

# How sellers decide on dynamic mechanisms: Information matters

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Nov 2024

# Optimal Dynamic Mechanism Design

- ▶ To maximize the revenues (payoff), the seller (principle) sets rules of allocations and prices over multi-period as the buyer (agent) receives private information over time.
  - ▶ **Repeated selling of perishable goods**
  - ▶ Long-term principal-agent relationship
- ▶ Dynamic mechanism improves revenues and the efficiency (Baron & Besanko, 1984).
- ▶ How do human sellers choose a dynamic mechanism?

# Two Simple Dynamic Mechanisms

- ▶ Non-Clairvoyant: not using future information  $\Rightarrow$  general form.

## (RS) Repeated Static Mechanism (Myerson, 1981)

- ▶ 100% optimal static revenue (intra-period)
- ▶ 0% optimal dynamic revenue (inter-period)

## (NC) Non-Clairvoyant Dynamic Mechanism (Mirrokni et al., 2020)

- ▶ 50% optimal static revenue (intra-period)
- ▶ 50% optimal dynamic revenue (inter-period)

## NC Cannot Always Outperform RS

- ▶ Relative size of inter-period revenue matters (Gui and Houser, 2024).

# Research Question

## How do Sellers Decide on Dynamic Mechanisms?

- ▶ How do Sellers choose between NC and RS?
- ▶ Can Sellers make good decisions and improve revenues?

## What Information Sellers Use in Deciding on Mechanism?

- ▶ **Relative Simplicity:** NC is harder: set more prices.
- ▶ **Demand Knowledge:** NC outperforms RS in some scenarios.
- ▶ **Feedback:** NC gets less revenue as buyers might quit the later period.

# Experimental Procedure

## Settings

- ▶ **Clairvoyant environment:**  $F_1, F_2$  known for Sellers at the beginning.
- ▶ 10 Rounds + 2 Practice Rounds, feedback on each round, each period.
- ▶ Fixed role, re-match for each round.
- ▶ Risk task and ambiguity task (random ordered) at the end.

## Choosing from Two Mechanisms in each Round

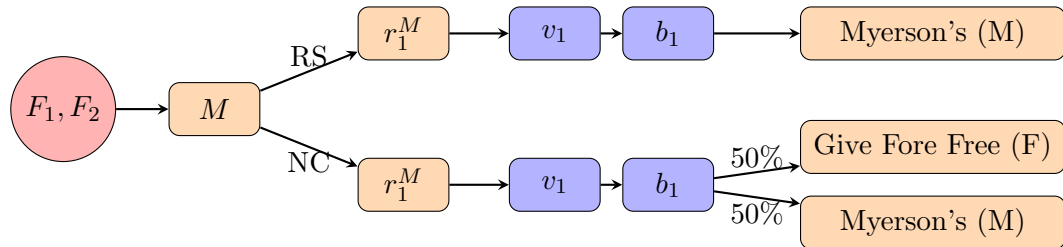
- ▶ Non-Clairvoyant Dynamic Mechanism (NC)
- ▶ Repeated Static Mechanism (RS)

# Experimental Task in each Round

## Period 1

1. Seller chooses mechanism,  $\mathbf{M}$  (=NC or RS), buyer is informed
2. Seller sets reserve price  $\mathbf{r}_1^{\mathbf{M}}$  for Period 1, Buyer makes a bid  $\mathbf{b}_1(\mathbf{v}_1)$ .
  - ▶ in RS: buyer pays  $r_1^M$  if  $b_1 \geq r_1^M$
  - ▶ in NC: buyer has 50% chance to get free item

## Choose Mechanism for two Periods



# Experimental Task in each Round

## Period 2

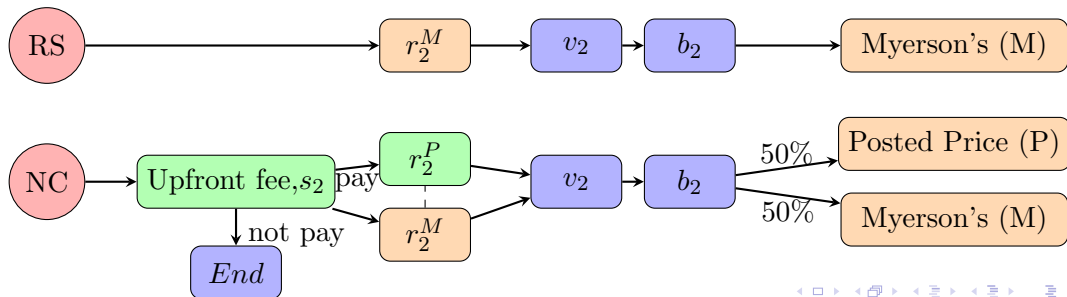
1. Seller sets reserve price  $r_2^M$  for Period 2

**Treatment Set 2:**  $s_2, r_2^P$  will be set by the Computer optimally

**Treatment Set 4:**  $s_2, r_2^P$  will be set by the Seller

2. Buyer chooses to pay the upfront fee  $s_2$  or not

Buyer makes a bid  $b_2(v_2)$  in RS or in NC if entering in the market



# Experimental Design: Different Information

## Relative Simplicity - Two Treatments (Between-subject)

- ▶ Treatment Set 2: Automated Posted Price Auction (green area)
- ▶ Treatment Set 4: Sellers set 4 prices in NC and 2 prices in RS.

## Scenario-specific Demand - Twelve Scenarios (3 Categories) (Within-subject)

- ▶ 4 NC Better:  $Rev^{NC} > Rev^{RS}$
- ▶ 4 RS Better:  $Rev^{NC} < Rev^{RS}$
- ▶ 4 Same:  $Rev^{NC} = Rev^{RS}$
- ▶ For each Session: 2 Same in Practice Stage + 2 in Tail Stage  
Early Stage (4 rounds): 2 NC Better + 2 RS Better  
Later Stage (4 rounds): 2 NC Better + 2 RS Better

## Feedback - Revenue of each Round



## NC Better ( $Rev^{NC} > Rev^{RS}$ )

### Period 2 is Profitable

- ▶ Inter-period dynamic revenue is more important in NC
- ▶  $\exists$  “target buyers” (high valuation but low probability) in Period 2

$$REV^{RS} = 4, \quad REV^{NC} = 4.5 \quad \uparrow 12.5\%$$

$$F_A = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{16}), (32, \frac{1}{16})\}, \quad \mathbb{E}_A = 6.$$

1.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{2})\}, \quad F_2 = F_A$
2.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{4})\}, \quad F_2 = F_A$
3.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{8})\}, \quad F_2 = F_A$
4.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{16}), (32, \frac{1}{16})\}, \quad F_2 = F_A$

## RS Better ( $Rev^{NC} < Rev^{RS}$ )

Period 2 is not Profitable

- ▶ Intra-period revenue is more important in NC
- ▶ *e.g.*, Constant valuation,  $v_2 = 0$  in Period 2.

$$REV^{RS} = 4, \quad REV^{NC} = 3.5 \quad \downarrow 12.5\%$$

$$F_B = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{2}), \}, \quad \mathbb{E}_B = 3.$$

1.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{2})\}, \quad F_2 = F_B$
2.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{4})\}, \quad F_2 = F_B$
3.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{8})\}, \quad F_2 = F_B$
4.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{16}), (32, \frac{1}{16})\}, \quad F_2 = F_B$

Same ( $Rev^{NC} = Rev^{RS}$ )

Inter- is as important as Intra- revenue

- ▶  $\Longleftrightarrow Rev^P = Rev^{M_1} + Rev^{M_2}$
- ▶ e.g., Constant valuation,  $v_1 = c_1 = 0$  in Period 1,  $v_2 = c_2 \geq 0$  in Period 2.

$$REV^{RS} = REV^{NC} = 4$$

$$F_C = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{4})\}, \quad \mathbb{E}_C = 4.$$

1.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{2})\}, \quad F_2 = F_C$
2.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{4})\}, \quad F_2 = F_C$
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## Summary of Theoretical Revenue (Total)

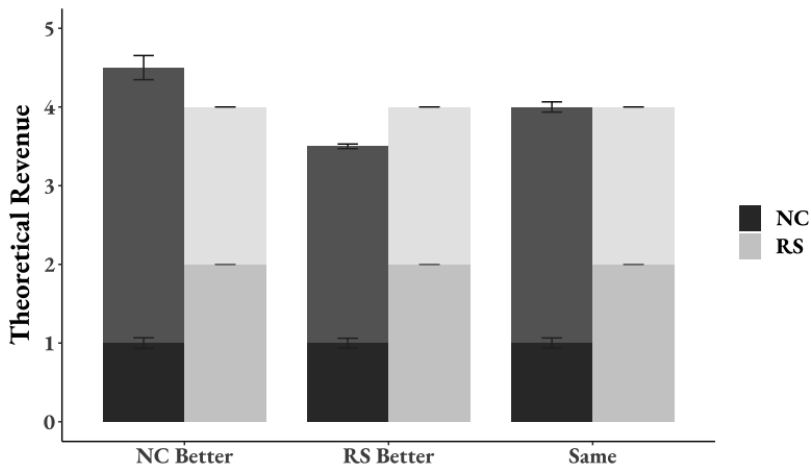


Figure 1: Theoretical Revenue (Total)

# Hypotheses

## Relative Simplicity

- ▶ H1: Sellers choose more NC in the treatment Set 2.

## Demand Knowledge

- ▶ H2: Sellers Choose NC more (less) in NC Better (B) in the Later Stage.
- ▶  $\Rightarrow$  Sellers choose correct mechanism more in Later stage.

## Feedback on Revenues

- ▶ H3: Sellers choose NC more (less) when past revenue from NC is high (low).

## \* Buyers' Behaviors

- ▶ H4: Buyers participate Period 2 more in the treatment Set 2.

# Experiments

- ▶ 256 George Mason Students. October to November 2022.

Table 1: Summary Statistic

Treatment Role	Set 2		Set 4	
	Sellers	Buyers	Sellers	Buyers
Age	22.6	22.2	21.2	22.5
Gender (Male=1)	0.59	0.62	0.52	0.50
Risk aversion	3.14	3.95	3.90	3.70
Ambiguity	3.30	3.02	3.67	3.32
Observation	64	64	64	64

# Result 1. Relative Simplicity does not Matter

R1. Sellers do not choose NC more in the treatment Set 2. H1 is not supported.

- ▶ Early Stage: no difference from 50% in neither Treatment.
- ▶ Later Stage: Significant less than 50% in the treatment Set 2 ( $p < 0.01$ ).
- ▶ No treatment difference in either stage.

## Sellers do not Choose NC More in General

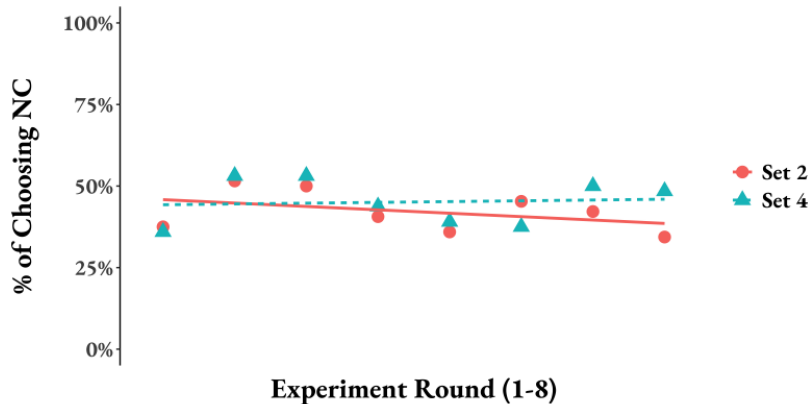


Figure 2: % of Choosing NC



## Result 2. Demand Knowledge Matters

R2. Sellers choose NC less in RS Better in Later Stage. H2 is supported.

- ▶ NC Better: No difference from 50%
- ▶ RS Better: Significant less than 50% ( $p < 0.01$  in Set 2,  $p < 0.01$  in Set 4).
- ▶ More correct mechanisms in the Later stage ( $p = 0.01$ , in Set 2,  $p < 0.01$  in Set 4)

## R2. Sellers Choose NC Less in RS Better

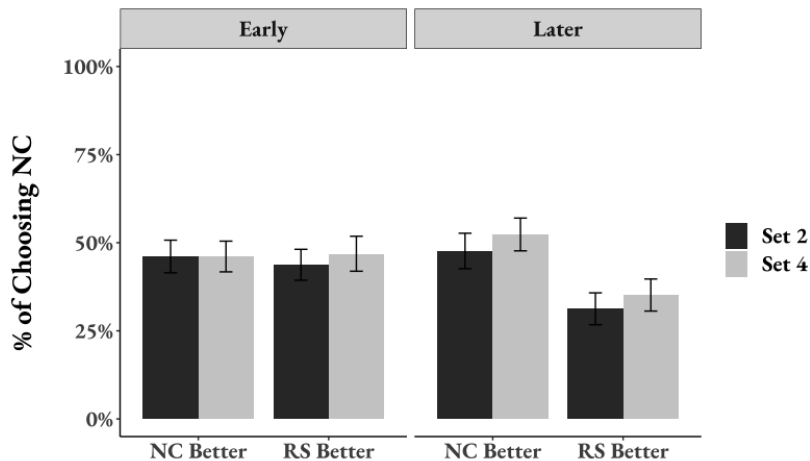


Figure 3: % of Choosing NC by Group of Scenario

## % of Choosing Correct Mechanism ↑

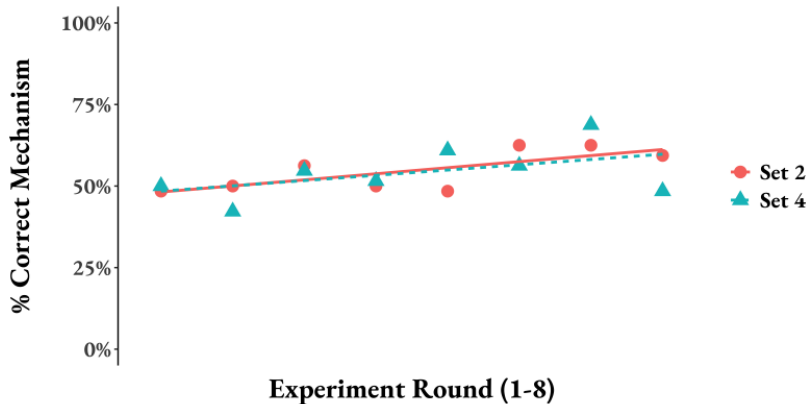


Figure 4: % of Choosing correct Mechanism

## Result 3. Feedback on Revenue Matters

R3. Sellers choose NC less if past revenue from NC is low. H3 is supported.

- ▶ Persist NC more in Later rounds.
- ▶ Less likely to choose NC if last round NC got less than 3 points.

# Reaction to Negative Feedback

Table 2: Regression of Choosing NC

	DV: Choosing NC	
	(1)	(2)
$\beta_1$ : Last (payoff < Ave, NC)	-0.21*** (0.07)	-0.21*** (0.07)
$\beta_2$ : Later * Treatment Set 4	0.01 (0.04)	-0.00 (0.04)
$\beta_3$ : Later * RS Better Scenarios	-0.18*** (0.06)	-0.18*** (0.06)
$\beta_4$ : Later * Last (NC)	0.16* (0.09)	0.17* (0.09)
$\beta_5$ : Later * Last (Correct = NC)	0.02 (0.10)	0.01 (0.10)
Constant	0.46*** (0.04)	0.31** (0.13)
Controls	No	Yes
R <sup>2</sup>	0.05	0.05
Num. obs.	1024	1024

# Revenue Improvement by choosing Correct Mechanism

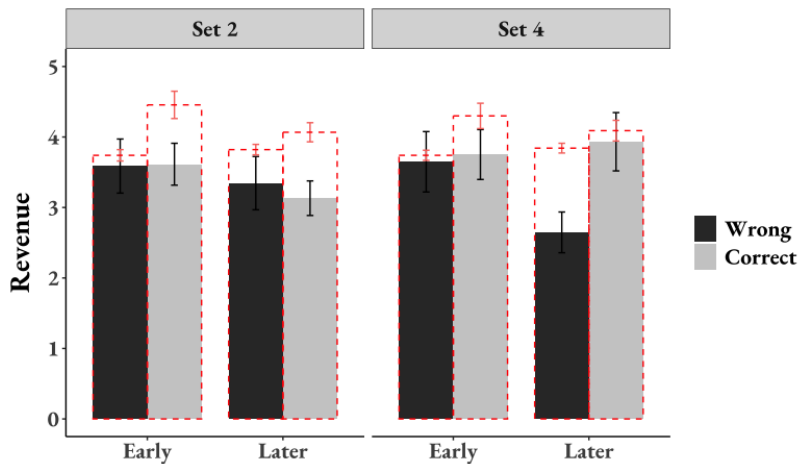


Figure 5: Experimental Revenue

## Result 4. Buyers Reacting to high entry fee.

R4. Buyers participate less in the treatment Set 4. H4 is supported.

- Explained by higher entry fee set by Sellers in the treatment Set 4.

## Sellers set higher than suggested entry fee

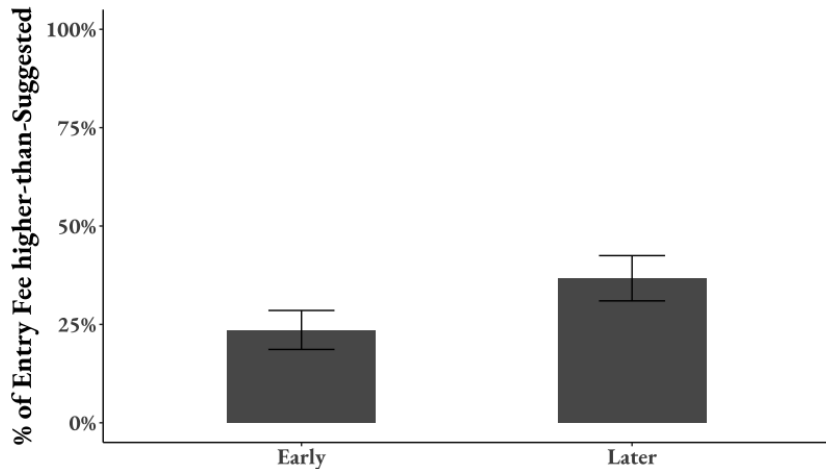


Figure 6: % of Setting Entry Fee Higher than Suggested



# Sellers Set higher Prices

## In Period 1

- ▶ “Go big or go home”.
- ▶ Aimed high, looking for a heavy bid
- ▶ You’d be surprised when I say - I based it off the charts.
- ▶ Random.

## In Period 2

- ▶ Again, attempted high roll, but failed greedily.
- ▶ Higher price didn’t work so I went lower.
- ▶ buyer bid for 1?? which makes no sense so I wanted to get some out of him and set the price to 6 as possible values could have been pretty high. Then set price to 4 as I would get it 50% of the time
- ▶ Set a low price, however, buyer decided not to purchase.

## Buyers participated less in the treatment Set 4

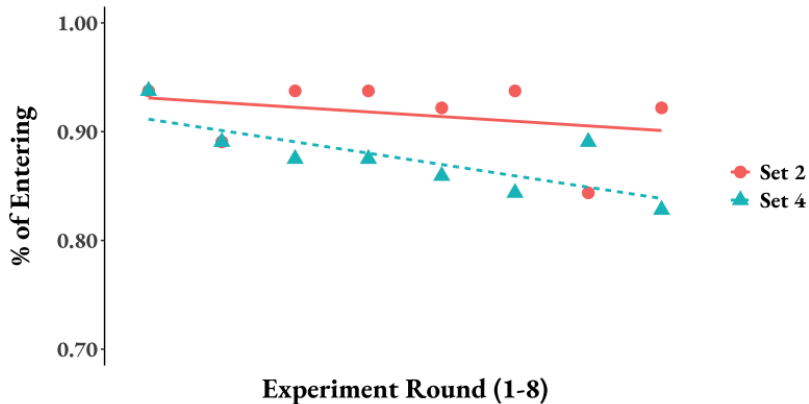


Figure 7: % of Entering Period 2

# High Entry Fee Deters Entering

Table 3: Probit Regression of Enter in Period 2

	DV: Enter in Period 2	
	(1)	(2)
$\beta_1$ : Entry Fee	-0.24*** (0.04)	-0.22*** (0.04)
$\beta_2$ : Set 4	-0.12 (0.20)	-0.19 (0.21)
$\beta_3$ : Set 4 * Later	-0.07 (0.23)	-0.18 (0.24)
$\beta_4$ : Later	-0.36** (0.16)	-0.35** (0.17)
Constant	2.04*** (0.23)	2.34*** (0.69)
Controls	No	Yes
Num. obs.	447	447

Table 4: Probit Regression of Enter in Period 2

# Conclusion

## Distributional knowledge and Feedback matter in Choosing Mechanism

- ▶ Sellers can find the optimal mechanism after gaining trading experience.
- ▶ Sellers abandon the mechanism with low revenue.
- ▶ Selling strategy evolves as selling condition or expectation changes.

## Discussion

- ▶ Decision Support Pool: appropriate expectation on Buyers behaviors.
- ▶ Experts: advice setting (lower) prices.

*Thank you!*