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The Dollar Auction Game: A laboratory comparison between Individuals and Groups

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Abstract: By means of a laboratory experiment, this paper aims at studying how individuals and groups behave in a simple game such as the dollar auction. This game is extremely interesting since it induces subjects to fall prey into the paradigm of escalation, which is driven by agents' commitment to higher and higher bids. Indeed, whenever each participant commits himself to a bid, the lower bidder, moved by the wish to win as well as to defend his prior investment, finds it in his best interest to place a higher bid to overcome his opponent. The latter mechanism may lead subjects to overbid, implying that the winner pays more than the auctioned value. The aim of the paper is to analyze bidder's behavior, comparing individuals vs. groups' decisions within the dollar auction framework. We find that groups are closer than individuals to the Nash equilibrium, and that experience reduces the escalation phenomenon, but it has a different impact on winners and losers.

Key Words: escalation; winner's curse.

JEL code: C91

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1. Introduction

It is a matter of fact that many decisions are made by groups. Economic, legal and political decisions are recurrent examples. In this sense, whether or not decisions performed by groups are different or superior relative to decisions undertaken by individuals in isolation is still an open question¹. This paper contributes to the debate studying whether groups' decisions outperform those of individuals by exploiting the well-known framework of the "Dollar Auction Game", proposed by Martin Shubikin (1971). The game involves a promoter who auctions off a dollar to the highest bidder. The dollar is auctioned through a modified "English Auction": both the winner and the second-highest bidder have to pay their own bid, but only the highest bidder obtains the dollar. Whenever both players submit a bid, these rules create potential for a perverse mechanism in which players commit themselves to higher and higher bids in order to preserve their prior investment. Then, a mutually reinforcing behaviour might be in place, leading players to fall prev into the overbidding trap, and implying the paradox that one-dollar is sold for more than its value.

In some way the dollar auction game also resembles the winner's curse (Thaler, 1988). The latter is a wellknown behavioral bias in which decision makers are naïve and fail to behave rationally in the attempt to acquire an item. As a consequence, and against theory predictions, people systematically end up with loss-making purchases² in common value auctions.

In its simplicity, the dollar auction game provides a meaningful representation of many economic scenarios, for example that involving companies in a competition for acquiring oil extraction rights. One further example is that of the so called "Concorde Trap": after the Second World War, US, England, France and the Soviet Union reached an agreement regarding the creation of a supersonic airplane, the "Concorde". It would have been the first high-speed flight. Even though, due to a sharp increase in production and management costs, the Concorde project was loss generating, the involved States did not break out the venture since they had invested too much in it to leave the table.

The latter as well as further evidence³ show that people decision-making is affected by non-negligible sunk costs, which are reflected in the effort to continue an investment beyond the rationality of a representative profit maximize agent. The origins of sunk costs are not clear at all. Teger (1980) argues that the sunk costs and self-commitment are, in some way, related to the feel of having invested too much to quit. Thaler (1980) explains sunk costs in the light of the prospect theory (Kahneman and Tversky, 1979). Indeed, starting from a reference point, after an unsuccessful investment has been made, people shift somewhere in the convex trunk of their utility function. At that point, while losses do not result in significant decreases in value, gains would lead to large increases in value. Then, risking negligible losses to seek significant gains seems a good deal. Staw (1981) shows that further commitment after bad decisions does occur since people are averse to admit that prior money was wasted.

It is pretty evident as the dollar auction is a powerful game in that it creates potential for detecting and measuring the impact of sunk costs on subjects' decision making. This is the reason why this article adopts this game to contribute to the diatribe regarding groups and individuals decisions. Are groups more inclined than individuals to avoid overbidding and self-commitment to bad investments? In this sense, are groups' decisions superior relative to

 $^{^1\}mathrm{Comprehensive}$ surveys comparing group and individual decision making can be found in Charness and Sutter (2012) and Temerario (2014)

²eBay auctions, mineral right auctions and corporate takeovers are widespread examples.

³See Arkes and Blunder (1985)

individuals' decisions? To the best of our knowledge, this work is the first shedding light on individuals and group decisions in such a framework.

In a class experiment involving seventy people, Murnighan (2002) auctioned off a 20\$ bill. He reported that, after intensive subjects' activity, the last two bidders did not stop bidding at the break-even point (20\$) but they kept on bidding, with the aim of driving the other bidder out. At the end of the game, the winner paid 54\$ for a 20\$ bill. Murnighan (2002) also reported other very extreme class experiments in which the winner ended up with paying 2,000\$ for a 20\$ dollar auctioned value.

O'Neill (1986) shows that upper-bounded bids prevent subjects from falling prey into the escalation phenomenon, since they anticipate the contingency of incurring in a loss.

In a revisited version of the dollar auction game, Migheli (2012) finds that, although escalation does not occur, some participants are willing to pay more than the value of thecoin. He argues that, probably, loss-related costs are counter balanced by the "intangible reward of glory and fame" coming from winning the auctioned good.

As far as individual vs. team decision-making is concerned, a wide range of experimental games has been used to investigate whether groups perform better then individuals in isolation. Charness and Sutter (2012) and Temerario (2014) provide comprehensive surveys in this area. Results in this field are far from being convergent. Some studies (see for example Kocher and Sutter, 2005) show that groups outperform individuals in a wide set of situations. In line with this strand, Cooper and Kagel (2005, 2009) report that groups outcomes in strategic task are sharply better than those of the most skilled member of the group. Blinder and Morgan (2005), involving teams and stand-alone individuals in both a statistical urn problem and a monetary policy experiment, support the evidence that not only teams perform better than individuals but, surprisingly, groups decision making is not slower than that of individuals. Then, two heads are better than one. On the other side, some literature (see Kerr at al. 1996) reports no evident differences between teams and individuals decisions. Sutter et al. (2009) compare three-member groups and individuals decision in an English auction framework with private and common value. Contrarily to previously mentioned literature, they find that groups fall into the winner's curse trap more frequently than individuals, thus earning lower profits. Sutter et al. (2009) relate their achievement to a different approach of groups and individuals toward competition, arguing that competition among groups is more ruthless than competition among individuals. In a recent contribution, Casari et al. (2015) compare three-member groups and individuals' performance in an "Acquiring a Company" task. In order to track the main forces leading the different choices of groups and individuals, the team decision making process is split up into three steps: first, each subject presents an individual proposal; secondly, subjects go through a group chat step and, as a last step, the decision itself takes place. Moreover, the difficulty level⁴ of the task is changed to provide insights on when groups outperform individuals. Casari et al. (2015) show that results are crucially task dependent. While in the simple task groups perform better than individuals, since they reduce the winner's curse and place better bids than stand-alone individuals, in the difficult task individuals decisions are superior relative to those made by groups. This achievement is explained by the evidence that disagreement within a group was generally resolved with the median (and not with the best, i.e. the "truth wins rule") proposal. Then, in the easy task case, groups make better decisions just because the subjects with the wrong answer are the minority. This result provides the interesting evidence that the choice of having individuals or groups as decision unit is strongly context dependent.

⁴While in the easy level of the task themajority of subjects can solve the problem, in the difficult version only a minority can succeed.

Shupp and Williams (2008) evaluate risk aversion using price data elicited by a willingness to pay mechanism for risky prospects. They find that the variance of risk preferences is generally smaller for groups than individuals and the average group is more risk averse than the average individuals in high-risk situations. Morone and Morone (2014) estimated and compared subjects and dyads preferences toward risk⁵. In addition, Morone et al. (2012) show that subject choices in the first-period play a key role in determining subjects' behavior in the repeated game, so experience teaches something and the initial choice is crucial in tracking the pathway to follow, and overall, groups behave more rationally, in the sense that they were always closer to the Nash equilibrium.

Leavitt (1989) showed that collective decisions should be more efficient than individual ones. He recognizes three causes to support this idea: it satisfies the human's need of social membership, groups seem to be more creative than individuals and they are able to correct their mistake, putting toghether different information. In the same year, the psychologist Irving L. Janis came up with a different pattern. He showed that sometime groups convey to a sub-optimal equilibrium caused by impulsive choices. He called his theory "Groupthink". It is a phenomenon in which the need for agreement and conformity in the group results in an irrational decision-making outcome; this behaviour can be particularly dangerous if coupled with escalation phenomenon.

In the next section we report the experimental design, in section 3 some theoretical background, and our results are presented. Finally section 4 concludes.

2. Experimental design and lab procedure

The experiment was conducted at the ESSE laboratory in Bari and programmed in z-Tree (Fischbacher, 2007). 48 subjects participated to the experiment, and they were randomly allocated to two different treatments, i.e. an individuals' treatment (IT) and a two-member groups' treatment (GT). 32 subjects played the group treatment. They were randomly gathered into 16 two-member groups and randomly matched with the same opponent during the whole lab session. Then, eight different auctions took place in the GT. 16 subjects got involved in the IT. Each of them was randomly matched with an opponent and played the whole auction session in isolation against the same opponent. Then, also in the IT, eight different auctions were played. The rules wereas follows: the highest bidder obtained 10 ECU⁶, the lowest bidder lost and had to pay his latest bid. The auction was initially opened for 30 seconds; whenever each subject posted a bid, the time auction was restored for other 30 seconds and, if nobody raised up his bid until the end of the 30 seconds, the auction was stopped and players had to pay according with their latest bids. In each treatment subjects played the game over 10 periods. Thereafter, one period was randomly selected for payment; the exchange rate was 1 ECU to $0.1 \in$. At the end of each lab session, all participants filled in a questioner. The

3. Theoretical aspects and Experimental Results

The auction is set up as a sequential game. The Nash Equilibrium is reached respectively whenever the faster bidder offers the minimum request to obtain the euro (i.e. 0.1 ECU) and the opponent leaves the auction. Whether subjects rise up their bids, they will lose the track of their losses.

⁵ For a thoughtful survey see Temerario (2014).

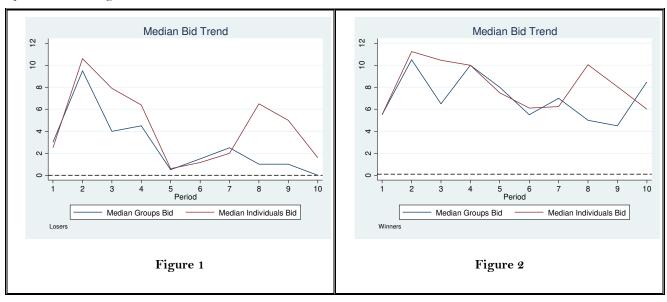
⁶ The ECU (Experimental Currency Unit) is the currency used to conduct the experiment.

Our research pursues a dual scope. As a first point, we aim at testing whether the sub-game perfect Nash equilibrium (SPNE) solution of the game holds. Since the game theoretical optimal strategy is different for losers (L) and winners (W), our analysis keeps the two categories separated. Thereafter, within each category, we focus on both groups and individual decision-making. The second goal is to assess whether groups and individuals' choices do rely on different underlying criteria, i.e. whether group decisions are on average superior to individual decisions⁷.

Following the above schedule, we first test whether our experimental data confirm the theoretical Nash equilibrium prediction of the game. Tables B1, B2, B3, B4 in Appendix B report the bid distribution for each dyad over the ten periods. Data are grouped by winners and losers and, within each of the latter categories, by groups and individuals. Accordingly, descriptive statistics are illustrated in Box-Plot C1, and Box-Plot C2 in Appendix C. As far as losers are concerned, a one sample t-test (n = 80) is performed to assess whether their bids are, on average, equal to 0 ECU. Similarly, as far as winners are concerned, the one sample t-test (n = 80) is carried out to test whether winners' proposals are, on average, equal to the minimum allowed value, i.e.0.1 ECU. The analysis is worked out for both groups and individuals. The results show that theory fails. Indeed, both losers and winners' average bids are significantly diverse from the respective theoretical prediction. This achievement holds⁸ for both stand alone and dyad choices (LGB: t = 7.45, *** p<0.01; LIB: t = 6.51, *** p<0.01; WGB: t = 12.64, *** p<0.01; WIB: t = 8.70, *** p<0.01).

Result 1: Both groups and individuals do not behave in accordance with the theoretical prediction of the game.

Anyway, our first result pushes us further to investigate if subjects' decisions differ depending on whether people play individually or in groups of two elements. In Figure 1 and Figure 2 we report the period-by-period median bid for both losers and winners, controlling for individual and group play. The black dash line represents the theoretical prediction of the game.



⁷ This evidence is largely supported in the related literature. See, for example, Blinder and Morgan (2005)

⁸ Full t-tests are reported in Appendix D

Interestingly, we can see that, both in the losers and winners instances, groups median bids are most of the times lower and, hence, closer to the Nash equilibrium than individuals median bids. We perform a two-sample-mean comparison test (n = 80, see Table E1 and E2, Appendix E) to assess whether subjects' bids are, on average, lower when the game is played in groups. The results are affirmative. Taking into consideration losers bids, we find that groups bids are, on average, significantly lower than individuals' bids (** p<0.05). The same result is detected when winners' bids are accounted for (** p<0.05).

Furthermore, by employing the root mean squared error (RMSE) index, we test the error margin between subjects' bids and the theoretical prediction of the game. The employed measure computes the average squared deviation of actual bids over the theoretical equilibrium value. In other words, the RMSE is used as an index of proximity of the observed bids from theoretical value.

$$RMSE = \left[\frac{1}{n}\sum_{i=1}^{n} (B_{o} - B_{N})^{2}\right]^{\frac{1}{2}}$$

where:

- B_0 represents the observed bid
- B_{N} represents the Nash equilibrium predicted bid
- *n* stands for the number of observations within each period

Then, the more actual bids approximate the equilibrium value, the lower the RMSE is.

Accounting for losers, Figure 3 shows the RMSE distribution box-plot of actual bids toward the equilibrium bid. As we can note, when the game is played in groups, the RMSE distribution is downward shifted. Then, groups bids track the theoretical equilibrium closer than individuals bids. A non-parametric Mann-Whitney U test (n = 10, table 1) shows that this difference is statistically significant (* p<0.1).

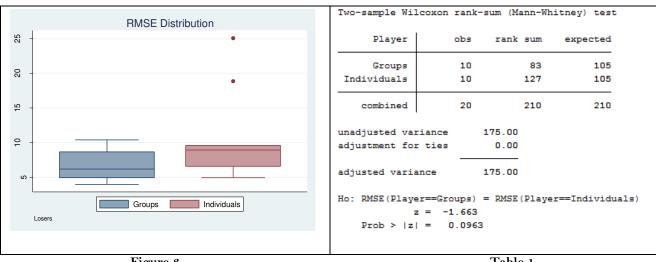
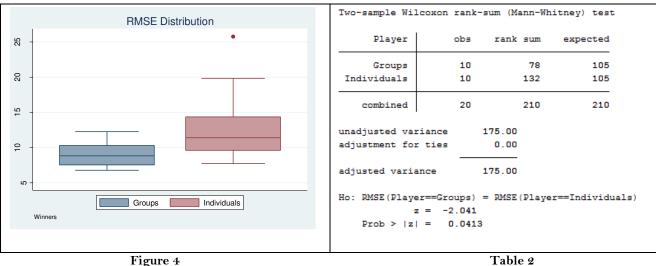


Figure 3

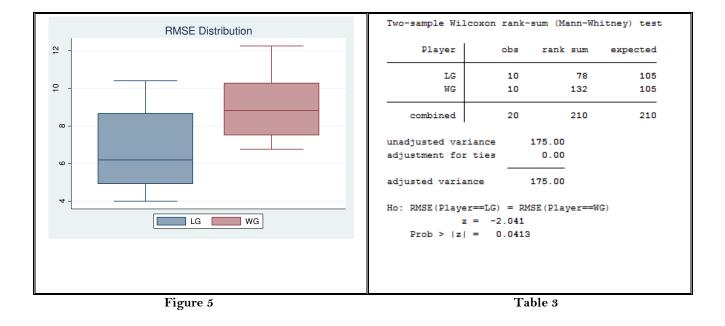
Table 1

Similarly, as far as winners are concerned, figure 4 shows that groups behaviour approximates the theoretical prediction much better than individuals' behaviour. A non-parametric Mann-Whitney U test (n = 10, table 2) shows that this result is statistically significant (** p<0.05).



Result 2: In both the winners and losers instances, groups bid closer to the theoretical equilibrium than individuals.

In Figure 5, Table 3 - Figure 6, Table 4 we report the comparison between losers vs. winners in the two treatments. Indeed losers are closer to the Nash equilibrium in both treatments. This result is in line with the winner's curse.



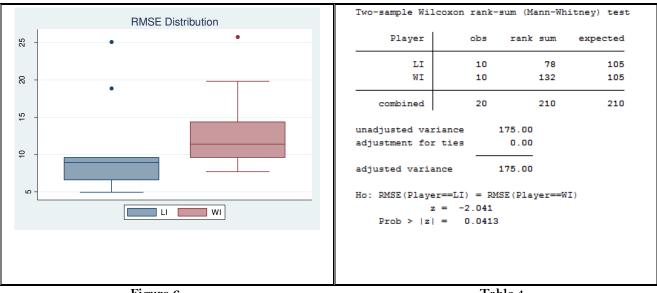


Figure 6

Table 4

At the end of the treatment, subjects answer to three simple questions: (1) "Have you offered more or less than one euro?"; (2) "Why do you follow this pattern?"; (3) "Does experience lead you to switch to another strategy?". We report information gathered in the questioner into a "statements cloud" (see Figure 7), in order to put in evidence the key words and the most meaningful statements. They report that inexperience is crucial in the first periods because they are unaware about the mechanism of the game and the strategy of their opponents. Finally, in some cases a learning process slowed down the size of bids and cleaned their decision-making process up, but sometimes subjects took a wrong position. A sunk cost effect (in most cases they reveal that they are moved by their wish to minimize the losses) blind the right equilibrium.

The helps us to change our strategy To reduce our losses we gave up because profits were being Fédicéd we never opponent To reduce our losses the wish to reduce losses, but inexperience led us to offer more than 10 ECU to the tempore my gate We directed that our opponent knew the mechanism of the game We were conditioned by time we have the state of the strategy We overbid because in this way we reduce our losses for more than to ECU, the test is you win the auction. We play moved by our wishes to reduce losses for more than to ECU, the test is test if you win the auction. We needled our strategy according to our opponent's behavior We medified our strategy according to our opponent's behavior we medified our strategy according to our opponent's behavior

Figure 7: statements cloud

4. Conclusion

In this paper we compared individuals and groups behavior in a particular auction, i.e. the dollar auction. We got two results; (i) both groups and individuals do not behave in accordance with the theoretical prediction of the game, but (ii) groups bid closer to the theoretical equilibrium than individuals. The escalation of commitment is a topic that regards our daily routine and each decision making process, indeed a great part of our choice is influenced by a sunk cost effects. It is difficult to find a break event point between the cost of our past and the actual benefits of our future gains (or losses). However a collective mechanism takes to a gradual dimming of this problem and it partially succeeds to avoid the paradox of the game.

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Appendix A

Box A1: Instructions

In this experiment you take part in an auction to win 1 euro. You compete against one opponent. The auction rules are as follows:

-The promoter auctions off 10 ECU (Experimental Currency Unit);

-The auction follows a competitive mechanism, it will start from a price of 0 ECU;

- In order to win the auction you have to submit at least one bid;

-You have 30 seconds to raise up your latest bid, you can raise your bid by 0.1 ECU at a time, there are no upper limits.

-If you do not raise up your bid within 30 seconds since your opponent's bid has been posted, the auction ends. Both the winner and the loser have to pay their latest bid to the auctioneer, but only the winner obtains 10 ECU. You will play the auction 10 times (periods) against the same opponent. Thereafter, one period will be randomly selected and you will be paid what you earned in that period. The exchange rate is 1 ECU = 0.1 €.

At the end of the experiment you will be asked to fill in a questionnaire.



A screenshot of the experiment layout.

Box A2: Questioner

Questions:

"Have you offered more or less than one euro?"

"Why do you follow this pattern?"

"Does experience leadyou to switch to anotherstrategy?"

Most meaningful answers:

"Replication over periods helped us to change our strategy."

"We never bidded more than 8 ECU."

"We met a risk averse opponent, then we never over-bidded."

"We over-bidded because of our opponent's strategy."

"We wished to reduce our losses, but inexperience led us to offer more than 10 ECU."

"We have been influenced by our opponent raising his bids"

"We played in order to reduce our losses."

"From time to time we changed our strategy."

"We never changed our strategy."

"We understood that our opponent knew the mechanism of the game."

"We over-bidded because, in this way, we reduced our losses."

"We played moved by thewish to reduce our losses."

"We were conditioned by time."

"We modified our strategy according to our opponent's behavior."

"We needed more attempts to understand the game."

"Experience helped us to improve ur strategy."

"We never changed our strategy."

"I bidded to increase my gain."

"The other player realized that, if he had bet 9 ECU and I had replied with 10 ECU, it would have been cheaper to lose 1 ECU raising up his bid to 11 ECU than losing 9 ECU leaving the auction."

"I played moved by the purpose to earn more, paying the promoter as few as possible."

"For more than 10 ECU, the loss is lower if you win the auction."

"I observed my opponent's attitude to commit himself to higher and higher bids, for this reason I stopped bidding when my losses were small."

"My aim was to win in order to avoid losses."

"My strategy consisted in bidding as few as possible to tone down losses, but my opponent did not accept it."

Appen	dix	B
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			Loser	rs Groups Bids (LGB)			
Period	Dyad 1	Dyad 2	Dyad 3	Dyad 4	Dyad 5	Dyad 6	Dyad 7	Dyad 8
1	6	6	3	0	10	1	3	2
2	3	14	10	0	11.1	12	4	9
3	5	8	5	3	12.2	2	0	0
4	5	22	10	2	3.6	4	15	0
5	5	1	11	0	0	2	0	0
6	5	1	20	2	0	0	1	13
7	7	1	4	12	0	0	0	18
8	6	1	8	0	0	1	1	12
9	1	1	11	2	0	0	1	0
10	7	1	17	0	0	0	0	0

Table	B1
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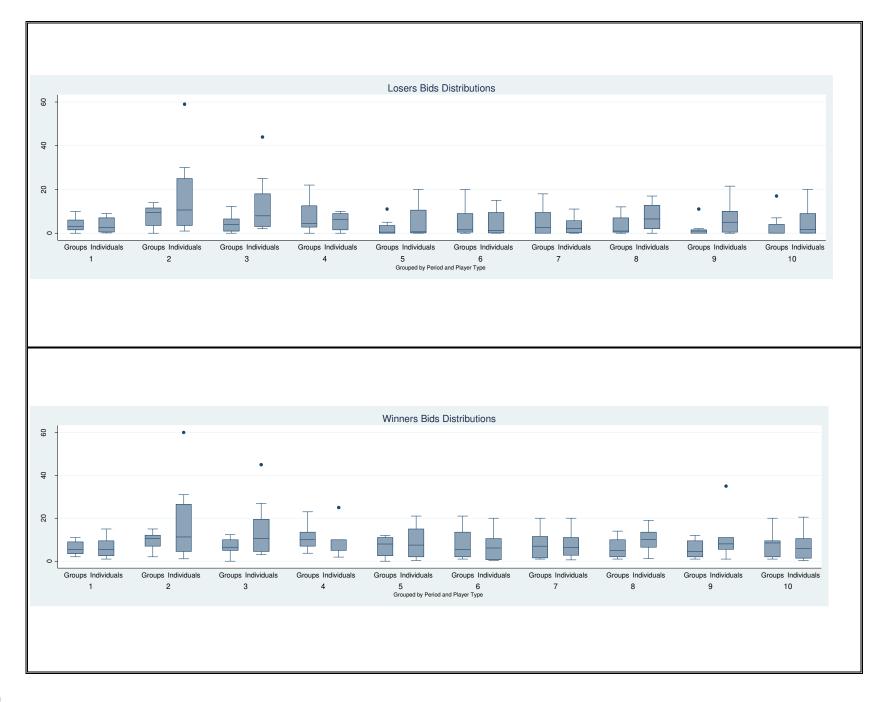
			Losers	Individuals Bid	s (LIB)			
Period	Dyad 1	Dyad 2	Dyad 3	Dyad 4	Dyad 5	Dyad 6	Dyad 7	Dyad 8
1	0.1	9	8	1	2	3	0.1	6
2	59	30	4	20	10	3	1	11.2
3	44	11	2	2	4	5	10.8	25
4	1.9	0	9	6.8	6	9	1.1	10
5	0.2	0	4	17	20	1	0.2	0
6	0.3	0	0	15	10	9	0.2	2
7	0.5	0	2	0	2	11	2	9.4
8	11	0	3	10	3	17	14.5	1
9	10	0	1	21.5	10	4	0	6
10	0.1	0	10	20.1	0	8	1.5	1.7

			Winner	rs Groups Bids	(WGB)			
Period	Dyad 1	Dyad 2	Dyad 3	Dyad 4	Dyad 5	Dyad 6	Dyad 7	Dyad 8
1	7	8	4	2	11	10	4	3
2	4	15	11	2	11.2	13	10	10
3	6	10	6	7	12.3	4	0	10
4	6	23	11	10	3.7	10	16	8
5	6	4	12	10	1	10	0	12
6	6	2	21	5	1	10	2	17
7	8	2	6	13	1	10	1	20
8	7	2	10	3	1	10	2	14
9	5	2	12	4	1	9	2	10
10	8	2	20	2	1	9	9	10

Ta	ble	B3
	~	~ ~

			Winners	Individuals Bio	ls (WIB)			
Period	Dyad 1	Dyad 2	Dyad 3	Dyad 4	Dyad 5	Dyad 6	Dyad 7	Dyad 8
1	15	10	9	2	3	4	1	7
2	60	31	5	22	11	4	1.1	11.5
3	45	12	3	3	10	6	10.9	27
4	1.9	10	10	10	8	10	2	25
5	0.3	10	5	20	21	3	1	10
6	0.4	10	1	20	11	10	0.5	2.2
7	0.6	10	3	20	3	12	2.5	9.5
8	12	10	9	10.1	4	19	15	1.2
9	11	10	5	35	11	6	1	6.1
10	0.2	10	11	20.5	1	10	2	1.8

Table	B4
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Appendix C

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Appendix D

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One-sample t test
                   Mean Std. Err. Std. Dev. [95% Conf. Interval]
           Obs
Variable
          80 4.54875 .6098765 5.454901 3.334821 5.762679
   LGB
                                                    t = 7.4585
   mean = mean(LGB)
Ho: mean = 0
                                       degrees of freedom = 79
                          Ha: mean != O
   Ha: mean < O
                                                  Ha: mean > 0
 Pr(T < t) = 1.0000 Pr(|T| > |t|) = 0.0000 Pr(T > t) = 0.0000
One-sample t test
         Obs Mean Std. Err. Std. Dev. [95% Conf. Interval]
Variable
  LIB
          80 7.1775 1.101518 9.852276 4.984983 9.370017
   mean = mean(LIB)
                                                     t = 6.5160
Ho: mean = 0
                                       degrees of freedom =
                                                           79
   Ha: mean < O
                          Ha: mean != 0
                                                    Ha: mean > 0
Pr(T < t) = 1.0000 Pr(|T| > |t|) = 0.0000 Pr(T > t) = 0.0000
 One-sample t test
 Variable
           Obs
                    Mean Std. Err. Std. Dev. [95% Conf. Interval]
                    7.54 .588429 5.263069 6.368761 8.711239
   WGB
            80
   mean = mean(WGB)
                                                    t = 12.6438
 Ho: mean = 0.1
                                       degrees of freedom =
                                                           79
  Ha: mean < 0.1
                         Ha: mean != 0.1
                                                 Ha: mean > 0.1
                     \Pr(|T| > |t|) = 0.0000 \qquad \Pr(T > t) = 0.0000
 Pr(T < t) = 1.0000
One-sample t test
Variable
            Obs
                    Mean Std. Err. Std. Dev. [95% Conf. Interval]
   WIB
            80
                  9.97875 1.134557 10.14779
                                               7.72047 12.23703
   mean = mean(WIB)
                                                   t = 8.7071
Ho: mean = 0.1
                                       degrees of freedom =
                                                            79
  Ha: mean < 0.1
                         Ha: mean != 0.1
                                                 Ha: mean > 0.1
                     Pr(|T| > |t|) = 0.0000
 Pr(T < t) = 1.0000
                                                Pr(T > t) = 0.0000
```

Appendix E

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf.	Interval]
LGB	80	4.54875	. 6098765	5.454901	3.334821	5.762679
LIB	80	7.1775	1.101518	9.852276	4.984983	9.370017
combined	160	5.863125	.6361568	8.046818	4.606718	7.119532
diff		-2.62875	1.259083		-5.120973	1365267
diff =	mean(LGB)	- mean(LIB)			t	= -2.0878
Ho: diff =	• 0		Satterthwai	te's degrees	of freedom =	= 123.274
Ha: di	ff < 0		Ha: diff !=	0	Ha: d:	iff > 0
Pr(T < t)	= 0.0194	Pr(T > t = 0	0.0389	$\Pr(T > t)$	= 0.9806
			Table E1			
Two-sample	t test wi	th unequal v				
Two-sample Variable	t test wi Obs	th unequal v Mean	variances		[95% Conf	. Interval]
			Variances Std. Err.	Std. Dev.	[95% Conf 6.368761	
Variable	Obs	Mean	Variances Std. Err.	Std. Dev.		
Variable WGB	රාය 80	Mean 7.54 9.97875	Variances Std. Err. .588429	Std. Dev. 5.263069 10.14779	6.368761 7.72047	8.711239 12.23703
Variable WGB WIB	Obs 80 80	Mean 7.54 9.97875 8.759375	Zariances Std. Err. .588429 1.134557	Std. Dev. 5.263069 10.14779	6.368761 7.72047	8.711239 12.23703 10.03191
Variable WGB WIB combined diff	Obs 80 80 160	Mean 7.54 9.97875 8.759375	7ariances Std. Err. .588429 1.134557 .6443217 1.278072	Std. Dev. 5.263069 10.14779	6.368761 7.72047 7.486842 -4.969542	8.711239 12.23703 10.03191
Variable WGB WIB combined diff	Obs 80 80 160	Mean 7.54 9.97875 8.759375 -2.43875	<pre>Zariances Std. Err588429 1.134557 .6443217 1.278072</pre>	Std. Dev. 5.263069 10.14779 8.150096	6.368761 7.72047 7.486842 -4.969542	8.711235 12.23703 10.03191 .0920417 = -1.9081
Variable WGB WIB combined diff	Obs 80 80 160	Mean 7.54 9.97875 8.759375 -2.43875	<pre>Zariances Std. Err588429 1.134557 .6443217 1.278072</pre>	Std. Dev. 5.263069 10.14779 8.150096	6.368761 7.72047 7.486842 -4.969542 t	8.7112 12.237 10.031 .09204 = -1.90