

Taking Turns

By Greg Leo*

Two individuals face a regular task that requires the effort of only one. They take turns but sometimes arrange to swap obligations. These swaps account for their changing, private costs. While seemingly primitive, flexible turn-taking is surprisingly efficient, even relative to what can be achieved by mechanisms using monetary transfers. I model and experimentally evaluate a simple form of flexible turn-taking and then present a second form that is both consistent with patterns of subject behavior and approximately second-best in a benchmark case. (JEL D82, C72, C73)

Turn-taking is a fundamental social behavior and a major developmental milestone for children (see [Sheridan, Sharma and Cockerill, 2014](#)). It has also been observed in animal species ([Portugal, 2014](#), [Harcourt et al., 2010](#)). Turn-taking is a fair and natural arrangement in settings where one individual's effort is needed to complete a regularly occurring and mutually beneficial task. Examples include a parent waking to calm a crying baby, a monitor keeping watch in a dangerous environment, or a doctor on-call for a late-night emergency.

Under stochastic, private costs, rote turn-taking can result in inefficient assignment. A flexible turn-taking arrangement, with the possibility of swapping turns, overcomes some of this inefficiency. I formalize¹, a simple model of flexible turn-taking as a dynamic economic mechanism referred to as *recurring rotation*. A

*Leo: Vanderbilt University Department of Economics. VU Station B #351819 2301 Vanderbilt Place Nashville, TN 37235-1819. g.leo@vanderbilt.edu. I am grateful to have received constructive comments from Ted Bergstrom, Rod Garratt, Cheng-Zhong Qin, Ryan Oprea, David Miller, Gary Charness, Emanuel Vespa, Marco Castillo, and Ragan Petrie, as well as seminar participants at The University of Arizona, The University of California Santa Cruz, Virginia Commonwealth University, Tulane University, George Mason University, The University of California Santa Barbara, The 2014 Association for Public Economic Theory International Meeting and The 2014 Economic Science Association North American Meeting.

¹Specifically, I model recurring rotation as a *Perfect Public Equilibrium (PPE)* of a repeated, private cost version of the volunteer's dilemma game ([Diekmann, 1985](#)).

benefit of modeling this in the framework of mechanism design is that flexible turn-taking can be compared to alternative solutions. The result of this comparison is surprising. Despite its simplicity, recurring rotation achieves impressive efficiency even relative to what may be achieved in this environment by mechanisms that use monetary transfer. For instance, with uniformly distributed costs, recurring rotation captures about three-quarters of the achievable efficiency². I present recurring rotation in section I.

Peeling back the gentle exterior of this mechanism reveals a complex set of incentives. The theoretical properties presented here are the properties of this indirect mechanism *in equilibrium*. However, given the complexity, it is natural to ask whether the empirical properties are likely to match the theoretical properties.

In section II, I report the results of a laboratory experiment designed test the empirical properties and of evaluate actual behavior of subjects under the recurring rotation mechanism. Interestingly, although the mechanism achieved efficiency close to the theoretical prediction, subject behavior departed systematically from predictions. These patterns of unexpected behavior appear to arise from how subjects approach the decision problem induced by the incentives of the mechanism and cannot be explained by pro-social interest or strategic concerns. This lends to the possibility that the subjects used heuristics borrowed from their experience in turn-taking arrangements with a different structure.

In section III, I present an alternative arrangement: *obligation takeover*. Obligation takeover permits a “debt” of turns where recurring rotation allows only delay. While more complex, obligation takeover retains a familiar structure, and the asymmetries in subject behavior that appear anomalous under recurring rotation are part of equilibrium behavior in obligation takeover. Furthermore, in the benchmark case of uniformly distributed costs, obligation takeover achieves second-best efficiency for patient players. In this case, no improvement is available from any mechanism appropriate for this environment, even those using monetary transfers. This may attest to the durability of flexible turn-taking as a social arrangement and suggests further study of the how these arrangements are commonly applied in real-world applications.

To my knowledge, this paper is the first formal study of flexible turn-taking. It may be understood as both a normative and a positive exposition of these insti-

²The robustness of the incentives in recurring rotation, which make it appropriate for an informal interpersonal environment, implies it is part of a particular subclass of *Perfect Public Equilibrium* (*PPE*) known as *Ex-post Incentive Compatible Perfect Public Equilibrium* or *EPPPE*, and first-best is not achievable by an *EPPPE* in the environment studied here (Miller, 2012).

tutions. The mechanisms in this paper may be thought of as abstract models of the kinds of flexible turn-taking arrangements people actually use. The results also permit normative conclusions; flexible turn-taking can provide a large amount of efficiency without the use of money, but the incentives underlying these mechanisms are remarkably complex, despite the familiar exterior.

More broadly, this paper differs from previous research in characterizing the properties of natural and robust mechanisms in an explicitly interpersonal environment. For instance, while [Athey and Miller \(2007\)](#) and [Miller \(2012\)](#) provide theoretical results on robust dynamic mechanisms, they consider mechanisms that involve monetary transfers. Other authors construct mechanisms explicitly for dynamic settings *without* monetary transfers³, but do not focus on mechanisms with robust incentive properties, appropriate for interpersonal environments. While the favor trading literature (see for instance [Mobius, 2001](#)), is interpersonal in context, the environment involves a different kind of private information than that considered here. A more detailed comparison to existing literature and concluding discussion is provided in section IV.

I Recurring Rotation Mechanism

Alice and Bob have a household task that must be done every day. It only takes one person to do it. Neither likes the task, but on some days it is more inconvenient than on others. Every day, the cost for each of them is independently drawn from a common distribution. They have agreed on a turn-taking arrangement with possible swaps. It works like this...

The partner who did not complete the task yesterday is obligated to complete it today unless there is a mutual decision to swap. All else equal, neither wants to be the one who has to do it. However, since completing the task in this arrangement results in becoming non-obligated on the the next day, a partner with a low enough cost will prefer to complete the task immediately, delaying future obligation. Suppose Alice is obligated and has a high enough cost to prefer putting off the task- remaining obligated tomorrow. Bob has a low cost and prefers to complete the task immediately rather than be obligated tomorrow. They can make a mutually beneficial swap. This is the *recurring rotation* mechanism.

³See, for instance, the repeated Bertrand environment of [Athey and Bagwell \(2001\)](#), repeated auction collusion environments of [Aoyagi \(2003\)](#) and [Skrzypacz and Hopenhayn \(2004\)](#), and the more generalized environment of [Jackson and Sonnenschein \(2007\)](#).

In this section, I derive several results on the incentive properties, equilibrium, and efficiency of recurring rotation for general cost distributions. I also compare it to other mechanisms and provide more detailed numerical results for the uniform distribution.

A Environment

Two players engage in a repeated game. Each has discount factor β . In each period, a task must be completed. The cost of performing the task is given by θ_i for player i . This cost is private information and is drawn independently in each period from identical, and commonly known distribution $F(\theta_i)$ on the (normalized) domain $[0, 1]$. The value of having the task completed is fixed and normalized to 0 for both.

The players use a mechanism which determines who will complete the task in each period. If neither completes the task, they each get a payoff that is less than -1 . This ensures that, no matter what their costs, if either knew the other was not going to do the task, he or she would prefer to do the task rather than leave it undone and gives the stage game the structure of a Volunteer’s Dilemma (Diekmann, 1985).

The Volunteer’s Dilemma has asymmetric Nash equilibria in which any player completes the task with certainty. Thus, the mechanism simply fixes players’ beliefs about which Nash equilibrium should be played in a period. Since the repeated game then involves a repetition of Nash play at each stage, no player has incentive to deviate from the proposed assignment⁴.

The players can communicate, but there is no formal authority to structure communication, prevent them from learning about each other’s private information, or absorb budget imbalance. These assumptions capture the interpersonal nature of their relationship. This assumption does not affect the modeling of recurring rotation; it is robust to these features. Rather, this assumption restricts the class of available alternative mechanisms to those that are suitable for such informal, interpersonal situations. This is discussed further in subsection G.

B Recurring Rotation

The mechanism has two states that correspond to which player is obligated. From either player’s perspective, the states are labeled o for “I am **O**bligated” and n for

⁴To be complete, I assume punishment for this deviation involves repetition of the asymmetric Nash equilibrium in which the deviating player completes the task in perpetuity.

“I am Not Obligated”. Whoever completes the task becomes non-obligated in the next period. The players communicate by offering to do the task. If only one player offers, that player completes the task. If neither offer, the obligated completes the task. The disagreement that arises when both offer can be handled more generally. Various versions are possible. The *deferment* version settles the disagreement on the side of the obligated, while *option* settles it on the side of the non-obligated. A hybrid version that nests these two deterministic versions allows an arbitrary rule for deciding who completes the task when both offer. q will indicate this rule and corresponds to the probability that the obligated is assigned the task when both offer. $q = 0$ corresponds to *option* and $q = 1$ corresponds to *deferment*.

A summary of the mechanism in terms of which partner completes the task, conditional on their actions, is given below in figure 1.

		Non-Obligated Action	
		<i>Not Offer</i>	<i>Offer</i>
Obligated Action	<i>Not Offer</i>	Obligated	Non-Obligated
	<i>Offer</i>	Obligated	Non-Obligated
		$Pr = q$	$Pr = 1 - q$

Figure 1: Partner Completing Task by Joint Action Profile

Dominant Threshold Strategy

When should a player offer? A interesting feature of this mechanism is the symmetry of the decision problem faced in either state. A player’s best response is a threshold strategy. When cost is below this threshold, the player should offer. When cost is above, the player should not offer. The *same* threshold determines action in both states.

Since the players are ex-ante identical and other elements of the game are stationary, I focus on identical Markov strategies. Here, a strategy maps the state and players type θ_i into the action space $\{Offer, Do Not Offer\}$. Let $V(o)$ and $V(n)$ be the present values associated with being obligated and non-obligated respectively⁵. Regardless of the state, the present value of the player who executes the task with cost θ_i is $\beta V(n) - \theta_i$. The present value of not executing the task is $\beta V(o)$. These are the only two possible outcomes.

A player will want to do the task if and only if $\theta_i \leq \beta (V(n) - V(o))$. Whenever $\beta (V(n) - V(o)) \in [0, 1]$ there is a well-defined threshold cost $\tilde{\theta} = \beta (V(n) - V(o))$

⁵These are stationary and not individual specific due to the identical Markov strategy assumption.

below which a player, in either state, prefers to complete the task. Suppose a player has $\theta_i < \tilde{\theta}$. Whether the player is obligated or non-obligated, it is weakly dominant to offer since it maximizes the probability of completing the task- the best outcome when cost is below $\tilde{\theta}$. Likewise, a player with $\theta_i > \tilde{\theta}$ does not want to complete the task. In this situation it is weakly dominant to not offer since it minimizes the probability of completing the task- the worst outcome when cost is above $\tilde{\theta}$. Thus, when players act according to symmetric Markov strategies, their actions are determined solely by a single threshold $\tilde{\theta}$.

However, the value functions are dependent on the strategies of the players, and thus the threshold $\tilde{\theta}$. In equilibrium, players acting according to threshold $\tilde{\theta}$ must generate a discounted difference in state value $\beta (V(n) - V(o))$ that is equal to the threshold. If this is not the case, players have incentive to adjust their strategies. A consistent equilibrium within recurring rotation involves a threshold $\tilde{\theta}$ that solves this fixed point problem. To determine these consistent equilibria, it is first necessary to derive an explicit form of $\beta (V(n) - V(o))$ in terms of the threshold.

Equilibrium Condition

The calculation $\beta (V(n) - V(o))$ is simplified by noting that whenever the two players are on opposite sides of the threshold $\tilde{\theta}$, it does not matter who is obligated. Refer to figure 1. Regardless of who is obligated, if one player has a cost below threshold and the other has a cost above, only the player with the low cost will offer. That player will complete the task and move to the non-obligated state. Thus, the difference in the two state values *only depends on what happens when both are on the same side of the threshold*. With this simplification, $(V(n) - V(o))$ can be determined as follows.

Both Above:

When both players are above the threshold, the obligated always carries out the task at an average cost $E(\theta_i | \theta_i \geq \tilde{\theta})$ and becomes non-obligated which carries discounted continuation value $\beta V(n)$. The non-obligated does not carry out the task and becomes obligated, which has discounted value $\beta V(o)$. Thus, the difference in value of being non-obligated and obligated, conditional on both being above the threshold, is given by $[E(\theta_i | \theta_i \geq \tilde{\theta}) + \beta V(o) - \beta V(n)]$. This event happens with probability $(1 - F(\tilde{\theta}))^2$.

Both Below:

When both are below the threshold, the player chosen to do the task by the tie-breaking rule completes the task at average cost $E(\theta_i|\theta_i \leq \tilde{\theta})$ and becomes non-obligated, which has discounted continuation value $\beta V(n)$. The player who does not complete the task becomes obligated, which has discounted value $\beta V(o)$. Taking in to account the tie-breaking rule q , the difference in the value of being non-obligated and obligated, conditional on both being *below* the threshold, is given by $(2q-1)[E(\theta_i|\theta_i \leq \tilde{\theta}) + \beta V(o) - \beta V(n)]$. For instance, when $q = 1$, the obligated completes the task for sure and this term is: $[E(\theta_i|\theta_i \leq \tilde{\theta}) + \beta V(o) - \beta V(n)]$. When $q = 0$, the non-obligated completes the task for sure and this term is the negative of the previous. The event that both are below the threshold happens with probability $(1 - F(\tilde{\theta}))^2$.

Taking the sum of these two differences weighted by the probabilities of occurrence gives:

$$(1) \quad V(n) - V(o) = F(\tilde{\theta})^2 (2q-1) [E(\theta_i|\theta_i \leq \tilde{\theta}) + \beta V(o) - \beta V(n)] + (1 - F(\tilde{\theta}))^2 [E(\theta_i|\theta_i \geq \tilde{\theta}) + \beta V(o) - \beta V(n)]$$

In equilibrium, $\theta^* = \beta(V(n) - V(o))$. Imposing this relationship on equation 1 provides the following equilibrium condition:

$$(2) \quad \theta^* = \beta \left[(1 - F(\theta^*))^2 [E(\theta_i|\theta_i \geq \theta^*) - \theta^*] - F(\theta^*)^2 (2q-1) [\theta^* - E(\theta_i|\theta_i \leq \theta^*)] \right]$$

C Equilibrium- Uniqueness and Location

The equilibrium threshold θ^* solves the fixed-point problem in equation 2. Notice the right side of this relates weighted versions of the terms $E(\theta_i|\theta_i \geq \theta^*) - \theta^*$ and $\theta^* - E(\theta_i|\theta_i \leq \theta^*)$. These are, respectively, the difference in the average cost of execution and the continuation transfer associated with a deferment when both have cost above the threshold and the negative of the same difference when both have cost below the threshold⁶.

⁶The term $E(\theta_i|\theta_i \geq \theta^*) - \theta^*$ is familiar in survival analysis and is often referred to as “mean residual lifetime”. $\theta^* - E(\theta_i|\theta_i \leq \theta^*)$ is a related term referred to as “mean advantage over inferiors” by [Bagnoli and Bergstrom \(2005\)](#).

In solving for the equilibrium, it proves useful to convert the fixed-point problem into a root problem. Define:

$$(3) \quad g(\tilde{\theta}) = \frac{1}{\beta} \tilde{\theta} + F(\tilde{\theta})^2 (2q - 1) [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})] - (1 - F(\tilde{\theta}))^2 [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}]$$

A threshold $\tilde{\theta}$ is an equilibrium of recurring rotation if and only if $g(\tilde{\theta}) = 0$. The function $g(\tilde{\theta})$ has the three properties (for continuous and strictly increasing cost distributions), which imply that the equilibrium threshold θ^* must be unique, below half of the mean type, increasing in β , and decreasing in q . These properties are demonstrated graphically in figure 2.

Lemma 1. $g(0) < 0$.

Proof. $g(0) = -E(\theta_i | \theta_i \geq 0) = -E(\theta_i) < 0$. □

Lemma 2. g increases strictly over the interval $[0, E(\theta_i)]$.

Proof in online appendix subsection A

Lemma 3. g is strictly positive over the interval $(\frac{1}{2}E(\theta_i), 1]$.

Proof in online appendix subsection B

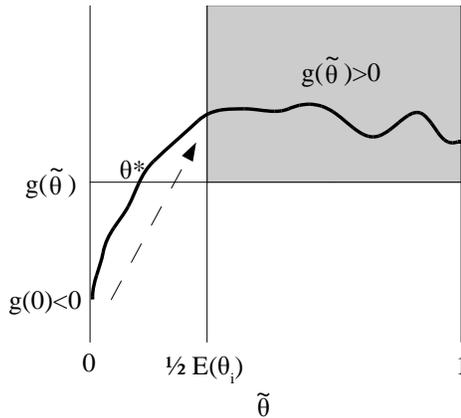


Figure 2: Properties of $g(\tilde{\theta})$.

Proposition 4. For any continuous and strictly increasing F and for any $q \in [0, 1]$ and $\beta \in [0, 1)$, the recurring rotation mechanism has a unique equilibrium that is below the half of the mean type.

Proof. This follows from the combination of lemmas 1,2,3. The equilibrium condition is $g(\tilde{\theta}) = 0$. $g(0) < 0$ and $g(E(\theta_i)) > 0$. Since g increases strictly over $[0, E(\theta_i)]$, it must cross 0 exactly once in the interval $[0, \frac{1}{2}E(\theta_i))$, and not again over $[\frac{1}{2}E(\theta_i), 1]$ since g remains strictly positive over this interval. \square

D Efficiency

Since no transfers are used, the efficiency can be measured by the average cost borne by the partners in executing the task. To simplify the expression of this average cost, notice that, under any version of the mechanism, if at least one offers, then the partner who does the task must have a cost below θ^* . Since the players' costs are independent, the average cost of execution when at least one offers is $E(\theta_i | \theta_i \leq \theta^*)$. When neither player offers, the obligated player is always chosen to do the task and must have cost above θ^* , executing the task at an average cost $E(\theta_i | \theta_i \geq \theta^*)$. Further, since the probability that at least one offers is $1 - (1 - F(\theta^*))^2$ and the probability that neither offer is $(1 - F(\theta^*))^2$, the expression for average cost of execution is $\bar{AC} = [1 - (1 - F(\theta^*))^2] E(\theta_i | \theta_i \leq \theta^*) + (1 - F(\theta^*))^2 E(\theta_i | \theta_i \geq \theta^*)$. After simplifying:

$$(4) \quad \bar{AC} = E(\theta_i) - F(\theta^*)(E(\theta_i) - E(\theta_i | \theta_i \leq \theta^*))$$

The derivative of \bar{AC} with respect to θ^* is $-f(\theta^*)E(\theta_i) + f(\theta^*)\theta^*$, which is negative as long as $\theta^* \leq E(\theta_i)$. Since all equilibria occur below the mean, the average cost is decreasing in the equilibrium threshold. This result is useful for comparative statics.

Lemma 5. *Higher equilibrium threshold implies lower average cost (higher efficiency).*

E Comparative Statics

Efficiency of recurring rotation is increasing in patience (β) and decreasing in the tie-breaking rule q . The option version ($q = 0$) of the mechanism is the most efficient for any cost distribution.

Corollary 6. θ^* and efficiency are increasing in β .

Proof.

$$(5) \quad \frac{\delta \theta^*}{\delta \beta} = - \frac{\frac{\delta g(\theta^*, \beta)}{\delta \beta}}{\frac{\delta g(\theta^*, \beta)}{\delta \theta^*}} = \frac{\frac{1}{\beta^2} \theta^*}{\frac{\delta g(\theta^*, \beta)}{\delta \theta^*}}$$

This is positive since $\frac{\delta g(\theta^*, \beta)}{\delta \theta^*} \geq 0$ at the equilibrium by Lemma 2 and Proposition 4. Increasing efficiency follows from lemma 5. \square

Corollary 7. θ^* and efficiency are decreasing in q .

Proof.

$$(6) \quad \frac{\delta \theta^*}{\delta q} = - \frac{\frac{\delta g(\theta^*, q)}{\delta q}}{\frac{\delta g(\theta^*, q)}{\delta \theta^*}} = - \frac{2F(\theta^*)^2 [\theta^* - E(\theta_i | \theta_i \leq \theta^*)]}{\frac{\delta g(\theta^*, q)}{\delta \theta^*}}$$

This is negative since $\theta^* - E(\theta_i | \theta_i \leq \theta^*) \geq 0$ and $\frac{\delta g(\theta^*, q)}{\delta \theta^*} \geq 0$, at the equilibrium, by Lemma 2 and Proposition 4. Increasing efficiency follows from lemma 5. \square

Corollary 6, that efficiency increases with patience, is rather intuitive and due to the fact that players are less willing to take high costs in the current period to change the potential outcomes of future periods when the future is discounted more heavily. This result also implies that patient players swap more often in equilibrium than less patient players. As β increases, turn-taking becomes less rigid.

Corollary 7, that the option version is most efficient, is less intuitive. Recall that the best-response threshold is equal to the discounted difference in the value of being non-obligated and obligated. Further, recall that the choice of q only matters in situations when both players *want* to complete the task. When q is small, the non-obligated is assigned the favorable outcome more often at the expense of the obligated player. This increases the difference between the non-obligated and obligated values for any fixed threshold and results in a higher equilibrium threshold.

It is an interesting result that the seemingly less natural and less 'cooperative' version of the mechanism is more efficient, especially since assignments in the option version depend only by the cost of the non-obligated player. In numerical tests, the difference in efficiency of the option and deferment versions tended to be small. This is because the versions differ only in how they handle the situation that both *want* to do the task- a relatively rare event in equilibrium. It is possible to put an analytical bound on this difference, see online appendix subsection C.

These results provide some general insight into the workings of the recurring rotation mechanism and the location of equilibria. Next, I look more deeply at the case of uniform costs.

F Example: Uniform Costs

When $\theta_i \sim U(0,1)$, θ^* is the solution to the cubic⁷ equation: $\theta^* = \frac{1}{2}\beta \left[(1 - \theta^*)^3 - (2q - 1)(\theta^*)^3 \right]$. For option ($q = 0$) and deferment ($q = 1$), respectively these are:

$$(7) \quad q = 0 : \theta^* = \frac{1}{2}\beta \left[(1 - \theta^*)^3 + (\theta^*)^3 \right]$$

$$(8) \quad q = 1 : \theta^* = \frac{1}{2}\beta \left[(1 - \theta^*)^3 - (\theta^*)^3 \right]$$

A graph of the solutions is show in figure 3.

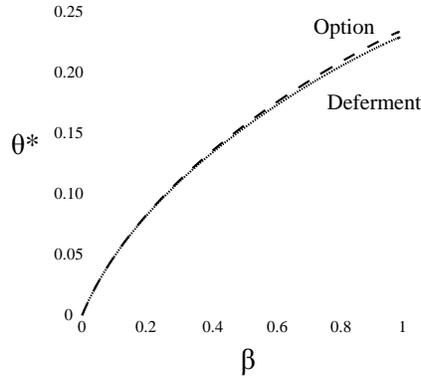


Figure 3: Equilibria under uniform distribution by discount factor.

Consistent with the analytical results of C, both versions have thresholds increasing in β and the equilibrium threshold in the option version is larger than the deferment version. As $\beta \rightarrow 1$, the equilibria approach 0.232 for option and 0.226

⁷At $q = 0$ the equation is quadratic.

for deferment. Using equation 4 together with the calculated equilibrium thresholds, the resulting average stage costs are 0.411 for option and 0.413 for deferment as $\beta \rightarrow 1$.

To put these costs in perspective, suppose players were to take turns in a rigid way or flip a coin. The average stage cost would be the mean of the distribution: $\frac{1}{2}$. Since they can achieve this stage cost without sharing *any* information about their costs, it represents a lower bound on the achievable efficiency⁸. Compared to this lower bound, recurring rotation provides a substantial efficiency improvement.

On the other hand, in a perfect information setting, they could always assign the task to the player with lower cost. This would yield an average stage cost equal to the expected value of the minimum of the partners' costs. In the uniform case, this would be $\frac{1}{3}$. Thus, in uniform environment, recurring rotation falls about halfway between the lower-bound and first-best efficiency. This is plotted on the line segment below.

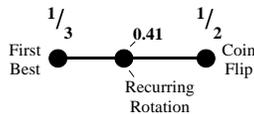


Figure 4: Efficiency Comparison

However, first-best is not the most appropriate comparison. Players do not have perfect information. Because of this, achieving first best would require a mechanism that perfectly reveals costs. In this environment, with no planner to enforce communication rules or absorb budget imbalance, there is not a suitable mechanism that achieves first-best even with monetary transfers. An explanation of this requires some discussion of the mechanism design literature.

G Finding the Appropriate Benchmark

In a repeated game, players may choose their actions based on the history of play. In recurring rotation, the relevant history is completely summarized by who is currently obligated. Because the only important aspect of the history is one that is commonly known, recurring rotation is known as a *perfect public equilibrium (PPE)*

⁸They could do even worse, but systematically assigning the task to the player with a higher cost would require some information transfer, which could be put to more beneficial use.

(Fudenberg, Levine and Maskin, 1994). There are other aspects of the history that are not used. For instance, a player *could* additionally consider past cost realizations, an aspect of the history that is not publicly known.

In a PPE, various *states* of the equilibrium provide different amounts of expected utility to players. For instance, in recurring rotation, it is less desirable to be obligated than non-obligated. A transition from one state to another in a *PPE* is a transfer - not of money but rather of continuation utility. These transfers can be used to incentivize the revelation of private information just as money transfers in many one-shot mechanisms. Because of this, a *PPE* may be considered a type of dynamic mechanism where incentives for revealing information are provided, at least in part, by continuation utility transfers (Miller, 2012).

In some settings, first-best efficiency is achievable by a *PPE*, even without money transfers (Fudenberg, Levine and Maskin, 1994; Athey and Bagwell, 2001). However, the whole class of *PPE* is not the appropriate benchmark. In recurring rotation, a player's incentive to report their cost truthfully, relative to the threshold θ^* , is weakly dominant, holding beliefs about future strategies fixed. The partners *stage beliefs* can be distorted in any way by cheap-talk, spying, etc. Their incentives remain the same. The induced set of dynamic, direct stage mechanisms have what Bergemann and Välimäki (2010) refer to as *periodic ex-post incentive compatibility*. That is, the incentives do not depend on history or beliefs about the current state of the other player, but may depend on future states.

Ex-post incentive compatibility provides robustness appropriate for an interpersonal setting (Chung and Ely, 2002). No planner is required to structure communication or isolate the partners to prevent changes in beliefs. The subclass of *PPE* with similarly robust incentives are what Miller (2012) refers to as *Ex Post Perfect Public Equilibrium (EPPPE)*.

Further, (Miller, 2012) provides an impossibility for this more restricted class of mechanisms. In an economically interesting class of environments, including the one studied here, first-best is not achievable by an EPPPE even with money transfers (under a no-subsidy condition on the ex-post budget)⁹! This means that, even if money transfers are allowed, but no subsidy can be provided for the players, first-best is not achievable by a robust mechanism in this environment. This leaves open the question of what *is* achievable by an EPPPE either with or without money transfers.

⁹Athey and Miller (2007) consider the repeated trade setting under weaker budget assumptions where first-best can sometimes be achieved, especially by patient players.

The efficiency achievable by a one-shot ex-post incentive compatible mechanism with transfers under the no-subsidy condition is an upper-bound on achievable efficiency in each round of an EPPPE without money transfers (Miller, 2012)¹⁰. This result simplifies the problem of bounding the optimal EPPPE efficiency to the problem of characterizing optimal one-shot ex-post incentive compatible mechanisms. While this is not a well-studied problem¹¹ due to the overwhelming variety of available mechanisms, Shao and Zhou (2013) provides a useful result for the uniform environment. They prove that, in the one-shot allocation of a valuable good between two players with uniformly distributed value¹², the highest welfare achievable by an ex-post incentive compatible mechanism under no-subsidy is $\frac{5}{8}$. Since the allocation problem may be thought of as the negative reflection of the assignment problem¹³, this implies that second-best optimal average stage cost in an EPPPE with monetary transfers is $\frac{3}{8}$. This can be used as a lower-bound on the achievable average stage costs of an EPPPE without monetary transfers as well¹⁴.

H Efficiency Comparison

The line segment below plots recurring rotation in relation to the second-best benchmark. The recurring rotation mechanism achieves about $\frac{3}{4}$ of the efficiency of an optimal ex-post¹⁵ incentive compatible mechanism with subsidy-free transfers in the uniform environment.

¹⁰An EPPPE without money transfers uses instead transfers of utility within the repeated game to provide incentives. In an equilibrium, these transfers must come from within a set of consistent continuation values (Abreu, Pearce and Stacchetti, 1986, 1990). Monetary transfers are not restricted in this way. Barring budget issues, players are free to transfer any amount. Consistency is not an issue. Because of this, the set of feasible transfers using money is at least as large as the set of feasible transfers in a dynamic mechanism without money. However, in a repeated setting where money transfers are allowed, all of the incentives can be handled by money without resorting to more complicated utility transfers.

¹¹Hagerty and Rogerson (1987) demonstrate that in the static trade setting, a posted price mechanism must be optimal under ex-post incentive compatibility, ex-post individual rationality (participation constraint), and strong ex-post budget balance (transfers must balance exactly).

¹²With support $[0, 1]$

¹³The problem of assigning a costly duty to do the task is the same as the problem of allocating the valuable right not to do the task.

¹⁴It is also possible to directly approximate a bound in this case for patient players using the separating hyperplane methods of Fudenberg, Levine and Maskin (1994). This procedure is also discussed in Miller (2012, p. 792). This numerical exercise also yields a bound on the average stage cost of $\frac{3}{8}$, and suggests that eschewing monetary transfers is not restrictive in this environment for patient players. In fact, the mechanism in section III approximates this efficiency.

¹⁵Since ex-post and dominant strategy incentive compatibility are equivalent for the private valuation environment, they are used interchangeably in this paper.

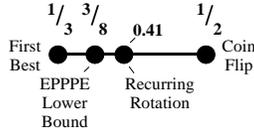


Figure 5: Efficiency Comparison

The efficiency lost over second-best comes from the fact that the equilibrium threshold is too low. For comparison, if players used threshold $\frac{1}{2}$, they would achieve second-best average stage cost of $\frac{3}{8}$. However, the incentives provided by recurring rotation are smaller than necessary to get players to implement this optimal threshold. This results in inefficient sorting of the players costs. This inefficiency is not a problem limited to the uniform case. In fact, the efficiency of recurring rotation is always bounded below second-best optimal. I discuss this further in online appendix subsection D.

A mechanism using a single threshold at the mean cost, with appropriate transfers, is always ex-post incentive compatible and optimal among single-threshold mechanisms. See lemmas 12 and 13 in subsection D of the online appendix. By comparing the efficiency of recurring rotation to this mean-threshold mechanism, it is possible to put a lower-bound on the efficiency lost by using recurring rotation instead of an optimal mechanism (see proposition 14 in the online appendix subsection D). However, despite the fact that recurring rotation is suboptimal relative to the best threshold mechanism, in online appendix subsection E, I demonstrate that there are some environments where the welfare loss is small.

To complement these results, in the next section I present results from an experiment designed to test the empirical properties of recurring rotation.

II Experiment

Given the surprising complexity of the incentives of recurring rotation, it is easy to imagine that actual behavior would deviate from theoretical predictions. In this section, I report the results of an experiment designed to evaluate the performance of the recurring rotation mechanism under true behavior.

This experiment put subjects in the scenario of playing an abstracted version of the repeated task game with uniformly distributed costs. The subjects used the deferment version of the recurring rotation mechanism to determine assignments. On

average, subjects chose threshold strategies higher than the predicted equilibrium threshold. Further, many subjects choose higher thresholds in the obligated state than the non-obligated state (“gap” strategies).

Despite variance in subject behavior, efficiency was close to what is expected in equilibrium. However, the pervasive use of gap strategies substantially decreased efficiency from what it would have been otherwise¹⁶. Further, the gaps remain in a treatment of the experiment designed to control strategic and pro-social variation by pairing subjects with computer partners. This suggests that the gaps may be due to heuristics developed under a turn-taking arrangement with different structure.

A Design Details

Data was collected at the University of California, Santa Barbara Experimental and Behavior Economics Laboratory using ZTREE (Fischbacher, 2007). The On-line Recruitment System for Economic Experiments was used to recruit subjects (Greiner, 2004). A total of 199 subjects participated in the experiment, including pilot sessions. Recruitment was not restricted by major or past participation. Equal numbers of men and women were invited to each session to provide rough gender balance. Each session of the experiment lasted about 50 minutes. No subjects participated in more than one session.

The experiment consisted of two key treatments. In the first, subjects were partnered with other subjects in their session. In the second, subjects were paired with a computer playing a known strategy- offering with probability $\frac{1}{2}$. In both treatments, the environment was the same and the subjects were aware of all of the details of the game.

In the person/person treatment there were four sessions. Three session used 16 subjects and one session used 18 subjects for a total of 66. Subjects were paired randomly with another subject from the session. One subject from each pair was randomly selected to start as the obligated partner. For the first 20 rounds, players found out their random cost, distributed uniformly between \$0 and \$3 in penny increments, before choosing an action. Simultaneously, the obligated player chose whether to ask his/her partner to do the task and the non-obligated chose whether to agree to do the task should his/her partner ask. After round 20, subjects instead submitted a threshold strategy before learning about their private costs¹⁷ - the ob-

¹⁶Efficiency is lower relative to what would have been achieved if subjects played non-gap strategies equal to the average of what they chose in the obligated and non-obligated states.

¹⁷This “strategy method” is a common experimental technique for efficiently collecting infor-

ligated choosing above what cost to ask and the non-obligated below what cost to agree¹⁸.

Below, I refer to the situation where the obligated asks the non-obligated to complete the task as the obligated partner asking for a “swap”.

After each round, subjects found out whether the obligated had asked and whether the non-obligated agreed or would have agreed (if the obligated did not ask)¹⁹. In each case, they also found out how much they paid in the round, if anything, and who would be obligated on the next round. Cumulative payoffs were not shown throughout the experiment, though no effort was made to prevent subjects from recording data with pen and paper (this was rare).

The probability of continuing with a partner in each round was $\frac{9}{10}$ to generate an effective discount factor of $\beta = \frac{9}{10}$. At the end of each round, players found out whether they would continue with the same partner or whether they had been paired with a new partner ($\frac{1}{10}$ probability) for the next round. Whenever a new partnership was formed, the software randomly chose one partner to be obligated. After 48 rounds²⁰, the game ended on the next partnership termination²¹.

Subjects were paid for every round of the game. This scheme was chosen for simplicity and to induce far-sighted behavior. The rationale for this choice is based on the experimental comparison of repeated game payment schemes in [Sherstyuk, Tarui and Saijo \(2013\)](#). Subjects started the experiment with \$5 and earned an additional \$3.20 every 4 rounds²². Subjects were reminded after each round that

mation about subjects’ underlying strategies. Rather than collecting a single data point about the subject’s strategy at the interim stage, this method elicits the entire strategy at the ex-ante stage. Importantly, the choice of when to collect this data does not affect the actual game in any way.

¹⁸The language used in the experiment is slightly different than the way the actions are described in the previous section. The obligated choose whether to ask the non-obligated to do the task and the non-obligated chose whether to agree conditional on being asked. In a pilot study, this language made the mechanism more transparent to subjects over having each player offer to do the task (or not) regardless of state. An obligated partner asking the non-obligated to do the task is equivalent to not offering to do the task in theoretical game. Similarly, the non-obligated partner agreeing to do the task is equivalent to offering in theoretical game.

¹⁹This design feature was chosen to ensure that the amount of information subjects received about their partner’s strategy was not conditional on actions.

²⁰The subjects did not know this cut-off, but rather that there was *some* specified cutoff. 48 was chosen based on the average play time of groups in pilot sessions.

²¹Although 58 is the expected number of rounds, actual average duration, in rounds, of the sessions were shorter. The sessions lasted 49, 51, 52, 58 rounds respectively.

²²Pilot-study data and theoretical efficiencies of the recurring rotation mechanism were used to select these payment parameters along with the range of the cost distribution to target an average subject payment of \$15 for an hour experiment while providing as much curvature on the pay-off function as possible without risking many subjects ending up below the \$5 lab minimum guaranteed payment. Subjects that suspected they might be below the minimum could face distorted incentives. Under these parameters, for 58 rounds, rigid turn-taking would result in an average payoff of \$7.90

they would receive \$3.20 every 4 rounds, but they did not know explicitly when this occurred. This was chosen to avoid focal-points and strategy distortion on the paying period. On average, players in the person/person experiment earned just over \$15.

In the computer treatment, all details were identical. However, instead of being paired with a new subject after partnership termination, the partnership was instead restarted. Each time this happened, the software chose randomly whether the subject or the computer would start as the obligated partner. Between the person/person and the person/computer treatments, partnership lengths were matched within sessions. That is, session one of the person/person treatment had the same partnership lengths and ultimate session length as session one of the person/computer treatment. A total of 63 subjects participated over 4 sessions. Three sessions included 16 subjects and one session included 15. Earnings in the computer treatment were higher with average of about \$20 per subject²³.

B Results: Person/Person Experiment

On average, selected thresholds were larger than those expected in a symmetric Markov equilibrium. Over all four sessions, the average chosen threshold was \$0.99 while equilibrium prediction is \$0.64. There was a great deal of variance in threshold choice. Histograms of average threshold choice by subject are available in the online appendix subsection J.

Average play did not vary much over the rounds²⁴. Changes in play appear to have settled down by round 35 (see online appendix subsection K), and I focus on rounds 35 – 48 for calculating steady-state properties of the mechanism. Over these rounds, the average threshold conditional on being obligated was \$1.11 and \$0.78 when non-obligated.

From a strategic standpoint, one of the most obvious questions to ask about subject behavior is how it compares to best-response. Due to the variance in subject

for the experiment while second-best play would result in an average of \$18.78 and equilibrium play in roughly \$15.21. The parameter choice was relatively accurate to the intended targets. Under these parameters, only 2 subjects ended up below \$5 in the person/person treatment. Both earned about \$3.50 in the experiment (and had payment rounded to \$5). At this level, even if the two subjects had been aware their cumulative earnings were below the \$5 limit, the round incentives were still meaningful since moving above the threshold might have taken only a few rounds.

²³This is not surprising. The computer's strategy of offering half of the time is quite favorable to the subjects.

²⁴CDFs of average thresholds over 7-round blocks by subject and split by obligated and non-obligated states are available in the online appendix subsection J.

behavior, a full best-response analysis is difficult. However, from an aggregate perspective, it appears subjects' biggest mistake relative to best-response was not asking for deferment often enough when obligated.

The 'average' subject chosen \$1.11 and \$0.78 when obliged and non-obligated respectively. Suppose two such 'average' subjects meet. With their strategies, the difference in continuation value of being obligated and non-obligated discounted by $\beta = 0.9$ is roughly \$0.63. For them, this is the best response threshold, regardless of state. In comparison, the average subject tended to offer swaps too often when non-obligated and, to a greater extent, reject swaps too often when obligated.

Furthermore, these gaps are pervasive at a subject level. I use the following model to test the significance of the threshold gap for each subject. A player i 's threshold $\theta_{i,t}$ in round t is estimated by an individual constant α_i plus individual-specific change δ_i when the player is obligated ($o_{i,t} = 1$ when subject i is obligated in period t).

$$(9) \quad \theta_{i,t} = \alpha_i + \delta_i o_{i,t} + \varepsilon_{i,t}$$

59 percent of subjects have a significantly positive difference between obligated and non-obligated threshold at the 5 percent level (against one-sided alternative), only 9 percent have a significantly negative difference. Thus, the bias towards playing larger thresholds when obligated is robust at the subject level. Below, I demonstrate that these gaps were costly.

C Efficiency Estimation

Several empirical facts have implications for the efficiency achieved by subjects. On average, subjects chose thresholds higher than equilibrium. There was substantial variation among subjects. Many subjects choose "gap" strategies. In this section, I estimate the empirical efficiency of recurring rotation using subject behavior.

For each subject, I calculate the average obligated and non-obligated thresholds (normalized to $[0, 1]$ for comparison to the previous section) over rounds 35 – 48. For every possible pair (with replacement) I calculate the long-run efficiency they would achieve using these strategies. Averaging over all of these pairs provides the estimated population efficiency. This procedure yields an estimate of 0.421²⁵ - only

²⁵This also corresponds closely to the actual achieved average cost for subjects in the experiment. Over rounds 35 – 48 the partner who ended up completing the task in each pair paid an average of

slightly worse than the equilibrium prediction: 0.416.

To get a measure of efficiency without gap strategies, I use the same procedure as above but assign each hypothetical player to use, in both states, their unconditional average threshold. This yields an average cost of 0.399, a substantial improvement over 0.416 estimated under gap strategies. This is not surprising since players tended to choose thresholds larger than the predicted equilibrium threshold. By lemma 5, *symmetric* thresholds closer to $\frac{1}{2}$ provide higher efficiency ²⁶.

For comparison, the difference between an average cost of 0.399 and 0.421 corresponds to an average of \$3 per pair given up over 50 rounds; about 10 percent of an average pair's experimental earnings. These estimates suggest that gap strategies are not just an unexpected phenomenon, but also an important factor in the empirical efficiency of recurring rotation.

What could account for these gaps? Perhaps subjects are responding to incentives that extend beyond the monetary incentives of the game. They may be pro-social. Alternatively, gaps could be a strategic response to subject's beliefs about their opponents' strategies; perhaps they are non-Markov. I analyze these two possibilities with subject data in online appendix subsection M. The evidence for either explanation is limited. However, a full analysis of these aspects could be very complex, and this treatment is not designed to test these possibilities directly.

On the other hand, the computer partner treatment controls for both pro-social and strategic aspects of the game. The computer's strategy is known, and explicitly Markov. The fact that gap strategies are prevalent in the computer treatment suggests that gaps cannot be explained by pro-social behavior or strategic response.

D Results: Person/Computer Treatment

In this treatment, subjects were paired with a computer known to offer to do the task with probability $\frac{1}{2}$ in both states. From a strategic point of view, this is equivalent to having the computer use a threshold of \$1.50, or $\frac{1}{2}$ when normalized to $[0, 1]$. The online appendix section J contains histograms as well as cumulative distribution plots of average thresholds by subject separated by obligated and non-obligated states. As in the person/person treatment, changes in play appear to have

\$1.26 or .42 when normalized to $[0, 1]$.

²⁶Figure 14 in the online appendix demonstrates visually why such a scenario leads to improved efficiency. The figure is a contour plot of the efficiency achieved when players choose arbitrary strategies with no gap. Contour lines are drawn at .416 (equilibrium efficiency for $\beta = .9$), .4, .39, and .38. Most pairs tend to fall in the darker blue area of the figure where the joint strategy profile yields efficiency better than equilibrium.

settled down by round 35. A formal test of this is provided in the online appendix subsection K.

For rounds 35 – 48 the average threshold chosen by subjects in the obligated state was \$0.83 and \$0.66 in the non-obligated state. Note that a gap remains at the aggregate level.

Best Response Analysis

As in the person/person treatment, subjects set average thresholds higher than predicted by best-response. In this case, the best response can be calculated from a modified form of equation 2 allowing different threshold values for two players $\tilde{\theta}_1$ and $\tilde{\theta}_2$.

$$(10) \quad \frac{1}{\beta} \tilde{\theta}_1 = F(\tilde{\theta}_1) F(\tilde{\theta}_2) [E(\theta_1 | \theta_1 \leq \tilde{\theta}_1) - \tilde{\theta}_1] \\ + (1 - F(\tilde{\theta}_1)) (1 - F(\tilde{\theta}_2)) [E(\theta_1 | \theta_1 \geq \tilde{\theta}_1) - \theta_1]$$

Normalized to $[0, 1]$, the computer implements threshold $\tilde{\theta}_2 = \frac{1}{2}$. Thus, the best response for player 1 is the solution to the following quadratic equation:

$$(11) \quad \tilde{\theta}_1 = \frac{9}{10} \left[\frac{(1 - \tilde{\theta}_1)^2}{4} - \frac{\tilde{\theta}_1^2}{4} \right]$$

The solution is $\frac{9}{58} \approx 0.156$ in the normalized game or about \$0.47 for the parameters faced by subjects. The aggregate obligated and non-obligated thresholds are significantly higher than the best response at better than the 1 percent level.

Gap Strategies

Using the same procedure as the person/person treatment, 49 percent of subjects have a significantly positive gap (at the 5 percent level against one-sided alternative), 3 percent have a significantly negative gap. Here, slightly fewer subjects show a significantly positive gap than in the person/person treatment where the figure was 59 percent. However, this still dwarfs the percentage with a significantly negative gap.

The persistence of these gaps, even against a computer that has a known strategy, suggests these patterns arise from the way subjects approach the decision problem,

rather than from unmodeled incentives. Although it is not clear precisely why these gap strategies are prevalent, one explanation is that the experience subjects draw from to make choices in the lab comes from their experience in situations where gaps are a more natural part of the arrangement. For instance, as described in the next section, gaps appear as part of equilibria in a slightly different form of flexible turn taking. It may be that recurring rotation, as a model of flexible turn-taking, is exceptional in terms of this symmetric threshold feature relative to the kinds of arrangements people use naturally. Further study into the precise arrangements people use to solve these problems may provide a deeper insight into this pattern.

III Obligation Takeover Mechanism

This simplicity of recurring rotation comes at the cost of efficiency. In recurring rotation, deferring shifts future obligations. However, this shift is not costly enough to get partners to use efficient thresholds. An alternative arrangement which makes deferment more costly would offer improved efficiency. One way to do this is to give the deferring player extra turns of obligation, rather than simply shifting existing obligations. These extra obligations come as 'debt'. This is a natural extension since non-monetary debt is common in interpersonal relationships, as demonstrated by the prevalence of the idiom "I owe you one".

In this section, I present a mechanism called *obligation takeover* which introduces this 'debt of turns'²⁷. The introduction of debt has two important effects. First, it improves efficiency. In fact, for patient players, obligation takeover achieves second-best efficiency under the uniform cost benchmark and that of the best threshold mechanism for any symmetric cost distribution. It also performs very well even for moderate patience levels.

Second, it causes asymmetries in the equilibrium that do not appear in recurring rotation. However, these asymmetries have a similar pattern to the "gaps" used by subjects in the recurring rotation experiment. The asymmetries are caused by the fact that the cost of acquiring additional debt for the obligated partner is larger than the associated benefit for the non-obligated.

²⁷This mechanism can be seen as extending (in the sense of integer accounting of debt) the "chips" mechanisms discussed in the favor trading literature (see: [Mobius, 2001](#); [Hauser and Hopenhayn, 2008](#); [Abdulkadiroglu and Bagwell, 2012](#)) to include ex-post incentive compatibility in this repeated volunteering environment.

A Mechanism Details

In obligation takeover, a deferment results in one extra period of obligation for the deferring partner. Specifically, this the deferring player “takes over” the nearest future obligation of her partner.

Assume the players start with a plan to alternate who is obligated. Let $\mathbf{p}^t \in \prod_{k=t}^{\infty} \{1, 2\}$ be a vector that represents the “plan” of obligation. At time t , the player identified by the first element of the vector \mathbf{p}^t is obligated. When the obligated completes the task, the plan continues: \mathbf{p}^{t+1} is equal to \mathbf{p}^t with the first element truncated. If, on the other hand, the non-obligated completes the task, the plan proceeds but with an additional period of obligation added at the nearest position for the obligated player.

For instance if the plan is alternating so that $\mathbf{p}^t = 1, 2, 1, 2, 1, 2, 1, 2, \dots$ and the obligated individual completes the task, then $\mathbf{p}^{t+1} = 2, 1, 2, 1, 2, 1, 2, 1, \dots$. On the other hand, if the non-obligated completes the task then $\mathbf{p}^{t+1} = 1, 1, 2, 1, 2, 1, 2, 1, \dots$. In this example the difference between the two resulting plans is precisely one extra period of obligation for player 1.

Under this arrangement, \mathbf{p}^t will always include some repetition of either 1 or 2 followed by alternation. Because of this, the state of the mechanism can be represented by a single variable $z \in \mathbb{Z}/0$. $|z|$ represents the number of repetitions of obligation and $sgn(z)$ represents for whom that repetition pertains. For instance, $\mathbf{p}^t = 1, 1, 1, 1, 2, 1, 2, \dots$ can be represented by state variable $z = 4$. While $\mathbf{p}^t = 2, 2, 2, 1, 2, 1, 2, \dots$ can be represented by $z = -3$.

In a sense, z represents the debt owed by one partner to the other. Assuming a positive z represents extra obligation for 1, then it is the case that player 1 will have to complete z more executions of the task than player 2 before 2 can even begin to accrue debt. In fact, not only does the absolute value of the state represent the debt owed, it also represents the number of extra executions carried out by the non-obligated player over the entire history of play! Because of this, the mechanism has an elegant way of bringing about fairness in the history of task-execution.

For computational reasons, I will assume that there is a limit on the number of obligations a player can accrue so that $|z| \leq \bar{z}$ - a debt limit. In a state where $|z| = \bar{z}$, the player currently obligated is forced to complete the task without the possible intervention of the non-obligated²⁸.

²⁸In equilibrium, the effect of the limit on efficiency is practically inconsequential as long as the limit is not very small. This is because the equilibrium strategies of the players make the probability of visiting larger states exponentially rare. A computational analysis of efficiency with limited states

B Equilibria

As in recurring rotation, assume available actions are to 'offer' or 'not offer' to do the task and that players implement symmetric Markov strategies. Let $V_i(z)$ be the discounted ex-ante utility for player i in state z . For any V function, there is a unique Markov strategy that is weakly dominant with respect to stage beliefs in each round. As in recurring rotation, there are only two outcomes in each round. If the obligated completes the task, the state decreases. If the non-obligated completes the task, the state increases.

Consider player 1 in a state $|z| \neq 1$. Regardless of whether player 1 is obligated, completing the task provides utility $-\theta_i + \beta V_1(z-1)$. Not completing the task provides utility $\beta V_1(z+1)$. Player 1 prefers to complete the task if and only if $\theta_1 < \beta (V_1(z-1) - V_1(z+1))$. Thus, $\tilde{\theta}_1(z) = \beta (V_1(z-1) - V_1(z+1))$ is the cost threshold that determines player 1's preference over completing the task in state z . A summary of the thresholds for each state is given below (for $|z| = 1$, a slight modification is needed since there is no state 0). Technically, the expressions below should include additional structure to account for the situation that the difference in state values is above 1 or below 0. To avoid unnecessary complication in the expressions below it should be understood that when $\beta (V_1(z-1) - V_1(z+1)) \geq 1$, the optimal strategy is $\tilde{\theta}(z) = 1$ and when $\beta (V_1(z-1) - V_1(z+1)) \leq 0$ the optimal strategy is $\tilde{\theta}(z) = 0$. Although the thresholds are written for player 1, by symmetry, player 2's thresholds are simply the reverse ($\tilde{\theta}_2(z) = \tilde{\theta}_1(-z)$).

$$(12) \quad |z| \neq 1 : \tilde{\theta}_1(z) = \beta (V_1(z-1) - V_1(z+1))$$

$$(13) \quad z = 1 : \tilde{\theta}_1(1) = \beta (V_1(-1) - V_1(2))$$

$$(14) \quad z = -1 : \tilde{\theta}_1(-1) = \beta (V_1(-2) - V_1(1))$$

As in recurring rotation, I focus on symmetric equilibria and drop the player index. For simplicity, the rule that determines task assignment, conditional on actions, is assumed to be identical to the deferment version of the recurring rotation mechanism. Swaps happen only by mutual agreement.

In state z , a player with cost below [above] $\tilde{\theta}(z)$ has a weakly dominant strategy to offer [not offer] since that action maximizes the probability of a partner getting her most favorable outcome. Thus, when players act according to symmetric

is provided in appendix subsection H.

Markov strategies, their actions are determined solely by the vector of thresholds.

However, just as in recurring rotation, the value function depends on the strategies of players. In equilibrium, the vector of threshold strategies must generate discounted state value differences equal to the thresholds. Expressions for the continuation values for each state as a function of the the vector of thresholds is derived in the online appendix subsection F. Imposing the equilibrium conditions, these can be modified into expression that relate only the thresholds. Again, to avoid complicating the expressions, it should be understood that the right-hand-side of these expressions are bounded between 0 and 1 (as follows from the equilibrium condition).

$$(15) \quad z > 1 : \tilde{\theta}(z) = \beta (\tilde{\theta}(z-1) + C(z+1, \tilde{\theta}) - C(z-1, \tilde{\theta}) + T(z+1, \tilde{\theta}) - T(z-1, \tilde{\theta}))$$

$$(16) \quad z < -1 : \tilde{\theta}(z) = \beta (\tilde{\theta}(z+1) + C(z+1, \tilde{\theta}) - C(z-1, \tilde{\theta}) + T(z-1, \tilde{\theta}) - T(z+1, \tilde{\theta}))$$

$$(17) \quad \tilde{\theta}(1) = \beta (C(2, \tilde{\theta}) - C(-1, \tilde{\theta}) + T(-1, \tilde{\theta}) + T(2, \tilde{\theta}))$$

$$(18) \quad \tilde{\theta}(-1) = \beta (C(1, \tilde{\theta}) - C(-2, \tilde{\theta}) + T(-2, \tilde{\theta}) + T(1, \tilde{\theta}))$$

Each of these fixed point equations involves functions C and T . C represents the average stage cost borne by the player in each state when both use threshold strategies $\tilde{\theta}$.

$$(19) \quad C(z, \tilde{\theta}) = \begin{cases} E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) F(\tilde{\theta}(-z)) + E(\theta_i) (1 - F(\tilde{\theta}(-z))) & z \geq 1 \\ E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) (1 - F(\tilde{\theta}(-z))) & z \leq -1 \end{cases}$$

T represents the expected transfer of continuation utility associated with a default in each state.

$$(20) \quad T(z, \tilde{\theta}) = \begin{cases} \tilde{\theta}(z) (1 - F(\tilde{\theta}(z))) F(\tilde{\theta}(-z)) & z \geq 1 \\ \tilde{\theta}(z) F(\tilde{\theta}(z)) (1 - F(\tilde{\theta}(-z))) & z \leq -1 \end{cases}$$

Proposition 8. *An equilibrium exists for obligation takeover.*

Proof. An equilibrium in obligation takeover is a solution to this multi-dimensional fixed-point problem above. For finite \bar{z} , this system continuously maps the $2(\bar{z} - 1)$ -dimensional unit cube into itself. Brouwer's fixed point theorem guarantees an equilibrium. When there is no limit on the number of states, the system maps the the infinite dimensional unit-cube $I^\infty = \prod_{i=1}^\infty [0, 1]$ into itself. I^∞ is a convex and

compact (by Tychonoff's theorem) subset of \mathbb{R}^∞ . Schauder's fixed point theorem guarantees an equilibrium. \square

Unlike in recurring rotation, the dimensionality of the problem makes finding closed form solutions difficult. Below, I present numerical approximations of the efficiency achieved by obligation takeover for the uniform cost benchmark over different discount factors. Online appendix subsection I contains results on the efficiency achieved by the mechanism for a range of symmetric beta distributions²⁹.

C Efficiency

Figure 6 plots the approximate efficiency achieved by obligation takeover with an unlimited number of states³⁰ for discount factors $\beta \in (0, 1)$. It is striking how much efficiency is generated even for relatively low discount factors. Though this plot is an approximation of the efficiency with unlimited states, computations in the online appendix subsection H suggests the efficiency lost by using finite limits tends to be small, due to the rarity of long chains of obligation. Further, for large β , the approximate efficiency (average cost) is nearly identical to second best: $\frac{3}{8}$. In fact for perfectly patient players, the efficiency of obligation takeover is identical to second best.

²⁹Unlike in recurring rotation, I do not prove that there is a unique equilibrium for this mechanism. Even if there are multiple equilibria, the efficiencies of the equilibria computed below still represent an approximate lower-bound on the achievable efficiency.

³⁰In order to determine the efficiency of the mechanism under fixed discount factor β , but without limit on the states, I utilize the following procedure: for each β , the equilibrium is approximated under a limited number of states and the efficiency is calculated for the computed equilibrium. After this, the efficiency is again calculated for the computed equilibrium with the number of states doubled. This process is repeated until the difference in efficiency of the two most recently computed equilibria converges to 0.

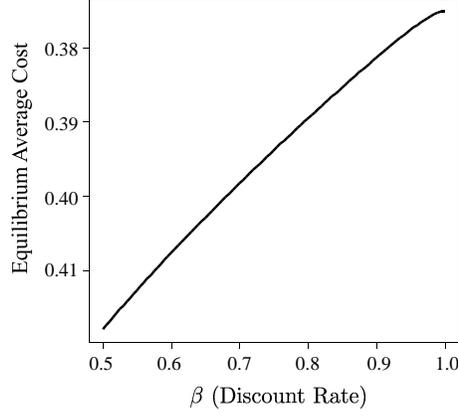


Figure 6: Average Cost in β of Obligation Takeover with Unlimited States

Proposition 9. For any continuous symmetric distribution, the efficiency of obligation takeover is equivalent to that of the best threshold mechanism for patient players.

The full proof appears in online appendix subsection G, an intuitive sketch is provided below.

When both players use thresholds equal to the mean, the difference in expected stage-cost of being obligated and non-obligated is:

$$(21) \quad F(E(\theta))^2 [E(\theta|\theta \leq E(\theta))] + (1 - F(E(\theta)))^2 [E(\theta|\theta \geq E(\theta))]$$

For symmetric distributions, this simplifies to $\frac{1}{2}E(\theta)$. By deferring, the obligated partner reaches a state in which he has *two* more obligations than if he did not defer. Similarly, by accepting a deferment, the non-obligated partner reaches a state in which he has *two* less obligations than he would have otherwise. Since players continue to use the same thresholds in every state, the difference in value of both of these state changes is exactly two times the difference in stage cost, which in this case is $2 \left(\frac{1}{2}E(\theta)\right) = E(\theta)$. Thus, for symmetric distributions, and for patient players, using thresholds at the mean generates differences in the state values that are consistent with using those thresholds- an equilibrium.

Corollary 10. For the uniform distribution, *Obligation Takeover* achieves second-best efficiency for patient players.

Proof. This follows from Proposition 9 and the fact that the threshold mechanism with threshold $E(\theta)$ is optimal among ex-post incentive compatible stage-mechanisms in this environment (see section G). \square

D Equilibrium Characterization

Here, I give a brief overview and intuitive explanations for some of the most striking characteristics that appear in equilibria of obligation takeover. In recurring rotation, there is little to say about the equilibrium beyond a single value, since even impatient partners use the same threshold in both states. This is not the case in obligation takeover. In equilibrium, different thresholds may be used in each state. A computed example equilibrium for $\beta = .9$ and $\bar{z} = 10$ is provided in figure 7 below:

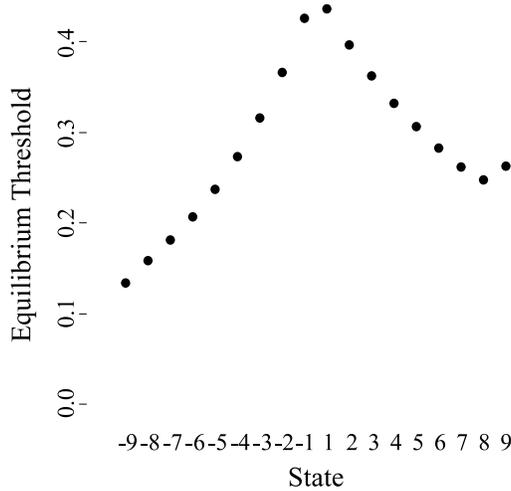


Figure 7: Equilibrium Threshold Values for $\beta = .9$ and $\bar{z} = 10$.

The various asymmetries in the equilibrium come from the fact that deferments carry different values to each partner in different states. This was not the case in recurring rotation. There, a deferment always has the same effect on the sequence of obligations. A deferment always leads to a state in which the currently obligated partner is again obligated. In obligation takeover, the sequence of obligations is affected differently by a deferment in each state. For instance, when a long string of obligations has already been accrued, an additional deferment only adds an additional obligation at the end of the already long chain. When the players discount,

the effect of this additional obligation is diminished relative to its effect if the chain of obligations had been short.

Because of this, the cost to the obligated player and the value to the non-obligated of having an additional obligation added are diminished in larger states. In equilibrium, the player's thresholds depend on these values or costs. This effect leads to thresholds that diminish as the partners move away from states 1 or -1 . This can be seen in the example above in figure 7.

A related effect leads to thresholds that are asymmetric between analogous obligated and non-obligated states. As additional obligations are accrued, the lower thresholds diminish the efficiency of the stage mechanisms. The effect of this weighs more heavily on the obligated player who is more likely to get stuck executing the task at high cost. Because of this, the obligated partner is willing to take higher costs to avoid additional obligations than the non-obligated is to accept them in any particular state, leading to the asymmetry across analogous states, which can also be seen in figure 7³¹. Interestingly, these obligated/non-obligated gaps are precisely the kind that appeared anomalous in subject behavior under recurring rotation.

This source of asymmetries can be seen directly in the equilibrium conditions. For instance, the following equation is derived from 13 and 14 and gives an expression for the gap between analogous obligated and non-obligated thresholds in state 1:

$$(22) \quad \theta_1^*(1) - \theta_1^*(-1) = \beta [(V(1) + V_1(-1)) - (V_1(2) + V_1(-2))]$$

$V(1) + V(-1)$ and $V(2) + V(-2)$ represent the total efficiency of the mechanism in states 1 and 2 respectively. In this case, the gap between obligated and non-obligated thresholds is simply the discounted difference in total efficiency of the mechanism in these two states.

³¹Note that in state 9, the obligated player uses a higher threshold than in state 8. This can be attributed to the fact that in state 10, the obligated player is forced to complete the task. The obligated player is willing to accept slightly larger costs to avoid this distortion.

IV Discussion

A Comparison to Existing Research

This work is related to several areas of the mechanism design literature. In the area of robust mechanism design with transfers, [Drexl and Kleiner \(2012\)](#); [Shao and Zhou \(2013\)](#) consider robust allocation of a valuable good in a one-shot environment with transfers, and construct optimal mechanisms. [Athey and Miller \(2007\)](#) consider a repeated trade setting and demonstrate that first-best efficiency can be achieved by robust mechanisms under relaxed budget balance conditions. [Miller \(2012\)](#) focuses on robust mechanisms for collusion of two firms in repeated settings with transfers. Like these papers, the mechanisms described here do not require a planner to enforce information structure. However, unlike in these, I explicitly analyze mechanisms where players do not use money transfers.

A procedure for constructing the set of achievable payoffs under EPPPE without transfers is discussed in [Miller \(2012, p. 792\)](#). However, this exercise provides little insight into the potential structure of such mechanisms. A primary contribution of this paper is to show that substantial efficiency can be achieved by robust and familiar turn-taking mechanisms.

Several papers also characterize or explicitly construct mechanisms using only continuation transfers in repeated settings without money transfers, but without focusing on robust mechanisms. [Athey and Bagwell \(2001\)](#) consider a repeated Bertrand environment with discrete cost-types and demonstrate that first-best profits can be achieved by impatient firms without money transfers through the use of promises about future market-share. The conclusion of their paper discusses the potential extension to interpersonal relationships that are the focus of this paper. In contrast to [Athey and Bagwell \(2001\)](#) however, I focus on a characterizing simple, *robust* mechanisms, in an environment with continuous type-space.

Several papers consider collusion in repeated auctions. [Aoyagi \(2003\)](#) considers repeated auctions with a type-space on the unit interval, and constructs highly efficient collusion mechanisms. However, these mechanisms require a coordinating institution. [Skrzypacz and Hopenhayn \(2004\)](#) also consider a repeated auction environment, but where communication and monitoring are restricted. Although the specifics of the environment are quite different, the mechanism for bid-rotation developed there is similar to the *obligation takeover* mechanism. Although these papers provide a great deal of insight into the details of using continuation transfers to incentivize mechanisms, the specifics of the environments are quite different

from those considered here³².

[Mobius \(2001\)](#) considers a model where two players can offer each-other “favors”. A favor is an opportunity for one player to offer a fixed benefit of b to another while incurring a fixed cost c . The ability for one player to offer the favor is privately known and arrives at random times. [Lau \(2011\)](#) extends the favor trading environment to random cost and benefits but with one-sided private information and comes closest in the favor-trading literature to the kind of private information in the present environment. In addition to these theoretic papers, [Roy \(2012\)](#) discusses an experimental implementation of the [Mobius \(2001\)](#) environment.

Though the type of private information is a different from that considered here, the mechanism [Mobius \(2001\)](#) develops, the “chips” mechanism, is similar to the obligation takeover mechanism in terms of integer accounting of obligations. [Hauser and Hopenhayn \(2008\)](#) consider alternative versions with improved efficiency in the favor trading environment, and [Abdulkadiroglu and Bagwell \(2012\)](#) derive optimal chips mechanisms. Chips mechanisms are further studied by [Olzowski and Safronov \(2012\)](#) who demonstrate the efficiency of a class of these mechanisms in a more generalized environment. The results on the optimality of obligation takeover in this paper can be seen as extending this line of research on the efficiency of chips mechanisms to instances where ex-post incentive compatibility is required.

The concept of “budgeting” also plays a role in the approximately efficient (first-best) mechanisms presented by [Jackson and Sonnenschein \(2007\)](#) for a general class of repeated private information problems. Their mechanisms force players’ private information reports to match the true underlying distribution over a long sequence of repeated instances. Ex-post incentive compatibility, which I focus on in this paper, does not permit mechanisms which allow a rich enough set of reports to achieve first best outcomes. Recurring rotation and obligation takeover can be seen as abstracted or simplified versions of this budgeting which meet the ex-post incentive requirement.

Other experiments are also related, especially those on turn-taking style cooperation in games without private information. [Cason, Lau and Mui \(2013\)](#) study a common-pool resource allocation game in which turn-taking is the efficient cooperative outcome. They find that subjects who successfully engage in turn-taking are

³²Similar to these papers in its use of auctions, [Guo, Conitzer and Reeves \(2009\)](#) develops a mechanism for repeated allocation without transfers using auctions of a fiat currency in a binary valuation environment.

able to teach this strategy to future partners. [Kuzmics, Palfrey and Rogers \(2012\)](#)³³ studies repeated allocation games in the laboratory and also finds that turn-taking is a prevalent form of behavior³⁴.

[Kaplan and Ruffle \(2012\)](#) study a repeated entry game with private information but where communication is not permitted. They demonstrate that the form of cooperation depends on the particular distribution of private information. With a low variance in private value, rigid turn-taking emerges. However, with larger variance, cut-off strategies based on private value emerge. In contrast, the mechanisms considered here combine both of these elements in an environment where the parties are permitted to communicate.

In addition to these, the current paper may be seen as a repeated extension of the volunteer's dilemma literature. See for instance ([Diekmann, 1985](#); [Weesie, 1994](#); [Weesie and Franzen, 1998](#); [Goeree, Holt and Moore, 2005](#)) and a similar model related to the volunteers dilemma ([Bliss and Nalebuff, 1984](#)).

An interesting line of research in linguistics, formally starting with [Sacks, Schegloff and Jefferson \(1974\)](#), studies turn-taking in conversation, which has been found to be a culturally universal system for organizing discussion ([Stivers et al., 2009](#)). Though, at first glance, this literature seems to be linked only in name to the institutions studied here, the role of turn-taking in conversation may be seen as a set of rules for allocation the scarce resource of opportunities to speak. When its function is considered in this way, it is more clear that conversational turn-taking is likely universal in the same way and for the same reasons that it is an important and universal institution for allocating indivisible goods, opportunities or tasks over time.

B Conclusion

This work represents, to my knowledge, the first formal analysis of flexible turn-taking. Whatever the precise form of flexible turn-taking people may use, it is likely to provide substantial efficiency, perhaps nearly optimal. If this is the case, it is not surprising that flexible turn-taking is such a durable institution. Nothing suitable would do much better.

³³[Kuzmics, Palfrey and Rogers \(2012\)](#) also gives an interesting theoretical interpretation of the Thue-Morse sequence in terms of symmetry of the continuation values for allocation orderings. A similar result is given in [Cooper and Dutle \(2013\)](#) for the example of structuring a fair duel.

³⁴In a computational model, [Neill \(2003\)](#) shows that turn-taking can be achieved in a noisy environment, even when agents use limited memory strategies.

However, the complexity of the underlying incentives in these mechanisms suggests the caution that actual behavior might depart from theoretical predictions in ways that impact the empirical properties. In an experiment, although the efficiency achieved by subjects under recurring rotation was close to that expected in theory, there were robust anomalies in subject behavior which cannot be attributed to pro-social behavior or strategic concerns. This suggests these patterns arise at in the decision process rather than because of unmodeled incentives. One explanation is that subjects may have used heuristics from forms of flexible turn-taking with slightly different structure. For instance, the strategies subjects implement in the experiment are consistent with the kind of “gaps” that appear theoretically in the obligation takeover mechanism.

Work confirming the precise type of flexible turn-taking people naturally implement would allow a more detailed comparison between these social arrangements and other mechanisms. Further, there are several interesting theoretical extensions possible including analysis of arrangements for larger groups, the effect of cost distributions that are asymmetric between players, the potential for players to condition on partially observable information, correlation between players and across time, and the effect of prosociality.

Further, while I have looked at turn-taking in the context of tasks assignment, the results of this paper extend immediately to the repeated allocation problem and should provide insight into analogous environments such as the repeated trade, and collusive environments considered elsewhere.

References

- Abdulkadiroglu, A, and Kyle Bagwell.** 2012. “The Optimal Chips Mechanism in a Model of Favors.” Working paper.
- Abreu, Dilip, David Pearce, and Ennio Stacchetti.** 1986. “Optimal cartel equilibria with imperfect monitoring.” *Journal of Economic Theory*, 39(1): 251–269.
- Abreu, Dilip, David Pearce, and Ennio Stacchetti.** 1990. “Toward a Theory of Discounted Repeated Games with Imperfect Monitoring.”
- Aoyagi, Masaki.** 2003. “Bid Rotation and Collusion in Repeated Auctions.” *Journal of Economic Theory*, 112(1): 79–105.
- Athey, Susan, and David A. Miller.** 2007. “Efficiency in repeated trade with hidden valuations.” *Theoretical Economics*, 2(3).
- Athey, Susan, and Kyle Bagwell.** 2001. “Optimal Collusion with Private Information.” *RAND Journal of Economics*, 32(3): 428–65.
- Bagnoli, Mark, and Ted Bergstrom.** 2005. “Log-concave Probability and Its Applications.” *Economic Theory*, 26(2): 445–469.
- Bergemann, Dirk, and Juuso Välimäki.** 2010. “The dynamic pivot mechanism.” *Econometrica*, 78(2): 771–789.
- Bliss, Christopher, and Barry Nalebuff.** 1984. “Dragon-slaying and Ballroom Dancing: The Private Supply of a Public Good.” *Journal of Public Economics*, 25(1-2): 1–12.
- Cason, Timothy N., Sau-Him Paul Lau, and Vai-Lam Mui.** 2013. “Learning, Teaching, and Turn Taking in the Repeated Assignment Game.” *Economic Theory*, 54(2): 335–357.
- Chung, Kim-Sau, and Jeffrey C. Ely.** 2002. “Ex-Post Incentive Compatible Mechanism Design.” Northwestern University, Center for Mathematical Studies in Economics and Management Science Discussion Papers 1339.
- Cooper, Joshua, and Aaron Dutle.** 2013. “Greedy Galois Games.” *The American Mathematical Monthly*, 120(5): 441–451.

- Diekmann, Andreas.** 1985. "Volunteer's Dilemma." *Journal of Conflict Resolution*, 29: 605–610.
- Drexl, Moritz, and Andreas Kleiner.** 2012. "Optimal Private Good Allocation: The Case for a Balanced Budget." University of Bonn, Germany Bonn Econ Discussion Papers.
- Fischbacher, Urs.** 2007. "z-Tree: Zurich toolbox for ready-made economic experiments." *Experimental Economics*, 10(2): 171–178.
- Fudenberg, Drew, David Levine, and Eric Maskin.** 1994. "The Folk Theorem with Imperfect Public Information." *Econometrica*, 62(5): 997–1039.
- Goeree, Jacob K., Charles A. Holt, and Angela K. Moore.** 2005. "An Experimental Examination of the Volunteer's Dilemma."
- Greiner, Ben.** 2004. "The Online Recruitment System ORSEE 2.0 - A Guide for the Organization of Experiments in Economics." University of Cologne, Department of Economics Working Paper Series in Economics 10.
- Guo, Mingyu, Vincent Conitzer, and Daniel M. Reeves.** 2009. "Competitive Repeated Allocation without Payments." In *Internet and Network Economics*. Vol. 5929 of *Lecture Notes in Computer Science*, , ed. Stefano Leonardi, 244–255. Springer Berlin Heidelberg.
- Hagerty, Kathleen M., and William P. Rogerson.** 1987. "Robust Trading Mechanisms." *Journal of Economic Theory*, 42(1): 94–107.
- Harcourt, Jennifer L., Gemma Sweetman, Andrea Manica, and Rufus A. Johnstone.** 2010. "Pairs of Fish Resolve Conflicts over Coordinated Movement by Taking Turns." *Current Biology*, 20(2): 156 – 160.
- Hauser, Christine, and Hugo Hopenhayn.** 2008. "Trading Favors: Optimal Exchange and Forgiveness." Collegio Carlo Alberto Carlo Alberto Notebooks 88.
- Jackson, Matthew O, and Hugo F Sonnenschein.** 2007. "Overcoming Incentive Constraints by Linking Decisions." *Econometrica*, 75(1): 241–257.
- Kaplan, Todd R., and Bradley J. Ruffle.** 2012. "Which Way to Cooperate." *The Economic Journal*, 122(563): 1042–1068.

- Kuzmics, Christoph, Thomas R. Palfrey, and Brian W. Rogers.** 2012. “Symmetric Play in Repeated Allocation Games.” *SSRN Electronic Journal*.
- Lau, C. Oscar.** 2011. “Soft Transactoins.” Department of Economics, Michigan State University.
- Miller, David A.** 2012. “Robust Collusion with Private Information.” *The Review of Economic Studies*, 79(2): 778–811.
- Mobius, Markus.** 2001. “Trading Favors.”
- Neill, Daniel B.** 2003. “Cooperation and Coordination in the Turn-taking Dilemma.” *TARK '03*, 231–244. New York, NY, USA:ACM.
- Olszewski, Wojciech, and Mikhail Safronov.** 2012. “Chip Strategies in Repeated Games with Incomplete Information.” Working paper, Department of Economics, Northwestern University.
- Portugal, Steven J.** 2014. “Abstract from Working Paper on Ibis Flights.” *private communication*.
- Roy, Nilanjan.** 2012. “Cooperation Without Immediate Reciprocity: An Experiment in Favor Exchange.” Caltech.
- Sacks, Harvey, Emanuel A Schegloff, and Gail Jefferson.** 1974. “A simplest systematics for the organization of turn-taking for conversation.” *language*, 696–735.
- Shao, Ran, and Lin Zhou.** 2013. “Optimal Allocation of an Indivisible Good.”
- Sheridan, Mary, Ajayand Sharma, and Helen Cockerill.** 2014. “From Birth to Five Years: Children’s Developmental Progress.” . 4th ed. Routledge.
- Sherstyuk, Katerina, Nori Tarui, and Tatsuyoshi Saijo.** 2013. “Payment Schemes in Infinite-Horizon Experimental Games.” *Experimental Economics*, 16(1): 125–153.
- Skrzypacz, Andrzej, and Hugo Hopenhayn.** 2004. “Tacit Collusion in Repeated Auctions.” *Journal of Economic Theory*, 114(1): 153–169.
- Stivers, Tanya, Nicholas J Enfield, Penelope Brown, Christina Englert, Makoto Hayashi, Trine Heinemann, Gertie Hoymann, Federico Rossano, Jan Peter**

- De Ruiter, Kyung-Eun Yoon, et al.** 2009. “Universals and cultural variation in turn-taking in conversation.” *Proceedings of the National Academy of Sciences*, 106(26): 10587–10592.
- Vespa, Emanuel.** 2011. “Cooperation in Dynamic Games: An Experimental Investigation.”
- Weesie, Jeroen.** 1994. “Incomplete Information and Timing in the Volunteer’s Dilemma: A Comparison of Four Models.” *Journal of Conflict Resolution*, 38: 557–585.
- Weesie, Jeroen, and Axel Franzen.** 1998. “Cost Sharing in a Volunteer’s Dilemma.” *Journal of Conflict Resolution*, 42: 600–618.

Online Appendix for *Taking Turns*

Greg Leo

V Appendix

A Proof of Lemma 2

Lemma 2: g increases strictly over the interval $[0, E(\theta_i)]$.

$$(23) \quad g(\tilde{\theta}) = \frac{1}{\beta} \tilde{\theta} + (2q-1) F(\tilde{\theta})^2 [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})] - (1 - F(\tilde{\theta}))^2 [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}]$$

The following two relationships are useful in re-writing this equation:

$$(24) \quad F(\tilde{\theta}) E(\theta_i | \theta_i \leq \tilde{\theta}) = F(\tilde{\theta}) \frac{\int_0^{\tilde{\theta}} \zeta f(\zeta) d\zeta}{F(\tilde{\theta})} = \int_0^{\tilde{\theta}} \zeta f(\zeta) d\zeta$$

$$(25) \quad (1 - F(\tilde{\theta})) E(\theta_i | \theta_i \geq \tilde{\theta}) = (1 - F(\tilde{\theta})) \frac{\int_{\tilde{\theta}}^1 \zeta f(\zeta) d\zeta}{1 - F(\tilde{\theta})} = \int_{\tilde{\theta}}^1 \zeta f(\zeta) d\zeta$$

Using these relationships, equation 23 can be rewritten:

$$(26) \quad g(\tilde{\theta}) = \tilde{\theta} \left(\frac{1}{\beta} + (2q-1) F(\tilde{\theta})^2 + (1 - F(\tilde{\theta}))^2 \right) - \left((2q-1) F(\tilde{\theta}) \int_0^{\tilde{\theta}} \zeta f(\zeta) d\zeta + (1 - F(\tilde{\theta})) \int_{\tilde{\theta}}^1 \zeta f(\zeta) d\zeta \right)$$

The derivative of the term on the first line of equation 26 with respect to $\tilde{\theta}$ is:

$$(27) \quad \left(\frac{1}{\beta} + (2q-1) F(\tilde{\theta})^2 + (1 - F(\tilde{\theta}))^2 \right) + 2(2q-1) \tilde{\theta} f(\tilde{\theta}) F(\tilde{\theta}) - 2\tilde{\theta} f(\tilde{\theta}) (1 - F(\tilde{\theta}))$$

The derivative of the term on the second line of equation 26 with respect to $\tilde{\theta}$ is:

$$(28) \quad - (2q-1) f(\tilde{\theta}) \int_0^{\tilde{\theta}} \zeta f(\zeta) d\zeta - (2q-1) \tilde{\theta} F(\tilde{\theta}) f(\tilde{\theta}) \\ + f(\tilde{\theta}) \int_{\tilde{\theta}}^1 \zeta f(\zeta) d\zeta + \tilde{\theta} (1-F(\tilde{\theta})) f(\tilde{\theta})$$

Together:

$$(29) \quad g'(\tilde{\theta}) = \left(\frac{1}{\beta} + (2q-1) F(\tilde{\theta})^2 + (1-F(\tilde{\theta}))^2 \right) \\ + 2(2q-1) \tilde{\theta} f(\tilde{\theta}) F(\tilde{\theta}) - 2\tilde{\theta} f(\tilde{\theta}) (1-F(\tilde{\theta})) \\ - (2q-1) f(\tilde{\theta}) \int_0^{\tilde{\theta}} \zeta f(\zeta) d\zeta - (2q-1) \tilde{\theta} F(\tilde{\theta}) f(\tilde{\theta}) \\ + f(\tilde{\theta}) \int_{\tilde{\theta}}^1 \zeta f(\zeta) d\zeta + \tilde{\theta} (1-F(\tilde{\theta})) f(\tilde{\theta})$$

This simplifies to:

$$(30) \quad g'(\tilde{\theta}) = (2q-1) \left[F(\tilde{\theta})^2 + f(\tilde{\theta}) F(\tilde{\theta}) [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})] \right] \\ + \frac{1}{\beta} + (1-F(\tilde{\theta}))^2 + f(\tilde{\theta}) (1-F(\tilde{\theta})) [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}]$$

Since the term multiplying $2q-1$ must be positive, we may bound the derivative below by choosing $\beta = 1$ and $q = 0$.

$$(31) \quad g'(\tilde{\theta}) \geq 2(1-F(\tilde{\theta})) - f(\tilde{\theta}) F(\tilde{\theta}) [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})] \\ + f(\tilde{\theta}) (1-F(\tilde{\theta})) [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}]$$

Thus, $g(\cdot)$ is increasing if:

$$(32) \quad 2(1-F(\tilde{\theta})) + f(\tilde{\theta}) (1-F(\tilde{\theta})) [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}] \\ \geq f(\tilde{\theta}) F(\tilde{\theta}) [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})]$$

Since $2(1 - F(\tilde{\theta})) \geq 0$, the following is sufficient:

$$(33) \quad f(\tilde{\theta})(1 - F(\tilde{\theta})) [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}] \geq f(\tilde{\theta})F(\tilde{\theta}) [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})]$$

Since, by assumption, density is strictly positive everywhere, this simplifies to:

$$(34) \quad (1 - F(\tilde{\theta})) [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}] \geq F(\tilde{\theta}) [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})]$$

Noting that $(1 - F(\tilde{\theta}))E(\theta_i | \theta_i \geq \tilde{\theta}) + F(\tilde{\theta})E(\theta_i | \theta_i \leq \tilde{\theta}) = E(\theta_i)$ (refer to 24 and 25), this simplifies to the following sufficient condition for increasing $g(\cdot)$.

$$(35) \quad E(\theta_i) \geq \tilde{\theta}$$

B Proof of Lemma 3

Proof. g is strictly positive for the interval $(\frac{1}{2}E(\theta_i), 1]$.

$$(36) \quad \begin{aligned} g(\tilde{\theta}) &= \frac{1}{\beta}\tilde{\theta} + (2q - 1)F(\tilde{\theta})^2 [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})] \\ &- (1 - F(\tilde{\theta}))^2 [E(\theta_i | \theta_i \geq \tilde{\theta}) - \tilde{\theta}] > 0 \end{aligned}$$

Since $[\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})] \geq 0$ and $\frac{1}{\beta}\tilde{\theta}$ is decreasing in β , let $q = 0$ and $\beta = 1$. This results in the following sufficient condition:

$$(37) \quad 2(1 - F(\tilde{\theta}))\tilde{\theta} + F(\tilde{\theta}) [E(\theta_i | \theta_i \leq \tilde{\theta})] > (1 - F(\tilde{\theta})) [E(\theta_i)]$$

When $\tilde{\theta} = 1$, this simplifies to $E(\theta_i) > 0$, which is true by assumption that density is strictly positive everywhere. Over the interval $(\frac{1}{2}E(\theta_i), 1)$, drop the positive term $(F(\tilde{\theta})) [E(\theta_i | \theta_i \leq \tilde{\theta})]$ to yield yet another sufficient condition:

$$(38) \quad 2(1 - F(\tilde{\theta}))\tilde{\theta} > (1 - F(\tilde{\theta})) [E(\theta_i)]$$

Since density is strictly positive everywhere, $1 - F(\tilde{\theta}) > 0$ over the interval $(\frac{1}{2}E(\theta_i), 1)$. Eliminating this term yields the desired result.

$$(39) \quad \tilde{\theta} > \frac{1}{2}E(\theta_i)$$

□

C Bound on Difference in Equilibrium Thresholds of Option and Deferment

Let θ_d^* and θ_o^* be the equilibria for the same parameters under deferment and option respectively.

Claim 11. For any distribution where $\frac{1}{2}E(\theta_i)$ is below the median, $\theta_o^* - \theta_d^* \leq 2F(\theta_d^*)^2 [\theta_d^* - E(\theta_i | \theta_i \leq \theta_d^*)]$.

Proof. Define $g_o(\tilde{\theta})$ and $g_d(\tilde{\theta})$ to be the option and deferment versions of the equilibrium root equation 3. From equation 3, the difference $g_d(\tilde{\theta}) - g_o(\tilde{\theta})$ is given by:

$$(40) \quad g_d(\tilde{\theta}) - g_o(\tilde{\theta}) = 2F(\tilde{\theta})^2 [\tilde{\theta} - E(\theta_i | \theta_i \leq \tilde{\theta})]$$

Evaluate this expression at the unique equilibrium point for the deferment version θ_d^* , at which $g_d(\theta_d^*) = 0$.

$$(41) \quad g_o(\theta_d^*) = -2F(\theta_d^*)^2 [\theta_d^* - E(\theta_i | \theta_i \leq \theta_d^*)]$$

The slope of g_o is greater than $2(1 - F(\tilde{\theta}))$ over the interval $[0, E(\theta_i)]$ by the results in A. Further, for any distribution in which $\frac{1}{2}E(\theta_i)$ is below the median, $2(1 - F(\tilde{\theta})) \geq 1$ for any $\tilde{\theta} \in [0, \frac{1}{2}E(\theta_i)]$. When this is true, the slope of g_o is greater than 1 over $[0, \frac{1}{2}E(\theta_i)]$. By proposition 4, all equilibria are in the interval $[0, \frac{1}{2}E(\theta_i)]$. Together, this implies:

$$(42) \quad g_o(\hat{\theta}) = 0 \text{ for some } \hat{\theta} \in \left[\theta_d^*, \theta_d^* + 2F(\theta_d^*)^2 [\theta_d^* - E(\theta_i | \theta_i \leq \theta_d^*)] \right]$$

Alternatively:

$$(43) \quad \theta_o^* \leq \theta_d^* + 2F(\theta_d^*)^2 [\theta_d^* - E(\theta_i | \theta_i \leq \theta_d^*)]$$

And thus,

$$(44) \quad \theta_o^* - \theta_d^* \leq 2F(\theta_d^*)^2 [\theta_d^* - E(\theta_i | \theta_i \leq \theta_d^*)]$$

□

This bound applies for any symmetric distribution, since the median is $E(\theta_i)$. When f is uniform, this bound is $\theta_o^* - \theta_d^* \leq \frac{1}{64}$. Notice that, for distributions with concave density, this term may be small. $2F(\frac{1}{2}E(\theta_1))^2$ is small due to low-tail weight, and $[\frac{1}{2}E(\theta_1) - E(\theta_i | \theta_i \leq \frac{1}{2}E(\theta_1))]$ is small due to the fact that density is shifted towards the mean. For instance, in the beta distribution with both shape parameters equal to 2, the difference is: $\theta_o^* - \theta_d^* \leq \frac{35}{8192} \approx 0.0043$. For distributions with convex density, the bound may be larger, indicating the possibility that the choice of tie-breaking rule might have a larger impact on play. For instance, the beta distributions with both shape parameters equal to $\frac{1}{2}$ has a bound: $\theta_o^* - \theta_d^* \leq 0.0364$.

D Suboptimality of Recurring Rotation

Lemma 12. Any threshold $\tilde{\theta}$ may be implemented by an ex-post incentive compatible mechanism using budget balanced money transfers.

Proof. Consider a “threshold mechanism” using transfers of $\frac{\tilde{\theta}}{2}$. The player completing the task will be paid $\frac{\tilde{\theta}}{2}$ by the other. Both players report whether they are above or below the threshold. If both are above or both are below, one is chosen at random to execute the task. Otherwise, the player reporting a low cost completes the task.

Note that a player prefers to complete the task and receive a payment of $\frac{\tilde{\theta}}{2}$ rather than pay $\frac{\tilde{\theta}}{2}$, as long as that player has a cost below $\tilde{\theta}$. Otherwise, the player prefers not to complete the task and pay $\frac{\tilde{\theta}}{2}$. Regardless of the other player’s choice,

offering to do the task maximizes a player's probability of doing the task. Not-offering maximizes the probability of not doing the task. Thus, a player with $\theta_i \leq \tilde{\theta}$ has a weakly dominant strategy of offering to do the task while a player with $\theta_i \geq \tilde{\theta}$ has a weakly dominant strategy of not-offering. Because of this, reporting costs according to the threshold is dominant strategy (ex-post) incentive compatible. \square

Lemma 13. For any $F(\cdot)$, the threshold mechanism that maximizes welfare (minimizes average stage cost) uses a threshold equal to the mean type.

Proof. By Lemma 12, any threshold mechanism can be implemented by an ex-post incentive compatible mechanism with budget-balanced transfers. These transfers are welfare neutral, and so the mechanism's welfare is given by its average stage cost. The average stage cost under a threshold mechanism is given by:

$$(45) \quad \bar{AC}(\tilde{\theta}) = F(\tilde{\theta}) E(\theta_i | \theta_i \leq \tilde{\theta}) + (1 - F(\tilde{\theta})) E(\theta_i)$$

It's derivative is:

$$(46) \quad \frac{\delta \bar{AC}(\tilde{\theta})}{\delta \tilde{\theta}} = f(\tilde{\theta}) \tilde{\theta} - f(\tilde{\theta}) E(\theta_i)$$

Since $\frac{\delta \bar{AC}(\tilde{\theta})}{\delta \tilde{\theta}}$ is negative below the mean and positive above the mean, $\bar{AC}(\tilde{\theta})$ is quasi-convex and reaches its global minimum at $\tilde{\theta} = E(\theta_i)$. \square

Proposition 14. For distributions with positive density everywhere, the average stage cost of recurring rotation is bounded above second best optimal. The stage average welfare gap is at least $\bar{AC}\left(\frac{E(\theta_i)}{2}\right) - \bar{AC}(E(\theta_i))$.

Proof. By lemmas Lemma 12 and Lemma 13, the optimal threshold mechanism incurring average stage cost $\bar{AC}(E(\theta_i))$ can be implemented by a robust mechanism. Thus, the second-best optimal average stage cost must be at least as small as $\bar{AC}(E(\theta_i))$.

By proposition 4, the equilibrium threshold in recurring rotation is always below half of the mean type. The average stage cost associated with using a threshold $\tilde{\theta}$ is $\bar{AC}(\tilde{\theta}) = E(\theta_i) - F(\tilde{\theta})(E(\theta_i) - E(\theta_i | \theta_i \leq \tilde{\theta}))$. This is strictly decreasing over $[0, E(\theta_i)]$ and so the average cost of recurring rotation must be larger than $\bar{AC}\left(\frac{E(\theta_i)}{2}\right)$.

Together, this implies that the gap in welfare (in terms of average stage cost) between recurring rotation and a second-best optimal mechanism is at least $\bar{AC}\left(\frac{E(\theta_i)}{2}\right) - \bar{AC}(E(\theta_i)) > 0$. \square

Using the result above for the uniform distribution predicts a gap of at least $\frac{1}{32} \approx 0.0278$. In fact, the gap for the option version of the mechanism as $\beta \rightarrow 1$ is approximately $0.411 - 0.375 = 0.036$ since $\frac{3}{8} = 0.375$ is known to be the second-best optimal cost.

I do not claim that the mean-threshold mechanism provides a tight bound on the welfare achievable by robust mechanisms for any distribution except uniform. In fact, (Miller, 2012, example 4) provides a counter-example under which the mean-threshold mechanism is not optimal under a particular asymmetric distribution where density is $\frac{2}{5}$ for $\theta_i \leq \frac{1}{2}$ and $\frac{8}{5}$ for $\theta_i > \frac{1}{2}$. However, several results suggest that the best threshold mechanism may still be a good benchmark.

General results showing that threshold mechanisms are optimal among the class of *deterministic* mechanisms for type distributions with monotonic hazard rate are given in Drexl and Kleiner (2012); Shao and Zhou (2013). In addition, Hagerty and Rogerson (1987) prove that a threshold, specifically a *posted price* mechanism, is optimal in a trade setting under dominant strategy incentive compatibility, ex-post individual rationality and strong ex-post budget balance. Further, Athey and Miller (2007) reports that in numerical tests where the threshold mechanism was not optimal, the computed optimal mechanism improved efficiency very little.

Using the best threshold rather than the true optimal for comparison has the benefit that the best threshold mechanism is well defined by Lemma 12 and Lemma 13 as the mean threshold mechanism and has an average stage cost that is easy to calculate. This provides the flexibility to make efficiency comparisons in environments with cost distributions that are not uniform. Despite the fact that recurring rotation is suboptimal relative to the best threshold mechanism, in appendix subsection E I demonstrate that there are some environments where the welfare loss is small.

E Recurring Rotation Under Symmetric Beta Distributions

Consider costs distributed beta with equal shape parameters both set to γ . I focus on the symmetric beta distributions for their ability to represent a wide variety of spread patterns with a single parameter while nesting the uniform distribution, and maintaining the same mean $E(\theta_i) = \frac{1}{2}$. $\gamma = 1$ is the uniform distribution. For $\gamma < 1$,

the distribution is U-shaped and for $\gamma > 1$, the distribution is unimodal and concave.

The table below reports the equilibrium threshold and associated average costs for recurring rotation for various parameter levels γ as $\beta \rightarrow 1$ using equation 2. This is accomplished by taking approximations of the equilibrium threshold over an increasing sequence of β until that sequence converges.

γ	Deferment $q = 1$		Option $q = 0$		Other Mechanisms	
	θ^*	Equilibrium $\bar{A}C$	θ^*	Equilibrium $\bar{A}C$	Best Threshold $\bar{A}C$	First Best $\bar{A}C$
$\frac{1}{300}$	0.168	0.253	0.250	0.251	0.251	0.250
$\frac{1}{100}$	0.169	0.255	0.248	0.254	0.253	0.250
$\frac{1}{10}$	0.186	0.290	0.240	0.286	0.279	0.255
$\frac{1}{5}$	0.198	0.318	0.235	0.313	0.301	0.266
$\frac{1}{3}$	0.207	0.345	0.232	0.341	0.322	0.281
$\frac{1}{2}$	0.215	0.369	0.231	0.366	0.341	0.297
1	0.226	0.412	0.232	0.411	0.375	$\frac{1}{3}$
2	0.236	0.451	0.237	0.451	0.406	0.371
3	0.240	0.470	0.241	0.469	0.422	0.392
5	0.245	0.486	0.245	0.486	0.438	0.414
10	0.249	0.498	0.249	0.498	0.456	0.438
50	0.250	0.500	0.250	0.500	0.480	0.472

Table 1: Computed Equilibrium Values for Beta Distribution

It is noted that the thresholds diverge most sharply among the U-shaped distributions (small γ). This was predicted to some extent by the bound given in equation 44. Despite this, the difference in average cost of the two mechanism versions remains small over the whole interval. The mechanism performs well relative to both best threshold and first-best at the extremes. When γ is large, the distribution becomes nearly a mass point at the mean. In this case, the problem of private costs disappears and *any* assignment results in nearly the same cost of execution. On the other hand, when γ is small, the distribution becomes nearly a binary distribution with either high or low cost. Private information is still a problem. However, any threshold mechanism will result in nearly first-best efficiency. When both are on the opposite side of the threshold, the efficient outcome is achieved. When both are on the same side, the two costs are nearly identical and so any assignment inefficiencies are small.

In this family, the recurring rotation mechanism compares least favorably to first-best and best threshold under cost distributions that have a relatively high en-

tropy (relatively flat). However, even in these cases, recurring rotation offers substantial welfare improvement over rigid turn-taking, which results in an average cost of $\frac{1}{2}$ for any distribution in the symmetric beta family.

F Calculating $V_1(z)$ For Obligation Takeover

The value function can be constructed from a set of threshold strategies. Below I derive expressions for the value function of player 1. The value function for player 2 is simply the reverse. For instance, $V_2(z) = V_1(-z)$ just as $\tilde{\theta}_2(z) = \tilde{\theta}_1(-z)$.

When obligated, ($z \geq 1$), Player 1 pays an average cost of $E(\theta_1)$ when player 2 has cost above threshold $\tilde{\theta}(-z)$. The probability of this occurrence is $(1 - F(\tilde{\theta}(-z)))$. On the other hand, player 2 is below this threshold with probability $F(\tilde{\theta}(-z))$. In this case, if player 1 has cost below threshold $\tilde{\theta}(z)$, then there is no swap and player 1 pays an average cost of $E(\theta_1 | \theta_1 \leq \tilde{\theta}(z))$. If player 1 is above threshold $\tilde{\theta}(z)$, there is a swap. Player 1 does not pay. Thus, the average stage cost for player 1 in state z is:

$$(47) \quad -E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) F(\tilde{\theta}(-z)) - E(\theta_i) (1 - F(\tilde{\theta}(-z)))$$

For $z > 1$, The state moves to $z - 1$ unless there is a swap (for $z = 1$, a slight correction is needed to account for the fact that there is no state 0). The discounted continuation value of this is $\beta(V(z - 1))$. Only if there is a swap, which happens with probability $(1 - F(\tilde{\theta}(z))) F(\tilde{\theta}(-z))$, does the state move to $z + 1$. The discounted continuation value of this is $\beta(V(z + 1))$. The continuation portion of the value function can thus be written:

$$(48) \quad \beta(V(z - 1)) - \beta(V_1(z - 1) - V_1(z + 1)) (1 - F(\tilde{\theta}(z))) F(\tilde{\theta}(-z))$$

In equilibrium however, $\beta(V_1(z - 1) - V_1(z + 1)) = \tilde{\theta}(z)$. Imposing this relationship, 48 can be rewritten:

$$(49) \quad \beta(V_1(z - 1)) - \tilde{\theta}(z) (1 - F(\tilde{\theta}(z))) F(\tilde{\theta}(-z))$$

Adding together expressions 47 and 49, a player's value function in an obligated

state can be written:

$$(50) \quad \forall z > 1 : V_1(z) = \beta (V_1(z-1)) - \tilde{\theta}(z) (1 - F(\tilde{\theta}(z))) F(\tilde{\theta}(-z)) \\ - E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) F(\tilde{\theta}(-z)) - E(\theta_i) (1 - F(\tilde{\theta}(-z)))$$

To simplify this expression, note that the term below represents the average cost paid by the obligated player in a stage with state z under strategy $\tilde{\theta}$. Denote this cost $C(z, \tilde{\theta})$.

$$(51) \quad C(z, \tilde{\theta}) = E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) F(\tilde{\theta}(-z)) + E(\theta_i) (1 - F(\tilde{\theta}(-z)))$$

The following term is the expected transfer of continuation utility associated with a deferment. This is denoted $T(z, \tilde{\theta})$.

$$(52) \quad T(z, \tilde{\theta}) = \tilde{\theta}(z) (1 - F(\tilde{\theta}(z))) F(\tilde{\theta}(-z))$$

Together, for $z > 1$:

$$(53) \quad \forall z > 1 : V_1(z) = \beta (V_1(z-1)) - C(z, \theta) - T(z, \theta)$$

In any state with $z < -1$ such that player 1 is non-obligated, the “status quo” is returning to a higher state, which has discounted value $\beta V_1(z+1)$. Only if player 2 is above threshold $\tilde{\theta}(-z)$ and player 1 is below threshold $\tilde{\theta}(z)$ does player 1 pay an average cost $E(\theta_1 | \theta_1 \leq \tilde{\theta}(z))$ and receive continuation utility transfer $\beta (V_1(z-1) - V_1(z+1)) = \tilde{\theta}(z)$ due to the state-change. In this way, a player’s utility in a non-obligated state can be written:

$$(54) \quad \forall z < -1 : V_1(z) = \beta (V_1(z+1)) + \tilde{\theta}(z) F(\tilde{\theta}(z)) (1 - F(\tilde{\theta}(-z))) \\ - E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) (1 - F(\tilde{\theta}(-z)))$$

As in the obligated states, the following terms represent the average stage cost paid by the non-obligated player and the expected continuation transfer associated with a deferment:

$$(55) \quad C(z, \tilde{\theta}) = E(\theta_i | \theta_i \leq \tilde{\theta}(z)) F(\tilde{\theta}(z)) (1 - F(\tilde{\theta}(-z)))$$

$$(56) \quad T(z, \tilde{\theta}) = \tilde{\theta}(z) F(\tilde{\theta}(z)) (1 - F(\tilde{\theta}(-z)))$$

Non-obligated continuation value can be written:

$$(57) \quad \forall z < -1 : V_1(z) = \beta(V(z+1)) - C(z, \theta) + T(z, \theta)$$

For states $z = 1$ and $z = -1$, a slight modification must be made to account for the fact that there is no state 0.

$$(58) \quad V_1(1) = \beta(V(-1)) - C(1, \theta) - T(1, \theta)$$

$$(59) \quad V_1(-1) = \beta(V(1)) - C(-1, \theta) + T(-1, \theta)$$

G Proof of Proposition 9

Letting $\tilde{\theta}(z) = E(\theta_i)$ for all z , for any $|z| \geq 2$, $C(z+1, \tilde{\theta}) = C(z-1, \tilde{\theta})$ and $T(z+1, \tilde{\theta}) = T(z-1, \tilde{\theta})$. Thus, the fixed point conditions are met for any state with $|z| \geq 2$. For $z = 1$ and $z = -1$, the following must be true:

$$(60) \quad E(\theta_i) = C(2, \tilde{\theta}) - C(-1, \tilde{\theta}) + T(-1, \tilde{\theta}) + T(2, \tilde{\theta})$$

$$(61) \quad E(\theta_i) = C(1, \tilde{\theta}) - C(-2, \tilde{\theta}) + T(-2, \tilde{\theta}) + T(1, \tilde{\theta})$$

When $\tilde{\theta}(z) = E(\theta_i)$ for all z , it is again true that $T(-1) = T(-2) = T(1) = T(2) = \frac{1}{4}E(\theta_i)$. While $C(2, \tilde{\theta}) = C(1, \tilde{\theta})$ and $C(-2, \tilde{\theta}) = C(-1, \tilde{\theta})$, C functions in states with opposite signs are not equal. The equilibrium condition simplifies to checking:

(62)

$$E(\theta_i) = E(\theta_i | \theta_i \leq E(\theta_i)) F(E(\theta_i)) F(E(\theta_i)) \\ + E(\theta_i | \theta_i \geq E(\theta_i)) (1 - F(E(\theta_i))) (1 - F(E(\theta_i))) + \frac{1}{2} E(\theta_i)$$

Since, $E(\theta_i) = E(\theta_i | \theta_i \leq E(\theta_i)) F(E(\theta_i)) + E(\theta_i | \theta_i \geq E(\theta_i)) (1 - F(E(\theta_i)))$,
this is:

(63)

$$E(\theta_i) = E(\theta_i | \theta_i \leq E(\theta_i)) F(E(\theta_i)) F(E(\theta_i)) + \\ + E(\theta_i | \theta_i \leq E(\theta_i)) F(E(\theta_i)) (1 - F(E(\theta_i))) + E(\theta_i | \theta_i \geq E(\theta_i)) (1 - F(E(\theta_i)))^2 \\ - E(\theta_i | \theta_i \leq E(\theta_i)) F(E(\theta_i)) (1 - F(E(\theta_i))) + \frac{1}{2} E(\theta_i)$$

Which simplifies to:

$$(64) \quad E(\theta_i) = E(\theta_i | \theta_i \leq E(\theta_i)) F(E(\theta_i)) F(E(\theta_i)) \\ + E(\theta_i | \theta_i \geq E(\theta_i)) (1 - F(E(\theta_i)))^2 + \frac{1}{2} E(\theta_i)$$

For any symmetric distribution, $F(E(\theta_i)) = 1 - F(E(\theta_i)) = \frac{1}{2}$. Thus, this simplifies to:

$$(65) \quad E(\theta_i) = \frac{1}{2} E(\theta_i) + \frac{1}{2} E(\theta_i) = E(\theta_i)$$

Thus, $\tilde{\theta}(z) = E(\theta_i)$ for all z , solves the fixed point problem for any symmetric distribution with patient players.

H Obligation Takeover - Uniform Distribution, Limited States

When players implement threshold strategies close to $\frac{1}{2}$, the probability of being in a state of absolute value $|z|$ is approximately $\frac{2}{3|z|}$. Because states with a large number of obligations are rarely visited, even if strategies in those states are inefficient, this distortion will not have much effect on the overall efficiency of the mechanism. As

long as the boundary is rarely reached, truncating the number of states only has a substantial effect on the mechanism to the extent that it distorts strategies in the lower states. In the uniform case, this distortion tends to be small.

The plot in figure below is a contour of the average equilibrium cost of the mechanism. On the y-axis is the number of states used and the x-axis is the discount factor β . The black lines are equi-cost (average joint cost) lines in .005 increments. Notice these are nearly vertical above $\bar{z} = 5$. This suggests that little is gained, especially for lower discount factors, by extending the limit beyond 5 obligations.

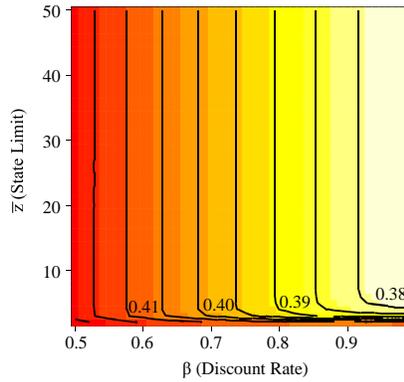


Figure 8: Average Cost Contour of Obligation Takeover (Deferment) in β and \bar{z} .

I Obligation Takeover with Unlimited States Under Symmetric Beta Distribution

Here, I consider beta distributions with identical shape parameters equal to γ . In the table below, the average cost of the obligation takeover mechanism is given for a variety of β and γ . To the right, I provide the average costs of the option version of the recurring rotation (*RR*) as $\beta \rightarrow 1$, the best threshold mechanism (*BT*) and the first best average cost (*FB*).

γ	Obligation Takeover					Other Mechanisms		
	$\beta = 0.5$	$\beta = 0.8$	$\beta = 0.9$	$\beta = 0.95$	$\beta = 0.99$	<i>RR</i>	<i>BT</i>	<i>FB</i>
$\frac{1}{300}$	0.252	0.251	0.251	0.251	0.251	0.251	0.251	0.250
$\frac{1}{100}$	0.255	0.254	0.254	0.253	0.253	0.254	0.253	0.250
$\frac{1}{10}$	0.294	0.284	0.281	0.280	0.279	0.286	0.279	0.255
$\frac{1}{5}$	0.323	0.308	0.303	0.302	0.301	0.313	0.301	0.266
$\frac{1}{3}$	0.351	0.331	0.326	0.323	0.322	0.341	0.322	0.281
$\frac{1}{2}$	0.376	0.352	0.346	0.343	0.341	0.366	0.341	0.297
1	0.418	0.389	0.381	0.377	0.375	0.411	0.375	$\frac{1}{3}$
2	0.454	0.423	0.414	0.409	0.407	0.451	0.406	0.371
3	0.471	0.440	0.430	0.426	0.422	0.469	0.422	0.392
5	0.487	0.458	0.448	0.443	0.439	0.486	0.438	0.414
10	0.498	0.476	0.466	0.460	0.457	0.498	0.456	0.438
50	0.500	0.498	0.490	0.485	0.481	0.500	0.480	0.472

Table 2: Computed Equilibrium Average Costs for Various Mechanisms Under Symmetric Beta Distribution

As is predicted by proposition 9, obligation takeover closely approximates the efficiency of the best threshold mechanism for β near 1. For the U-shaped distributions (small γ) the efficiency of the obligation takeover mechanism as well as the recurring rotation mechanism and the best threshold all come close to first-best. As discussed in subsection E of this appendix, this is because, for the extreme unimodal distributions, the assignment problem disappears, and for the U-shaped distributions, *any* threshold between the modes achieves nearly first best.

J Experiment Figures

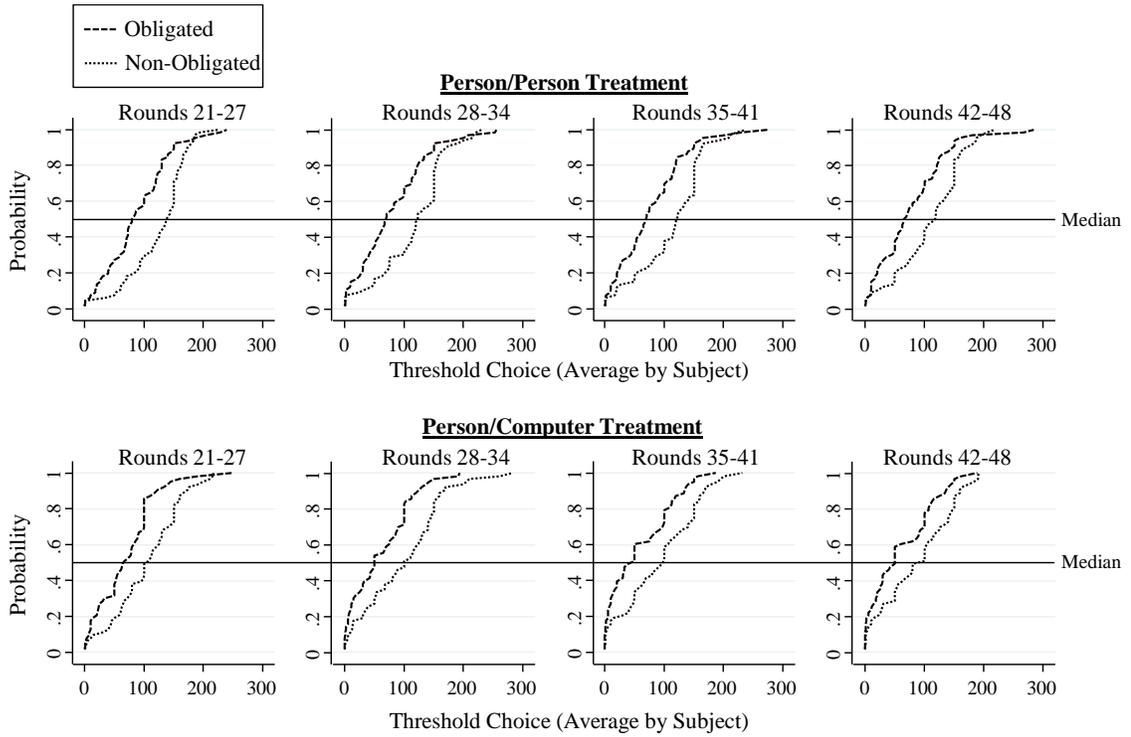


Figure 9: CDF of Average Threshold Choices (By Subject) Over 7 Round Blocks

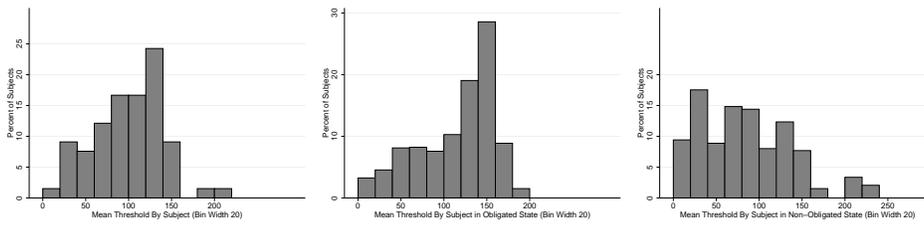


Figure 10: Histograms of average threshold choice by subject. Rounds 21-48. (Person/Person Treatment)

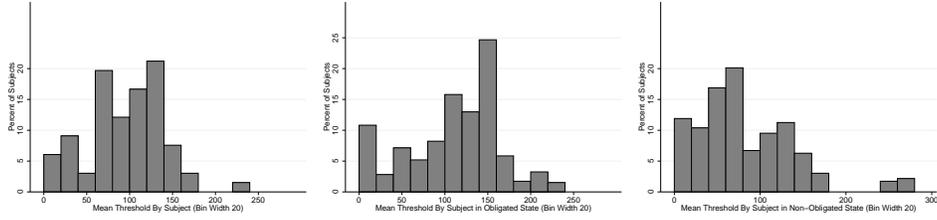


Figure 11: Histograms of average threshold choice by subject. Rounds 35-48. (Person/Person Treatment)

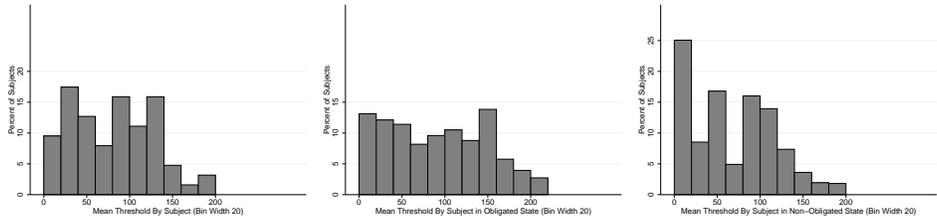


Figure 12: Histograms of average threshold choice by subject. Rounds 21-48. (Person/Computer Treatment)

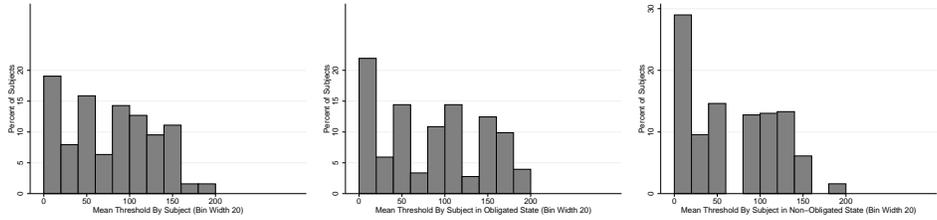


Figure 13: Histograms of average threshold choice by subject. Rounds 35-48. (Person/Computer Treatment)

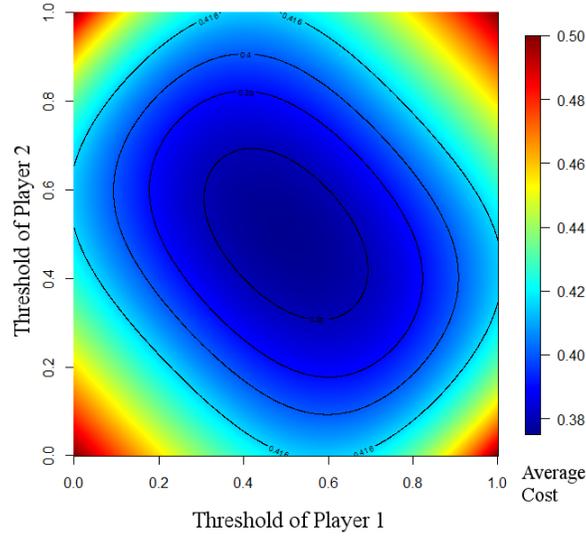


Figure 14: Disequilibrium Efficiency for Recurring Rotation (Deferment)

K Experiment Time Trend Analysis

Subjects' thresholds are estimated by an individual constant α_i plus some individual-specific change δ_i when the player is obligated ($o_{i,t} = 1$ when subject i is obligated in period t). Trends in average obligated and non-obligated thresholds are estimated using dummy variables for the same 7-period blocks in figure 9. The estimates of these trend coefficients are provided in the tables below.

$$\begin{aligned}
 \theta_{i,t} = & \alpha_i + \delta_i o_{i,t} + I_{(28 \leq t \leq 34)} \left(\gamma_{(28 \leq t \leq 34)}^n + o_{i,t} \gamma_{(28 \leq t \leq 34)}^o \right) \\
 (66) \quad & I_{(35 \leq t \leq 41)} \left(\gamma_{(35 \leq t \leq 41)}^n + o_{i,t} \gamma_{(35 \leq t \leq 41)}^o \right) + \\
 & I_{(42 \leq t \leq 48)} \left(\gamma_{(42 \leq t \leq 48)}^n + o_{i,t} \gamma_{(42 \leq t \leq 48)}^o \right) + \varepsilon_{i,t}
 \end{aligned}$$

In the person/person treatment, the average thresholds chosen by non-obligated players in rounds 35 – 41 and 42 – 48 were both significantly different from those chosen in rounds 21 – 28. Further, obligated players in rounds 42 – 48 chose thresholds significantly lower than obligated players in rounds 21 – 28. However, neither obligated nor non-obligated players' average thresholds differ significantly over any of the 7 round groupings after round 27. This suggests any learning and adjustment

in the experiment happened over the “action” rounds 1 – 20 and the early threshold submission rounds, but had settled down by round 35.

In the person/computer treatment, average non-obligated thresholds in rounds 28 – 41 were significantly different than in 21 – 27. While the non-obligated thresholds in rounds 42 – 48 were not significantly different, they were also not significantly different than those in rounds 28 – 34 and 35 – 41. Obligated thresholds in rounds 35 – 48 were significantly different than in rounds 21 – 27. Though the estimates of the obligated difference coefficients in rounds 35 – 41 and 42 – 48 are significantly different, the average threshold chosen by obligated players in those rounds measured by the sums $\gamma_{(35 \leq t \leq 41)}^n + \gamma_{(35 \leq t \leq 41)}^o$ and $\gamma_{(42 \leq t \leq 48)}^n + \gamma_{(42 \leq t \leq 48)}^o$ are not significantly different. This does not provide any overwhelming evidence against focusing on rounds 35 – 48, as in the person/person treatment.

Coefficient	Point Estimate	Standard Error ³⁵	<i>t</i> -statistic
$\gamma_{(28 \leq t \leq 34)}^n$	-8.69	5.24	-1.66
$\gamma_{(35 \leq t \leq 41)}^n$	-12.17	4.39	-2.77
$\gamma_{(42 \leq t \leq 48)}^n$	-10.43	5.02	-2.08
$\gamma_{(28 \leq t \leq 34)}^o$	1.29	8.37	0.15
$\gamma_{(35 \leq t \leq 41)}^o$	2.58	7.39	0.35
$\gamma_{(42 \leq t \leq 48)}^o$	-4.58	7.85	-0.58

Table 3: Coefficient Estimates for Model 66: person/person treatment.

Coefficient	Point Estimate	Standard Error ³⁶	<i>t</i> -statistic
$\gamma_{(28 \leq t \leq 34)}^n$	-9.16	4.02	-2.28
$\gamma_{(35 \leq t \leq 41)}^n$	-14.24	6.09	-2.34
$\gamma_{(42 \leq t \leq 48)}^n$	-8.32	5.15	-1.62
$\gamma_{(28 \leq t \leq 34)}^o$	1.73	5.74	0.30
$\gamma_{(35 \leq t \leq 41)}^o$	0.62	5.62	0.11
$\gamma_{(42 \leq t \leq 48)}^o$	-8.98	5.31	-1.69

Table 4: Coefficient Estimates for Model 66: person/computer treatment.

L Procedure for Empirical Efficiency

Letting $\tilde{\theta}_{i,o}$ be player i 's threshold strategy when player o is obligated, the equation for average cost under strategy profile $\tilde{\theta}$ in the state that i is obligated is given by:

$$(67) \quad \bar{AC}(\tilde{\theta}, i) = F(\tilde{\theta}_{i,i}) E(\theta_i | \theta_i \leq \tilde{\theta}_{i,i}) + F(\tilde{\theta}_{j,i}) [1 - F(\tilde{\theta}_{i,i})] E(\theta_j | \theta_j \leq \tilde{\theta}_{j,i}) \\ + [1 - F(\tilde{\theta}_{j,i})] [1 - F(\tilde{\theta}_{i,i})] E(\theta_i | \theta_i \geq \tilde{\theta}_{i,i})$$

Since the average stage cost depends on the state (who is obligated), the joint welfare in the long-run depends on the how often each state occurs. This was not the case when both players use the same strategy since the average stage cost is the same in both states and, in the long-run, each state is equally likely. The joint welfare of disequilibrium play with asymmetric strategies can be calculated by weighting the average cost in each state by the long-run visiting probabilities of the states. The sequence of which player is obligated follows a stationary Markov process and the probability player 1 is obligated in the long-run is:

$$(68) \quad \pi_1(\tilde{\theta}) = \frac{F(\tilde{\theta}_{1,2}) F(\tilde{\theta}_{2,2}) - F(\tilde{\theta}_{1,2}) + 1}{2 + F(\tilde{\theta}_{1,1}) F(\tilde{\theta}_{2,1}) - F(\tilde{\theta}_{2,1}) + F(\tilde{\theta}_{1,2}) F(\tilde{\theta}_{2,2}) - F(\tilde{\theta}_{1,2})}$$

The average stage cost under “gap” strategies $\tilde{\theta}$ is then given by:

$$(69) \quad \pi_1(\tilde{\theta}) \bar{AC}(\tilde{\theta}, 1) + \pi_2(\tilde{\theta}) \bar{AC}(\tilde{\theta}, 2)$$

M Analysis of Potential Explanations for “Gap” Strategies in Person/Person Treatment

Social Preferences

Suppose player j has a utility function that weighs personal outcomes equally with player i 's outcomes and that i plays according to some known threshold $\tilde{\theta}_i$ in both states. Conditional on i asking for a swap, player j will believe player i 's expected cost is $E(\theta_i | \theta_i \geq \tilde{\theta}_i)$. Player j will agree to a swap when $\theta_j < E(\theta_i | \theta_i \geq \tilde{\theta}_i)$. On the other hand, should i agree to a swap when obligated, j believes i has cost $E(\theta_i | \theta_i \leq \tilde{\theta}_i)$. j should only ask for a swap when $\theta_j > E(\theta_i | \theta_i \leq \tilde{\theta}_i)$. In this way,

gap strategies are consistent with altruism.

Further, these gaps can provide improved efficiency, even over second-best optimal. The efficiency of a single stage where player i is obligated and j is non-obligated is given in equation 67. For instance, in the uniform 0,1 environment, if both players choose threshold $\frac{1}{3}$ when obligated and $\frac{2}{3}$ when non-obligated, the average joint stage cost is 0.352 - an improvement over second best 0.375.

Equation 67 is plotted below for the uniform 0,1 environment. Grey points on this graph represent the average thresholds chosen by subjects (normalized to $[0, 1]$) in the obligated (x -axis) and non-obligated (y -axis) states for rounds 35 – 48. The value associated with the location of each point can be thought of as the efficiency that would be achieved if a subject was to be paired with partner playing the exact same strategy.

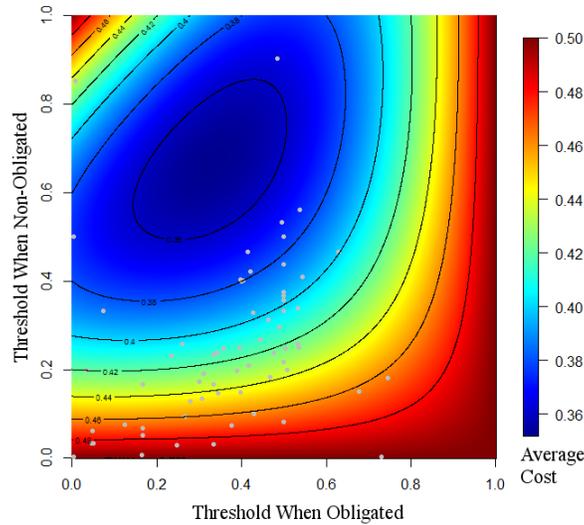


Figure 15: Average Stage Cost Under Asymmetric Strategies

Notice that greater efficiency (blue area) is achieved *above* the 45-degree line. In this area, the threshold is higher in the non-obligated state. This is also the direction of the gap implied by the example given above. Player j chooses a threshold that is higher in the non-obligated state ($E(\theta_i | \theta_i \geq \tilde{\theta}_i)$) than in the obligated state ($E(\theta_i | \theta_i \leq \tilde{\theta}_i)$). Instead, most subjects chose strategies in which their thresholds were higher in the *obligated* state. This explains why the gap

strategy is costly in terms of efficiency relative to symmetric equilibrium strategies.

Strategic Aspects

If a player believes that their partner's strategy is non-Markov, then the state of the mechanism is not enough to summarize the payoff relevant information. Gap strategies may be supported. For instance, suppose a player i believes player j will use a threshold that is different in the non-obligated state depending on how the players arrived at that state. For instance, j can become non-obligated because j was previously non-obligated but swapped with i or because j was actually obligated and no swap occurred. Suppose j uses a lower threshold when non-obligated after swapping with i . In this case, the cost of *remaining* obligated for i is higher than the cost of *becoming* obligated from a non-obligated state. Since the weakly dominant strategy is still to set threshold equal to the discounted cost associated with the change of states, i should respond with a higher threshold in the obligated state.

While this experiment was not explicitly designed to test the hypothesis that player's implement Markov strategies, it is possible to get a rough overview of non-Markov play by estimating players' threshold choices conditional both on the current state and aspects of the history of play - in this case, on the state in the previous round. Among Markov players, conditional on the current state, chosen strategies should be independent of the state in the previous round.

The model represented by equation 9 estimates a subject average threshold for the obligated and non-obligated state over rounds 35 – 48. This model is extended below in equation 70 to estimate average differences (over all subjects) in threshold choice dependent on the state of the previous round. Let $o_{i,t-1} = 1$ if the subject had the same partner last period and was obligated in that period. Then, $o_{i,t}o_{i,t-1} = 1$ if the subject was obligated both last period and the current period, while $(1 - o_{i,t})(1 - o_{i,t-1}) = 1$ if the subject was non-obligated in the last and current period (and had the same partner). In this way, it is possible to reject that, on average, player's strategies do not depend on the previous period's state by testing the hypotheses $\delta^0 = 0$ and $\delta^1 = 0$.

$$(70) \quad \theta_{i,t} = \alpha_i + \beta_i o_{i,t} + (\delta^0 (1 - o_{i,t})(1 - o_{i,t-1}) + \delta^1 o_{i,t}o_{i,t-1}) + \varepsilon_{i,t}$$

Although the estimates have meaningful magnitude, these hypotheses cannot be rejected. $\hat{\delta}^0 = -12.59$ with standard error³⁷ 11.68 ($t = -1.08$). $\hat{\delta}^1 = 14.31$ with standard error 8.88 ($t = 1.61$). A joint test of these coefficients is also insignificant. However, there is a minor endogeneity problem here. δ^0 and δ^1 in 70 measure the difference in the *average observation* conditional on state repetition. However, players implementing a low threshold when obligated are more likely to remain obligated, and players implementing a high threshold when non-obligated are more likely to remain non-obligated. Because of this, the observations in which $o_{i,t}o_{i,t-1} = 1$ are more likely to come from players already implementing low thresholds in the obligated state. Likewise, observations for which $(1 - o_{i,t})(1 - o_{i,t-1}) = 1$ are more likely to come from players implementing high thresholds in the non-obligated state. If there is a difference in the way these players change their behavior after a swap, then δ^0 and δ^1 do not measure how the *average player's* choice changes but rather some weighted average.

It is possible to instead calculate individual specific coefficients according to the regression expressed in equation 71. Over rounds, 35 – 48, when obligated twice in a row, 6 of the 66 subjects had an average threshold significantly higher than when obligated but not twice in a row. 4 had an average threshold significant lower. When non-obligated twice in a row, 6 significantly increased average threshold over being non-obligated but not twice in a row. 7 significantly decreased their average threshold. The remainder in each case did not have significantly different thresholds against a one sided alternative at the 5 percent level.

$$(71) \quad \theta_{i,t} = \alpha_i + \beta_i o_{i,t} + (\delta_i^0 (1 - o_{i,t})(1 - o_{i,t-1}) + \delta_i^1 o_{i,t}o_{i,t-1}) + \varepsilon_{i,t}$$

However, since the repeated states are already rarer, there is less data to work with. When the same model is estimated over rounds 21 – 48, 13 subjects significantly increased and 11 decreased their average thresholds when obligated twice in a row relative to when obligated but not twice in a row. 8 subjects significantly increased and 9 decreased their average threshold when non-obligated twice in a row. In general, while at least some subjects' behavior does not appear to be independent of this aspect of the history, it is not clear there is a systematic pattern.

Of course, this is only one possible source of history dependence, and subjects may have more sophisticated beliefs about their partners' strategies. A more de-

³⁷corrected for subject specific clusters

tailed analysis of Markov play could be a rather complex task (see [Vespa \(2011\)](#) for an experiment on testing the Markov assumption of subject behavior in a dynamic commons problem).

Experiment Instructions

In this experiment, you earn credits. You will receive \$1 for each 100 credits that you earn. You will start the experiment with 500 credits (\$5) and you will earn an additional 320 credits every 4 rounds of the experiment.

The experiment will take place over multiple rounds.

During these rounds you are paired with the same partner.

In each round, there is a task that must be performed either by you or your partner. The cost of performing the task will be X credits, where X is a random number that is equally likely to be any number between 0 and 300.

You will not know your partner's cost and your partner will not know your cost.

Who does the task is determined as follows. In any round, one of you is "obligated".

If you are obligated in a round, then you must perform the task, paying your random cost, unless you ask your partner to do it instead and he/she agrees to do so. If your partner agrees to do the task, then on the next round you will again be obligated. But if you do the task, then on the next round your partner will be obligated.

In each round, you will find out your random cost.

If you are obligated, you will be asked whether you would like to ask your partner to do the task.

If you are not obligated, you will be asked whether you will agree to do the task should your partner ask.

If you are obligated, you ask your partner to do the task and he/she agrees, your partner will complete the task, paying his/her cost, and you will remain obligated on the next round. Otherwise, you will complete the task, paying your cost, and become non-obligated on the next round.

If you are obligated, and you ask your partner to do the task, you will find out whether he/she agreed. If you do not ask your partner to do the task, you will still find out whether he/she would have agreed.

If you are non-obligated, you will find out whether your partner asked you to do the task.

In each case you will also find out your total earnings for the round, and whether you will be obligated on the next round.

After each round, there is a 90% chance you will continue playing with the same partner. There is a 10% chance you will be randomly paired with a new partner. In this case, one of you will be randomly chosen to start as the obligated partner.

The experiment will continue this way for a pre-specified number of rounds. After this, there is a 10% chance the experiment will end after each round (instead of a 10% chance of being matched with a new partner).