

Discrete clock auctions: an experimental study

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Abstract We analyze the implications of different pricing rules in discrete clock auctions. The two most common pricing rules are highest-rejected bid (HRB) and lowest-accepted bid (LAB). Under HRB, the winners pay the *lowest* price that clears the market; under LAB, the winners pay the *highest* price that clears the market. In theory, both the HRB and LAB auctions maximize revenues and are fully efficient in our setting. Our experimental results indicate that the LAB auction achieves higher revenues. This revenue result may explain the frequent use of LAB pricing. On the other hand, HRB is successful in eliciting true values of the bidders both theoretically and experimentally.

Keywords Clock auctions · Pricing rules · Market design · Experiments

JEL Classification D44 · C78 · L96

1 Introduction

A common method to auction diamonds, radio spectrum, electricity, gas, and other products is the discrete clock auction. The auctioneer names a price and each bidder responds with her desired quantity. If there is excess demand, the auctioneer then names a higher price. The process continues until there is no excess demand. This

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format is used for auctioning one or more products, where each product is either a single lot or many identical lots.

The clock auction is best thought of as a dynamic version of a sealed-bid uniform-price auction. In the uniform-price auction, the auctioneer collects a demand curve for each product from each bidder, forms the aggregate demand curve, and crosses it with the supply curve to determine the market clearing price and the quantity won by each bidder. The clock auction does the same thing, but gathers the demand curves from each bidder in a sequence of discrete rounds, and bidders receive information about excess demand at the end of each round. Indeed, the sealed-bid uniform-price auction is just a single-round clock auction. In both clock auctions and uniform-price auctions, two pricing rules are commonly used: highest-rejected bid (HRB) and lowest-accepted bid (LAB).

The LAB rule is used in treasury auctions in the US and elsewhere. These are sealed-bid uniform-price auctions. In contrast, clock auctions for diamonds, electricity and gas products in Europe have used the HRB rule.

Discrete rounds are used in practice to simplify communication, make the process robust to communication failures, and mitigate tacit collusion (Ausubel and Cramton 2004). An implication of discrete rounds is that the pricing rule matters. Another issue is whether the bidder specifies exit bids—the prices at which the bidder reduces the quantity demanded. In the limit as the size of the bid increment goes to zero, the distinction between pricing rules is irrelevant and exit bids are unnecessary. However, in practical auctions where the number of rounds often ranges from 4 to 10, discreteness matters. For example, in BHP Billiton's diamond auctions, to limit participation costs the auctions are scheduled to complete in a half-day, which entails about 4 to 10 comfortably paced rounds (Cramton et al. 2010). Similarly, medium-term electricity and gas auctions are scheduled to complete in one day to avoid overnight positions. These auctions typically last 6 to 12 rounds (Ausubel and Cramton 2004). In these settings, best practice is to have the bidder specify exit prices at which quantity reductions are desired. Exit bids make the discrete round auction much closer to a continuous auction and yet all the benefits of a discrete clock auction are retained. Exit bids are used in virtually all the major clock auctions in practice: diamonds, electricity, gas, and emission allowances (Ausubel and Cramton 2004, 2010; Cramton et al. 2010).

In this paper, we examine bidding behavior in HRB and LAB versions of discrete clock auctions. In each version, after each round the bidders learn the aggregate demand. To prevent bid-sniping, an activity rule requires that a bidder's quantity demanded cannot increase at higher prices. Bidders can only maintain or reduce quantity as the price rises. With many products, the simplest implementation is to require this monotonicity constraint for each product, as is done in diamond auctions. If the bidder reduces quantity in a round, the bidder names a price for each quantity reduction. The price of each reduction must be greater than the prior price and less than or equal to the current price. Each exit price is interpreted as the price at which the bidder is indifferent between the higher quantity and the lower quantity. If there is no excess demand at the current price, the supply is awarded to the highest bidders. The two versions differ in the pricing rule. Each winner pays the highest-rejected bid in HRB and the lowest accepted bid in LAB for the quantity won.

At first glance, it would seem that selecting the highest market clearing price (LAB) would result in greater revenue than selecting the lowest market clearing price

(HRB). The argument is incomplete, since the pricing rule influences behavior. LAB introduces an incentive for shading one's bid below value. In simple cases (assuming one-lot per product, symmetry, independence, and risk neutrality), the bid shading under LAB exactly offsets the revenue gain from selecting the higher clearing price. Revenue equivalence obtains, and the two pricing rules result in the same expected revenue—at least in theory when bidders seek to maximize profits.

The purpose of this paper is to examine the bidding behavior and outcomes, especially efficiency and revenue, under the two pricing rules experimentally. Our main hypothesis is that subjects will overbid under lowest-accepted bid, consistent with the first-price sealed-bid auction; whereas, under highest-rejected bid, bidders will bid truthfully. Thus, revenues under LAB will be higher than revenues under HRB. However, the efficiency comparison may go either way depending on how heterogeneous the bidding behavior in LAB is. Although the formats apply to the general case of auctioning many lots of multiple products, for simplicity we restrict attention to the case of auctioning a single item. This simple case is important in practice. For example, diamond auctions often sell large stones as individual lots, each with its own price clock. Many lots are sold simultaneously, but each lot is cleared separately and monotonicity of demand is enforced lot-by-lot.

It is a robust and puzzling finding of the experimental literature that in second-price sealed-bid auctions subjects bid more than their value although they bid truthfully in its dynamic counterpart, the English auction (Cooper and Fang 2008; Copping et al. 1980; Kagel et al. 1987; Kagel and Levin 1993). The HRB auction is somewhere between the English and second-price auctions and these two are the limit cases of HRB. HRB converges to the English auction (continuous clock) when the number of rounds approaches infinity. When there is only one round, HRB is the same as the second-price auction. The exit bid decision in HRB requires similar strategic thinking as in a second-price sealed-bid auction. On the other hand, whether to stay in the auction is the same binary decision as in the English auction. Harstad (2000) finds that experience with the English auction leads to less overbidding in second-price auctions. Moreover, he argues that the binary choice aspect of English auctions explains an important part of this learning. The HRB format allows subjects to make that kind of binary choice for early price increments and when they come to the price interval where they want to exit, they have this experience. In this respect, studying discrete clock auctions helps us understand what makes subjects overbid in second-price auctions and bid truthfully in English auctions.

The LAB pricing rule is analogous to a first-price sealed-bid auction. In particular, if there is only one round and one good, the LAB and first-price auctions are the same. Another robust finding of the experimental literature is that bidders in a first-price auction overbid compared to the risk-neutral Nash equilibrium prediction (see Cox et al. (1982, 1988), as the seminal papers; Kagel and Levin (2008) for a detailed survey).¹ Therefore, we expect to see overbidding in the LAB auction.

¹Risk aversion offers one explanation, but this has proven inadequate (see Kagel 1995). Several papers explain the overbidding phenomena with behavioral motives (for example, Goeree et al. 2002; Crawford and Iriberri 2007; Delgado et al. 2008; Engelbrecht-Wiggans and Katok 2007; Filiz-Ozbay and Ozbay 2007; Lange and Ratan 2009).

Ausubel et al. (2009) and Kagel and Levin (2001) are the most related papers to our work. Ausubel et al. (2009) compares a particular discrete clock with the sealed-bid auction in common value settings. The study considers reverse auctions where the bidders have liquidity needs. It tests whether dynamic or sealed bid formats lead to better price discovery (of the commonly valued goods) and identifies the format that allows bidders to satisfy the liquidity needs of the bidders. Kagel and Levin (2001) uses a continuous clock auction in a multi-unit setting; the aim of the study is to understand demand reduction in sealed-bid uniform-price and continuous clock auctions experimentally. Our study aims to understand the implications of different pricing rules in discrete clock auctions in private value settings.² No other paper to date makes this comparison.

Since we find experimentally that subjects do not deviate from the straightforward bidding strategy (bidding true value) in HRB, it is confirmed that dynamic formats make it easier for the bidders to recognize equilibrium strategies. Having a discrete clock makes HRB more practical than the English auction, and yet it still is successful in eliciting the true values in our experiment.

Our results indicate that revenues under LAB are significantly higher than under HRB. Thus, in settings where revenue is the predominant objective, the seller may favor LAB, but in settings where simplicity and gathering the true value of the bidders are of greater concern, then the seller may favor HRB. In terms of efficiency, we do not find any difference between the two formats experimentally.

In Sect. 2, we begin with a presentation of the theory. Equilibrium bidding strategies for the two versions of the discrete clock auction are characterized in Cramton and Sjarittanonta (2010). Here we summarize the results and provide the equilibrium bidding strategies for our experimental setting. The experimental design and the results are discussed in Sects. 3 and 4. Section 5 concludes.

2 Theory

There is one indivisible good for sale to $N > 1$ risk neutral bidders. Bidder i 's private value for the good is v_i where each v_i is independently drawn from a commonly known distribution F on $[\underline{v}, \bar{v}]$ with corresponding positive density function f . Assume that the hazard rate function associated with F is increasing. Bidder i 's payoff if she wins the good at a price p is $v_i - p$. Otherwise, it is equal to 0. The seller values the good at 0.

Before the auction starts, the seller announces a vector of bid levels, $P = (P_0, P_1, \dots, P_{T-1})$ where P_t is the price at round t and T is the number of bid levels. The clock price increases every round so that $\underline{v} = P_0 < P_1 < \dots < P_{T-1} = \bar{v}$. The auction begins in round one at a price P_1 .

²Although many auctioned goods have some common value characteristics, it is certainly the case that in electricity, gas, radio spectrum, and diamond auctions, each firm has private production and operation costs and may face different demands. We believe it is useful to begin with the private values case as it is also argued to be important in the literature. For example, Goeree et al. (2009) and Kagel and Levin (2001) study the issues related with spectrum auctions (such as demand reduction or preemptive bidding) in private value settings.

In each round t , each bidder chooses either to bid at the current clock price or to exit. The bidders who bid at the current clock price are called active bidders. At the beginning of each round, the bidders know the number of active bidders. When a bidder exits, she submits an exit bid. Once a bidder exits she cannot bid again. If more than one bidder stays in, the auction proceeds to the next round. If all but one bidder exit, then the bidder who stayed in wins the good. If all bidders exit, then the bidder who submitted the highest exit bid wins the good. The payments depend on the auction format.

2.1 Highest-rejected-bid (HRB)

In the HRB format, the final price is determined by the highest-rejected bid. If a bidder exits in round t , the bidder submits an exit bid—a price between P_{t-1} and P_t at which she wants to exit. If more than one bidder remains, the auction continues to another round; if all but one bidder exits, the remaining bidder wins and pays the highest exit bid; if all bidders exit, the bidder with the highest exit bid wins and pays the second-highest exit bid.

Proposition 1 *In the HRB auction, truthful bidding (bidding up to one's value) is a weakly-dominant strategy. The HRB auction is efficient and maximizes seller revenue.*

The dominant strategy result is standard and holds regardless of the number of bidders, the number of goods, or how values are drawn. All that is required is that each bidder demands only a single good. Thus, highest-rejected bid is the Vickrey price, thereby inducing truthful bidding. In our setting, efficiency is an immediate implication of truthful bidding. With the additional assumption of an increasing hazard rate, the optimal seller revenue can be achieved with the optimal reserve price. Nevertheless, among auctions with the same reserve price, the HRB auction yields the highest revenue due to the revenue equivalence theorem.

The HRB auction has extremely desirable properties in our setting. It is both efficient and maximizes seller revenues. Moreover, the bidding strategy is simple—just bid up to your true value—and is best regardless of what the other bidders are doing. Another important property of the HRB auction is that a bidder cannot lose at an affordable price as long as she bids her value. Hence, neither the winner nor the loser ever regrets having bid as they bid. The winner could not do better by exiting earlier; the loser could not do better by staying in longer.

2.2 Lowest-accepted-bid (LAB)

In the LAB format, the final price is determined by the lowest-accepted bid. If a bidder exits in round t , the bidder submits an exit bid—a price between P_{t-1} and P_t at which she wants to exit. If more than one bidder remains, the auction continues to another round; if all but one bidder exits, the remaining bidder wins and pays the current price; if all bidders exit, the bidder with the highest exit bid wins and pays her exit bid.

Table 1 Equilibrium strategy for LAB auction

Value	Exit round	Exit bid function
[50, 70)	1	$25 + \frac{1}{2}v_i$
[70, 90)	3	$35 + \frac{1}{2}v_i$
[90, 100]	5	$45 + \frac{1}{2}v_i$

Proposition 2 *In the symmetric equilibrium of the LAB auction, a bidder, who exits at P_t , bids $b(v_i, P_{t-1}) = E[Y|P_{t-1} < Y_k < v_i]$ where Y_k is the highest of $k - 1$ independently drawn values and k is the number of bidders present at P_{t-1} . The LAB auction is efficient and maximizes seller revenue.*

Although it is discontinuous, the equilibrium bid function in the LAB auction is monotonic in value. Therefore, the LAB auction is efficient in our symmetric setting. Similar to the HRB auction, the highest seller revenue in the LAB auction can be achieved with an optimal reserve price and among auctions with the same reserve price, the LAB auction yields maximum revenue. Nonetheless, one might favor the HRB auction because of its simple bidding strategy without bid shading. In the LAB auction, bidders must do a difficult equilibrium calculation to determine the optimal level of bid shading.

2.3 Theoretical predictions for the parameters used in the experiment

In our experiment, two bidders compete to buy a single good. Bidder i 's private value for the good is v_i where each v_i is independently drawn from the uniform distribution on [50, 100] and $\mathbf{P} = (50, 60, 70, 80, 90, 100)$.

By Proposition 1, in HRB the bidder submits her value as an exit bid.

By Proposition 2, in LAB the equilibrium strategy is given in Table 1. Any equilibrium exit should occur either in round 1, 3 or 5. Here is some intuition. Given the prior distribution, the equilibrium bid of a first price sealed bid auction is $25 + \frac{1}{2}v_i$. A bidder with a value higher than 70 would like to submit an exit bid higher than 60. Hence, any bidder with value less than 70 exits in round 1 and submits the exit bid $25 + \frac{1}{2}v_i$. Once, the price level increases in round 2, each active bidder updates her prior such that the values of the remaining bidders are 70 or more. Given the updated distribution (uniform between 70 and 100), the equilibrium bid is $35 + \frac{1}{2}v_i$. But this requires the lowest bid to be 70. Hence, no bidder drops out in the second round. Since no one drops out in round 2, nothing more is learned. In round 3, any bidder with a value less than 90 exits and submits the corresponding exit bid. Once, the price level increases to 90 in round 4, each active bidder updates that the lowest possible value should be 90. Since bidders with values higher than 90 would like to submit bids that are higher than 80, they will wait for round 5. Given the updated distribution (uniform between 90 and 100) the equilibrium bid is $45 + \frac{1}{2}v_i$. The active bidders wait for round 5 and submit the corresponding exit bids.

Ex ante symmetry—the fact that the bidders' values are drawn from the same distribution—is critical for Proposition 2. This allows for a symmetric equilibrium. Since the exit bid functions are the same and strictly increasing, the bidder with the

highest value wins and the outcome is efficient. Revenue maximization then follows from the revenue equivalence theorem. The assignment is the same as in the HRB auction, and both auctions give the bidder with a value of 50 a payoff of 0.

Ex post revenues differ between HRB and LAB. For each realization of values, the equilibrium revenue in HRB is equal to the second-highest value. However, the revenue in LAB, may not even be equal to the exit bid of the high value bidder. For a given price level, if one bidder stays in while the other one drops out, the winner pays the current price level rather than her intended equilibrium exit bid.

3 Experimental method

The experiments were run at the Experimental Economics Lab at the University of Maryland. There were 184 participants, all undergraduate students at the University of Maryland. The experiment involved two treatments: HRB and LAB. We conducted six sessions for each of the auction formats, HRB1 to HRB6 and LAB1 to LAB6. For each treatment, the numbers of participants in session one to six were 16, 14, 16, 16, 16 and 14, respectively. The random draws were balanced in the sense that we used the same sequence of random number “seed” values for each auction format, so the random value draws for HRB1 matched the random draws for LAB1. A new set of random draws was used for the second session in each format, etc.

No subject participated in more than one session. Participants were seated in isolated booths. Each session lasted about 80 minutes. Bidder instructions for each treatment are in the supplementary material. To test each subject’s understanding of the instructions, a subject had to answer a sequence of multiple choice questions. The auctions in a session did not begin until all the subjects answered all of the multiple choice questions correctly.

In each session, each subject participated in 21 auctions. The first auction was a practice auction. Each auction had two bidders, selected at random among the subjects. Bidders were randomly rematched after each auction. All bidding was anonymous. Bids were entered via computer. The experiment is programmed in *z-Tree* (Fischbacher 2007). At the conclusion of each auction, the bidder learned whether she won and the price paid by the winning bidder.

Bidders had independent private values for a fictitious good with values drawn uniformly from 50 to 100. The number of rounds was selected to approximate those of actual auctions. For example in diamond auctions for large stones, typically there are 4–6 rounds and increments are 10–20% of the price range. There were a maximum of five rounds in our discrete clock auctions and the price increments were 20% of the price range.

The winner in each auction earned her value minus the price paid in Experimental Currency Units (ECU). At the end of the experiment, total earnings were converted to US Dollars, at the conversion rate of 10 ECU = 1 US Dollar. Subjects also received a \$5 show-up fee. Cash payments were made at the conclusion of the experiment. The average subject payment was \$19.51.

Table 2 Outcomes of Treatment HRB and LAB

	HRB	LAB
Efficiency	91.09% (0.01)	90.87% (0.01)
Revenue per auction	67.03 (0.36)	69.91 (0.41)
Theoretical revenue	67.30 (0.40)	67.25 (0.36)

Standard errors are shown in the parentheses

4 Experimental results

As in any dynamic auction, bidding strategies of all bidders are not observable. For example, in the Dutch auction only the bid of the winner, or in the English auction only the bids of the losers are observable. Nonetheless, it is still possible to make revenue and efficiency comparisons, which are important in auction design. Later, we will study the bidding strategies based on the observable bids. Among 920 auctions we conducted per treatment, we observed 1092 and 1103 exit bids in HRB and LAB, respectively.

Table 2 shows the actual (what we observed in the experiment) efficiency and revenue, and the theoretical revenue (equilibrium revenue given the realization of values) of each treatment for the aggregate data. Both treatments, HRB and LAB, yield the efficient allocation with high frequency in each session (see Fig. 1) and nearly 100% of the gains from trade are realized in both treatments. Using the 6 independent sessions per treatment, the Mann-Whitney rank test shows no significant difference in efficiencies ($z = 0.40$, $p = 0.69$), but it shows a significant difference in revenues between treatments ($z = -2.88$, $p = 0.00$) (see Fig. 2). Furthermore, Mann-Whitney rank tests indicate that the actual revenue in LAB is significantly higher than the theoretical prediction ($z = -2.88$, $p = 0.00$) but there is no significant difference between the actual and theoretical revenues in HRB ($z = 1.44$, $p = 0.15$).

This evidence implies that subjects bid more aggressively than the equilibrium prediction in LAB while in HRB subjects bid close to the predictions of the theory. As a result the seller receives higher revenue in LAB. The bidding behaviors in HRB and LAB are investigated further in the next subsections.

One natural question is whether the relationship between actual and theoretical revenues in the two formats depends on the period or is persistent in the experiment. Figure 3 plots the average differences between the actual and theoretical revenues period by period. The actual revenues we observed in HRB are close to the prediction of the theory throughout the experiment. The regression of the difference between actual and theoretical revenues gives a significant coefficient for period and the constant is also significant. In HRB, actual revenue gets closer to the theoretical revenue: using the data from the six independent sessions of HRB, the Wilcoxon signed rank test shows that although the average difference between actual and theoretical revenues is significantly less than 0 in the first 10 periods ($z = -0.99$, $p < 0.05$), this difference is not significantly different from 0 in the last 10 periods ($z = 0.11$, $p = 0.92$). The actual revenues in LAB are persistently higher than the theoretical ones as can be seen in Fig. 3 and Table 3. In the regression for the difference between actual and

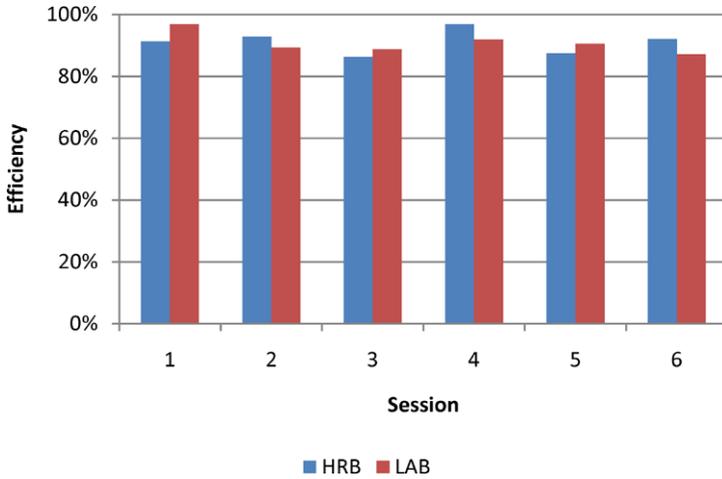


Fig. 1 Percentages of efficient allocations in each session

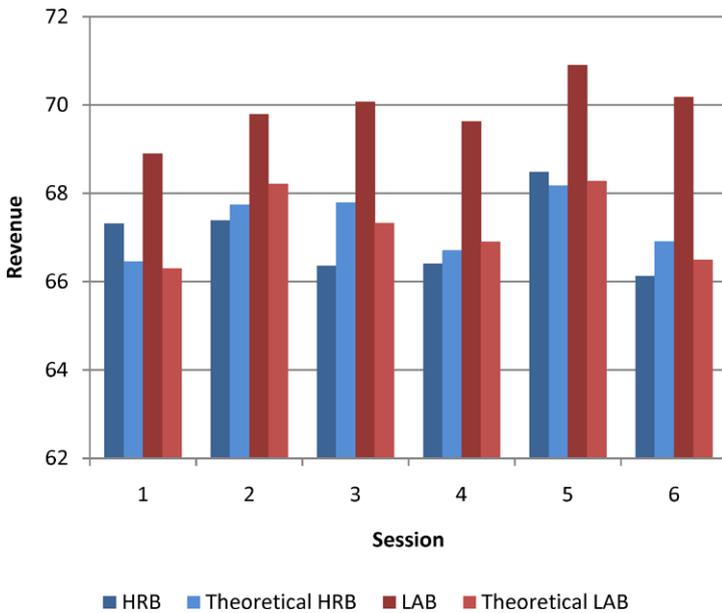


Fig. 2 Average actual and theoretical revenues in each session

theoretical revenues in LAB, only the constant was significant but not the coefficient of period. Moreover, the difference between actual and theoretical revenues is significantly higher than 0 both in the first 10 ($z = 2.20, p < 0.05$) and last 10 periods ($z = 2.20, p < 0.05$).

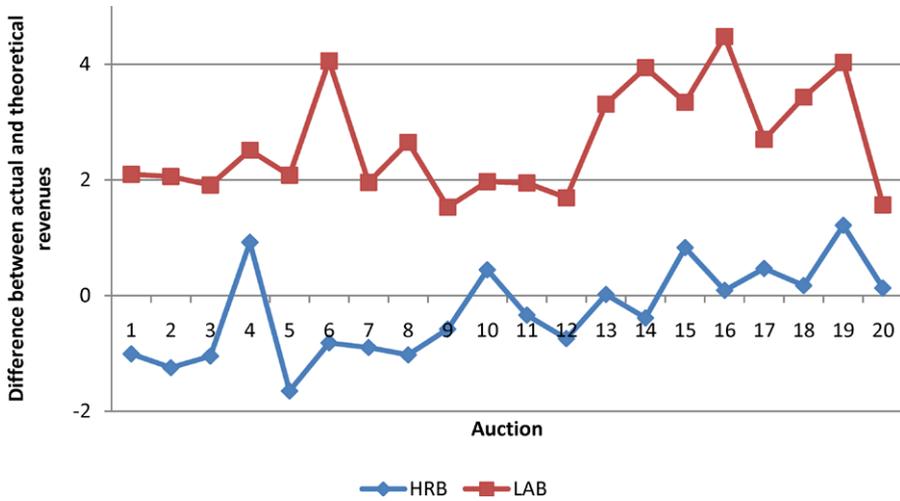


Fig. 3 Average differences between actual and theoretical revenues per period

Table 3 Regression of the difference between actual and theoretical revenues in HRB and LAB auctions

Independent variables	(HRB)	(LAB)
Constant	-1.2*** (0.347)	1.98*** (0.398)
Period	0.09*** (0.029)	0.06 (0.033)
Session fixed effect	YES	YES
<i>N</i>	920	920

*** Significant at 99% confidence interval. Standard errors are shown in the parentheses

4.1 Bidding behavior in treatment HRB

Subjects submitted a total of 1093 exit bids in 920 auctions. Figure 4 plots exit bids and corresponding values in HRB. The dashed line is the truthful bid function which is what the theory suggests. Although there were instances of bid shading and bidding above one's value, most of the exit bids were around the 45 degree line.

As shown in Fig. 5, 83.9% of exit bids were submitted in the round that the theory predicted and 13.6% of exit bids were submitted earlier than predicted. Among those who exited in the predicted round, 64.9% submitted exit bids within one percent of the exit bids suggested by the theory; 26.1% underbid and 9% overbid.

The vast majority of subjects were successful in waiting until the right round to exit. Since the binary decision of staying in or out is similar to the strategic thinking of the English auction, this finding is in line with the success of value bidding in English auction experiments. Harstad (2000) finds that subjects who gain experience in auctions with this kind of binary decision making perform better in discovering the value bidding strategy in second-price sealed-bid auctions. Staying in the HRB auction until the price reaches the correct level provides a similar experience as participating in an English auction.

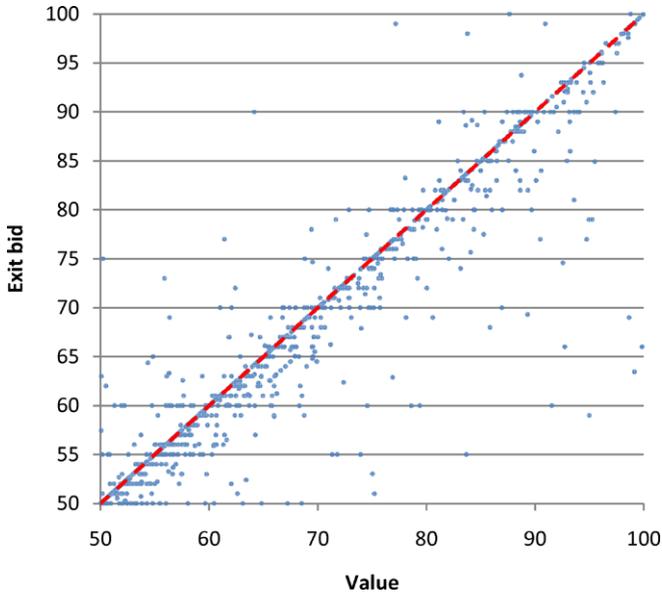


Fig. 4 Plot between values and exit bids in HRB (all data)

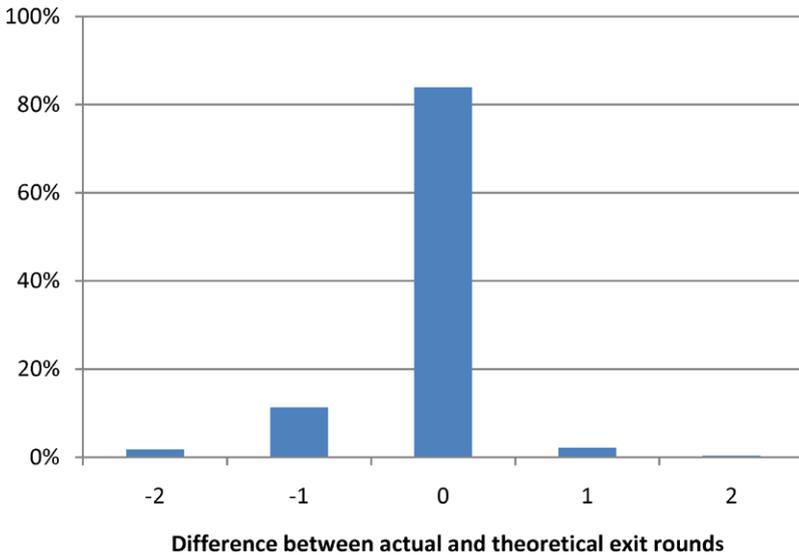


Fig. 5 Deviations from the predicted exit rounds in HRB

Fig. 6 Plot between values and exit bids in LAB

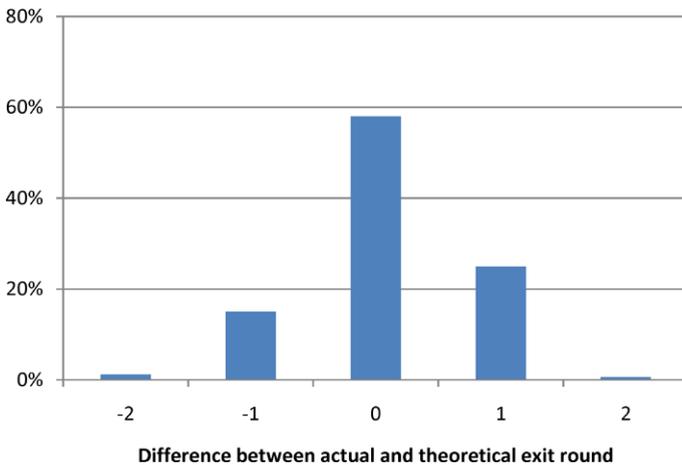
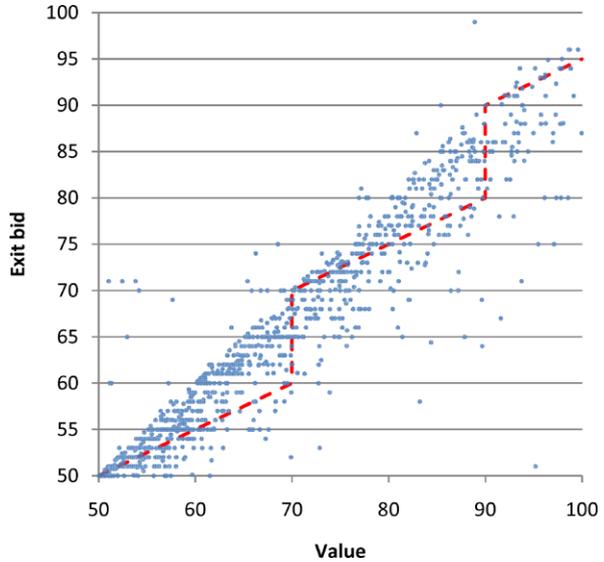


Fig. 7 Deviations from the predicted exit rounds in LAB

4.2 Bidding behavior in treatment LAB

A total of 1103 exit bids were submitted in 920 LAB auctions that are conducted. Figure 6 plots exit bids and corresponding values in LAB. The dashed line is the theoretical bid function as studied in Table 1.

Figure 7 shows that many subjects failed to exit in the round that the theory predicted. Only 58% exited in the correct round and 26% of the exit bids were submitted later than predicted. Even among the ones who exited in the predicted round, 44.4% overbid while 34.7% of those who exited in the predicted round did submit exit bids within one percent of the exit bids suggested by the theory and 20.9% underbid.

Table 4 Difference between actual and theoretical revenues by type of deviation

Type of upward deviation	Average difference between actual and theoretical revenues	Number of auctions
Winner exiting at the predicted round but submitting too high exit bid	1.44 (0.16)	71
Any bidder staying in for too long	9.85 (0.20)	260

Standard errors are shown in the parentheses

The high revenue in the LAB auction is caused by two types of upward deviations: (1) the winner submits a higher exit bid than predicted and (2) either bidder stays in too long. The first type involves a winner exiting at the round theory predicted and then submitting a too-high exit bid. The second type involves a winner, loser or both: a winner may drop out too late; or a loser who drops out too late extends the auction and thus raises the final price. Table 4 shows the difference between actual and theoretical revenues for each type of deviation. The relative effect of the second type of deviation is much stronger than the first. Furthermore, the second type of deviation was observed more frequently. These two types of deviation can take place in the same auction but there was only one auction with both types of deviation.

5 Conclusion

The pricing rule is of fundamental importance in practical auction design. Pricing impacts both the efficiency and the revenues of the auction. Although there is an immense literature on the pricing rule in static (sealed-bid) auctions—first-price vs. second-price in single unit auctions and pay-as-bid vs. uniform-price in multi-unit auctions—little is known about alternative pricing rules in the dynamic auctions commonly used in practice. We show how different pricing rules influence bidding behavior in discrete clock auctions in a simple setting.

Based on the standard theory in which bidders seek to maximize profits, the highest-rejected-bid (HRB) and the lowest accepted-bid (LAB) auctions both seem attractive. They maximize revenues and are fully efficient in our unit-demand setting. Despite this theoretical result, LAB pricing is often used in practice. In our experiments, both formats performed equally well and close to the theoretical predictions in terms of efficiency. However, the LAB auction yielded higher revenues than the HRB auction.

Although there is significant overbidding in LAB, this overbidding is not as severe as in LAB's limit case, the first-price sealed-bid auction. The dynamic auction, focusing initially on the binary in/out decision, appears to limit overbidding. This suggests an interesting tradeoff between price increments and revenues.

Developing strategies in HRB requires two things: when to exit as in an English auction, and what exit bid to submit as in a second-price sealed bid auction. Most bidders submit exit bids close to their values. The tendency of bidding above value found in second-price auctions is not observed here. We conjecture that this is due to

the experience bidders gain in deciding whether to exit each round. That reasoning process is similar to decision making in an English auction. Harstad (2000) finds that practicing English auctions improves performance in second-price sealed-bid auctions. Similarly, our experiment shows that the multiple-round implementation of the second-price auction eliminates the tendency to overbid when faced with second-price incentives. The dynamic implementation also limits the spread between the highest bid and the price to at most one bid increment.

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