

**AN EXPERIMENTAL NASH PROGRAM:
A COMPARISON OF STRUCTURED
VERSUS SEMI-STRUCTURED
BARGAINING EXPERIMENTS**

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An experimental Nash program: A comparison of structured versus semi-structured bargaining experiments

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Abstract

While market design advocates for the importance of good design to achieve desirable properties, experiments on coalition formation theory have shown fragility in proposed mechanisms to do so. We experimentally investigate the effectiveness of “structured” mechanisms that implement the Shapley value as an ex ante equilibrium outcome with those of corresponding “semi-structured” bargaining procedures. We find a significantly higher frequency of grand coalition formation and higher efficiency in the semi-structured than in the structured procedures regardless of whether they are demand-based or offer-based. While significant differences in the resulting allocations are observed between the two structured procedures, little difference is observed between the two semi-structured procedures. Finally, the possibility of free-form chat induces an equal division more frequently than occurs without it. Our results suggest that when it comes to bargaining and coalition formation, not having various restrictions imposed by different mechanisms may lead to more desirable outcomes.

JEL code: C70, C71, C92

Keywords: Nash Program, Bargaining procedures, Shapley value

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1 Introduction

For decades, economists, particularly game theorists, have played a crucial role in assisting legislators, regulators, lawyers, and judges in designing markets. They have been instrumental in developing complex markets in the classical theory of auctions (Vickrey, 1961; Milgrom and Weber, 1982a,b), in labor clearing houses for American doctors obtaining their first jobs (Roth and Peranson, 1999), markets for electric power (Wilson, 2002; Cramton, 2017) to name but a few. Additionally, they have proposed allocation procedures for markets that do not use prices, such as coalition formation (Kahan and Rapoport, 2014), school choice (Roth, 1985), live-donor kidney transplantation (Roth et al., 2004, 2007), and the job market for new economists (Coles et al., 2010). The common and well-established vision in the literature is that “*design is important because markets don’t always grow like weeds—some of them are hothouse orchids*” (Roth, 2002, p. 1373). Without external intervention, in fact, markets naturally struggle to provide desirable properties such as thickness, efficiency, safety, and simplicity.

Bargaining is one of the most ubiquitous and effective forms of market interaction between potentially conflicting or cooperating agents, and is the basis of many of the situations listed above. Thus, bargaining interactions, among others, may particularly benefit from the potential welfare gains of well-designed procedures. As such, the importance of bargaining research for enhancing the efficiency – and the many other desirable properties – of these interactions has been widely emphasized (e.g., Crawford, 1982).

Over the years, experimental economics has become a natural complement to theoretical work, aiding in understanding market failures and testing new solutions before proposing them to policymakers (Kagel and Levin, 1986; Kagel and Roth, 2000). Then,

the goals achieved through extensive theoretical research on mechanism design – including bargaining procedures design – have been well supported and documented by a substantial body of experimental literature (Brosig et al., 2003; Chen and Sönmez, 2006; Cox et al., 1988; Denton et al., 2001; Smith, 1967) and numerous field applications (Dickerson et al., 2012). However, there is a field in which experiments have shown a certain fragility in validating the proposed theoretical mechanisms and their appealing results: coalition formation theory (Abe et al., 2021; Okada and Riedl, 2005).

Coalitions are means-oriented, and frequently temporary, alliances among individuals or groups who may have different initial goals. In many situations, forming a coalition is advantageous to both individuals and groups. Coalition formation behavior is a pervasive aspect of social life (Gamson, 1961) and thus a crucial matter in economics (Kahan and Rapoport, 2014; Konishi and Ray, 2003). Theories of coalition formation have been developed in many models from different disciplines, such as mathematics in the theory of cooperative games, then widely adopted not only in economics and political science (Holler, 1982) but also in models in management (Stevenson et al., 1985), social psychology (Komorita and Kravitz, 1983), and computer science (Dang et al., 2006).

In this paper, we focus on coalition formation from the classical perspective of game theorists, with cooperative game theory serving as the means we use in our investigation. This theory emerged alongside noncooperative game theory after the foundational paper by John von Neumann (1928). Then, the active collaboration of John von Neumann and Oskar Morgenstern culminated in the renowned book *Theory of Games and Economic Behavior* (von Neumann and Morgenstern, 1944). Despite a common origin founded on the well-established assumption of the rationality of individuals, noncooperative game theory and cooperative game theory have advanced for decades on two different paths,

with the gap between the two becoming more and more apparent over the years.

However, there has been a significant effort to bridge these two branches of game theory. This strand of literature is known as the Nash (1953) program. This agenda aims to provide non-cooperative mechanisms to implement cooperative interactions (Serrano, 2005, 2008, 2014, 2021, for surveys). The theory predicts that at equilibrium, individuals, guided by the proposed strategic mechanisms in their interactions, will manage to cooperate efficiently and share the proceeds according to well-known cooperative solutions, such as the Shapley value. These mechanisms reconcile and integrate the two approaches by enabling players, within a specific allocation context, to engage in constrained bargaining processes and reach a stable outcome without requiring an external intermediary. The rationale behind this design is that if the same outcome can be reached through different paths, the solution is likely to be more robust. While many authors have contributed to the development of the Nash program, so in line with the previously illustrated common opinion on the importance of investing in research in bargaining, experimental investigations in the specific context of the Nash program have been scarce.

To address this gap in the literature, Chessa et al. (2022, 2023a,b) have conducted a series of experiments comparing different mechanisms, that are theoretically expected to implement full cooperation in games where rationality would lead individuals to choose cooperative strategies and share the proceeds according to the Shapley value, the most well-known cooperative solution (Shapley, 1953). Even if motivated by different research questions, the common pattern in all these experiments is that despite the implementation of structured bargaining procedures promising cooperation, the experimental results show that in many cases, individuals fail to cooperate. These results are in line with other experimental investigations of theoretical models for coalition formation

(Kahan and Rapoport, 1980a; Rapoport and Kahan, 1984).

A closer look at the structure of these bargaining procedures in the context of the Nash program suggests that this failure may be attributed to two factors. First, many of these mechanisms are expected to provide a fair allocation *a priori*, but in practice, the proposed divisions depend heavily on the specific implementation of the mechanism in a given scenario. This includes the order in which individuals participate in the mechanism, to the detriment of those who intervene last, with the first mover often being selected simply through a random choice (Chessa et al., 2023a,b). This feature then results in expected final shares that are clearly unfair to those individuals who are not lucky enough to be selected as first movers. Second, individuals explicitly retain the ability to reject a given final allocation even after agreeing to participate in a mechanism. This issue becomes even more evident in cases of unfair allocations, which are then rejected by one or more individuals, thus excluding any possibility of a complete future agreement.

In the present study, we aim to shed light on whether reducing the structural constraints in bargaining procedures can lead to higher levels of cooperation among individuals. More freedom in bargaining would grant individuals greater flexibility to adjust the proposed divisions and achieve a broader consensus. Indeed, previous studies have already highlighted the importance of shifting the focus toward more unstructured bargaining experiments and advocate for their revival as a future direction in experimental research (Güth, 2012; Karagözoğlu, 2019).

It is, however, difficult to design an “unstructured” computerized bargaining experiment, because the very design of a computer interface necessarily imposes some structure on the bargaining procedure. For example, Shinoda and Funaki (2019) conducted what they call a computerized “unstructured” three-player bargaining experiment. Par-

ticipants could freely propose a coalition and an associated allocation that was feasible among the members of the proposed coalition. Participants were also free to modify their proposal anytime during the negotiation, and to agree on the proposal made by another participant. A coalition was formed if all its members agreed. They also considered a treatment in which participants could freely send chat messages to others. Similarly, in three-player games that model negotiable conflicts involving two weak players and one strong player, Kahan and Rapoport (1980b) considered two communication conditions: a condition where subjects could exchange messages, and a condition where subjects were not allowed to send messages and were unaware that they had been denied this option. While these procedures are much less structured than those considered in Chessa et al. (2022, 2023a,b), there remain some constraints in what participants could do and how the coalition was formed.

We therefore consider both an offer-based and a demand-based (an alternative procedure where participants, instead of freely proposing a coalition with an associated allocation among its members, freely make their demand for them to join a coalition) “semi-structured” bargaining experiment. We call these experiments semi-structured (Duffy et al., 2021) because, as noted above, while they are much less structured compared to the structured bargaining experiments considered by Chessa et al. (2022, 2023a,b), there remains some structure in the bargaining procedures.

More specifically, we vary (a) whether or not participants can communicate freely via online chat during the negotiation. This dimension is motivated by Shinoda and Funaki (2019), who found that the grand coalition is more likely to be formed with than without a possibility of free-form communication among players through a chat window; We also vary (b) whether the negotiation is offer-based or demand-based. This second dimension is motivated by Chessa et al. (2023b), who found that demand-based

and offer-based mechanisms can result in outcomes satisfying much different properties. Then, we contrast (c) the results of these semi-structured experiments with the results of structured experiments of Chessa et al. (2023b), and this represents the third and key motivation for our analysis.

We find that semi-structured experiments, both offer-based and demand-based, result in a higher frequency of grand coalition formation and more efficiency than structured ones. Unlike Chessa et al. (2023b), who find significant differences in the outcomes of offer- versus demand-based procedures, we do not find significant differences between the two, regardless of the possibility of free-form communication. The latter result suggests that the exact procedure may have little impact on outcomes.

We further analyze the negotiation dynamics of offers and demands, first using two examples from semi-structured experiments to demonstrate that agents tend to converge toward efficient outcomes, as well as the Shapley value and equal division, during the negotiation process. We then show that this tendency arises primarily from the flexibility to repeatedly adjust offers and demands over time, a feature absent in structured mechanisms.

The rest of the paper is organized as follows. Section 2 illustrates the theoretical background, including the presentation of the structured mechanisms by Winter and Hart and Mas-Colell previously experimentally investigated in Chessa et al. (2023b). Section 3 presents the design of our new experiments and the newly defined semi-structured bargaining procedures. The results of our new experiments, with a comparison with our previous experiments, are presented and discussed in Section 4, while Section 5 concludes.

2 Theoretical background

In Section 2.1, we provide some theoretical definitions. Then, in Section 2.2, we present the two structured mechanisms that we analyzed in our previous experimental study (Chessa et al., 2023b) and that we question and compare with similar but semi-structured mechanisms in this paper.

2.1 Cooperative transferable utility games and solutions

Let $N = \{1, \dots, n\}$ be a finite set of *players*. Each subset $S \subseteq N$ is referred to as a *coalition*, with N called the *grand coalition*. A *cooperative transferable utility (TU) game* is defined as a pair (N, v) , where N is the set of players and $v : 2^N \rightarrow \mathbb{R}$, with $v(\emptyset) = 0$, is the *characteristic function*. This function assigns a *worth* $v(S)$ to each coalition $S \subseteq N$, representing the worth that members of S can achieve through cooperation. When the set of players N is fixed, we denote the game by v instead of (N, v) . The set of all games with player set N is denoted by $\mathcal{G}^N$. A player i is a *null player* in $v \in \mathcal{G}^N$ if $v(S) = v(S \setminus \{i\})$ for all $S \subseteq N$. A null player is a player who contributes nothing to any coalition.

The most famous solution concept for cooperative TU games, the *Shapley value*, fairly distributes the total contribution of all players in a system by averaging their marginal contributions across all possible participation orders. It is defined as follows:

$$\phi_i(v) = \sum_{S \subseteq N, i \in S} \frac{(|S| - 1)!(|N| - |S|)!}{|N|!} (v(S) - v(S \setminus \{i\})) \quad \forall i \in N.$$

In our analysis, we also consider a simpler solution concept, the *Equal Division solution*, which distributes the worth $v(N)$ equally among all players. It is defined as

follows:

$$ED_i(v) = \frac{v(N)}{n} \quad \forall i \in N.$$

This solution has been investigated as a compelling option for cooperative players when the worth of coalitions is not a primary consideration.

2.2 Winter and Hart and Mas-Colell mechanisms

Winter (1994) introduced a bargaining procedure based on sequential demands within strictly convex cooperative games.¹ In these games, cooperation becomes increasingly attractive, generating a snowball effect that leads to the formation of the grand coalition. In this model, players take turns publicly announcing their demands. Essentially, each player declares, “I am willing to join any coalition that offers me ...” and then waits for these demands to be satisfied by other players. The bargaining procedure begins with a randomly selected player from the set N ; say player i . This player publicly states her demand d_i and then selects a second player, who must also declare her demand. The game continues in this manner, with each player presenting a demand and then selecting another player to take their turn. If at any point a compatible demand is made – meaning there exists a coalition $S \subseteq N$ for which the total demand of the players in S does not exceed $v(S)$ – the first player to make such a demand selects the compatible coalition S . The players in S then receive their demands and exit the game, while the remaining players continue bargaining under the same rules applied to v restricted to $N \setminus S$. In a T -period implementation, where $T > 1$ and T is finite, if any players are left with unmet demands after the first period, the bargaining procedure is repeated in the second period with this subset of players. Their previous demands are canceled, and they incur a fixed

¹A game $v \in \mathcal{G}^N$ is *strictly convex* if $v(S) + v(T) < v(S \cup T) + v(S \cap T)$, for each $S, T \subseteq N$.

delay cost. This process continues until T periods have been completed. In Chessa et al. (2023b), we considered a one-period implementation in which players with unmet demands at the end of the first period receive their individual value.²

Winter (1994) demonstrated that this mechanism has a unique subgame perfect equilibrium, that assigns equal probabilities according to the principle of indifference. At this equilibrium, the grand coalition forms, and the *a priori* expected equilibrium payoff aligns with the Shapley value.

Hart and Mas-Colell (1996) introduced a bargaining procedure designed for monotonic cooperative games,³ which is a less stringent condition than the strict convexity required by Winter mechanism. In the following, we present a simplified version of the mechanism as implemented in Chessa et al. (2023b). In this mechanism, the bargaining procedure begins with a randomly selected proposer making an offer to the other players, framed as, “If you wish to form a coalition with me, I will give you ...” The other players, acting sequentially, may choose to either accept or reject the offer. Unanimity is required for the proposal to be accepted. A critical aspect of the model is determining what happens if no agreement is reached, leading the game to progress to the next stage. The more general mechanism proposed by Hart and Mas-Colell allows the proposer, even after a rejection, to remain in the game and continue to the next stage with a certain probability. In our analysis, we consider the special case where this probability is zero. If the proposal is rejected, the proposer exits the game with her individual value, and bargaining continues among the remaining players, with a new proposer randomly selected.

²Chessa et al. (2023a) compared a one-period implementation and a two-period implementation of the Winter mechanism, investigating scenarios with both low and high delay costs in the latter case. Their findings indicate that the three different implementations yield similar outcomes in terms of coalition formation, alignment with the Shapley value predictions, and satisfaction of the axioms.

³A game $v \in \mathcal{G}^N$ is *monotonic* if $v(S) \leq v(T)$ for each $S \subseteq T \subseteq N$.

Table 1: A three-player game

S	1	2	3	1,2	1,3	2,3	N
$v(S)$	10	20	20	50	50	60	100

Hart and Mas-Colell (1996) demonstrated that this game has a unique subgame perfect equilibrium. At this equilibrium, the grand coalition forms, and the *a priori* expected equilibrium payoff corresponds to the Shapley value.

We illustrate the two mechanisms using the strictly convex three-player game presented in Table 1. This game represents a meaningful example, as it is very similar to the games implemented in our experimental studies. This game satisfies the conditions required by both the Winter and H-MC mechanisms. The Shapley value for this game is represented by the vector $\phi(v) = (\frac{80}{3}, \frac{110}{3}, \frac{110}{3}) \approx (26.67, 36.67, 36.67)$, which corresponds to the *a priori* equilibrium payoff for both mechanisms.

Now, let us assume that player 1 is randomly selected as the first proposer in both mechanisms. Regardless of the subsequent order of players in the Winter mechanism, the proposer will receive an *a posteriori* equilibrium payoff of 40 in both mechanisms, which equals their marginal contribution to the grand coalition; that is, $v(N) - v(N \setminus \{1\})$. Suppose further that the order of players in the Winter mechanism is 1, 2, and 3. In this case, the *a posteriori* equilibrium payoff for the Winter mechanism is given by the vector $(40, 40, 20)$, where player 2 demands her marginal contribution $v(\{2, 3\}) - v(\{3\})$, and player 3 claims her individual value $v(\{3\})$. Conversely, in the case of the H-MC mechanism, the proposer offers the Shapley value of the reduced game to players 2 and 3. Consequently, the *a posteriori* equilibrium payoff is given by the vector $(40, 30, 30)$. Repeating the above argument for every order, we obtain the *a posteriori*

Table 2: *A posteriori* equilibrium payoffs

Order	Winter	H-MC
123	(40, 40, 20)	(40, 30, 30)
132	(40, 20, 40)	(40, 30, 30)
213	(30, 50, 20)	(20, 50, 30)
231	(10, 50, 40)	(20, 50, 30)
312	(30, 20, 50)	(20, 30, 50)
321	(10, 40, 50)	(20, 30, 50)
Shapley value $\phi(v)$	$(\frac{80}{3}, \frac{110}{3}, \frac{110}{3})$	$(\frac{80}{3}, \frac{110}{3}, \frac{110}{3})$

equilibrium payoffs summarized in Table 2.

We can observe how the payoff shares are strongly affected by the order in which players are asked to make their demand or offer. In particular, the player who moves first has a significant advantage in the corresponding *a posteriori* equilibrium, and this represents a clear drawback of these structured mechanisms. To stress this anomaly in an even more extreme case, consider the following three-player glove game (Shapley and Shubik, 1969), where players 1 and 2 each own a left-handed glove, and player 3 owns a right-handed glove. Only a matched pair is worth one unit of value, such that $v(\{1, 3\}) = v(\{2, 3\}) = v(\{1, 2, 3\}) = 1$ and zero otherwise. The *a posteriori* and the *a priori* – coinciding with the Shapley value – equilibrium payoffs, are summarized in Table 3.

Thus, under the Winter mechanism, for every player order, we observe that the last player is always expected to receive and accept a payoff of zero. This also happens when the last player is player 3, who has a central role in our glove game as she is the only one necessary for the formation of a coalition of nonzero worth.

In this paper, we posit that this could be one reason for the limited success in experimental implementation of structured mechanisms, as observed in Chessa et al. (2023b).

Table 3: *A posteriori* and *a priori* equilibrium payoffs in the glove game

Order	Winter	H-MC
123	$(0, 1, 0)$	$(0, \frac{1}{2}, \frac{1}{2})$
132	$(0, 0, 1)$	$(0, \frac{1}{2}, \frac{1}{2})$
213	$(1, 0, 0)$	$(\frac{1}{2}, 0, \frac{1}{2})$
231	$(0, 0, 1)$	$(\frac{1}{2}, 0, \frac{1}{2})$
312	$(0, 0, 1)$	$(0, 0, 1)$
321	$(0, 0, 1)$	$(0, 0, 1)$
Shapley value $\phi(v)$	$(\frac{1}{6}, \frac{1}{6}, \frac{4}{6})$	$(\frac{1}{6}, \frac{1}{6}, \frac{4}{6})$

Table 4: The games

S	1	2	3	4	1,2	1,3	1,4	2,3	2,4	3,4	1,2,3	1,2,4	1,3,4	2,3,4	N
$v_1(S)$	0	5	5	10	20	20	25	20	25	25	50	60	60	60	100
$v_2(S)$	0	20	20	30	20	20	30	45	55	60	45	55	60	100	100
$v_3(S)$								$= v_1(S) + v_2(S)$							
$v_4(S)$								$= 2v_1(S)$							

3 The experimental design

We first describe in Section 3.1 the four-player bargaining games we consider in our experiment. We then explain our four treatments in Section 3.2, based on our new-defined semi-structured bargaining procedures, which are presented in Section 3.3 and in Section 3.4. At each step, we systematically provide a comparison with the experimental design of the previous experiments in Chessa et al. (2023b).

3.1 The games

We consider the four four-player games shown in Table 4. These games are the same as those considered in Chessa et al. (2022, 2023a,b). This is to allow a direct comparison of the results in Chessa et al. (2023b). The Shapley values of the four games are presented

Table 5: The Shapley value of games 1, 2, 3, and 4

	$\phi_1(v)$	$\phi_2(v)$	$\phi_3(v)$	$\phi_4(v)$
Game 1	22.08	23.75	23.75	30.42
Game 2	0	28.33	30.83	40.83
Game 3	22.08	52.08	54.58	71.25
Game 4	44.16	47.5	47.5	60.83

in Table 5. The equal division solution is simply equal to $ED(v_k) = (25, 25, 25, 25)$ when $k = 1, 2$ and $ED(v_k) = (50, 50, 50, 50)$ when $k = 3, 4$.

Following Chessa et al. (2022, 2023a,b), each participant played all four games twice. The order of games was counter-balanced across sessions. That is, participants played these four games in one of the following four orders: 1234, 2143, 3412, and 4321. At the beginning of a new round, participants were randomly rematched into groups of four players, and their roles were randomly reassigned within the newly created groups.⁴

3.2 Treatments

In our 2×2 between-subjects design, we vary (a) whether participants communicate freely via online chat during the negotiation and (b) whether the negotiation is offer-based or demand-based. Then, (c) we contrast the results of these four semi-structured bargaining procedures with the results of the structured offer-based and demand-based experiment in Chessa et al. (2023b). In total, we thus present analyses related to compar-

⁴The reasons for these design choices given in Chessa et al. (2023b) are as follows. While letting participants play all four games, instead of just one, in each session and randomly reassigning their roles across rounds instead of fixing them might slow their learning how to play the game, (a) having within-session variations was needed for some of the analyses, and (b) random reassignment was implemented to avoid upsetting participants because of the existence of the null player in one of the four games.

isons of the results of six different treatments corresponding to six different bargaining procedures.

In the two semi-structured bargaining procedures with free-form communication, participants could freely send chat messages (except for messages that could identify them and those that could insult others). Messages could be seen by everyone in the same group and could be sent at any time during game play.

In all four semi-structured bargaining procedures, the total maximum duration of a negotiation was randomly determined to be between 300 and 360 seconds. Participants were informed that a negotiation could continue for at least 300 seconds, but its end would terminate at a randomly chosen moment during the 60 seconds following the 60-second mark. The negotiation could end earlier if a grand coalition was formed, or when only one player remained who was not in any coalition. By contrast, in the structured bargaining procedures in Chessa et al. (2023b) an overall time limit was not set, but one was established for each step of the negotiation. A time limit of 60 seconds was set for making a demand (Winter) or a offer (H-MC), and 30 seconds were allocated for choosing a coalition (Winter) or deciding whether to approve or reject a proposal (H-MC).

Fig. 1 represents the six bargaining procedures implemented in the six different treatments and analyzed in this paper, highlighting their main features and differences. On the left of the tree, we present the two structured mechanisms, Winter and H-MC. On the right are the four semi-structured mechanisms, described in greater detail in Sections 3.3 and 3.4. A key difference between structured and semi-structured mechanisms is highlighted here. In a structured mechanism, the bargaining is started by the individual who has been selected as first mover, randomly and by the algorithm. In semi-structured mechanisms, by contrast, any individual can decide to be the first mover.

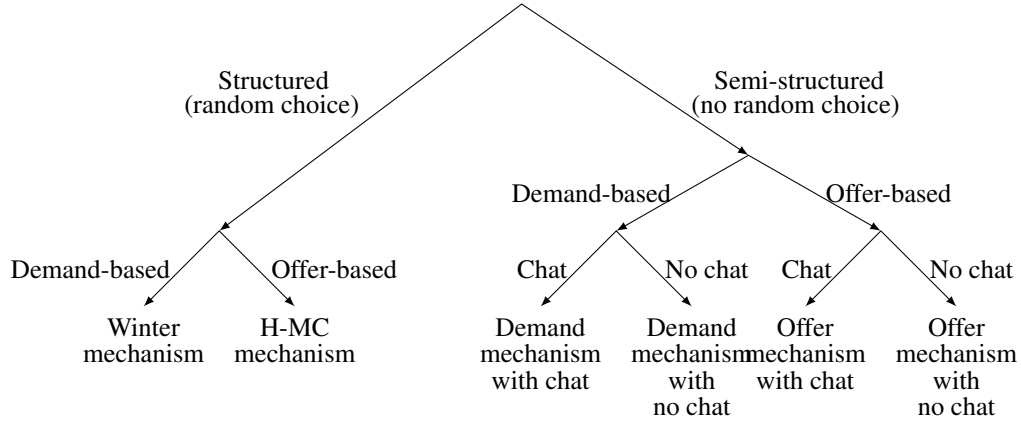


Figure 1: Representation of the six treatments based on the six different bargaining mechanisms.

We now describe the demand-based and offer-based bargaining procedures of our semi-structured mechanisms in detail. These procedures represent the unstructured version of the Winter and the H-MC mechanisms, respectively, presented in Section 2.2 and the object of the previous study by Chessa et al. (2023b).

3.3 Demand-based bargaining procedure

In this procedure, at any point during a negotiation, players are free to demand points they want to obtain. Note that in doing so, players are not proposing a coalition. Instead, they are expressing the points they want to receive for joining a coalition. Each player can make at most one demand at any time during the negotiation, but players are free to modify their demands at any time during a negotiation.

A coalition can be formed if the sum of the demands made by its members is no greater than its worth. When there is such a coalition for a player, the player is free to agree to form it. We exclude a single-player coalition, because it is the default outcome for the player when she ends the game without belonging to any coalition. Each player

can agree to form at most one coalition at any time during the negotiation. Thus, if players want to form a different coalition than the one she is currently agreeing to form, they need to withdraw from their current agreement before agreeing to form a new coalition.

A coalition is formed if all its members agree to form it. Once a coalition is formed, its members exit the negotiation and receive the points they have demanded. The negotiation continues with the remaining players. If only one player remains, the negotiation ends. A player without an agreed coalition at the end of the negotiation (either because of the time limit or because she is the only player left) obtain the singleton value.

3.4 Offer-based bargaining procedure

This procedure is similar to the one that Shinoda and Funaki (2019) call an “unstructured bargaining” protocol. That is, at any time during a negotiation, each player is free to propose or approve a coalition that includes her among players who remain in the game and an associated allocation within the coalition. Below, let proposing or approving a coalition mean both proposing or approving members of that coalition and the associated allocation among them. For example, at the beginning of a negotiation when all four players remain in the game, player 1 can propose $\{1,2\}$, $\{1,3\}$, $\{1,4\}$, $\{1,2,3\}$, $\{1,2,4\}$, $\{1,3,4\}$, or $\{1,2,3,4\}$. Note that a single-player coalition is not considered here, as it is the default outcome for the player when she ends the game without belonging to any coalition. Instead of proposing a coalition, a player can also approve a coalition that includes her and has been proposed by another player.

In our experiment, each player can propose or approve at most one coalition at any point in time. Thus, if a player has proposed a coalition but would like to approve the

one proposed by another player, the player has to withdraw her proposal. Similarly, if a player has approved a coalition proposed by another player but would like to propose a new one, the player has to first withdraw her approval.

If all the members of a proposed coalition approve it, the coalition is formed, and its members all exit the negotiation and receive the allocated points. The negotiation continues with the remaining players. If only one player remains, the negotiation ends. Players without an agreed coalition at the end of the negotiation (either because of the time limit or because she is the only player left) obtain the singleton value.

4 Results

The new experiments on semi-structured mechanisms were conducted at the Institute of Social and Economic Research (ISER), Osaka University, in May and June 2021 (offer-based) and May and June 2022 (demand-based). A total of 344 students, who had not previously participated in similar experiments were recruited as subjects. The experiment was computerized with z-Tree (Fischbacher, 2007), and participants were recruited using ORSEE (Greiner, 2015). The previous experiments on structured mechanisms presented in Chessa et al. (2023b) were conducted under identical conditions in January and February 2019 (Winter mechanism) and January and February 2022 (H-MC mechanism). They involved a total of 176 students. See Table 6 for the number of participants and the mean duration and mean payment in each treatment.

In both the structured and the semi-structured bargaining procedures, at the end of each experiment, two rounds (one from the first four rounds and another from the last four rounds) were randomly selected for payments. Participants received cash rewards based on the points they earned in these two selected rounds at an exchange rate of 20

Table 6: The number of participants, the mean duration, and the mean payment in four semi-structured treatments and Winter and H-MC from Chessa et al. (2023b).

Treatment	No. of participants	Mean duration	Mean payment	When
Demand-based Without chat (No chat)	88 (24x2 + 20x2)	1h34m	2810 JPY	May–June 2022
Demand-based With chat (chat)	84 (24x2,16,20)	1h22m	2860 JPY	May–June 2022
Offer-based Without chat (No chat)	88 (24x2 + 20x2)	1h36m	2810 JPY	June–July 2021
Offer-based With chat (chat)	84 (20x3 + 24x1)	1h36m	2900 JPY	June–July 2021
Winter	96 (24x4)	1h40m	2650 JPY	Jan–Feb. 2019
H-MC	80 (24, 20x2, 16)	1h45m	2850 JPY	Jan–Feb. 2022

JPY = 1 point, in addition to the 1500 JPY participation fee. The experiments lasted on average around 90 minutes for semi-structured mechanisms and around 100 to 105 minutes for structured mechanisms, including the instructions, comprehension quiz, and payment. Participants received a copy of the instruction slides, and a pre-recorded instruction video was played. The comprehension quiz was given on the computer screen after the explanation of the game. The user interface was explained during the practice rounds and referred to the handout about the computer screen. See Appendix A for English translations of the instruction materials and the comprehension quiz for both experiments.

Table 7 summarizes the duration of a negotiation, the frequency of complete break-

Table 7: Summary statistics of semi-structured bargaining experiment

Treatment	Duration ¹	Frequency of complete failure	Number of offers or demands ¹	Number of messages ²	Number of non-greeting messages	Duration ³	Time until the formation of the first coalition ³
Demand-based	119.27	0.045	7.95			109.82	97.75
Without chat (No chat)	(14.58)	(0.029)	(0.33)			(9.74)	(6.25)
Demand-based	96.90	0.018	6.31	2.03	1.69	92.59	80.50
With chat (chat)	(8.64)	(0.010)	(0.33)	(0.69)	(0.71)	(9.03)	(11.46)
Offer-based	185.33	0.131	4.88			163.94	152.55
Without chat (No chat)	(9.73)	(0.013)	(0.24)			(10.14)	(11.47)
Offer-based	179.23	0.119	3.33	4.68	3.16	159.79	156.23
With chat (chat)	(11.42)	(0.024)	(0.13)	(1.33)	(0.62)	(13.23)	(13.97)
Num. Obs.	688	688	688	336	336	634	634
R ²	0.675	0.108	0.773	0.403	0.312	0.656	0.657

Note: Standard errors are corrected for session-level clustering effects and shown in parentheses.

1: Includes those groups that did not form any coalition and were thus terminated by reaching the maximum duration.

2: Includes greeting messages.

3: Does not include groups that did not form any coalitions.

down (no coalition being formed), the number of offers or demands made within a negotiation, the number of messages sent during a negotiation (in treatments with chat), and the time until the first coalition was formed in the semi-structured experiments.⁵ Equivalent results for structured experiments are not reported here because they are not relevant, given the limited freedom for negotiation.

Observe that the average duration of a negotiation is significantly longer in the offer-based than in the demand-based bargaining procedures.⁶ The possibility of chat does not significantly affect the duration of the negotiation.⁷ Note that while more messages are sent under the offer-based than the demand-based bargaining procedures when chat is possible, the difference is not statistically significant.⁸ The complete failure of the negotiation is more frequently observed under the offer-based than the demand-based bargaining procedures,⁹ while the possibility of chat does not significantly affect the failure rate.¹⁰

The longer duration of a negotiation under the offer-based bargaining procedure is not just because there are more groups in which negotiation failed completely. Even among those groups where a coalition was formed, the negotiation took longer under the offer-based than the demand-based bargaining procedures.¹¹ The same is true for

⁵The table is created based on the estimated coefficients of the following linear regressions: $y_i = \beta_1 ONC_i + \beta_2 OC_i + \beta_3 DNC_i + \beta_4 DC_i + \mu_i$, where y_i is the statistic of interest in group i , ONC_i , OC_i , DNC_i , and DC_i are dummy variables that take a value of one if the treatment is offer-based no chat (ONC), offer-based with chat (OC), demand-based no chat (DNC), or demand-based with chat (DC), respectively and zero otherwise. Standard errors are corrected for within-session clustering effects. The statistical tests are based on the Wald test for the equality of the estimated coefficients of treatment dummies.

⁶ $p = 0.0019$ and $p < 0.0001$ for without chat and with chat, respectively (Wald test).

⁷ $p = 0.207$ and $p = 0.690$ for the demand-based and the offer-based bargaining procedures, respectively (Wald test).

⁸ $p = 0.1200$ and $p = 0.1619$ with and without greeting messages, respectively (Wald test).

⁹ $p = 0.0166$ and $p = 0.0013$ without chat and with chat, respectively (Wald test).

¹⁰ $p = 0.3834$ and $p = 0.6701$ demand-based and offer-based bargaining procedures, respectively (Wald test).

¹¹ $p = 0.0016$ and $p = 0.0011$ without chat and with chat, respectively (Wald test).

the time that elapsed before the first coalition was formed.¹²

The number of proposals made under the offer-based bargaining procedures is significantly smaller than the number of demands made under the demand-based bargaining procedures.¹³ The possibility of chat significantly reduces the number of offers or demands made.¹⁴ The number of both demands and offers is, however, larger than what was allowed under Winter and H-MC (with a maximum of four) considered in Chessa et al. (2023b). It is not unexpected that this difference in the number of demands or offers between the structured and semi-structured bargaining procedures would affect the outcomes.

We now turn to analyzing the outcomes of our experiments. Section 4.1 focuses on grand coalition formation and the efficiency of the final outcome. Section 4.2 delves into the negotiation dynamics; that is, whether demands and offers evolved during the negotiation and whether and how they affected the final outcome.¹⁵

4.1 Grand coalition formation and efficiency

For our four games, the structured mechanisms theoretically predict that the grand coalition will form and that full efficiency will be reached. Then, we first compare the frequency of cooperation and the level of efficiency across our four semi-structured mechanisms treatments, after which we compare these outcomes with the results from the experiments on the structured bargaining mechanisms reported in Chessa et al. (2023b).

¹² $p = 0.0008$ and $p = 0.0008$ without chat and with chat, respectively (Wald test).

¹³ $p < 0.0001$ and $p < 0.0001$ without chat and with chat, respectively (Wald test).

¹⁴ $p = 0.0030$ and $p = 0.0001$ for the demand-based and the offer-based bargaining procedure, respectively (Wald test).

¹⁵For completeness and a more exhaustive comparison with the results presented in Chessa et al. (2023b), we report additional analyses in Appendix C.

4.1.1 Grand coalition formation

Table 8 shows the frequencies of various coalitions being formed, focusing on the grand coalition and coalitions with three members. Our analysis thus centers on those groups that reached full cooperation or those groups that did not succeed in doing that, but showed a high level of cooperation.

On one hand, in games 1, 3 and 4, the grand coalition is the most frequently formed under semi-structured bargaining procedures, regardless of whether it is demand-based or offer-based and with or without chat. That is also the case under structured Winter and H-MC mechanisms, except for the Winter mechanism in game 3, where the coalition $\{2, 3, 4\}$ is the most frequently formed coalition. In game 2, on the other hand, instead of the grand coalition, the three-player coalition that excludes the null player $\{2, 3, 4\}$ is the most frequently formed coalition, except under the H-MC mechanism. Chessa et al. (2023b) noted that the proposers do not exclude the null player in game 2 to avoid the risk of the proposal being rejected and receiving a singleton payoff under H-MC. This result shows that such a risk imposed by an offer-based structured mechanism is eliminated under the offer-based semi-structured bargaining procedures.

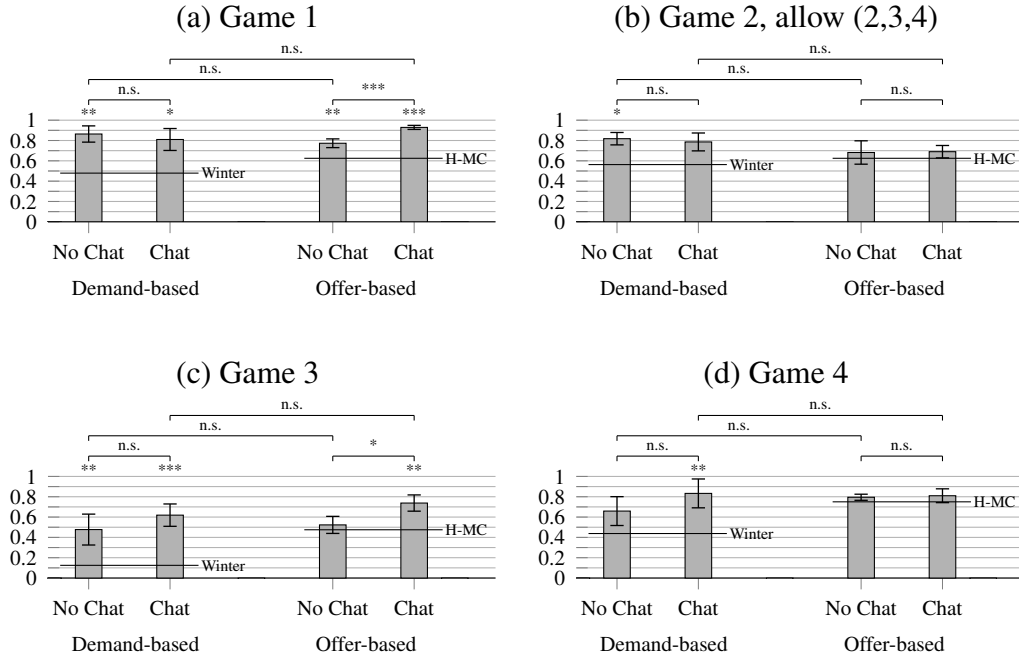
In Fig. 2, we show the results of comparisons across treatments of the frequencies of the grand coalition formation. To take into account the existence of the null player, we include the three-player coalition without the null player in game 2 ($\{2,3,4\}$) as a grand coalition. The horizontal lines with indications of Winter and H-MC are the experimental results of these two structured procedures reported in Chessa et al. (2023b).¹⁶

¹⁶The figure is created based on the estimated coefficients of the following linear regressions: $y_i = \beta_1 ONC_i + \beta_2 OC_i + \beta_3 DNC_i + \beta_4 DC_i + \beta_5 Winter_i + \beta_6 H - MC_i + \mu_i$, where y_i is a dummy variable that takes a value of one if the grand coalition is formed and zero otherwise. The equation is presented for group i . $Winter_i$, and $H - MC_i$ are dummy variables that take a value of one if the treatment is Winter, and H-MC, respectively, and zero otherwise. Other treatment dummies are the same as the one explained above. The standard errors are corrected for within-session clustering effects. The statistical

Table 8: Frequencies of coalitions being formed

Game 1						
Coalition	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
{1234}	38	34	23	34	39	25
{234},{1}	1	1	2	1	0	3
{134},{2}	1	1	10	0	0	4
{124},{3}	0	2	8	2	0	4
{123},{4}	1	0	0	1	0	2
Others	3	4	5	6	3	2
Total	44	42	48	44	42	40
Game 2						
Coalition	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
{1234}	3	5	0	3	9	18
{234},{1}	33	28	27	27	20	7
{134},{2}	0	1	0	0	0	1
{124},{3}	0	0	0	0	0	1
{123},{4}	0	0	0	0	0	0
Others	8	9	21	14	13	13
Total	44	42	48	44	42	40
Game 3						
Coalition	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
{1234}	21	26	6	23	31	19
{234},{1}	10	8	27	14	9	5
{134},{2}	3	1	4	0	1	3
{124},{3}	3	1	1	0	0	3
{123},{4}	0	0	2	1	0	1
Others	7	6	8	6	1	9
Total	44	42	48	44	42	40
Game 4						
Coalition	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
{1234}	29	35	21	35	34	30
{234},{1}	4	2	4	0	1	1
{134},{2}	3	1	7	2	0	2
{124},{3}	0	2	6	0	0	2
{123},{4}	4	2	5	0	0	0
Others	4	0	5	7	7	5
Total	44	42	48	44	42	40

Figure 2: Proportion of times the grand coalition formed



Note: Error bars show one standard error range; ***, **, and * indicate the proportion of times the grand coalition formed was significantly different between two treatments at the 1%, 5%, and 10% significance levels, respectively (Wald test); "n.s." indicates the absence of a significant difference at the 10% level between the two treatments being compared.

We observe that the grand coalition is formed more frequently under the semi-structured bargaining procedures than under the structured bargaining procedures, for both demand-based and offer-based bargaining procedures, regardless of the existence of chat in game 1 and when chat is allowed for game 3. In games 2 and 4, there is no significant difference between the semi-structured and structured offer-based bargaining procedures. For the demand-based procedures, the grand coalition is more frequently formed under the semi-structured ones than the structured approach, significantly so when chat is not allowed in game 2 and when chat is allowed in game 4. Among semi-structured bargaining procedures, there are cases where chat significantly facilitates the tests are based on the Wald test for the equality of the estimated coefficients of treatment dummies.

formation of the grand coalition (games 1 and 3 for the offer-based procedure). Conditional on whether free-from chat is possible, we do not observe significant differences between the demand-based and the offer-based semi-structured bargaining procedures in any of the four games.

4.1.2 Efficiency

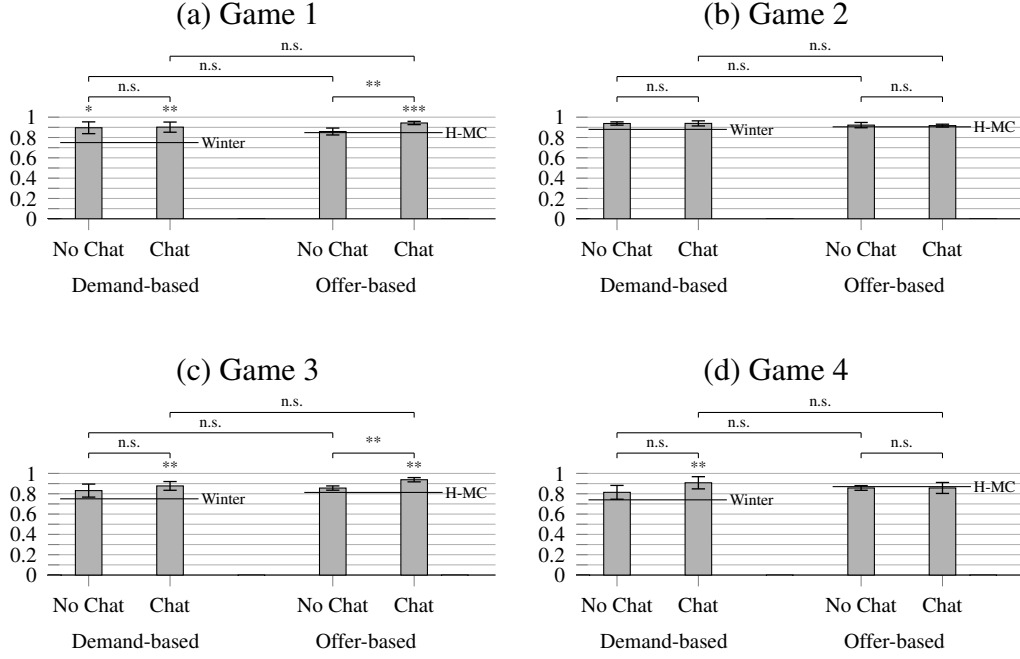
Fig. 3 shows the efficiency across treatments in each game. Efficiency is computed as the share of the sum of the points obtained by four players relative to the worth of the grand coalition.¹⁷

Because efficiency is higher when the grand coalition is formed more frequently, the results are similar to the frequencies of the grand coalition formation discussed above. However, it is important to note that even when the grand coalition is formed, the resulting share can still be inefficient. Thus, this additional analysis is of interest. In our results, efficiency is significantly higher for both demand- and offer-based approaches, under the semi-structured bargaining procedure with chat than under the structured bargaining procedure in games 1 and 3. For games 2 and 4, while the efficiency is not significantly different between structured and semi-structured offer-based bargaining procedures, in case of the demand-based bargaining procedure, it is higher under semi-structured than the structured bargaining procedure with chat in game 4.

We summarize the results of this section by stating that our experiments confirm the overall superior performance in terms of grand coalition formation and efficiency of semi-structured mechanisms than their structured counterparts.

¹⁷The figure is created based on the estimated coefficients of the linear regressions similar to the frequencies of the grand coalition formation, except that the dependent variable is now efficiency.

Figure 3: Efficiency



Note: Error bars show one standard error range; ***, **, and * indicate the proportion of times the grand coalition formed is significantly different between two treatments at the 1%, 5%, and 10% significance levels, respectively (Wald test); “n.s.” indicate the absence of a significant difference at the 10% level between the two treatments.

4.2 Negotiation dynamics

We have already noted that the number of offers and demands made in the semi-structured bargaining procedures are larger than were permitted by construction in the Winter and H-MC mechanisms. In this subsection, we analyze the content of these offers and demands to better understand how they evolved during the negotiation and whether and how they affected the final outcome.¹⁸

¹⁸It should be noted that our data is incomplete in that while all the offers and demands and their cancellations are recorded, only the most recent approval and withdrawal of approval by each player, and not any intermediate ones, are recorded.

4.2.1 Two examples of negotiation dynamics

We start by providing two specific examples of negotiation dynamics in semi-structured mechanisms, one for offer-based and the other for demand-based bargaining procedure without chat. These examples may help identify the possible advantage of semi-structured mechanisms, over structured ones.

Table 9 shows an example of the dynamics of the proposals made in game 1 of the offer-based semi-structured bargaining procedure without chat. “Time” and “Player” represent when and by whom an offer was made; “Coali.” and “ x_j ” ($j \in \{1, 2, 3, 4\}$) correspond to the proposed coalition and the allocation within it; “Dist.SV”, “Dist.ED”, and “Eff.” are the distance of the proposed allocation from the Shapley value, the distance from the equal division, and the efficiency of the offer, respectively.¹⁹ “Outcome” and “Out.time” show the eventual outcome and the timing of the offer.

In this group, all the offers involved the grand coalition, but the allocation within the coalition differed. First, player 4 made an offer giving himself slightly higher payoff than the others while reducing the payoff of player 3. Later, player 3 made a counter-offer by reducing player 1’s payoff while increasing her own as well as player 4’s payoff compared to the proposal by the player 4. This counter-offer was closer to the Shapley value and further away from the equal division than the offer made by player 4. Seeing this counter-offer, player 4 canceled her offer. Later, player 2 made a counter-offer to that of player 3 by reducing the payoff for player 4 while increasing her own and player 1’s payoff and maintaining the payoff for player 3. This offer was further away from the Shapley value than player 3’s existing offer while the distance from the equal division

¹⁹Dist.SV = $\sum_j \sqrt{(x_j - sv_j)^2}$, where sv_j is the Shapley value for j in the corresponding game. Dist.ED = $\sum_j \sqrt{(x_j - ed)^2}$ where $ed = 25$ in games 1 and 2, and $ed = 50$ in games 3 and 4. For game 2, when coalition $\{2,3,4\}$ is formed, Dist.ED = $\sum_{j \neq 1} \sqrt{(x_j - 100/3)^2}$

Table 9: An example of the dynamics of negotiation in game 1 of the offer-based semi-structured bargaining procedure without chat

Time	Player	Coali.	x_1	x_2	x_3	x_4	Dist.SV	Dist.ED	Eff.	Outcome	Out.time
22.562	4	1234	25	25	23	27	4.73	2.83	1.0	Cancel	120.75
112.25	3	1234	20	25	25	30	2.76	7.07	1.0	Cancel	206.78
161.20	2	1234	25	30	25	20	12.56	7.07	1.0	Cancel	192.30
198.63	2	1234	25	25	25	25	6.41	0	1.0	Approve	235.14

Table 10: An example of the dynamics of negotiation in game 1 of the demand-based semi-structured bargaining procedure without chat

Time	No.Act.	Player	No.Dem.	x_1	x_2	x_3	x_4	Dist.SV	Dist.ED	Eff.	Outcome
7.13	4	2		1	0	10	0	0			
7.72	4	3		2	0	10	25	0			
9.15	4	1		3	10	10	25	0			
10.72	4	4		4	10	10	25	27	18.66	21.31	0.72
15.25	4	2		3	10	0	25	27			
29.86	4	2		4	10	25	25	27	12.68	15.13	0.87
31.67	4	1		3	0	25	25	27			
41.52	4	1		4	20	25	25	27	4.38	5.39	0.97
											1234

was the same. Eventually, player 2 canceled this offer and made an offer that split the pie equally among four players. This offer was eventually approved by everyone.

Table 10 shows an example of the dynamics of the demands made in game 1 of the demand-based semi-structured bargaining procedure without chat. The columns are similar to those in Table 9. “No.Act.” and “No.Dem.” represent the number of active players and the number of demands submitted at the time, respectively; “Dist.SV”, “Dist.ED”, and “Eff.” are computed only when four demands are submitted; “Outcome” shows the coalition being formed. The specific demands made by the players are shown in bold. The cancellation of an existing demand is recorded as demanding 0.

In this group that eventually formed the grand coalition, there were three instances

where four demands were all on the table. Among these instances, we observe increasing efficiency and a corresponding decline in the distance to the Shapley value and the equal division over time. In the end, although it was possible for some players to demand a few more points, they decided to stop the negotiation.

In these two examples, in many instances, some improvements, in terms of either the efficiency or the distance from the Shapley value or the equal division, were observed during the negotiation. Note that in the corresponding structured bargaining procedures, such within-negotiation improvements are not possible by design.

We now analyze the differences across treatments more systematically by first looking at the properties of the first offers (the offer-based case) or the first instances in which the demands from all four players are on the table (the demand-based case) and their likelihood of approval. We then complement these analyses by investigating the relationships between properties of all the offers or the set of four demands and their likelihood of resulting in the grand coalition formation.

4.2.2 First offers and demands versus all offers and demands and the grand coalition formation

Tables B.1 and B.2 in Appendix B report the average (and standard deviation) of the characteristics of the first set of four demands (for demand-based bargaining procedures) and the offers (for offer-based bargaining procedures) in our four games.

From a careful observation of these tables, we can highlight the two key results of this analysis (which we do not report in full here for simplicity). Focusing on offer-based mechanisms, we observe that the grand coalition is formed less frequently based on the first offer in semi-structured bargaining procedures when compared to H-MC. As a

Table 11: Characteristics of the sets of demands (for demand-based procedures) and the offers (for offer-based procedures) and grand coalition formation

	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
Dist.SV	-0.0002 (0.0033)	0.0059** (0.0026)	-0.0116* (0.0060)	-0.0052*** (0.0013)	-0.0061*** (0.0018)	-0.0080*** (0.0024)
Dist.ED	-0.0072** (0.0031)	-0.0108*** (0.0022)	0.0023 (0.0054)	-0.0029** (0.0013)	-0.0084*** (0.0015)	-0.0035 (0.0027)
Eff.	1.314*** (0.342)	1.535*** (0.315)	1.435*** (0.351)			
Cons.	-0.712** (0.355)	-0.866*** (0.327)	-0.450 (0.370)	0.302*** (0.030)	0.497*** (0.040)	0.828*** (0.053)
N	298	245	112	662	442	158
R ²	0.141	0.304	0.372	0.040	0.093	0.157

Note. Dependent variable is whether the grand coalition is formed (= 1) or not (= 0); linear probability model pooling data from the games; standard errors are clustered at the group (negotiation) level and reported in parentheses.

*, **, and *** show statistical significance at the 10%, 5%, and 1% levels, respectively.

result, efficiency is also lower. Similarly, for demand-based mechanisms, the likelihood that the grand coalition is formed based on the first instance in which all players made their demand is lower in semi-structured bargaining procedures than in Winter (except for game 3).

These results confirm what was previously hypothesized; namely that the better performance of semi-structured mechanisms is a consequence of the ability to adjust offers and demands. In fact, this improved performance is not observed at the very beginning but only at the end of the negotiation. To conclude, we now ask in which direction these offers and demands should be adjusted by investigating which characteristics of the proposed allocations more easily lead to an agreement.

Table 11 shows the results of linear regressions where the dependent variable is whether the grand coalition (including $\{2, 3, 4\}$ in game 2) is formed (= 1) or not (= 0),

and independent variables are the characteristics of the set of demands or offers. We focus only on those sets of demands or offers where the formation of the grand coalition is possible (that is, it involves all the members of the grand coalition and is feasible).

We see that the “Dist.ED” (the distance to the equal division solution) is negatively correlated with grand coalition formation in both demand-based and offer-based semi-structured bargaining procedures. This suggests, as we have seen above through examples, that the dynamics of negotiation tend to favor allocations that are closer to the equal division under semi-structured bargaining. “Dist.SV” (the distance to the Shapley value) is also negatively correlated with the formation of the grand coalition in both Winter and H-MC, as well as in the two offer-based semi-structured bargaining procedures. In the demand-based semi-structured approach with chat, “Dist.SV” is instead positively correlated with the formation of the grand coalition. However, its magnitude is only about half of the effect of “Dist.ED”. Finally, efficiency is positively correlated with the formation of the grand coalition in all the demand-based bargaining procedures.

To summarize these results, we observe that proposals closer to equal division, closer to the Shapley value, and with greater efficiency are more likely to be accepted, thus facilitating the formation of the grand coalition. By design, semi-structured mechanisms provide more opportunities to adjust proposals in this direction, leading to better final outcomes in terms of coalition formation and overall efficiency.

5 Concluding remarks

Unstructured, or, in our case, semi-structured bargaining experiments, have been argued to more closely resemble real-world bargaining situations, suggesting that after many decades, it is time for a revival of unstructured bargaining experiments (Karagözoğlu,

2019). The present study seeks to contribute to this body of literature by experimentally comparing the outcomes of structured versus semi-structured bargaining procedures in the context of coalition formation. Specifically, it contrasts the experimental results of two mechanisms that implement the Shapley value (Shapley, 1953) as an *ex ante* equilibrium outcome, as considered in Chessa et al. (2023b): simplified versions of the demand-based mechanism proposed by Winter (1994) and the offer-based mechanism proposed by Hart and Mas-Colell (1996), with the outcomes of two corresponding but much less structured bargaining procedures. In doing so, this paper also contributes to the literature on the Nash program (Nash, 1953).

We found that semi-structured bargaining procedures led to a significantly higher frequency of grand coalition formation and greater efficiency than did structured procedures. This outcome occurs because participants in the former could explore a wider range of proposals and demands during negotiations, allowing them to adjust their offers toward proposals that are more likely to be accepted. A deeper analysis reveals that semi-structured bargaining procedures do not necessarily yield better outcomes when considering only the initial set of offers and demands. Instead, their advantage over structured bargaining procedures becomes evident toward the end of the negotiation process. Structured bargaining procedures, while theoretically predicted to lead to the formation of the grand coalition, result instead in an *a posteriori* expected payoff distribution that heavily favors randomly selected first movers at the expense of players who intervene later. These unfair proposals are often rejected, and structured mechanisms do not allow for adjustments, ultimately limiting their effectiveness.

Unlike the sharp differences between the outcomes of the demand-based and the offer-based structured bargaining reported by Chessa et al. (2023b) in terms of the frequency of grand coalition formation and efficiency, no significant differences between

the demand-based and offer-based semi-structured bargaining procedures arose in these dimensions. In terms of the design of bargaining experiments, this result is encouraging because it suggests that when the participants are less constrained in terms of when and how often they can act, the outcomes of the negotiations become similar regardless of whether the bargaining procedure is an offer-based or a demand-based one. Even though the possibility of communicating through free-form chat does not play a substantial role, it has been shown in some cases to facilitate better outcomes.

Our findings suggest that one should carefully consider the potential effects of various restrictions imposed by different mechanisms – such as who can act and when – while also accounting for behavioral biases and cognitive limitations. When it comes to bargaining and coalition formation, in fact, not having various restrictions imposed by different mechanisms and having the possibility of freely adjusting during the negotiation may lead to more desirable outcomes. More broadly, our results align with the extensive literature supporting Adam Smith’s “invisible hand” (Rothschild, 1994) and the possibility of cooperation without coercion (Friedman, 2016).

The directions for future research are many and challenging. First, in our experiment we used only the four games considered by Chessa et al. (2022, 2023a,b) in order to make a direct comparison with previous experimental investigations. As an initial step, future studies should consider more varieties of games (non-superadditive, non-convex, those with more than four players, or more general games such as partition function games (Thrall and Lucas, 1963)) to better understand the possible impacts of various behavioral biases, such as fairness consideration and loss aversion, in advancing the Nash program while incorporating the fruits of advances in behavioral and experimental economics. Testing our unstructured and semi-structured mechanisms for a wider range of games can confirm or call into question the results of this paper.

But we believe that the most challenging direction for future investigations is to explore whether alternative structured mechanisms can be designed to outperform semi-structured mechanisms, as observed in many other markets (see the numerous examples presented in the Introduction). It is in fact important to stress that in many situations rigorous structured mechanisms may be especially useful or even necessary. This is the case, for example, when unstructured or semi-structured bargaining is not feasible, either because players are dispersed or because they do not even know whom they are playing against, as for example in financial markets.

A natural approach would be to define and/or test mechanisms whose theoretical predictions lead to a fairer payoff distribution, not only *a priori* but also *a posteriori*, that is, given the specific implementation of the game. In this regard, it is important to note that Chessa et al. (2022) experimentally tested the bargaining mechanism proposed by Pérez-Castrillo and Wettstein (2001). This mechanism, despite predicting exactly the Shapley value at equilibrium, was found to perform even worse than H-MC, further amplifying the first-mover advantage. But this negative result was likely due to the complexity of the mechanism, and to the difficulty of the experimental subjects in understanding its dynamics. An alternative and simpler way to test a different algorithm would be to implement a mechanism à la Winter, where players are required to play following all possible orders, or à la H-MC, where all players take turns acting as the first mover. However, while this approach could be promising, it would lead to a much longer experiment in which additional factors would need to be carefully controlled, such as the order of the different sequences itself.

Acknowledgement

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A Instructions

English translations of the instructions for all the treatments of both the game and the experimental software along with the comprehension quiz, are available at https://osf.io/kqw6n/?view_only=ea41e1284a1347d09760fba82fda37ea. File names starting with demand-based are for the semi-structured demand-based treatments, and those starting with offer-based are for semi-structured the offer-based treatments. Those starting with Winter and H-MC are for Winter and H-MC, respectively.

B Additional material for Section 4.2

Tables B.1 and B.2 report the average (and the standard deviation) of the characteristics of the first set of four demands (for demand-based bargaining procedures) and the offers (for offer-based bargaining procedures) in our four games. We consider, in addition to “Dist.SV”, “Dist.ED”, and “Eff.” defined in Section 4.2, “All”, “Feasible”, and “GC”. For demand-based bargaining procedures, “All” takes a value of one if no player has exited the game before all four players have made their demands (offer-based) and zero otherwise. For offer-based bargaining procedures, it takes a value of one if positive points are offered to all four players, and zero otherwise.²⁰ “Feasible” takes a value of one when the set of demands are feasible, and zero otherwise. “GC” takes a value of one if the grand coalition is formed from this set of demands or offer, and zero otherwise.

Let us start from the offer-based bargaining procedures. Here, by design, all the offers are feasible. Comparing H-MC and the two offer-based semi-structured bargaining mechanisms (No Chat and Chat), we observe that more offers in the latter exclude some players when the mean value of “All” is smaller. As a result, efficiency is also lower in the latter compared to H-MC. As might be expected, the grand coalition is formed less frequently based on the first offer in the semi-structured bargaining procedures compared to H-MC. While “Dist.SV” and “Dist.ED” tend to be smaller when Chat is allowed than when it is not, there is no clear tendency between H-MC and No Chat.

For the demand-based bargaining procedures, compared to Winter, the likelihood of all four players (three non-null players in game 2) making their demands before some

²⁰For game 2, as we allow the coalition $\{2,3,4\}$ to be the grand coalition, “All” takes a value of one when all three players (except the null player) have made their demand before one of them exits the game, and zero otherwise for the demand-based approach. Similarly, for the offer-based approach, it takes a value of one when three non-null players are all offered positive points, and zero otherwise.

player exits the game is higher in the two semi-structured bargaining procedures (“All” is higher in No Chat and Chat compared to Winter), although many of these first sets of demands are jointly non-feasible (less so when chat is allowed). Even among those sets of demands that are jointly feasible, the likelihood is that the grand coalition is formed based on the first instances in which all players made their demand is lower in semi-structured bargaining procedures than in Winter, except for game 3. Furthermore, we observe that “Dist.SV” and “Dist.ED” are smaller under the semi-structured procedures than in Winter except for “Dist.SV” in game 3.

Table B.1: Characteristics of the first sets of four demands (for demand-based approaches) and the first offers (for offer-based approaches) in games 1 and 2

Game 1									
	No Chat ^a	No Chat ^b	Demand-based		Winter ^a	Winter ^b	Offer-based		H-MC
			Chat ^a	Chat ^b			No Chat	Chat	
All	0.98 (0.15)		0.93 (0.26)		0.69 (0.47)		0.80 (0.41)	0.90 (0.30)	1.00 (0.00)
Dist.SV	12.55 (6.71)	11.82 (6.35)	9.90 (6.38)	8.92 (4.83)	17.55 (11.10)	15.42 (5.47)	11.74 (9.98)	8.94 (7.36)	10.49 (5.83)
Dist.ED	12.38 (8.52)	11.12 (7.88)	6.56 (8.97)	4.44 (6.76)	17.27 (12.50)	14.85 (6.32)	8.38 (12.55)	4.90 (9.75)	7.92 (7.82)
Eff.	0.97 (0.19)	0.87 (0.14)	0.98 (0.11)	0.95 (0.09)	1.00 (0.18)	0.95 (0.08)	0.90 (0.21)	0.95 (0.15)	1.00 (0.00)
Feasible	0.65 (0.48)		0.82 (0.39)		0.88 (0.33)				
GC	0.26 (0.44)	0.39 (0.50)	0.56 (0.50)	0.69 (0.47)	0.70 (0.47)	0.79 (0.41)	0.27 (0.45)	0.55 (0.50)	0.63 (0.49)
N	44 (43)	28	42 (39)	32	48 (33)	29	44	42	40

Game 2									
	No Chat ^a	No Chat ^b	Demand-based		Winter ^a	Winter ^b	Offer-based		H-MC
			Chat ^a	Chat ^b			No Chat	Chat	
All	0.93 (0.25)		0.95 (0.22)		0.83 (0.50)		0.77 (0.42)	0.90 (0.30)	0.95 (0.22)
Dist.SV	9.87 (4.28)	8.3 (3.18)	10.28 (3.51)	10.29 (3.58)	13.93 (15.55)	7.94 (3.73)	20.12 (15.57)	16.27 (12.30)	29.22 (18.87)
Dist.ED	10.41 (6.25)	7.89 (3.98)	8.13 (5.43)	8.28 (5.73)	17.27 (17.62)	10.87 (5.51)	14.51 (17.76)	10.58 (11.38)	15.61 (17.84)
Eff.	1.01 (0.11)	0.94 (0.05)	0.95 (0.09)	0.93 (0.08)	1.08 (0.23)	0.99 (0.03)	0.90 (0.19)	0.96 (0.12)	1.00 (0.00)
Feasible	0.66 (0.48)		0.83 (0.38)		0.73 (0.45)				
GC	0.20 (0.41)	0.30 (0.47)	0.38 (0.49)	0.45 (0.51)	0.68 (0.47)	0.93 (0.26)	0.23 (0.42)	0.24 (0.43)	0.45 (0.50)
N	44 (41)	27	42 (40)	33	48 (40)	29	44	42	40

^a Dist.SV, Dist.ED, Eff., Feasible, and GC are computed only for those cases where all four (three in the case of game 2) demands are active (i.e., All = 1). The number of observations are reported in parentheses.

^b Dist.SV, Dist.ED, Dff., and GC are computed only for those cases where all four demands are active and those demands are jointly feasible.

Table B.2: Characteristics of the first sets of four demands (for demand-based approaches) and the first offers (for offer-based approaches) in games 3 and 4

Game 3									
	No Chat ^a	No Chat ^b	Demand-based		Winter ^a	Winter ^b	Offer-based		H-MC
			Chat ^a	Chat ^b			No Chat	Chat	
All	0.86 (0.35)		0.90 (0.30)		0.56 (0.50)		0.68 (0.47)	0.74 (0.45)	1.00 (0.00)
Dist.SV	31.39 (10.94)	30.32 (11.10)	33.41 (7.61)	33.49 (7.09)	32.28 (12.03)	32.18 (13.48)	36.29 (19.37)	35.66 (18.25)	36.72 (16.09)
Dist.ED	24.68 (15.59)	26.83 (17.67)	18.45 (18.85)	16.77 (19.66)	31.58 (15.12)	36.09 (13.80)	25.10 (27.56)	21.51 (26.11)	18.04 (19.16)
Eff.	1.02 (0.15)	0.91 (0.12)	0.96 (0.14)	0.92 (0.12)	0.94 (0.14)	0.90 (0.14)	0.88 (0.22)	0.89 (0.21)	1.00 (0.00)
Feasible	0.50 (0.51)		0.79 (0.41)		0.74 (0.45)				
GC	0.19 (0.40)	0.37 (0.50)	0.53 (0.51)	0.67 (0.48)	0.22 (0.42)	0.30 (0.47)	0.18 (0.39)	0.31 (0.47)	0.48 (0.51)
N	44 (38)	19	42 (38)	30	48 (27)	20	44	42	40

Game 4									
	No Chat ^a	No Chat ^b	Demand-based		Winter ^a	Winter ^b	Offer-based		H-MC
			Chat ^a	Chat ^b			No Chat	Chat	
All	0.86 (0.35)		0.98 (0.15)		0.71 (0.46)		0.86 (0.35)	0.90 (0.30)	1.00 (0.00)
Dist.SV	22.87 (12.31)	23.34 (12.21)	21.00 (13.48)	22.51 (14.20)	34.56 (11.63)	34.56 (11.63)	20.63 (21.25)	18.21 (16.13)	24.88 (24.41)
Dist.ED	21.76 (16.65)	20.49 (17.65)	15.99 (17.32)	16.29 (17.65)	36.07 (12.74)	36.07 (12.74)	14.27 (25.29)	9.67 (19.31)	20.67 (27.36)
Eff.	0.96 (0.17)	0.88 (0.14)	0.91 (0.17)	0.88 (0.17)	0.89 (0.17)	0.89 (0.17)	0.91 (0.23)	0.95 (0.16)	1.00 (0.00)
Feasible	0.68 (0.47)		0.83 (0.38)		1.00 (0.00)				
GC	0.38 (0.49)	0.58 (0.50)	0.44 (0.50)	0.53 (0.51)	0.62 (0.49)	0.62 (0.49)	0.20 (0.41)	0.43 (0.50)	0.75 (0.44)
N	44 (38)	26	42 (41)	34	48 (34)	34	44	42	40

^a Dist.SV, Dist.ED, Eff., Feasible, and GC are computed only for those cases where all four demands are active (i.e., All = 1). The number of relevant observations are reported in parentheses.

^b Dist.SV, Dist.ED, Dff., and GC are computed only for those cases where all four demands are active and those demands are jointly feasible.

C Additional results

For completeness and a more exhaustive comparison with the results presented in Chessa et al. (2023b), we report additional analyses here.

C.1 The realized allocations within the grand coalition

We investigate the realized allocations within the grand coalition (including $\{2,3,4\}$ in game 2). We focus on grand coalition because it is important to stress that potential differences with the theoretical prediction not only are the consequence of failing to form the grand coalition but also affect the final allocations even when players manage to form one.

We categorize various allocations into eight types, summarized with examples in Table C.1, depending on whether they are efficient, whether they respect equal sharing (i.e., $\pi_i = \pi_j$ for all $i, j \in GC$), and whether they satisfy a weak coalitional rationality (WCR) argument. The *core* is given by those allocations that are efficient and satisfy WCR at the same time. We recall that the formation of the grand coalition does not always coincide with efficiency. Surprisingly, in fact, players could form the grand coalition while “wasting” some value. In addition, investigating about equal sharing, particularly when the grand coalition is formed, is a crucial matter. Previous studies have shown the tendency of individuals to choose the equal sharing (Murnighan and Roth, 1977). It may be speculated that the possibility of communicating through a chat may lead even more often to equal sharing. Note that the equal division solution is an allocation that satisfies both efficiency and equal sharing. Finally, we check whether allocation x satisfies the WCR; that is, whether $x(S) \geq v(S) \forall S \subset N$. In other words, the sum of payoffs obtained by members of all possible coalitions under the realized

Table C.1: Eight allocation types

Type	Example	Description
1: Eff, Equal, WCR	{25,25,25,25} in game 1	$\sum_i \pi_i = v(N)$, $\pi_i = \pi_j$ for all $i, j \in GC$, WCR
2: Eff, Not Equal, WCR	{0,33,33,34} in game 2	$\sum_i \pi_i = v(N)$, $\pi_i \neq \pi_j$ for some $i, j \in GC$, WCR
3: Eff, Equal, Not WCR	{25,25,25,25} in game 2	$\sum_i \pi_i = v(N)$, $\pi_i = \pi_j$ for all $i, j \in GC$, Not WCR
4: Eff, Not Equal, Not WCR	{1,33,33,33} in game 2	$\sum_i \pi_i = v(N)$, $\pi_i \neq \pi_j$ for some $i, j \in GC$, Not WCR
5: Not Eff, Equal, WCR	{24,24,24,24} in game 1	$\sum_i \pi_i < v(N)$, $\pi_i = \pi_j$ for all $i, j \in GC$, WCR
6: Not Eff, Not Equal, WCR	{15,15,20,30} in game 1	$\sum_i \pi_i < v(N)$, $\pi_i \neq \pi_j$ for some $i, j \in GC$, WCR
7: Not Eff, Equal, Not WCR	{0,33,33,33} in game 2	$\sum_i \pi_i < v(N)$, $\pi_i = \pi_j$ for all $i, j \in GC$, Not WCR
8: Not Eff, Not Equal, Not WCR	{0,30,30,35} in game 2	$\sum_i \pi_i < v(N)$, $\pi_i \neq \pi_j$ for some $i, j \in GC$, Not WCR

Note: WCR: satisfies weak coalitional rationality. GC: grand coalition.

allocation is no less than their worth, except for the grand coalition. Investigating for WCR is interesting for checking if, regardless of the possible inefficiency of a proposed share, the allocation is still stable against the possible departure of smaller groups of players.

Tables C.2 and C.3 show the frequencies of observed allocation types among those formed the grand coalition in each treatment for games 1 and 4 and games 2 and 3, respectively. See Appendix C.2 for a list of the realized allocations among those formed the grand coalition in each game and treatment.

Table C.2: Frequencies of realized allocation types for games 1 and 4

Game 1						
Allocation Type	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
1: Eff, Equal, WCR	17	30	0	17	34	11
2: Eff, Not Equal, WCR	15	1	16	17	5	14
3: Eff, Equal, Not WCR	0	0	0	0	0	0
4: Eff, Not Equal, Not WCR	0	0	0	0	0	0
5: Not Eff, Equal, WCR	0	0	0	0	0	0
6: Not Eff, Not Equal, WCR	5	3	5	0	0	0
7: Not Eff, Equal, Not WCR	0	0	0	0	0	0
8: Not Eff, Not Equal, Not WCR	1	0	2	0	0	0
In the core	32	31	16	34	39	25
Not in the core	6	3	7	0	0	0
Total	38	34	23	34	39	25

Game 4						
Allocation Type	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
1: Eff, Equal, WCR	16	24	0	13	26	14
2: Eff, Not Equal, WCR	9	7	16	22	8	16
3: Eff, Equal, Not WCR	0	0	0	0	0	0
4: Eff, Not Equal, Not WCR	0	0	0	0	0	0
5: Not Eff, Equal, WCR	0	2	0	0	0	0
6: Not Eff, Not Equal, WCR	4	1	4	0	0	0
7: Not Eff, Equal, Not WCR	0	0	0	0	0	0
8: Not Eff, Not Equal, Not WCR	0	1	1	0	0	0
In the core	25	31	16	35	34	30
Not in the core	4	4	5	0	0	0
Total	29	35	21	35	34	30

For games 1 and 4 in Table C.2, we observe that most allocations belong to the core (i.e., types 1 and 2). Even among those cases that do not belong to the core because they are not efficient (only observed under demand-based approaches), only 1 or 2 fail to satisfy WCR. Furthermore, the equal division solution (that is, type 1 allocation in these games) accounts for almost half the cases in both the demand-based and offer-based

semi-structured procedures without chat. A similar observation can be made for H-MC.

The equal division solution is even more frequently observed in semi-structured procedures when chat is allowed. It is only under Winter that it is not observed. This prevalence of the equal division solution easily explains the reason behind the average payoff of player 4 being smaller than the Shapley value, as well as why the Shapley distance, due to the violation of additivity, tends to be larger with chat than without chat.

For games 2 and 3 shown in Table C.3, allocations that do not belong to the core (i.e., types 3 to 8) are more frequently observed, than in games 1 and 4. As in games 1 and 4, failure of the efficiency is observed only in demand-based approaches. Among the efficient allocations, however, there are cases, especially in offer-based approaches, in which WCR is not satisfied.

This difference between games 1 and 4 and games 2 and 3 is mainly due to the equal division solution not belonging to the core in the latter. To see this, one can check that the worth of a three-player coalition $\{2, 3, 4\}$ is greater than the sum of the payoff for these three players under the equal division solution ($75 < 100$ in case of game 2, and $150 < 160$ in case of game 3). However, in game 3, under a semi-structured procedure with chat, the equal division solution (type 3) is realized frequently for both the demand-based (18 out of 26) and offer-based procedures (11 out of 31). Thus, just as in games 1 and 4, the possibility of chat pushes the allocation toward the equal division solution.²¹

²¹Interestingly, we observe the following differences in the proposals or demands with and without chat in semi-structured procedures in game 3. In the offer-based case, coalition $\{2, 3, 4\}$ is proposed 29 times out of a total of 154 proposals with chat (18.8%); and without chat, it is proposed 59 times out of a total of 226 (26.1%). This difference, however, is not statistically significant ($p = 0.1269$, the proportion test). In the demand-based case, the grand coalition is feasible when four players have submitted demands in 45 out of 98 cases with chat (45.9%) while it is feasible in 40 out of 130 cases without chat (30.8%). This difference is significant ($p = 0.0276$, the proportion test). In addition, in the demand-based case, when players 2, 3, and 4 have submitted their demands, the sum of their demands is between 151 and 160 (thus, better than equal sharing among four and feasible) in 32 out of 157 cases with chat (20.4%), while it is

Table C.3: Frequencies of realized allocation types for games 2 and 3
Game 2

Allocation Type	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
1: Eff, Equal, WCR	0	0	0	0	0	0
2: Eff, Not Equal, WCR	21	18	25	27	20	7
3: Eff, Equal, Not WCR	0	2	0	0	2	2
4: Eff, Not Equal, Not WCR	2	2	0	3	7	16
5: Not Eff, Equal, WCR	0	0	0	0	0	0
6: Not Eff, Not Equal, WCR	0	0	0	0	0	0
7: Not Eff, Equal, Not WCR	6	3	0	0	0	0
8: Not Eff, Not Equal, Not WCR	7	8	2	0	0	0
In the core	21	18	25	27	20	7
Not in the core	15	15	2	3	9	18
Total	36	33	27	30	29	25

Game 3

Allocation Type	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
1: Eff, Equal, WCR	0	0	0	0	0	0
2: Eff, Not Equal, WCR	16	7	4	16	17	6
3: Eff, Equal, Not WCR	1	18	0	5	11	9
4: Eff, Not Equal, Not WCR	2	0	0	2	3	4
5: Not Eff, Equal, WCR	0	0	0	0	0	0
6: Not Eff, Not Equal, WCR	2	1	2	0	0	0
7: Not Eff, Equal, Not WCR	0	0	0	0	0	0
8: Not Eff, Not Equal, Not WCR	0	0	0	0	0	0
In the core	16	7	4	16	17	6
Not in the core	5	19	2	7	14	13
Total	21	26	6	23	31	19

However, the chat does not push the allocation toward the equal division solution when a null player exists (game 2). For this game, the most frequently observed allocations are type 2 - the efficient, non-equal ones that satisfy WCR (and thus belong feasible in 73 out of 195 cases without chat (37.4%). This difference is also significant ($p < 0.001$, the proportion test).

Table C.4: Frequencies of allocations when the coalition $\{2, 3, 4\}$ is formed in game 3

	Demand-based			Offer-based		
	No Chat	Chat	Winter	No Chat	Chat	H-MC
Better than the equal division	7	6	8	13	9	4
Others	3	2	19	1	0	1
Total	10	8	27	14	9	5

to the core) - except in H-MC. In these allocations, a coalition without the null player $\{2, 3, 4\}$ is formed, and the worth of the grand coalition is fully shared among the members, with one of the members obtaining slightly a higher payoff than the others (and is thus close to equal sharing; see Appendix C.2). In the semi-structured demand-based procedure without chat, there are also cases where these three players obtain 33 points, each leaving 1 point aside (type 6 allocation, which is equal but not efficient and does satisfy WCR).

C.1.1 Further investigation of game 3

Table 8 shows that the coalition $\{2, 3, 4\}$ was formed frequently in game 3 when players failed to form the grand coalition. As noted above, in this game, the worth of coalition $v(\{2, 3, 4\}) = 160$ is higher than the sum of payoffs to its three members under the equal division solution. At least some participants are aware of this, as a few discussions appear in the chat data, although they number only four in both the offer-based and the demand-based procedures. As a result, we have observed that without chat, the equal division solution is not frequently observed in this game. Indeed, except for Winter, in most cases when the coalition $\{2, 3, 4\}$ was formed, at least one member received more than 50 (indicated as “Better than the equal division”), as shown in Table C.4.

C.2 Allocations within Grand Coalition

In this appendix, we list the frequencies of the allocations realized within the grand coalition (including $\{2, 3, 4\}$ in game 2). In each table, type corresponds to the type of allocation described in Section C.1, allocation in the form of $\{\pi_1, \pi_2, \pi_3, \pi_4\}$, freq. is the frequency of the realization, efficiency is $\sum \pi_{i \in GC} / v(N)$, Dis. Eq. is the sum of the squared deviation from the equal devision within the grand coalition (GC) $\sum_{i \in GC} (\pi_i - \sum \pi_{i \in GC} / |GC|)^2$, $\in \text{Core} = 1$ when the allocation belongs to the core, and $= 0$ otherwise, No. Block is the number of coalitions that can potentially block the realized coalition by offering at least one of its members a better payoff, and Block is the largest coalition among such blocking coalitions. Allocations are sorted according to type, frequency, efficiency and distance from the equal devision.

Table C.5: Realized allocations. Game 1. Demand-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
1	25,25,25,25	17	1.000	0.000	1	0	
3	20,25,25,30	2	1.000	50.000	1	0	
3	25,25,20,30	2	1.000	50.000	1	0	
3	25,20,30,25	2	1.000	50.000	1	0	
3	25,25,30,20	1	1.000	50.000	1	0	
3	20,25,30,25	1	1.000	50.000	1	0	
3	20,20,30,30	1	1.000	100.000	1	0	
3	15,25,30,30	1	1.000	150.000	1	0	
3	30,30,15,25	1	1.000	150.000	1	0	
3	20,25,20,35	1	1.000	150.000	1	0	
3	20,15,30,35	1	1.000	250.000	1	0	
3	20,20,20,40	1	1.000	300.000	1	0	
3	10,30,20,40	1	1.000	500.000	1	0	
7	25,25,20,25	1	0.950	18.750	1	0	
7	20,25,25,25	1	0.950	18.750	1	0	
7	20,20,25,25	1	0.900	25.000	1	0	
7	20,25,25,27	1	0.970	26.750	1	0	
7	15,15,20,30	1	0.800	150.000	1	0	
8	15,15,25,5	1	0.600	200.000	0	6	124

Table C.6: Realized allocations. Game 1. Demand-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
1	25,25,25,25	30	1.000	0.000	1	0	
3	25,15,25,35	1	1.000	200.000	1	0	
7	25,23,25,25	1	0.980	3.000	1	0	
7	25,20,25,25	1	0.950	18.750	1	0	
7	15,20,30,25	1	0.900	125.000	1	0	

Table C.7: Realized allocations. Game 1. Demand-based. Winter

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
3	20,20,30,30	2	1.000	100.000	1	0	
3	29,26,20,25	1	1.000	42.000	1	0	
3	25,25,20,30	1	1.000	50.000	1	0	
3	25,30,20,25	1	1.000	50.000	1	0	
3	30,25,25,20	1	1.000	50.000	1	0	
3	20,25,30,25	1	1.000	50.000	1	0	
3	20,30,30,20	1	1.000	100.000	1	0	
3	34,23,23,20	1	1.000	114.000	1	0	
3	30,25,15,30	1	1.000	150.000	1	0	
3	30,30,15,25	1	1.000	150.000	1	0	
3	14,25,31,30	1	1.000	182.000	1	0	
3	30,30,10,30	1	1.000	300.000	1	0	
3	15,20,25,40	1	1.000	350.000	1	0	
3	5,30,30,35	1	1.000	550.000	1	0	
3	14,23,15,48	1	1.000	754.000	1	0	
7	25,25,20,15	1	0.850	68.750	1	0	
7	32,15,17,28	1	0.920	206.000	1	0	
7	38,21,15,25	1	0.990	284.750	1	0	
7	10,34,25,25	1	0.940	297.000	1	0	
7	35,10,30,20	1	0.950	368.750	1	0	
8	25,20,10,25	1	0.800	150.000	0	1	234
8	5,33,30,12	1	0.800	558.000	0	3	134

Table C.8: Realized allocations. Game 1. Offer-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
1	25,25,25,25	17	1.000	0.000	1	0	
3	20,25,25,30	3	1.000	50.000	1	0	
3	10,30,30,30	3	1.000	300.000	1	0	
3	22,26,26,26	2	1.000	12.000	1	0	
3	25,25,24,26	1	1.000	2.000	1	0	
3	24,26,26,24	1	1.000	4.000	1	0	
3	23,25,25,27	1	1.000	8.000	1	0	
3	21,26,26,27	1	1.000	22.000	1	0	
3	20,25,27,28	1	1.000	38.000	1	0	
3	19,27,27,27	1	1.000	48.000	1	0	
3	20,22,28,30	1	1.000	68.000	1	0	
3	20,21,30,29	1	1.000	82.000	1	0	
3	16,28,28,28	1	1.000	108.000	1	0	

Table C.9: Realized allocations. Game 1. Offer-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
1	25,25,25,25	34	1.000	0.000	1	0	
3	10,30,30,30	2	1.000	300.000	1	0	
3	20,25,25,30	1	1.000	50.000	1	0	
3	16,27,27,30	1	1.000	114.000	1	0	
3	8,30,30,32	1	1.000	388.000	1	0	

Table C.10: Realized allocations. Game 1. Offer-based. H-MC.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
1	25,25,25,25	11	1.000	0.000	1	0	
3	23,31,23,23	3	1.000	48.000	1	0	
3	20,25,25,30	2	1.000	50.000	1	0	
3	28,24,24,24	1	1.000	12.000	1	0	
3	21,25,25,29	1	1.000	32.000	1	0	
3	23,23,31,23	1	1.000	48.000	1	0	
3	22,22,31,25	1	1.000	54.000	1	0	
3	21,23,31,25	1	1.000	56.000	1	0	
3	34,22,22,22	1	1.000	108.000	1	0	
3	23,35,22,20	1	1.000	138.000	1	0	
3	15,25,25,35	1	1.000	200.000	1	0	
3	15,20,20,45	1	1.000	550.000	1	0	

Table C.11: Realized allocations. Game 2. Demand-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
3	0,30,30,40	5	1.000	66.667	1	0	
3	0,33,32,35	3	1.000	4.667	1	0	
3	0,30,35,35	3	1.000	16.667	1	0	
3	0,35,32,33	2	1.000	4.667	1	0	
3	0,29,31,40	2	1.000	68.667	1	0	
3	0,25,25,50	2	1.000	416.667	1	0	
3	0,33,33,34	1	1.000	0.667	1	0	
3	0,33,30,37	1	1.000	24.667	1	0	
3	0,33,27,40	1	1.000	84.667	1	0	
3	0,28,27,45	1	1.000	204.667	1	0	
4	10,20,30,40	1	1.000	500.000	0	1	234
4	3,28,31,38	1	1.000	698.000	0	1	234
6	0,33,33,33	4	0.990	0.000	0	1	234
6	0,30,30,30	2	0.900	0.000	0	1	234
8	0,30,25,35	1	0.900	50.000	0	1	234
8	0,25,27,35	1	0.870	56.000	0	1	234
8	0,35,25,35	1	0.950	66.667	0	1	234
8	0,20,30,35	1	0.850	116.667	0	1	234
8	0,30,22,40	1	0.920	162.667	0	1	234
8	0,22,25,50	1	0.970	472.667	0	1	234
8	1,33,33,30	1	0.970	726.750	0	1	234

Table C.12: Realized allocations. Game 2. Demand-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	25,25,25,25	2	1.000	0.000	0	3	234
3	0,35,30,35	5	1.000	16.667	1	0	
3	0,33,34,33	2	1.000	0.667	1	0	
3	0,33,33,34	2	1.000	0.667	1	0	
3	0,34,33,33	1	1.000	0.667	1	0	
3	0,32,33,35	1	1.000	4.667	1	0	
3	0,30,35,35	1	1.000	16.667	1	0	
3	0,30,30,40	1	1.000	66.667	1	0	
3	0,30,40,30	1	1.000	66.667	1	0	
3	0,25,40,35	1	1.000	116.667	1	0	
3	0,35,25,40	1	1.000	116.667	1	0	
3	0,25,35,40	1	1.000	116.667	1	0	
3	0,20,30,50	1	1.000	466.667	1	0	
4	25,20,25,30	1	1.000	50.000	0	2	234
4	1,33,33,33	1	1.000	768.000	0	1	234
6	0,33,33,33	2	0.990	0.000	0	1	234
6	0,30,30,30	1	0.900	0.000	0	1	234
8	0,30,33,30	1	0.930	6.000	0	1	234
8	0,30,35,33	1	0.980	12.667	0	1	234
8	0,33,30,35	1	0.980	12.667	0	1	234
8	0,30,32,35	1	0.970	12.667	0	1	234
8	0,30,30,35	1	0.950	16.667	0	1	234
8	0,35,25,35	1	0.950	66.667	0	1	234
8	0,22,30,35	1	0.870	86.000	0	1	234
8	7,25,27,40	1	0.990	552.750	0	1	234

Table C.13: Realized allocations. Game 2. Demand-based. Winter

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
3	0,30,30,40	3	1.000	66.667	1	0	
3	0,25,40,35	3	1.000	116.667	1	0	
3	0,35,25,40	3	1.000	116.667	1	0	
3	0,30,35,35	2	1.000	16.667	1	0	
3	0,25,35,40	2	1.000	116.667	1	0	
3	0,25,30,45	2	1.000	216.667	1	0	
3	0,34,34,32	1	1.000	2.667	1	0	
3	0,32,33,35	1	1.000	4.667	1	0	
3	0,30,37,33	1	1.000	24.667	1	0	
3	0,30,39,31	1	1.000	48.667	1	0	
3	0,32,29,39	1	1.000	52.667	1	0	
3	0,30,29,41	1	1.000	88.667	1	0	
3	0,25,41,34	1	1.000	128.667	1	0	
3	0,30,25,45	1	1.000	216.667	1	0	
3	0,29,25,46	1	1.000	248.667	1	0	
3	0,35,20,45	1	1.000	316.667	1	0	
8	0,25,30,40	1	0.950	116.667	0	1	234
8	0,25,25,42	1	0.920	192.667	0	1	234

Table C.14: Realized allocations. Game 2. Offer-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
3	0,30,30,40	10	1.000	66.667	1	0	
3	0,33,33,34	4	1.000	0.667	1	0	
3	0,30,35,35	3	1.000	16.667	1	0	
3	0,34,33,33	2	1.000	0.667	1	0	
3	0,31,31,38	2	1.000	32.667	1	0	
3	0,33,34,33	1	1.000	0.667	1	0	
3	0,32,34,34	1	1.000	2.667	1	0	
3	0,31,34,35	1	1.000	8.667	1	0	
3	0,34,31,35	1	1.000	8.667	1	0	
3	0,32,32,36	1	1.000	10.667	1	0	
3	0,29,29,42	1	1.000	112.667	1	0	
4	20,30,25,25	1	1.000	50.000	0	2	234
4	10,30,30,30	1	1.000	300.000	0	1	234
4	1,33,33,33	1	1.000	768.000	0	1	234

Table C.15: Realized allocations. Game 2. Offer-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	25,25,25,25	2	1.000	0.000	0	3	234
3	0,30,30,40	7	1.000	66.667	1	0	
3	0,33,33,34	6	1.000	0.667	1	0	
3	0,32,32,36	2	1.000	10.667	1	0	
3	0,34,33,33	1	1.000	0.667	1	0	
3	0,32,33,35	1	1.000	4.667	1	0	
3	0,33,32,35	1	1.000	4.667	1	0	
3	0,35,30,35	1	1.000	16.667	1	0	
3	0,30,35,35	1	1.000	16.667	1	0	
4	15,25,25,35	2	1.000	200.000	0	1	234
4	20,25,25,30	1	1.000	50.000	0	1	234
4	10,30,30,30	1	1.000	300.000	0	1	234
4	10,28,28,34	1	1.000	324.000	0	1	234
4	7,28,28,37	1	1.000	486.000	0	1	234
4	1,33,33,33	1	1.000	768.000	0	1	234

Table C.16: Realized allocations. Game 2. Offer-based. H–MC.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	25,25,25,25	2	1.000	0.000	0	3	234
3	0,33,33,34	2	1.000	0.667	1	0	
3	0,33,34,33	1	1.000	0.667	1	0	
3	0,30,35,35	1	1.000	16.667	1	0	
3	0,37,27,36	1	1.000	60.667	1	0	
3	0,30,30,40	1	1.000	66.667	1	0	
3	0,25,40,35	1	1.000	116.667	1	0	
4	15,25,25,35	3	1.000	200.000	0	1	234
4	24,24,28,24	1	1.000	12.000	0	3	234
4	25,20,25,30	1	1.000	50.000	0	2	234
4	21,23,24,32	1	1.000	70.000	0	1	234
4	21,21,27,31	1	1.000	72.000	0	2	234
4	21,23,23,33	1	1.000	88.000	0	1	234
4	22,22,22,34	1	1.000	108.000	0	2	234
4	15,25,30,30	1	1.000	150.000	0	1	234
4	20,22,22,36	1	1.000	164.000	0	2	234
4	20,20,20,40	1	1.000	300.000	0	2	234
4	11,27,27,35	1	1.000	304.000	0	1	234
4	10,25,30,35	1	1.000	350.000	0	1	234
4	10,27,27,36	1	1.000	354.000	0	1	234
4	10,20,40,30	1	1.000	500.000	0	2	234

Table C.17: Realized allocations. Game 3. Demand-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	50,50,50,50	1	1.000	0.000	0	1	234
3	40,50,50,60	3	1.000	200.000	1	0	
3	40,53,50,57	2	1.000	158.000	1	0	
3	40,60,50,50	2	1.000	200.000	1	0	
3	40,56,50,54	1	1.000	152.000	1	0	
3	40,55,40,65	1	1.000	450.000	1	0	
3	35,50,50,65	1	1.000	450.000	1	0	
3	35,45,53,67	1	1.000	548.000	1	0	
3	30,50,60,60	1	1.000	600.000	1	0	
3	40,50,40,70	1	1.000	600.000	1	0	
3	30,60,60,50	1	1.000	600.000	1	0	
3	25,60,45,70	1	1.000	1150.000	1	0	
3	20,50,70,60	1	1.000	1400.000	1	0	
4	50,50,40,60	1	1.000	200.000	0	1	234
4	47,40,53,60	1	1.000	218.000	0	1	234
7	30,50,50,60	1	0.950	475.000	1	0	
7	15,50,60,70	1	0.975	1718.750	1	0	

Table C.18: Realized allocations. Game 3. Demand-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	50,50,50,50	18	1.000	0.000	0	1	234
3	40,50,60,50	2	1.000	200.000	1	0	
3	30,60,50,60	1	1.000	600.000	1	0	
3	25,60,55,60	1	1.000	850.000	1	0	
3	25,50,60,65	1	1.000	950.000	1	0	
3	20,50,60,70	1	1.000	1400.000	1	0	
3	10,50,70,70	1	1.000	2400.000	1	0	
7	15,50,50,60	1	0.875	1168.750	1	0	

Table C.19: Realized allocations. Game 3. Demand-based. Winter.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
3	30,51,45,74	1	1.000	1002.000	1	0	
3	20,50,65,65	1	1.000	1350.000	1	0	
3	10,70,60,60	1	1.000	2200.000	1	0	
3	11,70,50,69	1	1.000	2282.000	1	0	
7	20,45,84,50	1	0.995	2080.750	1	0	
7	35,30,40,90	1	0.975	2318.750	1	0	

Table C.20: Realized allocations. Game 3. Offer-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
2	50,50,50,50	5	1.000	0.000	0	1	234
3	35,51,51,63	2	1.000	396.000	1	0	
3	25,55,55,65	2	1.000	900.000	1	0	
3	20,60,60,60	2	1.000	1200.000	1	0	
3	10,60,60,70	2	1.000	2200.000	1	0	
3	40,55,55,50	1	1.000	150.000	1	0	
3	38,55,49,58	1	1.000	234.000	1	0	
3	35,55,55,55	1	1.000	300.000	1	0	
3	31,54,54,61	1	1.000	514.000	1	0	
3	30,55,55,60	1	1.000	550.000	1	0	
3	30,50,50,70	1	1.000	800.000	1	0	
3	25,55,60,60	1	1.000	850.000	1	0	
3	25,60,55,60	1	1.000	850.000	1	0	
4	47,47,47,59	1	1.000	108.000	0	1	234
4	45,45,45,65	1	1.000	300.000	0	1	234

Table C.21: Realized allocations. Game 3. Offer-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	50,50,50,50	11	1.000	0.000	0	1	234
3	30,55,55,60	4	1.000	550.000	1	0	
3	40,50,50,60	3	1.000	200.000	1	0	
3	35,55,55,55	2	1.000	300.000	1	0	
3	35,50,50,65	2	1.000	450.000	1	0	
3	20,50,50,80	2	1.000	1800.000	1	0	
3	38,52,52,58	1	1.000	216.000	1	0	
3	35,52,53,60	1	1.000	338.000	1	0	
3	31,53,53,63	1	1.000	548.000	1	0	
3	25,55,55,65	1	1.000	900.000	1	0	
4	45,51,51,53	1	1.000	36.000	0	1	234
4	45,50,50,55	1	1.000	50.000	0	1	234
4	41,53,53,53	1	1.000	108.000	0	1	234

Table C.22: Realized allocations. Game 3. Offer-based. H-MC.

type	allocation	freq.	efficiency	Dis. Eq.	$\in$ Core	No. Block	Block
2	50,50,50,50	9	1.000	0.000	0	1	234
3	30,50,50,70	3	1.000	800.000	1	0	
3	40,50,50,60	2	1.000	200.000	1	0	
3	40,40,80,40	1	1.000	1200.000	1	0	
4	50,55,40,55	1	1.000	150.000	0	1	234
4	50,45,45,60	1	1.000	150.000	0	1	234
4	50,50,40,60	1	1.000	200.000	0	1	234
4	55,35,55,55	1	1.000	300.000	0	1	234

Table C.23: Realized allocations. Game 4. Demand-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
1	50,50,50,50	16	1.000	0.000	1	0	
3	40,50,50,60	2	1.000	200.000	1	0	
3	56,50,50,44	1	1.000	72.000	1	0	
3	50,40,50,60	1	1.000	200.000	1	0	
3	44,43,50,63	1	1.000	254.000	1	0	
3	40,60,40,60	1	1.000	400.000	1	0	
3	50,30,60,60	1	1.000	600.000	1	0	
3	35,40,50,75	1	1.000	950.000	1	0	
3	40,40,40,80	1	1.000	1200.000	1	0	
7	50,50,50,45	1	0.975	18.750	1	0	
7	45,40,40,55	1	0.900	150.000	1	0	
7	30,30,50,70	1	0.900	1100.000	1	0	
7	30,50,45,50	1	0.875	268.750	1	0	

Table C.24: Realized allocations. Game 4. Demand-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
1	50,50,50,50	24	1.000	0.000	1	0	
3	40,50,50,60	3	1.000	200.000	1	0	
3	45,52,53,50	1	1.000	38.000	1	0	
3	50,50,40,60	1	1.000	200.000	1	0	
3	30,50,60,60	1	1.000	600.000	1	0	
3	30,40,50,80	1	1.000	1400.000	1	0	
5	40,40,40,40	2	0.800	0.000	1	0	
7	35,50,50,50	1	0.925	168.750	1	0	
8	35,40,50,40	1	0.825	118.750	0	1	124

Table C.25: Realized allocations. Game 4. Demand-based. Winter.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
3	55,45,45,55	1	1.000	100.000	1	0	
3	60,40,50,50	1	1.000	200.000	1	0	
3	40,50,60,50	1	1.000	200.000	1	0	
3	50,40,62,48	1	1.000	248.000	1	0	
3	35,50,65,50	1	1.000	450.000	1	0	
3	40,40,70,50	1	1.000	600.000	1	0	
3	30,50,60,60	1	1.000	600.000	1	0	
3	50,63,30,57	1	1.000	618.000	1	0	
3	25,60,60,55	1	1.000	850.000	1	0	
3	30,40,70,60	1	1.000	1000.000	1	0	
3	66,30,35,69	1	1.000	1242.000	1	0	
3	30,79,61,30	1	1.000	1762.000	1	0	
3	50,50,20,80	1	1.000	1800.000	1	0	
3	40,80,20,60	1	1.000	2000.000	1	0	
3	40,30,40,90	1	1.000	2200.000	1	0	
3	53,13,54,80	1	1.000	2294.000	1	0	
7	70,45,50,25	1	0.950	1025.000	1	0	
7	40,40,30,80	1	0.950	1475.000	1	0	
7	60,66,25,35	1	0.930	1157.000	1	0	
7	50,50,10,60	1	0.850	1475.000	1	0	
8	15,50,40,50	1	0.775	818.750	0	2	134

Table C.26: Realized allocations. Game 4. Offer-based. No Chat.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
1	50,50,50,50	13	1.000	0.000	1	0	
3	40,50,50,60	6	1.000	200.000	1	0	
3	47,51,51,51	2	1.000	12.000	1	0	
3	35,55,55,55	2	1.000	300.000	1	0	
3	49,51,50,50	1	1.000	2.000	1	0	
3	48,50,51,51	1	1.000	6.000	1	0	
3	51,51,51,47	1	1.000	12.000	1	0	
3	52,45,51,52	1	1.000	34.000	1	0	
3	44,52,52,52	1	1.000	48.000	1	0	
3	50,45,55,50	1	1.000	50.000	1	0	
3	50,50,45,55	1	1.000	50.000	1	0	
3	43,51,51,55	1	1.000	76.000	1	0	
3	42,52,52,54	1	1.000	88.000	1	0	
3	40,50,55,55	1	1.000	150.000	1	0	
3	35,50,50,65	1	1.000	450.000	1	0	
3	20,65,50,65	1	1.000	1350.000	1	0	

Table C.27: Realized allocations. Game 4. Offer-based. With Chat.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
1	50,50,50,50	26	1.000	0.000	1	0	
3	40,50,50,60	4	1.000	200.000	1	0	
3	45,50,50,55	2	1.000	50.000	1	0	
3	45,55,50,50	1	1.000	50.000	1	0	
3	30,55,55,60	1	1.000	550.000	1	0	

Table C.28: Realized allocations. Game 4. Offer-based. H-MC.

type	allocation	freq.	efficiency	Dis. Eq.	∈ Core	No. Block	Block
1	50,50,50,50	14	1.000	0.000	1	0	
3	40,50,50,60	3	1.000	200.000	1	0	
3	45,65,45,45	2	1.000	300.000	1	0	
3	55,50,50,45	1	1.000	50.000	1	0	
3	47,59,47,47	1	1.000	108.000	1	0	
3	55,55,40,50	1	1.000	150.000	1	0	
3	35,55,55,55	1	1.000	300.000	1	0	
3	40,44,68,48	1	1.000	464.000	1	0	
3	50,40,40,70	1	1.000	600.000	1	0	
3	41,73,41,45	1	1.000	716.000	1	0	
3	40,75,40,45	1	1.000	850.000	1	0	
3	80,40,40,40	1	1.000	1200.000	1	0	
3	30,35,60,75	1	1.000	1350.000	1	0	
3	30,40,80,50	1	1.000	1400.000	1	0	

C.3 Allocations

In this appendix, we turn our attention to an individual-level analysis; that is, we investigate the experimental payoff obtained by each player, depending on her role in the games.

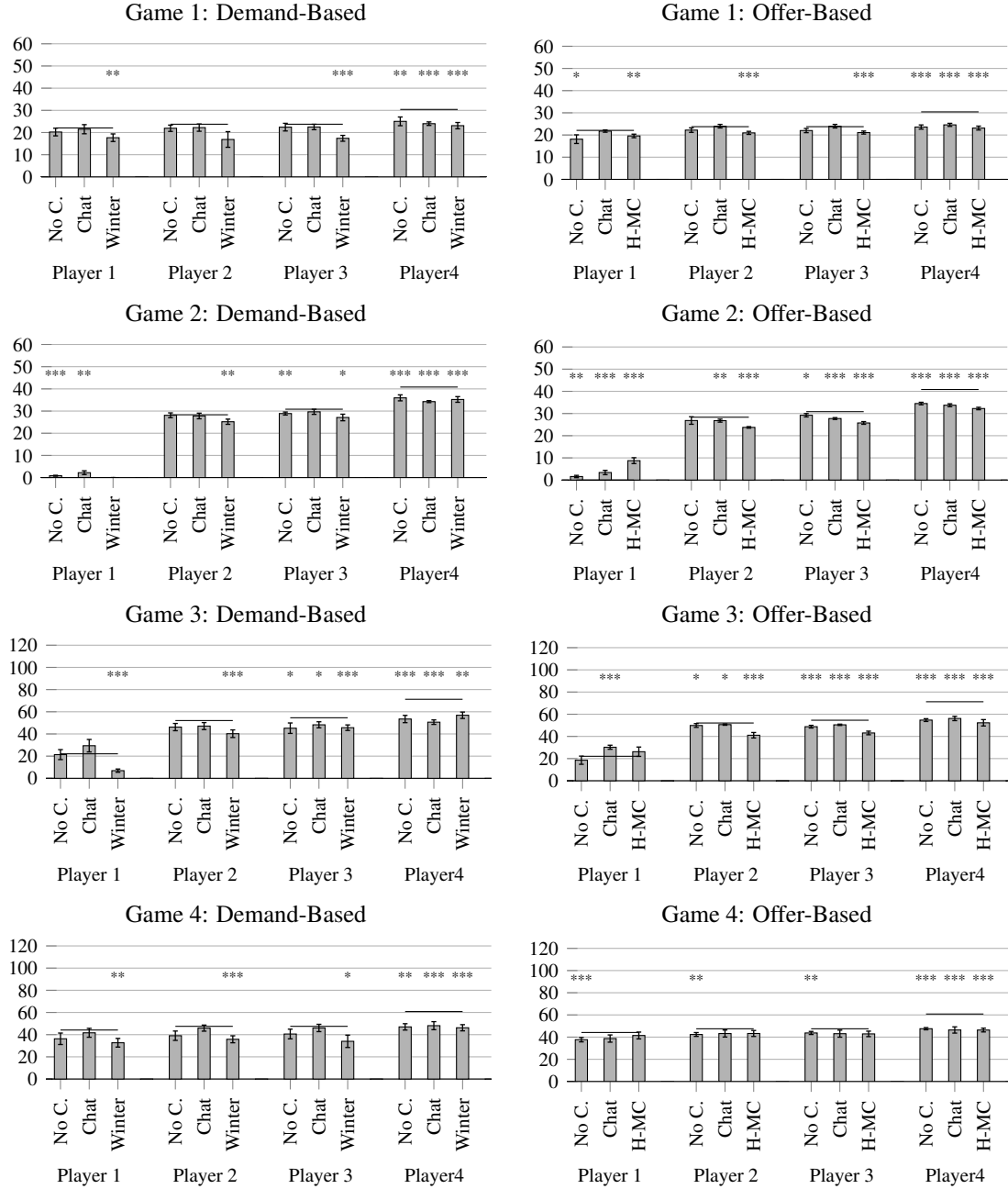
Fig. C.1 shows the mean payoffs for each player in the four games. Treatments are divided into demand-based (without chat (No C.), with chat (chat), and Winter) and offer-based (without chat (No C.), with chat (chat), and H-MC). The horizontal lines indicate the Shapley values for each player in each game. The means and standard errors are obtained by running a set of ordinary least squares regressions for the following system of equations for each treatment separately:

$$\begin{aligned}
 \pi_1 &= a_1g_1 + a_2g_2 + a_3g_3 + a_4g_4 + u_1 \\
 \pi_2 &= b_1g_1 + b_2g_2 + b_3g_3 + b_4g_4 + u_2 \\
 \pi_3 &= c_1g_1 + c_2g_2 + c_3g_3 + c_4g_4 + u_3 \\
 \pi_4 &= d_1g_1 + d_2g_2 + d_3g_3 + d_4g_4 + u_4,
 \end{aligned} \tag{1}$$

where π_i is the payoff of player i , g_j is a dummy variable that takes a value of one if the game $j \in \{1, 2, 3, 4\}$ is played and zero otherwise. Because participants play all four games twice, we correct the standard errors for within-group clustering effects. Note that the estimated coefficients a_j , b_j , c_j , and d_j are the average payoffs in game j for players 1, 2, 3, and 4, respectively.

We observe that the mean payoff of player 4 is significantly lower than the Shapley value, at at least the 5% significance level, in all the games and treatments. In terms of treatment differences, we observe that the average payoffs tend to be lower under Winter and H-MC than under the corresponding semi-structured ones. As a result, the

Figure C.1: Mean payoffs for each player in each treatment



Note: The horizontal lines indicate the Shapley values; error bars show one standard error range; ***, **, and * indicate the average payoff being significantly different from the Shapley value at 1%, 5%, and 10 % significance levels (Wald test), respectively.

average payoffs deviate significantly from the Shapley values for more players in more games under the structured than under the semi-structured procedure.

Furthermore, there is a notable difference in terms of the average payoff of the null player i.e., player 1 in game 2, between the structured and semi-structured procedures. On one hand, in Winter, the average payoff of the null player is zero and equal to the Shapley value. On the other hand, in H-MC, the average payoff of the null player is much higher. The average payoffs of the null player under the semi-structured experiments are between Winter and H-MC, and are higher when chat is possible. As we show next, the null player property tends to fail by a larger extent under the offer-based experiment than under the demand-based experiment, regardless of whether it is structured (Winter vs. H-MC, as shown by Chessa et al., 2023b) or semi-structured, although the difference is smaller in the latter. We now investigate the sources of the deviation of the realized allocation of the Shapley value.

We observe that both the higher average payoff for the null player (the one with a zero Shapley value) and the generally lower payoff for the fourth player (the one with the highest Shapley value) align with the observable tendency toward an equal distribution of shares. This matter, largely documented in the literature, is discussed further in Section C.1.

C.4 Shapley distance

To understand in greater detail why the realized allocations diverge so significantly from the Shapley value prediction, in this appendix we apply the approach of Aguiar et al. (2018) to decompose the deviation²² into the failure of efficiency, symmetry, additivity,

²²Rapoport (1987) considers alternative metrics to measure the distance between payoff vectors and solution concepts, such as the Bonacich's error measure (Bonacich, 1979) and the net rate of success

and the null player property.²³ The same method of decomposition is used in Chessa et al. (2022, 2023a,b) in experiments on structured mechanisms. Although the procedure of the decomposition is presented in these previous papers, in order for this appendix to be self-contained, we present it again here.

Let π be the realized allocation (i.e., vector of payoffs) in a game. First, we find an allocation, π^{sym} , that satisfies the symmetry and is closest to π . We do so by summing the payoffs obtained by symmetric players s (players 2 and 3 in games 1 and 4²⁴) and divide it equally among them. That is, in games 1 and 4, $\pi_s^{sym} = \sum_{s \in \{2,3\}} \pi_s / 2$. For other players k , $\pi_k^{sym} = \pi_k$. Second, we find a new allocation, $\pi^{sym,eff}$, that satisfies efficiency and is closest to π^{sym} . For each player $i = 1, 2, 3, 4$, $\pi_i^{sym,eff} = \pi_i^{sym} + [v(N) - \sum_{j \in N} \pi_j] / 4$. Third, we find yet another allocation, $\pi^{sym,eff,null}$, that satisfies null player property and is closest to $\pi^{sym,eff}$. For a null player n (player 1 in game 2), $\pi_n^{sym,eff,null} = 0$. For other players j in the game, $\pi_j^{sym,eff,null} = \pi_j^{sym,eff} + \pi_n^{sym,eff} / 3$. That is, three other players in the game equally share the $\pi_n^{sym,eff}$ of the null player. If there is no null player, $\pi_i^{sym,eff,null} = \pi_i^{sym,eff}$ for all i .

Let $e_i^{sym} = \pi_i - \pi_i^{sym}$, $e_i^{eff} = \pi_i^{sym} - \pi_i^{sym,eff}$, $e_i^{null} = \pi_i^{sym,eff} - \pi_i^{sym,eff,null}$, and $e_i^{add} = \pi_i^{sym,eff,null} - \phi_i(v)$ for all i .

Aguiar et al. (2018, Theorem 3) shows that an allocation π from game v can be decomposed as $\pi = \phi(v) + e^{sym} + e^{eff} + e^{null} + e^{add}$. Therefore, the Shapley error, $e^\phi = \pi - \phi(v)$, is $e^\phi = e^{sym} + e^{eff} + e^{null} + e^{add}$, and the Shapley distance, $\|e^\phi\|^2$, can

(Selten and Krischker, 1983).

²³Aguiar et al. (2018) decompose the deviation into the failures of efficiency, symmetry, and marginality. While these three components are orthogonal to each other, in our decomposition, the failures due to additivity and the null player property are not orthogonal.

²⁴Players i and j are *symmetric* in $v \in \mathcal{G}^N$ if $v(S \cup \{i\}) = v(S \cup \{j\})$ for all $S \subseteq N \setminus \{i, j\}$.

be decomposed into

$$||e^\phi||^2 = ||e^{sym}||^2 + ||e^{eff}||^2 + ||e^{null}||^2 + ||e^{add}||^2 + 2 \langle e^{add}, e^{null} \rangle,$$

where $\langle \cdot, \cdot \rangle$ is the scalar product and for any vector $y \in \mathbb{R}^n$, $||y||^2 = \langle y, y \rangle = \sum_{i \in N} y_i^2$. As noted above, in general, vectors e^{null} and e^{add} are not orthogonal so that $\langle e^{add}, e^{null} \rangle$ is not equal to zero. Its magnitude, however, is much smaller than the remaining components in our experimental data.

We perform the Shapley distance decomposition of each realized allocation and the corresponding Shapley value, to compute the average distance, pooling data of all groups and all games, to compare across four treatments by regressing each one onto four treatment dummies (without constant).²⁵ The regression results are presented in Table C.29.

Column $||e^\phi||^2$ of Table C.29 shows that the Shapley distance is lower under semi-structured bargaining than under Winter or H-MC. As one can observe from Table C.30, the difference between the demand-based approach and Winter is statistically significant (at the 5% and 1% levels for with chat vs. Winter and without chat vs. Winter, respectively). Tables C.29 and C.30 further show that this significant difference of $||e^\phi||^2$ between the demand-based approach with chat and Winter is due to the significant difference in $||e^{sym}||^2$, $||e^{eff}||^2$, and $||e^{null}||^2$, and the difference between the demand-based approach without chat and Winter is due to the difference in $||e^{sym}||^2$ and $||e^{add}||^2$.

²⁵More specifically, we run the following regressions, $||e_i||^2 = \beta_1 ONC_i + \beta_2 OC_i + \beta_3 DNC_i + \beta_4 DC_i + \beta_5 Winter_i + \beta_6 H - MC_i + \mu_i$ where $||e_i||^2$ is the decomposed distance in group i , ONC_i , OC_i , DNC_i , DC_i , $Winter_i$, and $H - MC_i$ are dummy variables that take a value of one if the treatment is offer-based with no chat (ONC), offer-based with chat (OC), demand-based with no chat (DNC), demand-based with chat (DC), Winter, and H-MC, respectively, and zero otherwise. The Standard errors are corrected to account for within-session clustering effects. The statistical tests are based on the Wald test for the equality of the estimated coefficients of treatment dummies.

Table C.29: Result of Shapley distance decomposition. Based on pooling the data of all groups and all games

	$ e^{sym} ^2$	$ e^{eff} ^2$	$ e^{null} ^2$	$ e^{add} ^2$	$ e^{\phi} ^2$
Demand-Based	16.03	448.09	7.03	216.50	687.59
No chat	(5.18)	(124.53)	(1.34)	(19.85)	(142.08)
Demand-Based	10.53	241.08	19.45	281.07	552.09
With chat	(4.85)	(74.98)	(6.20)	(27.93)	(95.84)
Offer-Based	7.98	478.02	13.49	164.30	663.74
No chat	(1.74)	(72.65)	(2.67)	(11.45)	(63.12)
Offer-Based	0.07	363.13	25.64	211.55	600.37
With chat	(0.07)	(117.09)	(6.79)	(13.20)	(113.00)
Winter	85.18	606.81	7.28	321.49	1020.68
	(18.14)	(96.94)	(1.79)	(16.58)	(69.08)
H-MC	38.19	429.96	63.97	270.84	802.88
	(12.18)	(51.08)	(7.91)	(19.80)	(59.98)
No. Obs	1040	1040	1040	1040	1040
R^2	0.121	0.146	0.079	0.343	0.300
p-value*	0.0001	0.1660	0.0115	0.0025	0.759

Note: Standard errors are corrected for session-level clustering effects and shown in parentheses; $< e^{add}, e^{null} >$ are not reported in the table as they are negligible (the mean values are 0.0020, 0.0031, 0.0029, 0.0046, 0.0026, 0.0093 for demand-based no chat, demand-based with chat, offer-based no chat, offer-based with chat, Winter, and H-MC, respectively.).

*: The null hypothesis is the estimated coefficients of four treatment dummies (excluding Winter and H-MC) are the same (Wald test).

Other significant differences between semi-structured and structured experiments are $||e^{sym}||^2$, $||e^{null}||^2$, and $||e^{add}||^2$ between offer-based bargaining procedures (with and without chat) and H-MC. As Fig. C.1 shows, the average payoff of the null player was particularly high in H-MC Chessa et al. (2023b) conjectured that this was caused by proposers trying to avoid their proposal being rejected by the null player. Indeed, in the version of H-MC considered in Chessa et al. (2023b), it was not possible for the proposer to propose a coalition excluding the null player unless the null player has already been

Table C.30: p-values for pair-wise comparisons

$ e^{sym} ^2$		$ e^{eff} ^2$	
Test	p-value	Test	p-value
Demand, No Chat vs. Demand, With Chat	0.446	Demand, No Chat vs. Demand, With Chat	0.168
Offer, No Chat vs. Offer, With Chat	0.0001***	Offer, No Chat vs. Offer, With Chat	0.413
Demand, No Chat vs. Offer, No Chat	0.154	Demand, No Chat vs. Offer, No Chat	0.837
Demand, With Chat vs. Offer, With Chat	0.042**	Demand, With Chat vs. Offer, With Chat	0.389
Demand, No Chat vs. Winter	0.0013***	Demand, No Chat vs. Winter	0.325
Demand, With Chat vs. Winter	0.0006***	Demand, With Chat vs. Winter	0.0066***
Offer, No Chat vs. H-MC	0.0220**	Offer, No Chat vs. H-MC	0.594
Offer, With Chat vs. H-MC	0.0047***	Offer, With Chat vs. H-MC	0.606

$ e^{null} ^2$		$ e^{add} ^2$	
Test	p-value	Test	p-value
Demand, No Chat vs. Demand, With Chat	0.062**	Demand, No Chat vs. Demand, With Chat	0.072*
Offer, No Chat vs. Offer, With Chat	0.109	Offer, No Chat vs. Offer, With Chat	0.013**
Demand, No Chat vs. Offer, No Chat	0.041**	Demand, No Chat vs. Offer, No Chat	0.032**
Demand, With Chat vs. Offer, With Chat	0.508	Demand, With Chat vs. Offer, With Chat	0.034**
Demand, No Chat vs. Winter	0.910	Demand, No Chat vs. Winter	0.0005***
Demand, With Chat vs. Winter	0.072*	Demand, With Chat vs. Winter	0.226
Offer, No Chat vs. H-MC	0.0001***	Offer, No Chat vs. H-MC	0.0001***
Offer, With Chat vs. H-MC	0.0012***	Offer, With Chat vs. H-MC	0.0204**

$ e^{\phi} ^2$	
Test	p-value
Demand, No Chat vs. Demand, With Chat	0.437
Offer, No Chat vs. Offer, With Chat	0.629
Demand, No Chat vs. Offer, No Chat	0.879
Demand, With Chat vs. Offer, With Chat	0.748
Demand, No Chat vs. Winter	0.046**
Demand, With Chat vs. Winter	0.0006***
Offer, No Chat vs. H-MC	0.124
Offer, With Chat vs. H-MC	0.127

removed from the game. In the semi-structured offer-based bargaining procedures, it is possible for non-null players to propose such a coalition without fear of the null player rejecting it. The possibility of chat makes it more likely for the null player property to be violated under both the demand-based and the offer-based bargaining procedures, although the differences from the treatment without chat are significant only for the former.²⁶

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²⁶In the same vein, Kahan and Rapoport (1980b) demonstrated that the presence or absence of messages affects their discrepancy score index from the bargaining sets defined as the mean absolute deviation of each player’s payoff in a winning coalition from that player’s quota.

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