

# How sellers decide on mechanism: Information matters

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# Background

## Non-Clairvoyant Environment

- ▶ More practical: future distribution is not available in designing mechanism.
- ▶ (NC) Non-Clairvoyant Dynamic Mechanism (Mirrokni et al., 2020):  
50% optimal inter-period revenue and 50% optimal intra-period revenue
- ▶ (RS) Repeated Static Mechanism (Myerson, 1981):  
0% optimal inter-period revenue and 100% optimal intra-period revenue

## NC Cannot Always Outperform RS

- ▶ Relative size of inter-period revenue matters.
- ▶ Experiments support theoretical revenue predictions (Gui and Houser, 2022).

# Research Question

## How do Sellers Decide on Mechanism?

- ▶ How do Sellers choose between NC and RS?
- ▶ Can Sellers make good decision and improve payoff?

## What Information Sellers Use in Deciding on Mechanism?

- ▶ **Mechanism Features:** NC requires to set more prices.
- ▶ **Current Conditions:** NC is optimal for some conditions.
- ▶ **Past Experiences:** NC gets less revenue as Buyers might quit the second period.

# Experimental Procedure

## Settings

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

## Choosing from Two Mechanisms in each Round

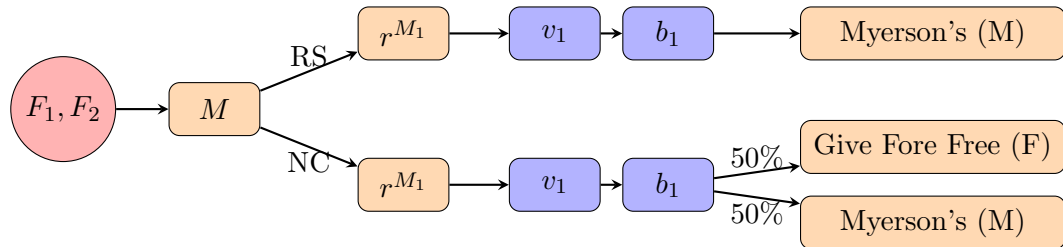
- ▶ Non-Clairvoyant Dynamic Mechanism (NC)
- ▶ Repeated Static Mechanism (RS)

# Experimental Task in each Round

## Period 1

1. Seller chooses mechanism,  $\mathbf{M}$  (=NC or RS), buyer is informed
2. Seller sets reserve price  $\mathbf{r}^{M_1}$  for Period 1, Buyer makes a bid  $\mathbf{b}_1(\mathbf{v}_1)$ .
  - ▶ in RS: buyer pays  $r^{M_1}$  if  $b_1 > r^{M_1}$
  - ▶ in NC: buyer has 50% chance to get free item

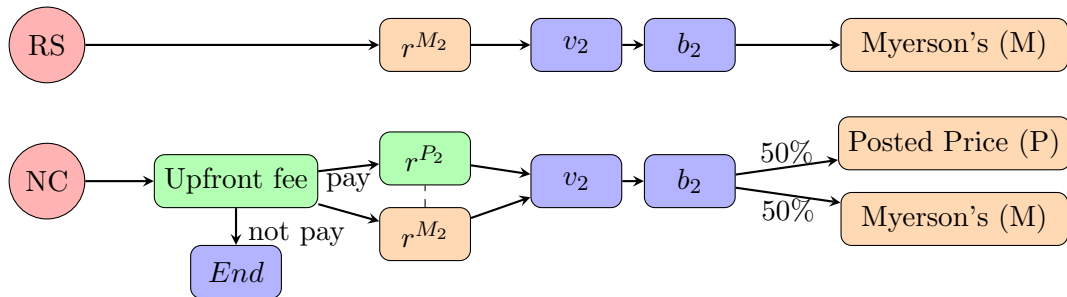
## Choose Mechanism for two Periods



# Experimental Task in each Round

## Period 2

1. Seller sets reserve price  $r^{M_2}$  for Period 2  
(for NC,  $e_2, r^{P_2}$  will be set by computer optimally)
2. Buyer chooses to pay the upfront fee  $u_2$  or not  
Buyer makes a bid  $b_2(v_2)$  in RS or in NC if entering in the market



# Experimental Design: Different Information

## Mechanism Features - Two Treatments (Between-subject)

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

## Current Conditions - Ten Scenarios (3 Groups) (Within-subject)

- ▶ 4 Scenarios A:  $NC > RS$
- ▶ 4 Scenarios B:  $NC < RS$
- ▶ 4 Scenarios C:  $NC = RS$
- ▶ For each Session: 2 Scenarios C in Practice Stage + 2 in Tail Stage  
2 Scenarios A + 2 Scenarios B in Early Stage (4 rounds)  
2 Scenarios A + 2 Scenarios B in Later Stage (4 rounds)

## Past Experiences - Feedback each Round

## Scenarios A ( $NC > RS$ )

Inter-period revenue is more important

- ▶  $\mathbb{E}_2$  is greater than  $Rev^M$  in the second period
- ▶  $\exists$  “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.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{16}), (32, \frac{1}{16})\}, \quad F_2 = F_A$



## Scenarios B ( $NC < RS$ )

Intra-period revenue is more important

- ▶  $\mathbb{E}_2$  is not great enough while  $Rev^M$  can achieve at least half of  $\mathbb{E}_2$
- ▶ *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.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{16}), (32, \frac{1}{16})\}, \quad F_2 = F_B$

## Scenarios C ( $NC = RS$ )

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.

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

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

1.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{2})\}, \quad F_2 = F_C$
2.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{4})\}, \quad F_2 = F_C$
3.  $F_1 = \{v, p(v)\} = \{(2, \frac{1}{2}), (4, \frac{1}{4}), (8, \frac{1}{8}), (16, \frac{1}{8})\}, \quad F_2 = F_C$
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## Summary of Theoretical Revenue (Period 1)

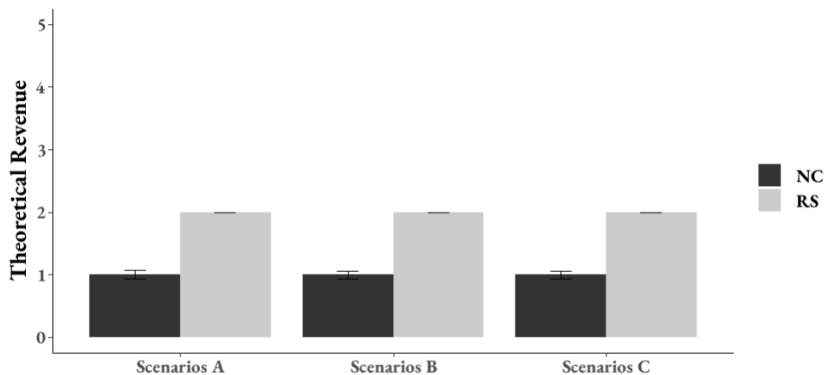


Figure 1: Theoretical Revenue (Period 1)

## Summary of Theoretical Revenue (Total)

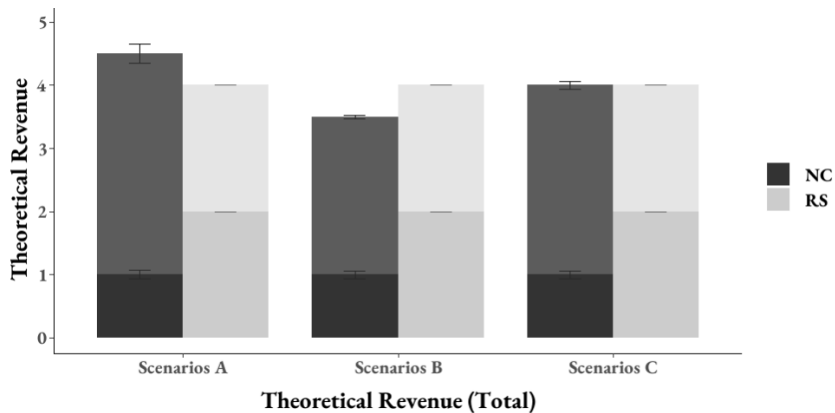


Figure 2: Theoretical Revenue (Total)

# Hypotheses

## Mechanism Features

- ▶ H1: Sellers choose more NC in Partial Treatment.

## Current Conditions

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

## Past Experiences (Revenues)

- ▶ H3.1: Sellers get more revenue in NC (RS) in Scenarios A(B).
- ▶ H3.2: Sellers choose NC more (less) when past revenue from NC is high (low).

# Experiments

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

Treatment Role	Partial		Full	
	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

Table 1: Summary Statistic

# Result 1. Mechanism Features Do not Matter

R1. Sellers do not choose NC more in Partial.

- ▶ Early Stage: no difference from 50% in either Treatment.
- ▶ Later Stage: Significant less than 50% in Partial ( $p < 0.01$ ).
- ▶ No treatment difference in either stage.

## R1. Sellers do not Choose NC More in Full

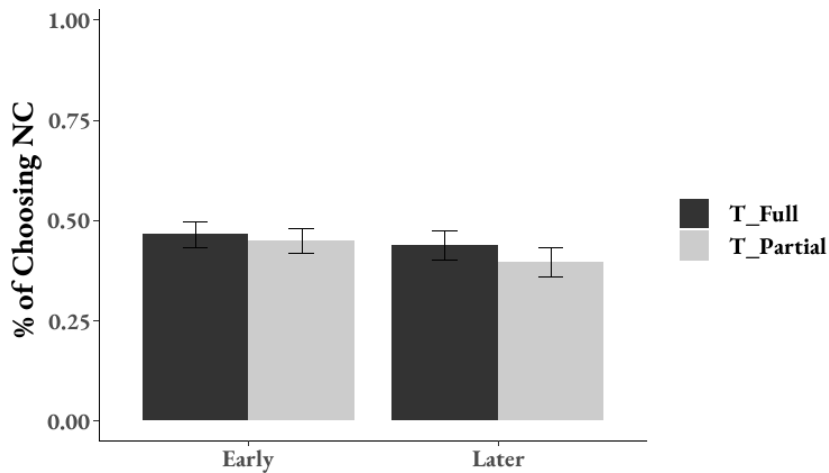


Figure 3: % of Choosing NC



## Sellers do not Choose NC More in General

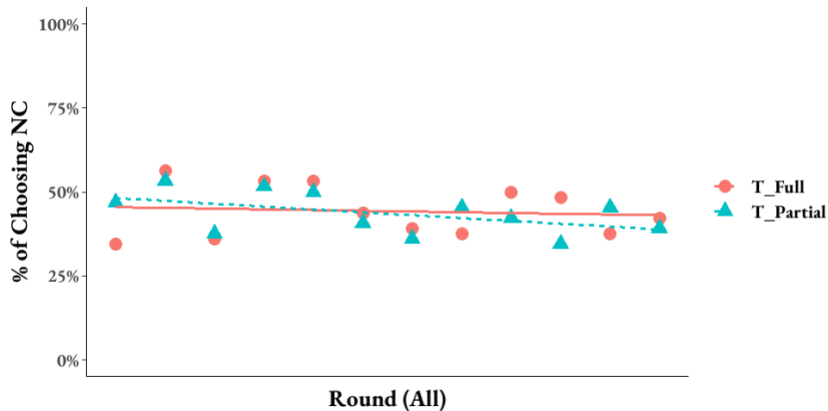


Figure 4: % of Choosing NC

## Result 2. Current Conditions Matter

### R2. Sellers choose NC less in Scenarios B in Later Stage

- ▶ Scenarios A: No difference from 50%
- ▶ Scenarios B: Significant less than 50% ( $p < .01, p < 0.01$ ).

### ⇒ Sellers choose optimal mechanism more in Later Stage

- ▶ Early stage: no difference from 50%.
- ▶ Later stage: Significant greater from 50% ( $p = .01, p < 0.01$ ).
- ▶ No treatment difference in either stage.

## R2. Sellers Choose NC Less in Scenarios B

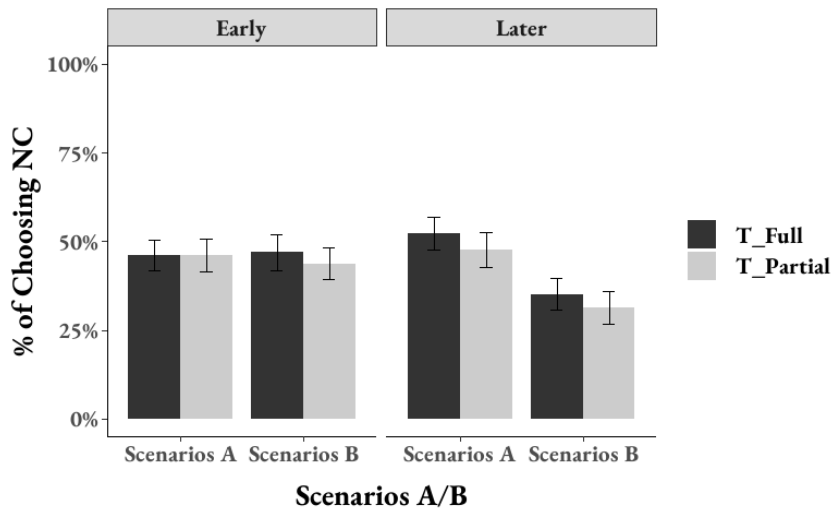


Figure 5: % of Choosing NC by Group of Scenario

## Sellers Choose Optimal Mechanism more in Scenarios B

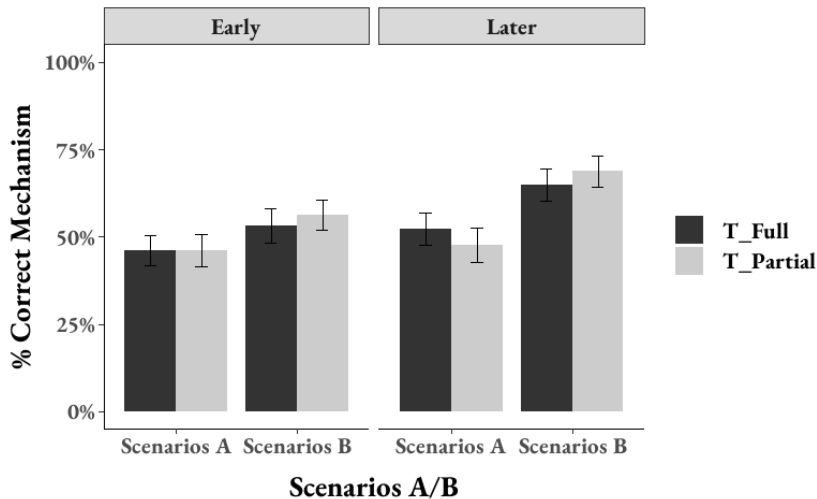


Figure 6: % of Correct Mechanism by Group of Scenario

## % of Choosing Correct Mechanism ↑

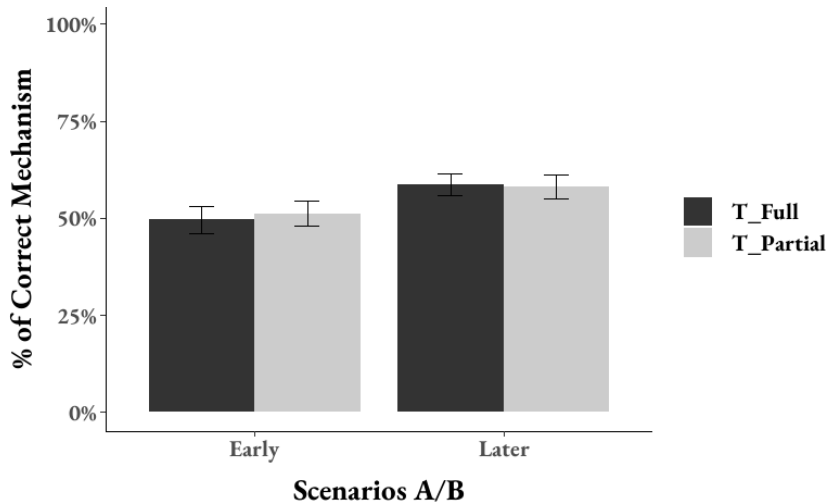


Figure 7: % of Choosing correct Mechanism

## % of Choosing Correct Mechanism $\uparrow$

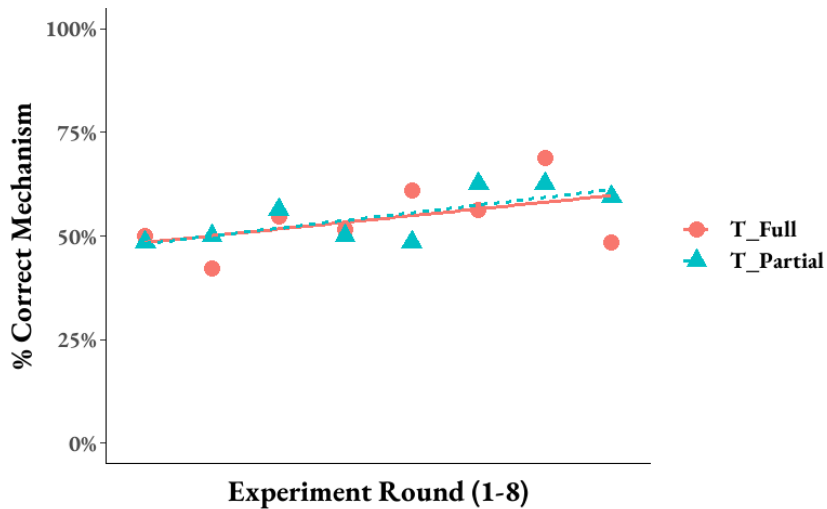


Figure 8: % of Choosing correct Mechanism

## Result 3. Past Experiences Matter

### R3.1. Sellers get less revenue from NC in Scenarios B.

- ▶ In Early stage: no difference from RS
- ▶ In Later stage: significantly less than RS ( $p < 0.01, p = 0.06$ ).
- ▶ In Scenarios A, Sellers do not get more revenue from NC.

### R3.2. Sellers choose NC less if past revenue from NC is low.

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

## Theoretical Revenue by Scenarios in Treatment (Full)

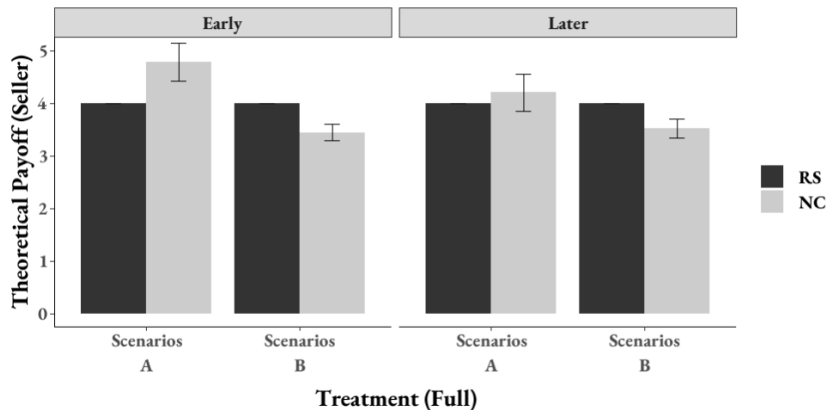


Figure 9: Theoretical Revenue by Scenarios in Treatment (Full)



### R3-1. Sellers do not Get More Revenue from NC in Scenarios A (Full)

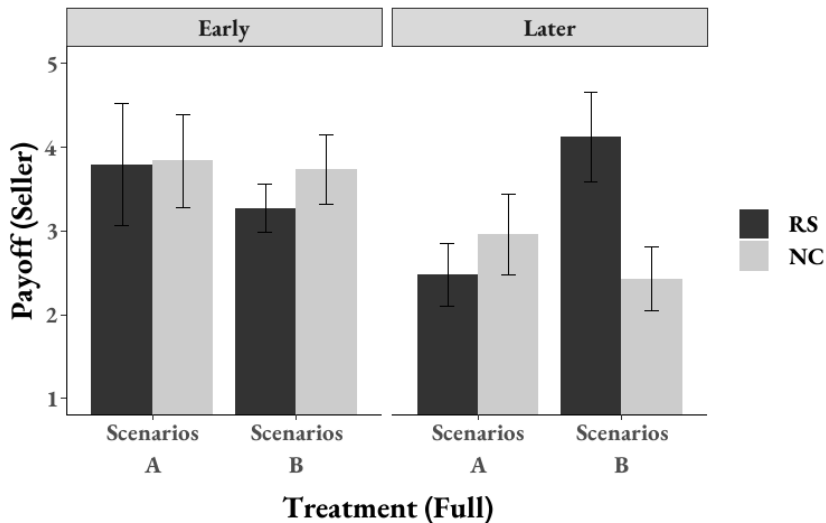


Figure 10: Seller's Payoff by Scenarios in Treatment (Full)

## Theoretical Revenue by Scenarios in Treatment (Partial)

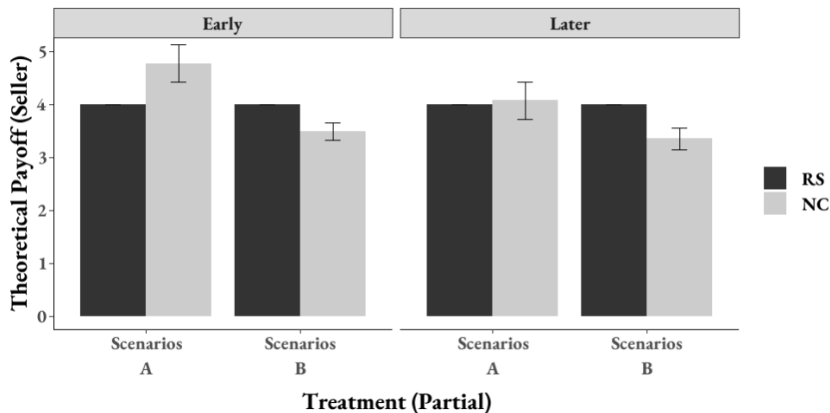


Figure 11: Theoretical Revenue by Scenarios in Treatment (Partial)

## R3-2. Sellers Get less Revenue from NC in Scenarios A (Partial)

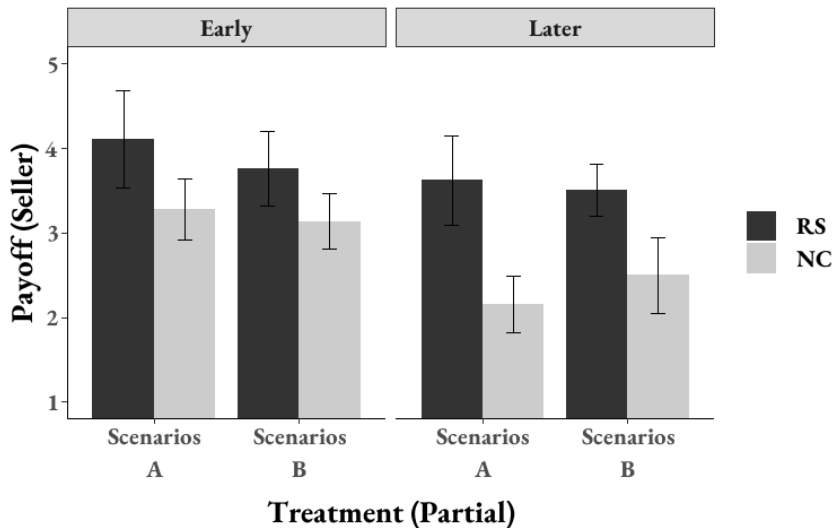


Figure 12: Seller's Payoff by Scenarios in Treatment (Partial)

# Why Sellers' Revenues are not improving?

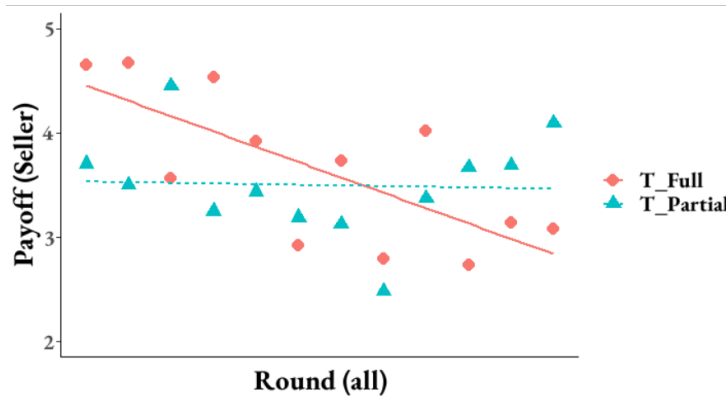


Figure 13: Seller's payoff

# 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.

## Higher Entry Fee in Period 2 in Treatment (Full)

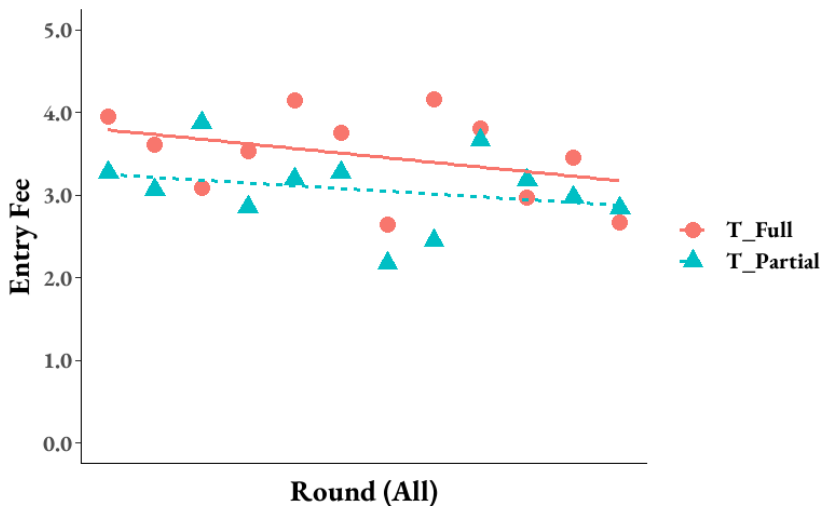


Figure 14: Entry Fee in Period 2

## Sellers set entry fee higher than suggested

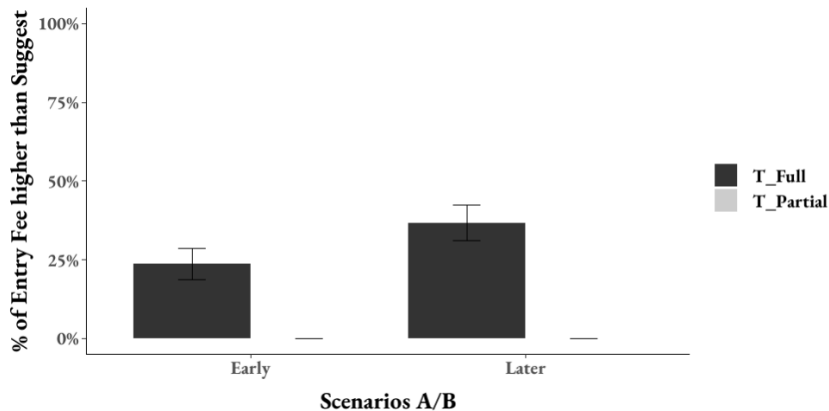


Figure 15: % of Setting Entry Fee Higher than suggested

## % of Entering Period 2

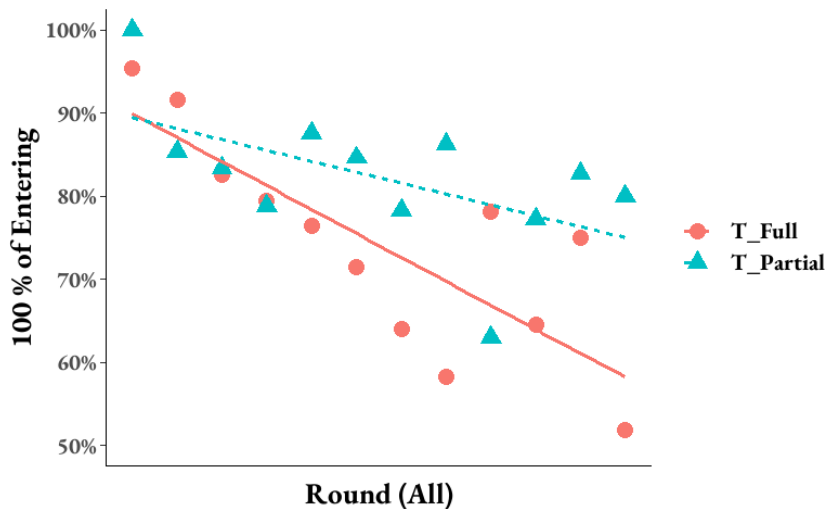


Figure 16: % of Entering Period 2



## High Entry Fee Deters Entering

	DV: Enter in Period 2	
	(1)	(2)
T_Partial	0.15 (0.16)	0.28 (0.17)
Entry Fee	-0.24*** (0.04)	-0.22*** (0.04)
Later	-0.40*** (0.11)	-0.44*** (0.12)
Scenarios B	-0.50*** (0.15)	-0.53*** (0.16)
Constant	1.90*** (0.23)	2.13*** (0.63)
Controls	No	Yes
Num. obs.	447	447

Table 2: Probit Regression of Enter in Period 2

## Bid/Value in Period 1 in Treatment (Full)

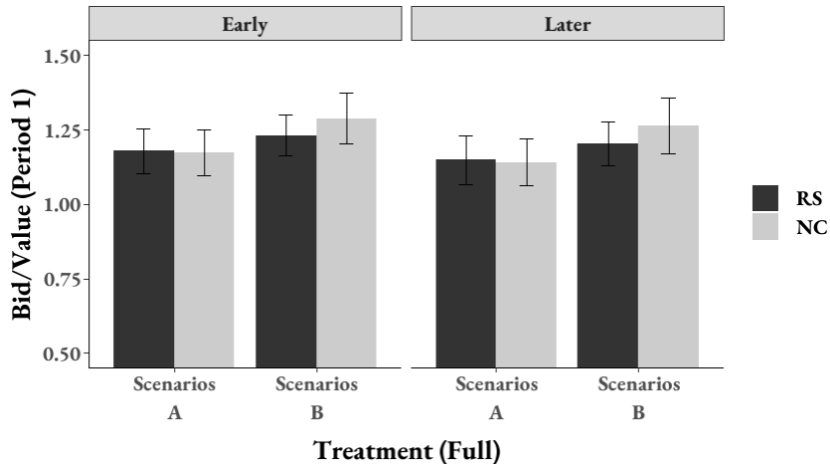


Figure 17: Bid/Value in Period 1 in Treatment (Full)

## Bid/Value in Period 2 in Treatment (Full)

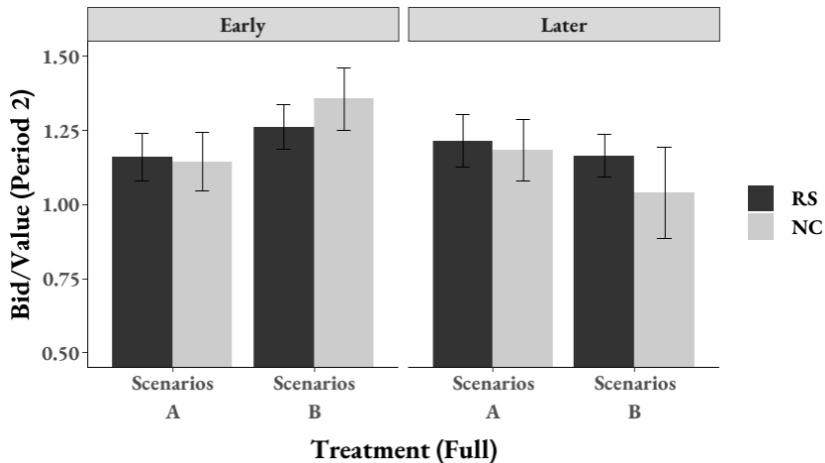


Figure 18: Bid/Value in Period 1 in Treatment (Full)

## R3.2: Sellers Choose NC less if NC Had Low Revenue in Last Round

	DV: Choosing NC	
	(1)	(2)
Last (payoff<3, NC)	-0.26*** (0.08)	-0.29** (0.12)
Later * Scenarios B	-0.18*** (0.06)	-0.25*** (0.09)
Later * Treatment(Partial)	-0.03 (0.06)	-0.12 (0.08)
Later * Last (NC)	0.18* (0.09)	0.35*** (0.13)
Later * Last (Correct = NC)	0.01 (0.10)	-0.11 (0.16)
Later * Last(Enter=1)	0.15 (0.13)	0.25 (0.19)
Controls	No	Yes
Adj. R <sup>2</sup>	0.04	0.06
Num. obs.	1024	1024

Table 3: Regression of Choosing NC

# Conclusion

## Current Conditions and Past Payoff matters in Choosing Mechanism

- ▶ Sellers can find optimal mechanism after gaining experience with the environment.
- ▶ Sellers abandon mechanism with low revenue.

## Discussion

- ▶ Sellers in real life adjust selling strategies as selling condition (or expectation) changes.
- ▶ Decision Support Pool: appropriate expectation on Buyers behaviors
- ▶ Experts: advice in setting (lower) prices.

*Thank you!*

# How do Sellers make decision?

Sellers set higher entry fee in Full Treatment.

- ▶ In Treatment (Full): Entry fee is higher in Early Stage ( $p = .03$ ).
- ▶ In Later stage: Entry fee is higher in Treatment (Full) ( $p = .04$ )
- ▶ Compared with RS, seller set high prices in NC.

## Higher Entry Fee in Period 2 in Treatment (Full)

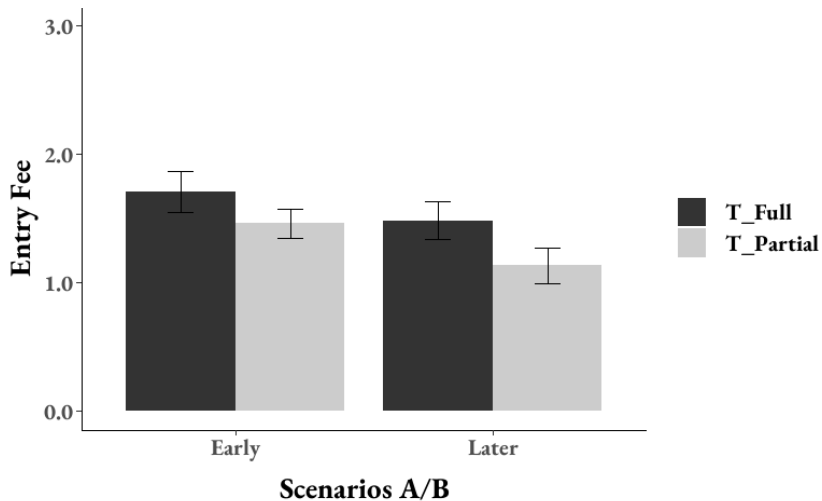


Figure 19: Entry Fee in Period 2



## Higher Reserve Price (Myersion) in Period 1 in Treatment (Full)

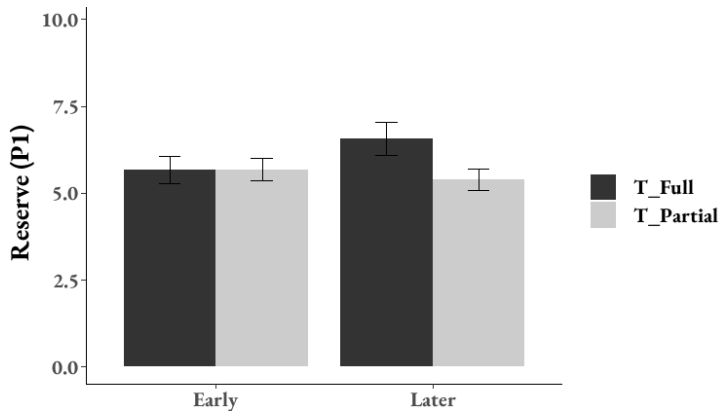


Figure 20: Reserve Price in Period 1

## Higher Reserve Price (Myersion) in Period 2 in Treatment (Full)

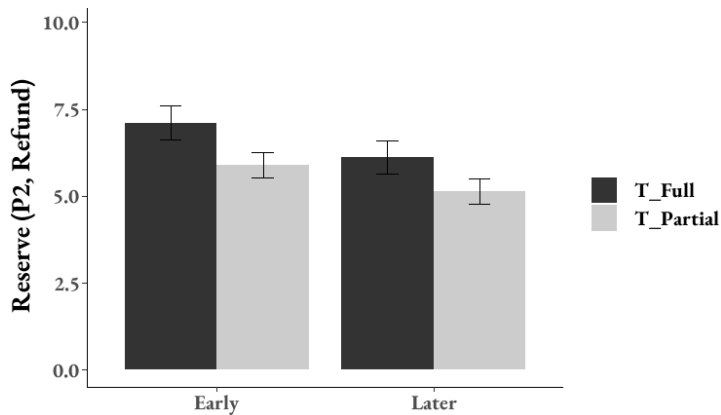


Figure 21: Reserve Price in Period 2 (Refund)

## Higher Reserve Price (Posted Price) in Period 2 in Treatment (Full)

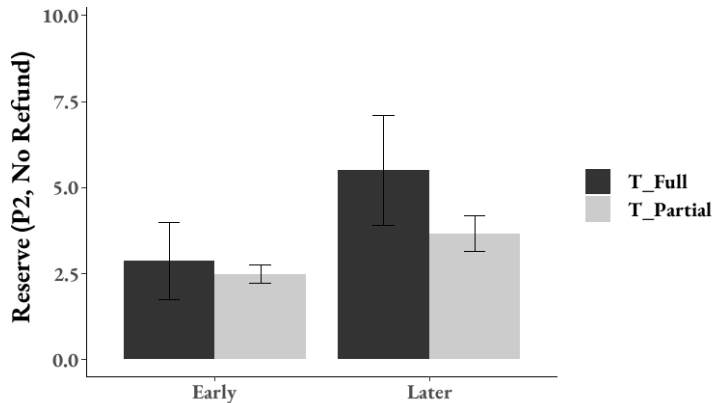


Figure 22: Reserve Price in Period 2 (No Refund)

## Seller's Payoff not Increasing

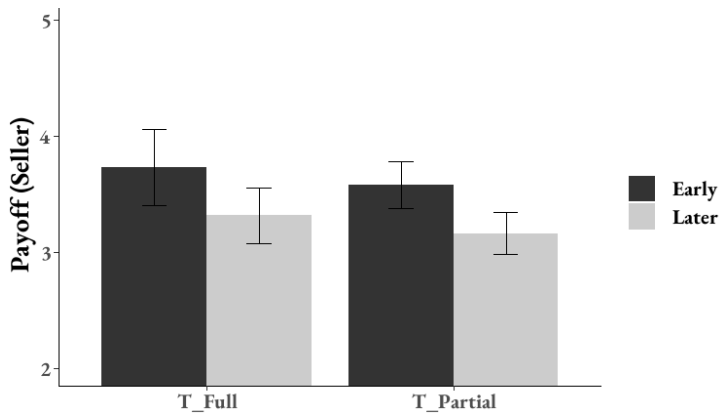


Figure 23: Seller's payoff

# Seller's Payoff Inconsistent with Theoretical Prediction

	DV: Seller's Payoff	
	(1)	(2)
Correct	-0.26 (0.32)	-0.16 (0.30)
T_Partial	-0.17 (0.20)	-0.13 (0.22)
Later	-0.41* (0.22)	-0.36* (0.21)
Scenarios B	0.06 (0.28)	0.02 (0.29)
Correct*Scenarios B	0.79** (0.39)	0.91** (0.41)
Value 1	0.29*** (0.05)	0.29*** (0.04)
Value 2	0.16*** (0.03)	0.17*** (0.03)
Controls	No	Yes
Adj. R <sup>2</sup>	0.23	0.26
N Clusters	128	128

Table 4: Regression of Seller's payoff

# How do Buyer make decision?

Buyers quit more in Later stage and in Scenarios B.

- ▶ Treatment (Full): significant more quit in Later stage ( $p = 0.02$ ).
- ▶ Treatment difference can be explained by high entry fee in Treatment (Full).
- ▶ Buyers quit more in Scenarios B.

Buyers do not overbid less in NC.

## Bid/Value in Period 1 in Treatment (Partial)

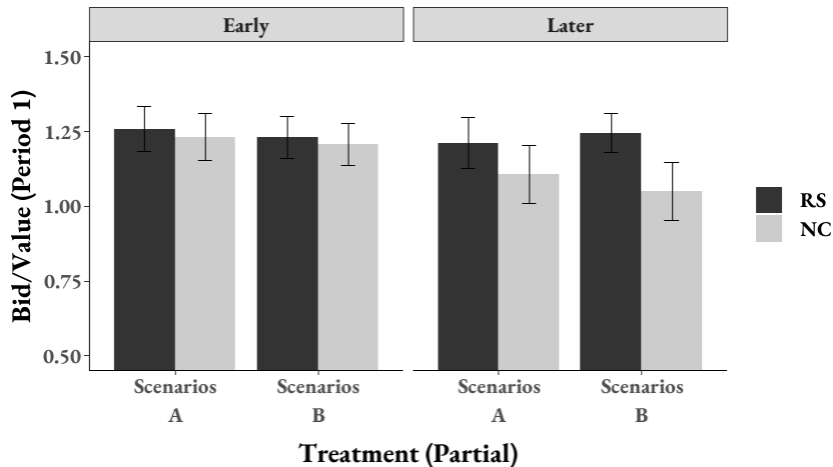


Figure 24: Bid/Value in Period 1 in Treatment (Partial)

## Bid/Value in Period 2 in Treatment (Partial)

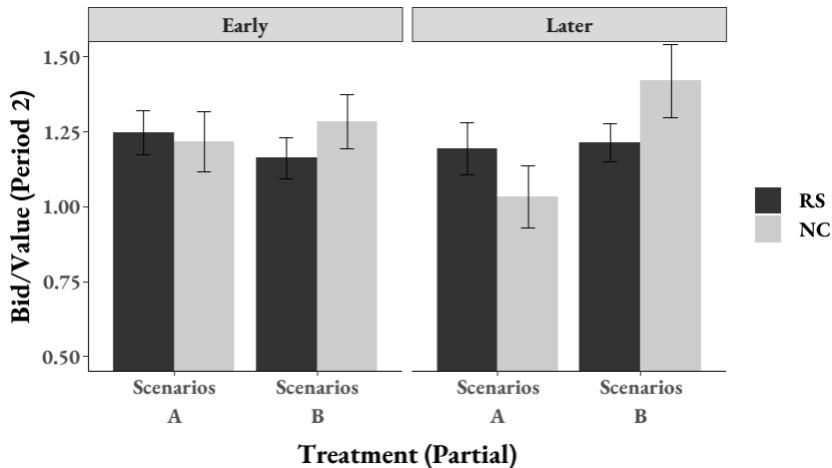


Figure 25: Bid/Value in Period 1 in Treatment (Partial)



# Buyer's Payoff

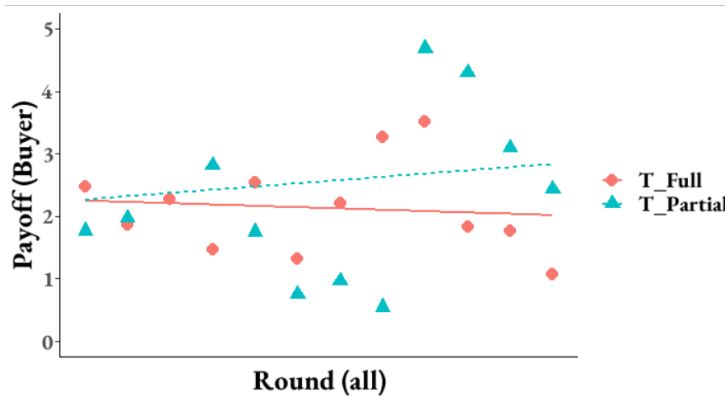


Figure 26: % of Entering Period 2

# Learning : Test Summary

Sellers learn the optimal mechanism in Later stage.

- ▶ In treatment (Partial): not significant ( $p = .06$ , one-sided t-test)  
51% correct in Early stage (vs. 0.5,  $p = .93$ )  
58% correct in Early stage (vs. 0.5,  $p < .01$ )
- ▶ In treatment (Full): significant ( $p = .02$ , one-sided t-test)  
48% correct in Early stage (vs. 0.5,  $p = .72$ )  
58% correct in Early stage (vs. 0.5,  $p = .01$ )
- ▶ No difference between treatment ( $p = .63$ ,  $p = .46$ , one-sided)