), Roberts and Sweeting (2013), Gentry and Stroup (2019), ...

▷ New focus on how competition affects what kinds of info buyers acquire.

### OUTLINE OF TALK

#### 1. The Model

2. How Competition Shapes Information



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- 1. The Model
- 2. How Competition Shapes Information
- 3. How Information Shapes Competition
  - Revenue Distortions
  - Market Design Solutions



# The Model

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Utility of buyer *i* in state  $(\nu_i, \nu_{-i})$  under bid profile  $(b_j)_j$  is

$$U(\nu_i, b_i, b_{-i}) \equiv \left(\nu_i - \max_{j \neq i} b_j\right) \mathbb{1} \left\{ b_i = \max_j b_j \right\}.$$

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Second-price auction: info about  $\tilde{v}_{-i}$  only useful to decide what to learn about  $\tilde{v}_i$ .

*Structure of signals:* a signal of  $\tilde{v} \iff$  a convex partition  $\Pi = \{\pi_l\}_{l=1}^{L}$  of *V*.



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Buyers have **no** private info ex-ante but, before bidding, can acquire **two signals**:

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 $cost(signal) = \mathbb{E}[reduction in uncertainty of buyer's belief]$ 

There exists a concave function  $H : \Delta V \longrightarrow \mathbb{R}_+$  s.t.

$$c(\Pi, \text{prior}) = H(\text{prior}) - \mathbb{E}[H(\text{posterior}(\cdot|\Pi))].$$

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Overall cost is  $c(\Pi_i^{other}, \text{prior of max}_j v_j) + \mathbb{E}_{\Pi_i^{self} \mid \pi_i^{other}} \left[ c(\Pi_i^{self}, \text{prior of } v_i) \right]$ .

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#### After having acquired info, each buyer submits a bid $\sigma_i(\pi_i)$ .

A **NE** is a strategy profile  $(\prod_{i}^{other}, \prod_{i}^{self}, \sigma_i)_i$  s.t. *i*'s eq strategy solves

$$\max_{(\widehat{\Pi}_{i}^{other},\widehat{\Pi}_{i}^{self},\widehat{\sigma}_{i})} \mathbb{E}_{\nu_{i},\widehat{\pi}_{i},\pi_{-i}} \left[ U(\nu_{i},\widehat{\sigma}_{i}(\widehat{\pi}_{i}),\sigma_{-i}(\pi_{-i})) \mid \widehat{\pi}_{i} \right] - \lambda \cot\left(\widehat{\Pi}_{i}^{other},\widehat{\Pi}_{i}^{self}\right) \quad \forall i.$$

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*Equilibrium Selection:* Focus on **symmetric** equilibria satisfying a trembling-hand-like refinement.

1. Measure of uncertainty H is strongly concave, and sufficiently so. $\bullet$  details $\approx$  cost of signal is sufficiently convex in partition fineness.

**2**. Prior dist. of  $v_i$  is sufficiently uncertain.

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#### **2**. Prior dist. of $v_i$ is sufficiently uncertain.

 $\approx$  prior cannot put almost all weight on a couple of realizations.

E.g., set *H* to be the **entropy** and suppose entropy of prior is not too small.

 $\left[ \text{ entropy } H(\Pr(\cdot)) = -\sum_{v_i} \Pr(v_i) \log(\Pr(v_i)) \right]$ 

How Competition Shapes Information

# CAN BUYERS LEARN THEIR VALUATIONS FULLY?

#### Proposition

There exist **no** sequence of equilibria  $\{(\Pi_{\lambda}^{other}, \Pi_{\lambda}^{self}, \sigma_{\lambda})\}_{\lambda}$  such that

$$\Pr\left(\Pi_{\lambda}^{self} = \{\{v_i\}_{v_i \in V}\}\right) \longrightarrow 1 \text{ as } \lambda \longrightarrow 0.$$

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▷ **Cost-efficient** to first get some info about  $v_{-i}$ , and only learn  $v_i$  when worth it.

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▷ If buyers can only learn about  $v_i$ , they become fully informed for  $\lambda$  small enough.

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Alternative learning strategy for *i*:

1. Learns whether  $\max_{j} v_{j} \leq v^{*}$  for some threshold  $v^{*}$ .

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## WHY NOT?

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Alternative strategy yields same gross payoff, but strictly lower info costs.

▷ Assumptions on the cost and prior are key. ••••••

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# **Revenue Distortions**



How does it affect the performance of the auction?

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#### **Theorem 2**

Let  $N \ge 3$ . There exists L > 0 such that, for small enough information cost  $\lambda$ , the revenue generated in any equilibrium<sup>\*</sup> is bounded away from the expected second-highest valuation:

$$\mathbb{E}\left[ \text{ equilibrium revenue } \left| \left( \Pi_{\lambda}^{other}, \Pi_{\lambda}^{self}, \sigma_{\lambda} \right) \right] < \mathbb{E}\left[ \nu_{(2)} \right] - L.$$

▷ Info acquisition **distorts revenue**, even for small information costs.

## Revenue Loss-Cont'd



uniform example entry

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uniform example > entry

# Revenue Loss-Cont'd



For every interval of the eq. partition:

▷ If  $v_{(1)}$ ,  $v_{(2)}$  fall in same interval, then price  $\approx v_{(2)}$ . (Same as in standard model.) ▷ If not, losing buyers fail to learn  $v_i$  and bid  $\approx \mathbb{E}[v_i|v_i \leq \hat{v}] < \mathbb{E}[v_{(2)}|v_{(2)} \leq \hat{v}]$ .

uniform example

Implications for Market Design

# THE VALUE OF COMPETITION

Why do sellers use auctions? Competitive pressure between buyers helps sellers find a good **price**.

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Bulow & Klemperer (1996) show the value of competition is significant:

revenue of second-price auction with N + 1 buyers

revenue of opt. mechanism with *N* buyers

▷ Careful design of markets may not be that important absent additional frictions.

>

Learning frictions reduce value of competition between buyers.

#### **Theorem 3**

*There exists*  $\overline{N}$  *such that, for all*  $N \ge \overline{N}$  *and for*  $\lambda$  *small enough:* 

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▷ Reserve price even **more valuable** as more likely to trim low bids.

<

 $\triangleright$  Opt. reserve price depends on number of buyers *N*.

uniform example
The seller is hurt by buyers' incentives to learn about the competition.

*Solution:* **Randomize** access to the auction to maintain **uncertainty** over who will be competing:

 $\widetilde{M}$  random set of selected buyers.

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▷ Relates to the use of NDAs in practice.

▷ Might result in **misallocation** of the good if highest-valuation buyer is excluded.

#### **Theorem 4**

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Seller wants buyers to sign NDAs. Buyers want to disclose or signal a high value.

# Conclusion

### Concluding Remarks



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# CONCLUDING REMARKS



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TAKE AWAY: competition more effective if *designed* (w/ reserve price) or *uncertain*.

# **CONCLUDING REMARKS**



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Thank you! (agathep@stanford.edu)

# **Additional Material**

A1. *H* is sufficiently concave  $\approx$  cost is **sufficiently convex** in partition fineness.

*Formally*—there exists m > 0 such that , for all  $q, q' \in \Delta V$  and all  $t \in [0, 1]$ ,

$$H(tq + (1-t)q') - tH(q) + (1-t)H(q') \ge mt(1-t)||q-q'||^2$$

and *m* is sufficiently large.

A2. *Prior* is sufficiently uncertain, i.e.,  $\sum_{v_i} Pr(v_i)^2$  is small enough.

# When is it Cost-Efficient?



#### Lemma

*There exists*  $\overline{\Sigma}$  *and*  $\overline{m}$  *s. t. if*  $\sum_{v} [p(v)]^2 \leq \overline{\Sigma}$  *and*  $m \geq \overline{m}$ *, then* 

$$c\left(\prod_{v^*}^{other}, p_{1:N-1}\right) + \Pr\left(\max_{j} v_j < v^*\right) c\left(\prod_{v^*}^{self}, p\right) + \Pr\left(\max_{j} v_j > v^*\right) c\left(\prod_{v^*}^{self}, p\right)$$

is strictly lower than choosing  $\Pi^{self} = \{\{v_i\}_{v_i \in V}\}$  for some  $v^* \in V$ .

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back

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STEP 4. Info partition about others **must** be informative  $\Pi_{\lambda}^{other} \neq \{V\}$ .

$$V = \left\{\frac{1}{K}, \frac{2}{K}, \dots, \frac{K-1}{K}, \frac{K}{K}\right\}$$
 with  $Pr(v_i) = \frac{1}{K}$  for all  $v_i$ . Cost is based on entropy.



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**—** *Posted price:* seller chooses a **fixed** price; if multiple buyers want to buy, one is chosen randomly.

So far, participation in the auction was *free*. Now suppose buyers must incur an entry cost  $\kappa > 0$  to bid. Entry decision occurs at the end of info acquisition stage.

- ▷ If learns that another has a higher valuation, might not want to enter at all...
- $\triangleright$  ... which *worsens* the revenue loss.

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...which *worsens* the revenue loss.

*Standard model:* For small entry cost  $\kappa$ , almost all buyers enter.

*Our model:* Several buyers enter only if their valuations fall in a similar range.

▷ Presence of high-value buyer **deters entry** of lower-value buyers.

#### Equilibrium signal about $\max_j v_j$ partitions V into:



## Extra Buyer vs. Reserve Price-Cont'd

Back to our uniform example with 
$$V = \left\{\frac{1}{K}, \frac{2}{K}, \dots, \frac{K-1}{K}, \frac{K}{K}\right\}$$
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## Extra Buyer vs. Reserve Price-Cont'd

Back to our uniform example with  $V = \left\{\frac{1}{K}, \frac{2}{K}, \dots, \frac{K-1}{K}, \frac{K}{K}\right\}$ .



#### **Optimal Reserve Price**

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Back to our uniform example with 
$$V = \left\{\frac{1}{K}, \frac{2}{K}, \dots, \frac{K-1}{K}, \frac{K}{K}\right\}$$
.

0.9 0.8 Optimal reserve price costly info costless info 0.2 0.1 3 7 8 9 10 2 4 5 6 Number of buyers N

Buyers have an additional *strategic* reason to learn about their competitors:

 $\triangleright$  Winning buyer wants to bid just above the 2<sup>*nd*</sup>-highest bid.

#### Proposition

The Revenue Equivalence Theorem no longer holds in our model.

If it held, revenue loss in SPA  $\implies$  in some states, winning bid  $b_{(1)} \ll v_{(2)}$ :

$$\xrightarrow{b_{(1)}} \xrightarrow{v_{(2)}} \xrightarrow{v_{(1)}}$$

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# Role of Price Discovery-Model Extension

 $\omega \in \Omega$ , with  $\Omega$  finite, is the common component (*what the seller wants to learn*).

Buyers' valuations are drawn i.i.d. from full-support  $p_{\omega} \in \Delta V$ , given  $\omega$ .

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### Proposition

There exist environments  $\{p_{\omega}\}_{\omega}$  with  $p_{\omega} \neq p_{\omega'}$  for all  $\omega$ ,  $\omega'$ , s. t. the auction **does not** reveal the common component as  $N \longrightarrow \infty$ , even when  $\lambda$  is arbitrarily small.

### Does the Auction Reveal $\omega$ ?

Let 
$$\Omega = {\underline{\omega}, \overline{\omega}}$$
 and  $\mu_0(\overline{\omega}) = 0.5$ .



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In previous lit., price discovery often thought of as  $price \longrightarrow \nu_{(1)}$  as  $N \longrightarrow \infty$ .

[Wilson (1977), Milgrom (1979), Pesendorfer & Swinkels (1997, 2000),...]

In our model, price converges more slowly because losing buyers often fail to learn and price  $\ll \nu_{(2)}$ .

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In our model, price converges **more slowly** because losing buyers often fail to learn and price  $\ll \nu_{(2)}$ .

▷ Can be problematic if auction prices serve as benchmarks.

Need larger auctions to find a "correct" price.

# PRICE CONVERGES MORE SLOWLY

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#### Same example as before.

