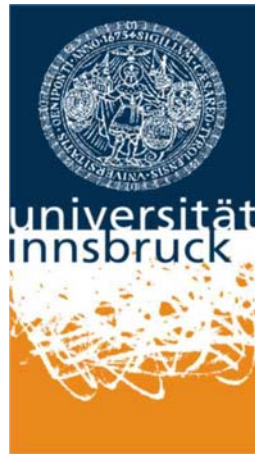


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**Strategic sophistication of individuals and teams
in experimental normal-form games**

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Strategic sophistication of individuals and teams in experimental normal-form games^{*}

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Abstract

We present an experiment on strategic thinking and behavior of individuals and teams in one-shot normal-form games. Besides making choices, decision makers have to state their first- and second-order beliefs. We find that teams play the Nash strategy significantly more often, and their choices are more often consistent by being a best reply to first order beliefs. We identify the complexity of a game and the payoffs in equilibrium as determining the likelihood of consistent behavior according to textbook rationality. Using a mixture model, the estimated probability to play strategically is 62% for teams, but only 40% for individuals.

JEL-classification: C72, C91, C92

Keywords: Strategic sophistication, beliefs, experiment, team decision making, individual decision making

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1. Introduction

Strategic sophistication refers to the extent to which players consider the structure of a game and the other players' incentives in the game before deciding on their strategy. There is a large literature that examines the strategic sophistication of individuals by means of experimental normal-form games (see, e.g., Stahl and Wilson, 1994, 1995; Haruvy, Stahl and Wilson, 1999; Costa-Gomes, Crawford and Broseta, 2001; Weizsäcker, 2003; Bhatt and Camerer, 2005; Crawford and Iriberri, 2007a; Costa-Gomes and Weizsäcker, 2008; Fehr, Kübler and Danz, 2008; Rey-Biel, 2009). In a nutshell, the main insight from this literature is the fact that strategic sophistication is often limited, such that many subjects ignore the incentives and the rationality of the other players (e.g., Costa-Gomes et al., 2001, Weizsäcker, 2003) or fail to best reply with their choices to their own stated beliefs (Costa-Gomes and Weizsäcker, 2008).

In this paper, we present an experiment in which we examine the strategic sophistication of teams (of three subjects each) and compare team decision making with individual decisions. Generally speaking, investigating team decision making seems interesting and important because many economically important decisions are taken by teams rather than by individuals. One can think of company boards, management teams, committees, or central bank boards as relevant economic agents in a variety of contexts where strategic sophistication plays a role, such as in decisions on market entry, technology races, company takeovers, or how to optimally intervene through monetary policy instruments in financial markets during a crisis. In addition to this genuine relevance of team decision making, our paper contributes to the literature on strategic sophistication and the literature on team decision-making in the following ways.

We let teams and individuals make choices in 18 different, one-shot normal-form games that have been designed by Costa-Gomes et al. (2001) to study the strategic sophistication of individuals. In addition to asking for choices of individuals and teams, we elicit their first order beliefs about their opponent's choice, and their second-order beliefs about the opponent's first order belief. We examine the differences between individual and team decision making by analyzing the following aspects of strategic sophistication: (i) the relative frequency of equilibrium play and beliefs about equilibrium play, (ii) the degree of a decision maker's consistency and the consistency expected from the opponent, where consistency means that choices are a best reply to beliefs, (iii) the determinants of standard textbook-rationality by which a decision maker plays the equilibrium strategy and expects the opponent to do so as well, and (iv) the distribution of eight different strategic and non-strategic types

which is estimated separately for teams and individuals by a mixture model in which each decision maker's type is drawn from a common prior distribution over the eight types and remains constant for all 18 games.

Our estimations on the determinants of textbook-rationality examine how the complexity of a game and how the distribution of payoffs in equilibrium affect the likelihood of a decision maker playing Nash and expecting the opponent to do the same. This adds to the literature on strategic sophistication by shedding more light on which factors of a game promote or hinder equilibrium choices and beliefs. A recent paper by Rey-Biel (2009) has investigated the influence of constant-sum games versus variable-sum games on the predictive power of Nash equilibrium, finding that the likelihood of observing Nash equilibrium choices is higher in constant-sum games than variable-sum games. We contribute to this issue by focusing on the role of the decision maker (being either an individual or a team). By considering not only first-order, but also second-order expectations we are also able to investigate how rational a decision maker expects his opponent to be, something which has not been explored in detail in the literature on strategic sophistication.

We also contribute to the flourishing research on team decision making. While the basic bottom-line of team decision making-research seems to be that teams are “more rational” than individuals in strategic games – meaning that team behavior is in the aggregate typically closer to standard game theoretic predictions than individual behavior (see, e.g., Bornstein and Yaniv, 1998; Bornstein, Kugler and Ziegelmeyer, 2004; Cooper and Kagel, 2005; Kocher and Sutter, 2005; Charness and Jackson, 2007)¹ – we are not aware of any research on team decision making that classifies single teams as of a particular strategic or non-strategic type and compares the distribution of different types across individuals and teams. The closest paper with respect to classifying teams as more or less strategic is by Cooper and Kagel (2005) who investigate individual and team decision-making in a signaling game where a market incumbent has to decide on a production quantity (as a signal for its type) before a potential market entrant decides on whether or not to enter the market. Cooper and Kagel (2005) find that teams are more frequently playing strategically by choosing higher production levels to signal a low-cost type market incumbent, thus making it less likely for outsiders to enter the market. While their study – in particular the excellent analysis of chat communication in the two-person teams – adds valuable insights about the strategic thinking

¹ The paper by Cason and Mui (1997) is often misinterpreted as showing that teams are more generous than individuals in a dictator game. However, Cason and Mui (1997) did not find that teams in general are more generous than individuals, but only reported more other-regarding team choices when team members differed in their individual dictator game choices.

of teams, it does not provide a detailed classification of types on the level of single teams as will be done in our paper. Furthermore, Cooper and Kagel (2005) have not elicited higher-order beliefs, leaving open the question if the different behavior of individuals and teams is due either to differences in choice behavior or to different beliefs which then influence choices. We will analyze the strategic sophistication of teams by considering not only choices, but also their first- and second-order beliefs. Eliciting beliefs will allow us to check whether teams are more likely to best reply to their own first-order beliefs, and whether teams expect their opponents to be best responding as well (by matching first-order beliefs with second-order beliefs). None of this has been studied systematically and in depth in the team decision-making literature before.² Our paper can therefore provide a fine-grained picture of the (bounded) rationality and strategic sophistication of teams and how this compares to individual decision making.

Based on a total of 192 experimental participants, we find that teams play Nash-strategies significantly more often, and their choices are more often a best reply to their first order beliefs. Hence, teams are more often consistent than individuals.³ Teams also expect their opponents to be more consistent, i.e. first-order beliefs are more often a best reply to second-order beliefs. Neither teams nor individuals are found to discriminate in their behavior and beliefs with respect to the type of decision-maker in the opponent's role. This is most likely due to our experimental setup which does not allow for learning.⁴ This finding implies that teams expect (mistakenly) their opponents to be more consistent in general, and not only in situations where a team faces another team in the opponent's role. Addressing the determinants of textbook-rationality (by playing and expecting equilibrium strategies) we find that the complexity of a game has a considerable impact on textbook-rationality, as has the distribution of payoffs in equilibrium. Using a mixture model, the estimated probability for teams to play strategically is 62%. The modal type of team decision making is a strategic *DI-*

² Second-order beliefs in normal-form games have been studied with individuals as decision makers by Bhatt and Camerer (2005) and Costa-Gomes and Weizsäcker (2008). The latter authors mention to have run some treatments where they had asked for second-order beliefs, but they do not report these treatments and their results in the paper. The role of second-order beliefs on actual behavior (of individuals) in extensive form games (such as a lost wallet-game or a trust game) is studied in Dufwenberg and Gneezy (2000) or Charness and Dufwenberg (2006), for instance.

³ The degree of consistency – by choices being a best reply to first-order beliefs – that we find for individuals is remarkably similar to the level reported in Costa-Gomes and Weizsäcker (2008) or found in the early phases of the experiment of Fehr et al. (2008) where they had feedback. The level of consistency reported in these studies and in ours is somewhat lower than in Rey-Biel's (2009) study, which might be caused by the fact that Rey-Biel (2009) also included games with weak dominance relationships (which should make consistent behavior more likely) and constant-sum games.

⁴ We are not interested in the issue of learning in team decision making here, since this has been studied elsewhere. The interested reader is referred to Kocher and Sutter (2005) or Feri, Irlenbusch and Sutter (forthcoming) on how teams learn (in beauty-contest games and coordination games, respectively).

type which applies one step of deleting strategies that are dominated by pure strategies and then plays best reply to a uniform prior over the opponent's remaining strategies. Individuals are classified as strategic significantly less often, i.e., in only 40% of cases. Their modal type is a non-strategic *optimistic* type (which chooses the strategy that maximizes the maximum possible payoff).

The rest of the paper is organized as follows. In section 2 we present the experimental design, covering the normal-form games used in the experiment, the experimental treatments and procedures. Section 3 reports the experimental results. It starts with an aggregate analysis of choices, first- and second-order beliefs of individuals and teams. Then we examine the consistency of players' choices with own first-order beliefs, and the expected consistency of their opponents by matching players' first- and second-order beliefs. We continue with a probit regression that determines the factors that lead players to play the Nash strategy and to expect their opponent to play Nash as well. In the final subsection we use a maximum-likelihood error-rate model to classify individuals and teams into either of eight different strategic and non-strategic types. Section 4 discusses our results and concludes the paper.

2. Experimental design

2.1. The 18 normal-form games

Figure 1 presents the 18 normal-form games that we used in our experiment. They are taken from Costa-Gomes et al. (2001), because these games are very well designed to study strategic thinking and behavior. Out of the 18 games, 16 games are pairs of isomorphic games (which are identical for row and column players except for transformation of player roles and small, uniform payoff shifts). We applied the same order of games as in Costa-Gomes et al. (2001). This is indicated in Figure 1 as “game # x ”, with $x \in \{1, \dots, 18\}$. In brackets after the game number we refer to two different types of games. “**D**”-games (“**ND**”-games) are those in which one (none) of the two players has a strictly dominant strategy. In total, there are 10 **D**-games and 8 **ND**-games. Note that in all games the unique, pure strategy Nash equilibrium is Pareto-dominated by another strategy combination of row and column players.

Figure 1 about here

The number of rows (columns) indicates the number of available strategies for row (column) players. The 18 games have different levels of complexity, interpreted as the required number of rounds of iterated pure-strategy dominance a row or column player needs to identify the own equilibrium choice. The notation $[xR, yR]$ refers to the number of rounds needed for the row (x) and column player (y). In **D**-games, we have $x, y \in \{1, 2\}$, while in **ND**-games $x, y \in \{2, 3, \infty\}$, where $[\infty R, \infty R]$ denotes non-dominance solvable games.

2.2. Treatments, decisions and payments

We have implemented three different treatments which are distinguished by which types of decision makers are interacting with each other in the normal-form games. Note that the treatment – and thus the type of decision maker in the opponent’s role – was common knowledge in the experiment (see the experimental instructions in the Supplementary Material).

- **Ind**: In the individual treatment both row and column players were individuals.
- **Team**: In the team treatment both row and column players were teams of three subjects each. Team members sat together in a sound-proof cabin and could communicate with each other in order to reach a team-decision. There were no restrictions on communication and no regulations how a team should reach an agreement, except that only one decision could be entered by a team on the computer.
- **Mixed**: In the mixed treatment a team of three subjects was interacting with an individual. In half of the **Mixed**-sessions the team was in the role of the row-player and the individual in the role of the column-player, while the reverse order applied in the other half of sessions.

The **Mixed** treatment has been included because it allows examining whether individuals and teams condition their choices and beliefs on the type of decision maker (individual or team) in the opponent’s role. To avoid any influence of the kind of presentation on the data, we presented all games to all players in a way that they saw themselves as a row player. In each single game each player (either individual or team) had to make three different decisions:

- **Choice**: This was a player’s choice of a row in a given game.

- **First-order belief (FOB):** Each player had to indicate the action expected from the opponent. From the viewpoint of the player this meant to select a column as the expected choice of the opponent.
- **Second-order belief (SOB):** Players stated a belief about their opponent's first-order belief. This meant to indicate a row that matches the player's belief about the opponent's expectation of the player's choice.

Each player had to make all three decisions in each of the 18 games, yielding a total of 54 decisions to be taken during the experiment. We proceeded game by game with all three decisions, but the order of the three decisions for a given game was randomly determined.⁵ In order to suppress learning, partners changed after each game (random matching) and there was no feedback until the experiment was finished. All details of this procedure were common knowledge to participants.

In order to keep the per-capita incentives constant for individuals and teams the payoffs described in the following were paid to each member of a team, and participants knew this. At the end of the experiment each player needed to draw a card from a deck of cards which were showing numbers from 1 to 18. This card determined the game that was relevant for payment. Players then received full information about their own and their opponent's choices and beliefs in this game. After that players were asked to draw another card from a deck of cards showing numbers 1 (for choice), 2 (for first-order belief), or 3 (for second-order belief). If a player was paid for her choice, the player received 30 Euro Cents for each experimental point earned in the game as a consequence from the own and the opponent's choice (as shown in Figure 1). Players who were paid for their first-order or second-order belief received 15 Euros if their belief was correct, but nothing if it was wrong. The all-or-nothing feature for paying beliefs was intended to make the decisions on first- and second-order beliefs as salient as possible. Our approach with respect to eliciting and paying beliefs was also chosen for practical reasons, though. Given the relatively high number of decisions (54) to be taken in the experiment we opted for eliciting point beliefs rather than asking for a probability distribution of beliefs over all available strategies. In particular asking teams for a probability distribution would have been very time-consuming and would probably have exhausted

⁵ Note that Costa-Gomes and Weizsäcker (2008) do not find order effects depending on whether to elicit beliefs before choices or the other way round.

participants' attention and patience in an experiment that already took on average 2 hours with our procedure.⁶

2.3. Experimental procedure

In each treatment we conducted four sessions that were run in the video-lab of the Max Planck Institute of Economics in Jena.⁷ This lab has eight sound-proof cabins which are connected via a computer network (using z-Tree by Fischbacher, 2007). Hence, we could invite eight decision makers for each session. Four decision makers were row players and four decision makers were column players. Depending upon the treatment, a decision maker was either an individual subject or a team of three subjects. When recruiting subjects with the help of ORSEE (Greiner, 2004), assignment to treatments was randomly determined. In the treatments with team decision making, subjects were randomly assigned to a particular team at the beginning of the experiment. Teams stayed together for the entire experiment. The matching of row and column players was random and changed after each game. Note that for a particular game (with its three decisions) the matching was kept constant. To make sure that participants understood the instructions and the experimental tasks, they needed to answer individually a questionnaire before the experiment in which they were tested how choices and first- and second-order beliefs mapped into outcomes and payoffs. All participants were able to answer all questions correctly.

In sum, we had a total of 192 participants, recruited from undergraduate students of the University of Jena. For each game in each treatment we obtained 32 different observations. No subject was allowed to participate in more than one session. On average, a session lasted 2 hours and subjects earned 16.5 Euro, including a show-up fee of 4 Euro.

⁶ Haruvy (2002), Costa-Gomes and Weizsäcker (2008) or Rey-Biel (2009) elicited (more informative) probability distributions of the opponent's choices (i.e., a player's first-order beliefs). While in the study of Haruvy (2002) the computer then calculated the best response for the player's first order belief and implemented the best response automatically as the choice of the player, the subjects in the studies of Costa-Gomes and Weizsäcker (2008) and Rey-Biel (2009) could make their choices and state their first-order beliefs independently. Our approach of asking for a single strategy as a belief and rewarding beliefs if they are exactly right induces subjects to report the mode of their (implicit) distribution of probabilities for the different available strategies.

⁷ Participants were not videotaped and they were told so. Individual decision makers were also seated alone in soundproof cabins in order to keep the conditions of individual and team decision making as identical as possible. We decided not to videotape team discussions because we wanted to minimize the possibility that differences between individuals and teams might be driven by the fact that teams were taped and thus observed.

3. Experimental results

3.1. Pooling of data

In the following analysis we pool the data of row and column players, because 16 out of 18 games are isomorphic games which are practically identical for row and column players expect for small payoff shifts and the transformation of player roles (see Costa-Gomes et al., 2001, for details). All analyses presented below are based on all 18 games.

We have also examined whether players' choices and beliefs depend on the type of decision maker they face as their opponent. In other words, we tested whether the distribution of choices and first- and second-order beliefs differs for individuals between treatments **Ind** and **Mixed**, respectively for teams between treatments **Team** and **Mixed**. In none of these distribution tests we find any significant difference. This result indicates that decision makers do not condition their behavior and their beliefs on the type of decision maker in the opponent role. This is most likely due to giving no feedback until the end of the experiment, which prevents teams or individuals to learn about the decision patterns of their opponents and thus calibrate their decisions on the experienced decision patterns of their opponents. We summarize as our first result:

Result 1: Neither individuals nor teams condition their choices and beliefs on the type of decision maker (individual or team) in the opponent role.

This result allows us to pool individual decisions from the treatments **Ind** and **Mixed**, as well as the team decisions taken in either **Team** or **Mixed**. Hence, when we talk of individual (team) decisions in the following, we refer to all decisions made by individuals (teams) in the different treatments. Whether individuals and teams make different choices and state different beliefs (independent of the type of opponent they are facing) is subject of the following analysis.

3.2. Choices, first- and second-order beliefs of individuals and teams

In Table 1 we report the relative frequency of different strategies – chosen either by teams or individuals – across all games, and separately for **D**-games and **ND**-games.⁸ We consider the strategy that leads to the Nash-equilibrium (“Nash”), the strategy that could yield an outcome that Pareto-dominates the Nash-equilibrium (“Pareto”), and other strategies

⁸ Tables S1 to S6 in the supplementary material present the relative frequencies of each single strategy in each single game, both for individuals and teams, and for choices, first- and second-order beliefs.

(“Other”). The upper panel reports relative frequencies for own choices, the middle and lower panel for first-order, respectively second-order, beliefs.

Table 1 about here

We compare the relative frequencies of the different strategies chosen by individuals and teams by considering for each single decision maker the relative frequency of choosing a particular strategy across all games (or across *D*- and *ND*-games separately) and then apply a two-sided Mann-Whitney U-test to the resulting data. We see from the upper panel of Table 1 that teams are significantly more often playing the Nash-strategy than individuals (49.30% vs. 40.97%; $p < 0.05$).⁹ Individuals choose more often what we call the “Pareto”-strategy of trying to improve on the Nash-equilibrium choice (53.47% vs. 45.72%; $p < 0.05$). These differences between individuals and teams are mainly driven by the 10 games with a dominant strategy. Only in *D*-games teams are significantly more often playing Nash, and less often playing Pareto, while there is no significant difference between individuals and teams in the games where neither player has a dominant strategy. This is a first indication that the complexity of a game has an impact on how the type of decision maker influences behavior. Table 2 below will shed more light on this issue.

The middle panel in Table 1 considers first-order beliefs. Although teams expect their opponents to play Nash more often, respectively Pareto less often, than individuals, none of these differences is significant according to conventional levels. Only when considering second-order beliefs (in the lower panel) we find for the set of *D*-games again the same type of differences between individuals and teams as we do for choice-data.¹⁰ Actually, teams think that their opponents expect them to play Nash more often, respectively Pareto less often, than individual decision makers think about their opponents’ expectations.

⁹ If players chose their strategies randomly, then the expected relative frequency of observing Nash equilibrium choices would be 43%. In Figure F1 of the supplementary material we show that players’ choices of both individuals and teams are significantly different from random, though. In this Figure F1 we plot the relative frequency of playing Nash in x out of 18 games (with x ranging on the horizontal axis from zero to 18) and compare the distributions of actually observed relative frequencies and theoretically expected ones in case of random play. The upper row of Figure F1 plots choices, first- and second-order beliefs for individuals, and the lower row for teams. In all six cases we find that the actually observed distribution is significantly different from a random one (Kolmogorov-Smirnov one sample test, $p < 0.05$ in all cases).

¹⁰ In the supplementary material we provide in Tables S7 to S10 the relative frequency with which individuals and teams were correct in their first- and second-order beliefs. On average, teams were marginally more often correct in their beliefs (62.7% vs. 61.0% for first-order beliefs; 69.1% vs. 68.6% for second-order beliefs), but the difference is far from being significant.

Comparing across panels in Table 1, while holding the type of decision maker constant, we find that the relative frequency of playing Nash decreases from top to bottom, i.e. is highest for own choices, intermediate for first-order beliefs and lowest for second-order beliefs. This holds true both for individuals and teams and the descending order is significant for both types of decision makers ($p < 0.05$; Page tests). Hence, equilibrium play is less likely according to beliefs than according to own choices and less likely the higher the order of beliefs.¹¹ The reverse pattern applies for (potentially) Pareto improving strategies which increase from choices to first-order beliefs, and from first-order to second-order beliefs.

Table 2 about here

In Table 2 we examine the influence of a game's complexity on choices and beliefs in more detail. We present the relative frequencies of chosen strategies separately for games with and without a dominant strategy (**D**- vs. **ND**-games) and for each type of game we report data contingent on the number of rounds of iterated pure-strategy dominance a player needs to identify the own equilibrium choice. The first column shows that in games with a dominant strategy those players with the dominant strategy play Nash in about 80% of cases. There is no significant difference between individuals and teams, neither for choices nor for first- or second-order beliefs.

However, the second column reveals a very strong difference between individuals and teams when they are in the role of the player without the dominant strategy (when the opponent has one). While teams choose the Nash-strategy in almost 60% of cases then, individuals play Nash in less than 40%. The reverse holds true for the Pareto strategy. Given the high probability of players with a dominant strategy who actually choose the dominant strategy (see first column), the behavior of teams in column 2R(D) can be judged as more rational, since playing what we call the Pareto strategy when the opponent plays Nash would lead to lower payoffs than playing Nash. These differences in choices between individuals and teams (when they need two rounds of iterated pure-strategy dominance to identify the own equilibrium choice) is again mirrored in second-order beliefs. Teams also think that their opponents expect them to play Nash almost double as often as individuals think of their opponents (25.42% vs. 13.75%). The differences in first-order beliefs are qualitatively similar, but fail significance at conventional levels.

¹¹ This pattern is consistent with a model of noisy introspection by Goeree and Holt (2004) in which they predict more noise (and less equilibrium play) the higher the order of beliefs.

The third column in Table 2 considers games without a dominant strategy for either of the players, and it clearly confirms the findings from the second column. When two rounds of iterated dominance are needed, teams play significantly more often Nash – and less often Pareto – than individuals. The same pattern is found for first- and second-order beliefs (with the differences being weakly significant for first-order beliefs). The final two columns in Table 2 show that for more complex situations (with more than two rounds of iterated dominance) there are no significant differences in strategy choices between individuals and teams. Thus, the evidence from Table 2 suggests that for very straightforward decisions (when a player has a dominant strategy) and for more complex decisions (with more than 2 rounds of iterated dominance) there are no differences between individuals and teams. Yet, for decisions with 2 rounds of iterated dominance teams choose (and think that they are expected to choose) more often the equilibrium-strategy.

Table 3 about here

The pattern observed in Table 2 is reflected from a different perspective in Table 3 which reports the relative frequency of equilibrium play in the different types of games, classified in Table 3 by the available number of strategies for row and column players and the presence or absence of a dominant strategy for a player or its opponent. Table 3 reveals the most prominent differences between individuals and teams in the games where the opponent has a dominant strategy. We summarize the findings from Table 2 and Table 3 in our second result:

Result 2: Overall, teams play the equilibrium strategy significantly more often than individuals. Taking a game's complexity into account we find that individuals and teams do not differ in the likelihood of playing equilibrium when they have a dominant strategy or need more than two steps of iterated dominance. In situations where two steps are required, though, teams play equilibrium much more often. These differences in choice-patterns of individuals and teams are also reflected in their beliefs, in particular in second-order beliefs.

3.3. Consistency of choices and beliefs

We classify a player's choice as consistent with beliefs if the choice is a best reply to the player's own first-order belief. Likewise, we define the opponent's expected consistency as to whether the first-order belief is a best reply to a player's second-order belief. The latter determines basically a player's first-order belief about the opponent's consistency.

Table 4 about here

Table 4 reports in the upper part the players' own consistency and in the lower part the expected consistency of the opponent player.¹² Consistent with earlier findings by Costa-Gomes and Weizsäcker (2008) or Fehr et al. (2008) we find that only about one half of the individual decisions (55.79%) are a best-reply to the individual's own first-order (point) belief.¹³ However, we find that teams play best-reply significantly more often (65.86%; $p < 0.05$, two-sided Mann-Whitney U-test). Again we see from the column “*D*-games” that this difference is predominantly driven by the games where one party has a dominant strategy (and where the main difference is again due to the player who does *not* have the dominant strategy). The lower part of Table 4 indicates that teams also expect their opponents to be more often consistent by playing best reply. However, this difference is not significant. We summarize these results in

Result 3: Teams make more often consistent choices than individuals, such that the choices of a team are more frequently a best reply to their first-order beliefs about the opponent's behavior. The level of consistency observed for individuals confirms earlier experimental findings. Hence, team decision making is a means for inducing more consistent behavior.

3.4. A closer look at the consistency of choices and beliefs

In this subsection we examine the consistency of choices and beliefs in more detail by having a look at the two most frequent types of consistency. One type is a straightforward combination of playing Nash and expecting the opponent to play Nash as well. We call this the “Nash-consistency” (Nash-CON). The other type is when the combination of a player's choice and first-order belief would yield the maximal payoff that is available for the player in a specific game. We call this the “Maximax-consistency” (Max-CON). Note that Max-CON

¹² In Figure F2 of the supplementary material we plot the theoretically expected relative frequency of being consistent in x out of 18 games under the assumption of random play (with x ranging from zero to 18 on the horizontal axis) versus the actually observed relative frequency of being consistent. We note a clear rightward shift – in particular for teams – of the actually observed distribution, clearly rejecting the null-hypothesis of randomness in consistency ($p < 0.01$ for individuals and teams, both for own consistency and the opponent's expected consistency; Kolmogorov-Smirnov one-sample tests).

¹³ Fehr et al. (2008) show that the relative frequency of best responses to own beliefs increases with repetition *if* feedback is given after each game. However, in their study it also starts out with slightly more than 50% in the first round – which is comparable to the level we observe without feedback – and converges to only around 75% in the final round of their experiment.

cannot coincide with the Nash-equilibrium in our games. The relative frequency of these two consistency types can be examined for a player himself, but also for the player's expectation about the opponent's consistency type.

Table 5 about here

Table 5 reports the relative frequency of Nash-CON and Max-CON. For the player's own consistency (see upper part) we find a very strong difference between individuals and teams with respect to Nash-CON. Again, the difference is mainly driven by our **D**-games. While only 27% of individual choices and beliefs can be classified as Nash-CON, it is 45% of team choices and beliefs in the 10 **D**-games where one player has a dominant strategy ($p < 0.05$, two-sided Mann-Whitney U-test). Interestingly, for individuals it is even the case that Max-CON is (marginally) more frequent than Nash-CON, while for teams the ratio is around 2:3. From the lower part of Table 5 we note that teams expect their opponents significantly more often to be of the Nash-CON-type than individuals, while individuals expect opponents more often to be of the Max-CON-type.¹⁴

Result 4: Teams play the Nash-strategy with the expectation that the opponent also plays Nash much more often than individuals do. Individuals' choices and beliefs more often focus (optimistically) on the maximum available payoff in a given game. Teams also expect their opponents more often to play and expect Nash (first- and second-order beliefs) than individuals do.

3.5. The determinants of consistency-type Nash-CON

As a final step in analyzing the consistency of decisions we examine the determinants that increase the likelihood of the "Nash-consistency". Note that this is precisely the type of consistency standard game theory would predict for rational and payoff-maximizing decision makers. Hence, estimating the factors that make "Nash-consistency" more likely does not only reveal which factors promote this type of consistency, but also which ones prevent decision makers from playing the equilibrium strategy and expecting their opponent to do the same. In Table 6 we report a probit estimation of a player's Nash-CON on various factors that

¹⁴ In Figure F3 of the supplementary material we compare the relative frequency of Nash-consistency if players' choices were random to the actually observed relative frequencies. Again, the actually observed distributions are significantly different from random both for individuals and teams as well as for the own Nash-consistency and the opponent's expected Nash-consistency ($p < 0.01$; Kolmogorov-Smirnov one-sample tests).

are explained in the following. Since each decision maker had to make decisions in 18 different games, we cluster the standard errors on the decision maker.

Table 6 about here

Panel [A] of Table 6 shows that the consistency type Nash-CON is more likely if players have a dominant strategy themselves (“dominant strategy”) or if their opponent has one (“opponent dominant strategy”). The next three independent variables are all insignificant. Making decisions in a team (“team player”), having a team as opponent player (“opponent = team”) or playing as a team against another team (“team * opponent = team”) does not influence the likelihood of “Nash-consistency”. The latter two terms confirm that the type of decision maker in the opponent role does not matter for decision making. However, the interaction of team decision making and having a dominant strategy (“team * dominant strategy”) or the opponent having a dominant strategy (“team * opponent dominant strategy”) has a significantly positive effect each. This is in line with our previous findings that decisions in the *D*-games mainly drive the differences between individuals and teams. The latter two independent variables show that team decision making in games where a dominant strategy is available increases the likelihood of decisions and beliefs to be in accordance with standard game theoretic predictions.

The final four variables in panel [A] of Table 6 consider the magnitude and distribution of payoffs in a game in different ways. The variable “risk with maxstrategy” measures the absolute difference between the decision maker’s payoff in the Nash-equilibrium and the decision maker’s payoff in case he chose the strategy in which the game’s highest payoff is possible while the opponent played Nash. In other words, this variable indicates how much money is at risk when a decision maker tries (unsuccessfully) to reach his maximal possible payoff instead of playing Nash. The estimation shows that Nash-CON gets weakly significantly more likely the more money is at risk if a player wants to deviate to the strategy that could maximize his payoff.¹⁵ The variable “risk with maxstrategy opponent” is constructed like “risk with maxstrategy”, but measures the opponent’s potential losses from

¹⁵ The estimation presented in Table 6 remains qualitatively identical if we used the relative – rather than the absolute – losses from deviating from the Nash strategy. In fact, “risk with maxstrategy” becomes significant even at the 5%-level, while all other significant variables in Table 6 keep their significance level with the alternative specification. We have opted for absolute losses since this fits better to the final two independent variables.

deviating from Nash. This variable has no significant influence on a player's likelihood to be Nash-consistent.

The final two variables examine the influence of inequality aversion in the Nash equilibrium on the consistency of decisions. Since subjects have been shown to respond differently to advantageous and disadvantageous inequality (Fehr and Schmidt, 1999), we distinguish between “advantageous inequality in Nash” (defined as $\max \{ \text{own payoff} - \text{opponent's payoff in Nash}, 0 \}$) and “disadvantageous inequality in Nash” (defined as $\max \{ \text{opponent's payoff} - \text{own payoff in Nash}, 0 \}$). Table 6 shows that advantageous inequality does not influence a player's likelihood to be Nash-CON. However, if disadvantageous inequality increases in the Nash equilibrium, a player is less likely of the Nash-CON-type. Hence, distributional preferences such as inequality aversion have a significant influence on consistent behavior in our experiment. It is important to note that interacting any of the final four variables in Table 6 with the variable “team player” does not yield any significant effects. Hence, payoff distributions affect the Nash-consistency of individuals and teams to the same extent.

Panel [B] of Table 6 reports the marginal effects of team decision making (compared to individual decision making) for specific combinations of the binary variables from panel [A], evaluated at the means of the four ultimate variables in panel [A]. If a decision maker has no dominant strategy, while the opponent has one, the likelihood of observing Nash-consistency is more than 20 percentage points higher when teams instead of individuals make choices and state their first-order beliefs. If a decision maker has a dominant strategy, while the opponent doesn't, then teams are also more likely to be Nash-consistent, however the effect fails significance.

Table 7 about here

Table 7 reports a similar probit regression as Table 6, but takes as the dependent variable a player's expectation about the opponent being a Nash-CON-type. The variables used in Table 7 are defined from the point of view of the player who states first- and second-order beliefs. Basically, we find very similar results as in Table 6, with a few noteworthy exceptions though. Most strikingly, “opponent dominant strategy” is not significant in itself. However, “teams * opponent dominant strategy” is significantly positive. Taken together, this shows that individuals fail to recognize that if an opponent has a dominant strategy then the opponent is more likely to be Nash-consistent (as it has become clear from Table 6). Teams,

however, expect (correctly) opponents who have a dominant strategy to be more often Nash-consistent. These implications are fully in line with our previous findings on the differences between individuals and teams when the degree of complexity is intermediate (i.e., requiring two steps of reasoning). Finally, both “advantageous inequality in Nash” and “disadvantageous inequality in Nash” are significantly negative in Table 7, indicating that players expect their opponents to be less often consistent even when the Nash-equilibrium would be to their (i.e., the opponent’s) advantage – which is something which players themselves do not consider as far as their own consistency is concerned (see Table 6). It is noteworthy that the latter four variables again show no significant interaction effect with the decision maker being either a team or an individual. Panel [B] of Table 7 reports the marginal effects of team decision making (compared to individual decision making) on the the likelihood of expecting the opponent to be Nash-consistent. The results are very similar to those discussed above for Table 6 and hence we dispense with a detailed discussion. We summarize the findings in this subsection as follows:

Result 5: Teams are more likely than individuals to be a Nash-consistency type – and expect their opponents to be of this type also more often – when the game has a dominant strategy. Inequality in payoffs in the Nash equilibrium makes consistent decisions (of choices and first-order beliefs, respectively of first- and second-order beliefs) less likely. Hence, distributional preferences affect the degree of standard game-theoretic rationality.

3.6. Estimation of strategic and non-strategic types

In the final subsection we present a maximum likelihood error-rate analysis of players’ choices following the econometric model used in Costa-Gomes et al (2001). This econometric model is a mixture model in which each player’s type is drawn from a common prior distribution over eight types and remains constant for all 18 games. The eight types that we consider can be classified into non-strategic and strategic types and they are defined as follows:¹⁶

Non-strategic types: (1) An *altruistic* type tries to maximize the sum of payoffs to himself and the opponent, implicitly assuming that the opponent is also *altruistic*.¹⁷ (2) A

¹⁶ We follow Costa-Gomes et al. (2001) in the selection of types to be considered. As they indicate, the definition of types is largely based on earlier work by Stahl and Wilson (1994, 1995).

¹⁷ In keeping with the literature (see Costa-Gomes et al., 2001) we denote this type *altruistic*, although *efficiency-loving* would probably be a more appropriate term. Note that efficiency-loving has been identified to

pessimistic type chooses the strategy that secures the best of all worst outcomes. Hence, a *pessimistic* type plays maximin. (3) An *optimistic* type chooses the strategy that maximizes the maximum possible payoff, thus ignoring the incentives of the opponent player.¹⁸

Strategic types: (4) Type *L2* plays best response to *optimistic (naïve)* types. (5) Type *D1* applies one step of deleting strategies that are dominated by pure strategies and then plays best reply to a uniform prior over the opponent's remaining strategies. (6) Type *D2* applies two steps of deleting dominated strategies and then best responds to the opponent's remaining strategies. (7) An *equilibrium* type makes equilibrium choices (which are unique in our games). (8) A *sophisticated* type plays best reply to the probability distribution of opponents' strategies by taking the actually observed distribution of strategies in the experiment's subject pool as the estimated probability distribution.¹⁹

For the estimation of the mixture model let $i = 1, \dots, N$ index the different players (individuals or teams), let $k = 1, \dots, K$ index our types, and let $c = 2, 3, \text{ or } 4$ be the number of a player's possible decisions in a given game. We assume that a type- k player normally makes a type k decision, but in each game he makes an error with probability $\varepsilon_k \in [0,1]$, type k 's *error rate*, in which case he makes each of his c decisions with probability $1/c$. For a type- k player, the probability of a type k decision is then $1 - \frac{(c-1)}{c}\varepsilon_k$. Hence, the probability of any single non-type k decision is $\frac{\varepsilon_k}{c}$. We assume errors are independently and identically distributed across games and players.

The likelihood function can be constructed as follows. Let T^c denote the total number of games in which players have c decisions; in our design we have $T^2 = 11$, $T^3 = 6$, and $T^4 = 1$. Then let x_k^{ic} denote the number of player i 's decisions that equal type k 's in games in which he has c decisions with $x_k^i = (x_k^{i2}, x_k^{i3}, x_k^{i4})$, $x^i = (x_1^i, \dots, x_K^i)$, and $x = (x^1, \dots, x^N)$. Let p_k denote players' common prior k -type probability, and $\sum_{k=1}^K p_k = 1$ and $p = (p_1, \dots, p_K)$. Let

be important in shaping subjects' behavior in non-strategic allocation tasks (Charness and Rabin, 2002). Hence, it seems interesting to study how efficiency-loving influences behavior in a strategic game.

¹⁸ As noted by Costa-Gomes et al. (2001), it is impossible to distinguish an *optimistic* type from a *naïve* type in the 18 games. A *naïve* type assigns equal probabilities to the opponent's strategies and best responds to this naïve belief. While a *naïve* type might reflect strategic decision making with diffuse beliefs, Costa-Gomes et al. (2001) describe *naïve* types as non-strategic. We follow their approach, but talk about *optimistic* types, which are non-strategic for sure.

¹⁹ The best responses in each game for each type are indicated in Figure F4 in the supplementary material. Note that the sophisticated type may have different strategies as best response, dependent on whether a player plays against an individual (denoted as S^I in Figure F4) or against a team (S^T). Types *D1* and *D2* may yield the same prediction, which is indicated as *D* in Figure F4.

ε_k denote the k -type error rate and $\varepsilon = (\varepsilon_1, \dots, \varepsilon_K)$. Given that a game has one type- k decision and $c-1$ non-type- k decisions, the probability of observing a particular sample with x_k^i type- k decisions when player i is type k can be written as:

$$L_k^i(\varepsilon_k | x_k^i) = \prod_{c=2,3,4} \left[1 - \frac{(1-c)}{c} \varepsilon_k \right]^{x_k^i} \left[\frac{\varepsilon_k}{c} \right]^{T^c - x_k^i} \quad (1)$$

Weighting the right-hand side by p_k , summing over k , taking logarithms, and summing over i yields the log-likelihood function for the entire sample:

$$\ln L(p, \varepsilon | x) = \sum_{i=1}^n \ln \sum_{k=1}^K p_k L_k^i(\varepsilon_k | x_k^i) \quad (2)$$

We provide two separate estimations of equation (2), one with individuals as decision makers and one with teams. Table 8 presents the main estimation results for individual choices (left-hand side) and team choices (right-hand side).²⁰

Table 8 about here

In the analysis of the results we interpret the estimated probability p_k as the probability to find a player of type k in the population under observation. From the bottom row of Table 8 we can see that individuals are estimated as strategic types in only 40% of cases, while 62% of teams are classified under any of the strategic types, and the difference is significant.²¹ There are three types with a significantly positive probability both for individuals and teams, and a fourth type which is only significant for teams. The three significant types common to both types of decision makers are the non-strategic types *altruistic* and *optimistic*, and the strategic type *DI*. While the probability to find an *altruistic* type does not differ significantly

²⁰ We have also run estimations with five types only where types (4), (5) and (6) introduced above have been dropped (like Costa-Gomes et al., 2001, did in their initial approach). The results with respect to the estimated frequencies of strategic, respectively non-strategic, play remain practically the same as those presented in Table 8. First- and second-order beliefs are noisier than choices, which makes estimating standard errors in some cases impossible when we use eight types. With five types, we can show that first-order and second-order beliefs are more often strategic for teams than for individuals. Additional results are available upon request.

²¹ At first sight it might seem surprising why the *equilibrium* type is so rare, while Table 1 has shown about 50% of equilibrium play by teams across all games. For the classification of types it is important what a decision maker chooses when, for example, *DI* and *equilibrium* do not suggest the same choice (as is the case in 6 of our games). In these games, the large majority of teams opt for the *DI*-strategy rather than the *equilibrium*-choice.

between individuals and teams, there are clear differences concerning the other two types. Individuals are significantly more often of the non-strategic *optimistic* type than teams. Teams, however, are significantly more likely to be of the strategic *D1*-type. The fourth (weakly) significant type – which applies only to teams – is the strategic type *L2*. We summarize the findings in this subsection in our final result.

Result 6: The likelihood of teams to be of any of five different strategic types (equilibrium, sophisticated, D1, D2, L2) is about 62%, while it is only 40% for individuals. The latter are predominantly acting non-strategically (as altruistic, pessimistic, or in particular optimistic types).

4. Conclusion

In this paper we have studied the strategic sophistication of individuals and teams in 18 different normal-form games taken from Costa-Gomes et al. (2001). We have found that teams are more likely than individuals to play these games strategically – meaning that teams consider more often the structure of the game and the incentives of their opponents when making decisions. We have estimated the likelihood of strategic play for teams to be 62%, while it is only 40% for individuals. The modal type of team decision making is a strategic *D1*-type which applies one step of deleting strategies that are dominated by pure strategies and then plays best reply to a uniform prior over the opponent's remaining strategies. On the contrary, the modal type of individual decision making is a non-strategic *optimistic* type which picks the strategy that maximizes the maximum possible payoff.

A detailed analysis of individual and team behavior has shown that teams play the Nash equilibrium significantly more often than individuals. Interestingly, this difference between individuals and teams is mainly driven by games with an intermediate degree of complexity. When games are very straightforward (when the decision maker has a dominant strategy) or when they are relatively complex (by requiring three or more steps of iterated dominance to figure out the equilibrium choice) there is no significant difference between individuals and teams.

Matching choices and first-order beliefs we have found that teams are more likely to play a best reply with their choices to their own first-order beliefs. While this kind of consistency of choices and beliefs is only found in 56% of individual decisions, it prevails in 66% of team decisions. It seems noteworthy that the relative frequency of consistency of individual decisions is very similar to the one reported in a recent study by Costa-Gomes and

Weizsäcker (2008). Our findings, however, indicate that this relatively low level of consistency can be significantly improved through team decision making. This is our first main contribution to the flourishing literature on strategic sophistication.

Another important feature of our paper is our consideration of how first- and second-order beliefs are related. This yields insights into how individuals or teams perceive the consistency of their opponents. Similar to the findings for the relation between choices and first-order beliefs, we find also a difference between individuals and teams with respect to how consistent are first- and second-order beliefs. Teams expect their opponents – irrespective of whether this is an individual or a team – to play and expect the Nash strategy more often than individuals expect that from their opponents. Obviously, this insight contributes to a deeper understanding of why individuals and teams differ already with respect to how own choices relate to first-order beliefs. If a decision maker expects his opponent to be someone who is likely to play optimally by best responding to his beliefs – and in particular to play the Nash equilibrium and expect it from the other player – then such a decision maker has a stronger incentive to play Nash (and expect it from the opponent) than a decision maker who perceives the opponent as less consistent.

Our paper also adds to the literature on strategic sophistication through our analysis of the determinants of “Nash-consistency”, which we consider a novel approach in this literature. The term Nash-consistency refers to a situation where a decision maker plays Nash *and* expects the opponent to play Nash. While our analysis has shown that team decision making – in situations where a dominant strategy for any of the players is available – makes Nash-consistency more likely, our analysis also sheds light on the reasons for the surprisingly low level of Nash-consistency in the first place. In particular, we find that Nash-consistency also depends on the potential monetary losses if a player deviates from Nash and on the distribution of payoffs in the Nash equilibrium. If payoffs in the equilibrium are too much different then consistency between choices and first-order beliefs becomes less likely. It is easily conceivable, for instance, that inequality averse subjects (Fehr and Schmidt, 1999; Bolton and Ockenfels, 2000) might expect their opponent to play the equilibrium strategy (which can be a rational expectation), but nevertheless refrain from their equilibrium choice for reasons of payoff asymmetry. Our analysis has also revealed an influence of what we have called the risk of not picking the equilibrium choice. If there is more money at risk when a decision maker tries to reach his maximal possible payoff instead of playing Nash, then Nash-consistency becomes more likely to be observed. It seems noteworthy that the effects of distributional preferences and the money at risk when deviating from Nash are important both

for individual and team decision making. Accordingly, our findings add to the explanation of Costa-Gomes and Weizsäcker (2008) for the low levels of consistency of choices and first-order beliefs. They have shown that subjects perceive a game differently when making choices than when stating beliefs (with stated beliefs revealing deeper strategic thinking than subjects' actions). In addition to addressing the issue of the consistency of choices in relation to first-order beliefs, our approach of eliciting also second-order beliefs allows to assess the expected level of consistency of the opponent player. This analysis has largely revealed the same general patterns as the analysis of a player's own consistency, meaning that teams expect others to be more often consistent.

Given that our paper addresses differences between individuals and teams it is also important to consider this paper's contribution to the literature on team decision making. First of all, it offers a classification of teams according to eight different strategic and non-strategic types of decision making. While the existing literature on team decision making has mainly focused on whether or not teams play equilibrium strategies more often than individuals (see, e.g., Bornstein et al., 2004; Cooper and Kagel, 2005), our experiment can also answer this question (affirmatively), but goes one step further by showing the distributional differences between individuals and teams with respect to different types of strategic and non-strategic behavior. In particular, the main difference (with respect to strategic types) originates from teams being more likely a *DI*-type, as explained above. This is an important difference as it shows that the higher strategic sophistication of teams is particularly driven by teams thinking one step further than individuals. This is in line with the findings of Kocher and Sutter (2005) where they have found in experimental guessing games that the modal depth of reasoning of teams is one level higher than the one of individuals.²² Second, we do not only elicit actual choices, but also first- and second-order beliefs. This has not been done in the team decision making literature before. Our approach allows checking the consistency of choices and first-order beliefs, respectively the expected consistency of one's opponent (by relating first- and second-order beliefs). We have found that teams are more consistent than individuals in this respect, and that they also expect their opponents to be more consistent than individuals expect their opponents to be. The latter feature of team behavior reinforces the consistency of teams, as it makes it more profitable for teams to be consistent themselves, because not playing Nash against an opponent who picks the equilibrium strategy leads to smaller gains

²² On the limited depths of reasoning (of individuals) in guessing games, information cascade experiments or auctions see, e.g., Kübler and Weizsäcker (2004), Gneezy (2005), Costa-Gomes and Crawford (2006), or Crawford and Iriberry (2007b).

than choosing the equilibrium strategy. In fact, we find that teams earn significantly more money than individuals in all games where one player has a dominant strategy (i.e., the *D*-games).

Previous studies on team decision making shed light on the possible reasons for our findings of higher strategic sophistication of teams. Cooper and Kagel (2005), for instance, have shown that putting oneself into the shoes of the opponent in a signaling game increases the likelihood of strategic play (being defined in their game as the relative frequency with which an incumbent monopolist chooses a larger quantity than a myopic best reply to the threat of market entry by an opponent would require). It seems very likely that our teams have also been influenced by such perspective taking, as suggested by the differences in higher-order beliefs between individuals and teams. Once such perspective taking is proposed, it seems hard to ignore it subsequently (as has been shown in the analysis of Cooper and Kagel, 2005).²³ For example, imagine that one team member draws the others' attention to how the opponent perceives the game, then the team might think more carefully about how to react optimally to the expected choice of the opponent, making it more likely that the strategic types prevail in a team.

Summing up, our paper has provided compelling evidence that teams are more sophisticated, i.e., more strategic in playing normal-form games than individuals. Given the surprisingly low level of strategic sophistication of individuals reported in previous papers – and replicated in this one – we consider the finding that team decision making increases the relative frequency of equilibrium play and the level of consistency in best responding to own beliefs an important contribution to the literature on strategic sophistication. Additionally, our paper is the first in the team decision making literature that provides a fine-grained classification of teams into eight different strategic and non-strategic types of decision makers, thereby adding important details to the bottom-line emerging from the team decision making literature that “teams are more rational” than individuals.

²³ The issue of perspective taking is related to the sharing and aggregating of information in teams. For a theoretical treatment of information aggregation in groups see Sobel (2006).

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Tables and Figures

Table 1: Choices and beliefs of individuals and teams (relative frequencies in %)

			<i>All games</i>	<i>D-games</i>	<i>ND-games</i>
<i>Choice</i>	Nash	Individuals	40.97 **	57.71 **	20.05
		Teams	49.30	70.42	22.92
	Pareto	Individuals	53.47 **	41.25 **	68.75
		Teams	45.72	28.96	66.67
	Other	Individuals	5.56	1.04	11.20
		Teams	4.98	0.62	10.41
<i>First-order beliefs</i>	Nash	Individuals	37.26	55.63	14.32
		Teams	40.04	60.62	14.32
	Pareto	Individuals	57.06	43.33	74.22
		Teams	55.21	38.75	75.78
	Other	Individuals	5.68	1.04	11.46
		Teams	4.75	0.63	9.90
<i>Second-order beliefs</i>	Nash	Individuals	30.55	47.71 **	9.11
		Teams	33.79	53.96	8.59
	Pareto	Individuals	62.62	50.21 *	78.13
		Teams	61.11	45.42	80.73
	Other	Individuals	6.83	2.08 *	12.76
		Teams	5.10	0.62	10.68

** (*) significant difference between individuals (in a given row) and teams (in the next row) at $p < 0.05$ ($p < 0.1$); two-sided Mann-Whitney U-test.

Table 2: Complexity of the game, choices and beliefs (relative frequencies in %)

		Complexity (game type) ^a	1R (D)	2R (D)	2R (ND)	3R (ND)	∞R (ND)
Choice	Nash	Individuals	76.67	38.75 **	31.25 *	14.58	17.19
		Teams	81.67	59.17	44.79	10.42	18.23
	Pareto	Individuals	22.50	60.00 **	68.75 *	56.25	75.00
		Teams	17.91	40.00	55.21	57.29	77.08
	Other	Individuals	0.83	1.25	0.00	29.17	7.81
		Teams	0.42	0.83	0.00	32.29	4.69
First-order beliefs	Nash	Individuals	88.75	22.50	21.87 *	7.29	14.06
		Teams	87.92	33.33	33.33	6.25	8.86
	Pareto	Individuals	10.42	76.25	78.13 *	56.25	81.25 *
		Teams	10.83	66.67	66.67	57.29	89.58
	Other	Individuals	0.83 *	1.25 *	0.00	36.46	4.69
		Teams	1.25	0.00	0.00	36.46	1.56
Second-order beliefs	Nash	Individuals	81.67	13.75 **	9.37	9.37	8.85
		Teams	82.50	25.42	17.71	4.17	6.25
	Pareto	Individuals	16.25	84.17 **	90.63	50.00	85.94
		Teams	17.08	73.75	82.29	57.29	91.67
	Other	Individuals	2.08 *	2.08 *	0.00	40.63	5.21
		Teams	0.42 *	0.83	0.00	38.54	2.08

** (*) significant difference between individuals and teams at $p < 0.05$ ($p < 0.1$); two-sided Mann-Whitney U-test.

^a The columns separate behavior according to (i) the different number of rounds (**R**) of iterated pure-strategy dominance a player needs to identify the own equilibrium choice, and (ii) the presence (**D**) or absence (**ND**) of a dominant strategy in the game.

Table 3: Percentages of decisions that comply with equilibrium, contingent on game-type

Complexity (game # for rows // game # for columns)		Individuals	Teams
Choice	<i>1 round of dominance to identify own equilibrium choice</i>		
	2x2 with dominant decision (#3, #13 // #1, #12)	78.13	78.13
	2x3 with dominant decision (#16 // #11)	75.00	83.33
	3x2 with dominant decision (#9 // #7)	81.25	79.17
	4x2 with dominant decision (#17 // #18)	70.83**	89.58
	<i>2 rounds of dominance</i>		
	2x2, partner has dominant decision (#1, #12 // #3, #13)	39.58**	60.42
	2x3, partner has dominant decision (#7 // #9)	27.08*	45.83
	3x2, partner has dominant decision (#11 // #16)	37.50**	66.67
	2x4, partner has dominant decision (#18 // #17)	50.00	62.50
	2x3 with 2 rounds of dominance (#2, #14 // #6, # 15)	31.25*	44.79
	<i>3 rounds of dominance</i>		
	3x2 with 3 rounds of dominance (#6, #15 // #2, #14)	14.58	10.42
	<i>No dominance</i>		
	2x3, unique equilibrium, no dominance (#8, #10 // # 4, #5)	18.75	22.92
	3x2, unique equilibrium, no dominance (#4, #5 // #8, #10)	15.63	13.54
First order beliefs	<i>1 round of dominance to identify own equilibrium choice</i>		
	2x2 with dominant decision (#3, #13 // #1, #12)	87.50	89.58
	2x3 with dominant decision (#16 // #11)	91.67	93.75
	3x2 with dominant decision (#9 // #7)	91.67	83.33
	4x2 with dominant decision (#17 // #18)	85.42	83.33
	<i>2 rounds of dominance</i>		
	2x2, partner has dominant decision (#1, #12 // #3, #13)	22.92	30.21
	2x3, partner has dominant decision (#7 // #9)	29.17	25.00
	3x2, partner has dominant decision (#11 // #16)	14.58**	35.42
	2x4, partner has dominant decision (#18 // #17)	22.92**	45.83
	2x3 with 2 rounds of dominance (#2, #14 // #6, # 15)	21.87*	33.33
	<i>3 rounds of dominance</i>		
	3x2 with 3 rounds of dominance (#6, #15 // #2, #14)	7.29	6.25
	<i>No dominance</i>		
	2x3, unique equilibrium, no dominance (#8, #10 // # 4, #5)	19.79	14.58
	3x2, unique equilibrium, no dominance (#4, #5 // #8, #10)	8.33	3.13
Second order beliefs	<i>1 round of dominance to identify own equilibrium choice</i>		
	2x2 with dominant decision (#3, #13 // #1, #12)	85.42	85.42
	2x3 with dominant decision (#16 // #11)	85.42	83.33
	3x2 with dominant decision (#9 // #7)	79.17	75.00
	4x2 with dominant decision (#17 // #18)	72.92	83.33
	<i>2 rounds of dominance</i>		
	2x2, partner has dominant decision (#1, #12 // #3, #13)	18.75	27.08
	2x3, partner has dominant decision (#7 // #9)	6.25	16.67
	3x2, partner has dominant decision (#11 // #16)	6.25**	27.08
	2x4, partner has dominant decision (#18 // #17)	18.75	29.17
	2x3 with 2 rounds of dominance (#2, #14 // #6, # 15)	9.37	17.71
	<i>3 rounds of dominance</i>		
	3x2 with 3 rounds of dominance (#6, #15 // #2, #14)	9.37	4.17
	<i>No dominance</i>		
	2x3, unique equilibrium, no dominance (#8, #10 // # 4, #5)	8.33	8.33
	3x2, unique equilibrium, no dominance (#4, #5 // #8, #10)	9.37	4.17

** (*) significant difference between individuals and teams for a given game type at $p < 0.05$

($p < 0.1$); two-sided Mann-Whitney U-test.

Table 4: Consistency of decisions (relative frequency of best reply in %)

		<i>All games</i>	<i>D-games</i>	<i>ND-games</i>
<i>Choice is best reply to first-order belief</i>	Individuals	55.79**	59.79 **	50.78
	Teams	65.86	75.21	54.17
<i>First-order belief is best reply to second-order belief</i>	Individuals	59.26	61.67	56.25
	Teams	62.73	67.08	57.29

** (*) significant difference between individuals and teams at $p < 0.05$ ($p < 0.1$); two-sided Mann-Whitney U-test.

Table 5: Relative frequency of consistency-types Nash-CON and Max-CON

			<i>all games</i>	<i>D-games</i>	<i>ND-games</i>
<i>Player's own consistency</i>	Nash-CON	Individuals	18.75**	27.71**	7.55
		Teams	27.66	45.00	5.99
	Max-CON	Individuals	32.41	31.46	33.59
		Teams	31.37	29.58	33.59
<i>Expected consistency of opponent</i>	Nash-CON	Individuals	10.53**	15.42**	4.43
		Teams	17.36	28.75	3.13
	Max-CON	Individuals	44.56*	44.79**	44.27
		Teams	40.16	37.71	43.23

** (*) significant difference between individuals and teams at $p < 0.05$ ($p < 0.1$); two-sided Mann-Whitney U-test.

Table 6: Determinants of a player's consistency-type Nash-CON

[A] Probit regression

<i>Variable</i>	<i>Coefficient</i>	<i>Std. error</i>	<i>p-value</i>
dominant strategy	1.313	0.281	0.000
opponent dominant strategy	0.935	0.257	0.000
team player	-0.145	0.287	0.614
opponent = team	-0.053	0.242	0.826
team * opponent = team	0.053	0.329	0.873
team * dominant strategy	0.508	0.209	0.015
team * opponent dominant strategy	0.703	0.221	0.001
risk with maxstrategy	0.011	0.007	0.092
risk with maxstrategy opponent	0.000	0.007	0.970
advantageous inequality in Nash	-0.013	0.012	0.292
disadvantageous inequality in Nash	-0.038	0.013	0.003
Constant	-1.339	0.405	0.001

N = 1.728; standard errors clustered for the 96 decision makers (48 individuals, 48 teams).

[B] Marginal effects of team decision making contingent on binary independent variables

Dominant strategy	Opponent dominant strategy	Opponent = Team	<i>Marginal effect</i>	<i>Std. error</i>	<i>p-value</i>
Yes	No	Yes	0.163	0.100	0.105
Yes	No	No	0.143	0.102	0.162
No	Yes	Yes	0.221	0.089	0.013
No	Yes	No	0.204	0.090	0.023
No	No	Yes	-0.009	0.025	0.728
No	No	No	-0.014	0.029	0.628

Marginal effects evaluated at the mean of independent variables (risk with maxstrategy, risk with maxstrategy opponent, advantageous inequality in Nash, disadvantageous inequality in Nash) and contingent on the three binary variables indicated in the top row (and their interaction effect with team decision making).

Table 7: Determinants of the opponent's expected consistency-type being Nash-CON

[A] Probit regression

<i>Variable</i>	<i>Coefficient</i>	<i>Std. error</i>	<i>p-value</i>
dominant strategy	1.594	0.369	0.000
opponent dominant strategy	0.414	0.289	0.152
team player	-0.196	0.364	0.589
opponent =team	-0.023	0.256	0.930
team * opponent = team	0.045	0.340	0.895
team * dominant strategy	0.581	0.269	0.031
team * opponent dominant strategy	0.720	0.229	0.002
risk with maxstrategy	0.020	0.007	0.007
risk with maxstrategy opponent	-0.007	0.006	0.239
advantageous inequality in Nash	-0.036	0.013	0.008
disadvantageous inequality in Nash	-0.037	0.014	0.010
Constant	-1.465	0.417	0.000

N = 1.728; standard errors clustered for the 96 decision makers (48 individuals, 48 teams).

[B] Marginal effects of team decision making contingent on binary independent variables

Dominant strategy	Opponent dominant strategy	Opponent = Team	<i>Marginal effect</i>	<i>Std. error</i>	<i>p-value</i>
Yes	No	Yes	0.108	0.048	0.024
Yes	No	No	0.099	0.052	0.059
No	Yes	Yes	0.169	0.095	0.077
No	Yes	No	0.151	0.107	0.158
No	No	Yes	-0.008	0.015	0.599
No	No	No	-0.011	0.022	0.630

Marginal effects evaluated at the mean of independent variables (risk with maxstrategy, risk with maxstrategy opponent, advantageous inequality in Nash, disadvantageous inequality in Nash) and contingent on the three binary variables indicated in the top row (and their interaction effect with team decision making).

Table 8: Estimated types of individuals and teams – Own choices

Type	Individuals		Teams	
	p_k	ε_k	p_k	ε_k
Altruistic	.122 (.051)**	.232 (.073)***	.121 (.066)*	.445 (.158)***
Pessimistic	.070 (.047)	.553 (.128)***	.005 (.212)	.999 (.000)***
Optimistic	.407 (.082)***	.220 (.044)***	.252 (.067)***	.208 (.048)***
Equilibrium	.021 (.021)	.097 (.094)	.020 (.028)	.318 (.171)*
Sophisticated	.058 (.051)	.311 (.117)***	.020 (.021)	.000 (.000)
D1	.223 (.080)***	.504 (.065)***	.472 (.087)***	.315 (.038)***
D2	.062 (.042)	.242 (.101)**	.002 (.201)	.999 (.001)***
L2	.038 (.049)	.376 (.147)**	.108 (.060)*	.270 (.076)***
Sum Strategic	.402 (.076)***		.622 (.221)***	

(standard errors in parenthesis, *** significant at 1% level, ** significant at 5% level, * significant at 10% level)

Figure 1. The 18 normal-form games

game # 3 (D)		[1R, 2R]	
	1	2	
1	72, 93	31, 46	
2	84, 52	55, 79	

game # 13 (D)		[1R, 2R]	
	1	2	
1	94, 23	38, 57	
2	45, 89	14, 18	

game # 1 (D)		[2R, 1R]	
	1	2	
1	75, 51	42, 27	
2	48, 80	89, 68	

game # 12 (D)		[2R, 1R]	
	1	2	
1	21, 92	87, 43	
2	55, 36	16, 12	

game # 7 (D)		[2R, 1R]		
	1	2	3	
1	59, 58	46, 83	85, 61	
2	38, 29	70, 52	37, 23	

game # 11 (D)		[2R, 1R]	
	1	2	
1	31, 32	68, 46	
2	72, 43	47, 61	
3	91, 65	43, 84	

game # 9 (D)		[1R, 2R]	
	1	2	
1	28, 37	57, 58	
2	22, 36	60, 84	
3	51, 69	82, 45	

game # 16 (D)		[1R, 2R]		
	1	2	3	
1	42, 64	57, 43	80, 39	
2	28, 27	39, 68	61, 87	

game # 6 (ND)		[3R, 2R]	
	1	2	
1	53, 86	24, 19	
2	79, 57	42, 73	
3	28, 23	71, 50	

game # 15 (ND)		[3R, 2R]	
	1	2	
1	76, 93	25, 12	
2	43, 40	74, 62	
3	94, 16	59, 37	

game # 14 (ND)		[2R, 3R]		
	1	2	3	
1	21, 26	52, 73	75, 44	
2	88, 55	25, 30	59, 81	

game # 2 (ND)		[2R, 3R]		
	1	2	3	
1	42, 45	95, 78	18, 96	
2	64, 76	14, 27	39, 61	

game # 8 (ND)		[∞R, ∞R]		
	1	2	3	
1	87, 32	18, 37	63, 76	
2	65, 89	96, 63	24, 30	

game # 10 (ND)		[∞R, ∞R]		
	1	2	3	
1	67, 91	95, 64	31, 35	
2	89, 49	23, 53	56, 78	

game # 5 (ND)		[∞R, ∞R]	
	1	2	
1	72, 59	26, 20	
2	33, 14	59, 92	
3	28, 83	85, 61	

game # 4 (ND)		[∞R, ∞R]	
	1	2	
1	46, 16	57, 88	
2	71, 49	28, 24	
3	42, 82	84, 60	

game # 17 (D)		[1R, 2R]	
	1	2	
1	22, 14	57, 55	
2	30, 42	28, 37	
3	15, 60	61, 88	
4	45, 66	82, 31	

game # 18 (D)		[2R, 1R]			
	1	2	3	4	
1	56, 58	38, 29	89, 62	32, 86	
2	15, 23	43, 31	61, 16	67, 46	

Supplementary material

to

“Strategic sophistication of individuals and teams in experimental normal-form games”

- Tables S1 to S10 (*pages 2 to 11*)
- Figures F1 to F4 (*pages 12 to 15*)
- Experimental instructions and control questionnaire (*pages 16 to 21*)

Table S1. Relative frequencies of choices in each game – Individuals

game # 3 (D)				game # 13 (D)			
	(70.83%)	(29.17%)		(45.83%)	(54.17%)		
(37.5%)	72, 93	31, 46		(87.5%)	94, 23	38, 57	
(62.5%)	84, 52	55, 79		(12.5%)	45, 89	14, 18	

game # 1 (D)				game # 12 (D)			
	(75%)	(25%)		(87.5%)	(12.5%)		
(37.5%)	75, 51	42, 27		(62.5%)	21, 92	87, 43	
(62.5%)	48, 80	89, 68		(37.5%)	55, 36	16, 12	

game # 7 (D)				game # 11 (D)			
	(0%)	(91.67%)	(8.33%)	(16.67%)	(83.33%)		
(83.33%)	59, 58	46, 83	85, 61	(29.17%)	31, 32	68, 46	
(16.67%)	38, 29	70, 52	37, 23	(8.33%)	72, 43	47, 61	
				(62.5%)	91, 65	43, 84	

game # 9 (D)				game # 16 (D)			
	(37.5%)	(62.5%)		(45.83%)	(4.17%)	(50%)	
(0%)	28, 37	57, 58		(66.67%)	42, 64	57, 43	80, 39
(29.17%)	22, 36	60, 84		(33.33%)	28, 27	39, 68	61, 87
(70.83%)	51, 69	82, 45					

game # 6 (ND)				game # 15 (ND)			
	(75%)	(25%)		(54.17%)	(45.83%)		
(20.83%)	53, 86	24, 19		(33.33%)	76, 93	25, 12	
(79.17%)	79, 57	42, 73		(12.5%)	43, 40	74, 62	
(0%)	28, 23	71, 50		(54.17%)	94, 16	59, 37	

game # 14 (ND)				game # 2 (ND)			
	(0%)	(29.17%)	(70.83%)	(16.67%)	(41.67%)	(41.67%)	
(25%)	21, 26	52, 73	75, 44	(70.83%)	42, 45	95, 78	18, 96
(75%)	88, 55	25, 30	59, 81	(29.17%)	64, 76	14, 27	39, 61

game # 8 (ND)				game # 10 (ND)			
	(54.17%)	(12.5%)	(33.33%)	(79.17%)	(8.33%)	(12.5%)	
(25%)	87, 32	18, 37	63, 76	(83.33%)	67, 91	95, 64	31, 35
(75%)	65, 89	96, 63	24, 30	(16.67%)	89, 49	23, 53	56, 78

game # 5 (ND)				game # 4 (ND)			
	(20.83%)	(79.17%)		(12.5%)	(87.5%)		
(4.17%)	72, 59	26, 20		(16.67%)	46, 16	57, 88	
(25%)	33, 14	59, 92		(12.5%)	71, 49	28, 24	
(70.83%)	28, 83	85, 61		(70.83%)	42, 82	84, 60	

game # 17 (D)				game # 18 (D)			
	(58.33%)	(41.67%)		(0%)	(4.17%)	(20.83%)	(75%)
(0%)	22, 14	57, 55		(41.67%)	56, 58	38, 29	89, 62
(4.17%)	30, 42	28, 37		(58.33%)	15, 23	43, 31	61, 16
(29.17%)	15, 60	61, 88					82, 46
(66.67%)	45, 66	82, 31					

Table S2. Relative frequencies of choices in each game – Teams

game # 3			(D)		
	(58.33%)	(41.67%)			
(33.33%)	72, 93	31, 46			
(66.67%)	84, 52	55, 79			

game # 13			(D)		
	(25%)	(75%)			
(91.67%)	94, 23	38, 57			
(8.33%)	45, 89	14, 18			

game # 1			(D)		
	(66.67%)	(33.33%)			
(54.17%)	75, 51	42, 27			
(45.83%)	48, 80	89, 68			

game # 12			(D)		
	(87.5%)	(12.5%)			
(29.17%)	21, 92	87, 43			
(70.83%)	55, 36	16, 12			

game # 7			(D)		
	(4.17%)	(83.33%)	(12.5%)		
(54.17%)	59, 58	46, 83	85, 61		
(45.83%)	38, 29	70, 52	37, 23		

game # 11			(D)		
	(20.83%)	(79.17%)			
(62.5%)	31, 32	68, 46			
(4.17%)	72, 43	47, 61			
(33.33%)	91, 65	43, 84			

game # 9			(D)		
	(45.83%)	(54.17%)			
(0%)	28, 37	57, 58			
(25%)	22, 36	60, 84			
(75%)	51, 69	82, 45			

game # 16			(D)		
	(70.83%)	(4.17%)	(25%)		
(87.5%)	42, 64	57, 43	80, 39		
(12.5%)	28, 27	39, 68	61, 87		

game # 6			(ND)		
	(58.33%)	(41.67%)			
(0%)	53, 86	24, 19			
(83.33%)	79, 57	42, 73			
(16.67%)	28, 23	71, 50			

game # 15			(ND)		
	(29.17%)	(70.83%)			
(25%)	76, 93	25, 12			
(8.33%)	43, 40	74, 62			
(66.67%)	94, 16	59, 37			

game # 14			(ND)		
	(0%)	(8.33%)	(91.67%)		
(25%)	21, 26	52, 73	75, 44		
(75%)	88, 55	25, 30	59, 81		

game # 8			(ND)		
	(75%)	(8.33%)	(16.67%)		
(25%)	87, 32	18, 37	63, 76		
(75%)	65, 89	96, 63	24, 30		

game # 5			(ND)		
	(20.83%)	(79.17%)			
(16.67%)	72, 59	26, 20			
(12.5%)	33, 14	59, 92			
(70.83%)	28, 83	85, 61			

game # 4			(ND)		
	(25%)	(75%)			
(4.17%)	46, 16	57, 88			
(12.5%)	71, 49	28, 24			
(83.33%)	42, 82	84, 60			

game # 17			(D)		
	(66.67%)	(33.33%)			
(0%)	22, 14	57, 55			
(0%)	30, 42	28, 37			
(0%)	15, 60	61, 88			
(100%)	45, 66	82, 31			

game # 18			(D)		
	(0%)	(0%)	(20.83%)	(79.17%)	
(41.67%)	56, 58	38, 29	89, 62	32, 86	
(58.33%)	15, 23	43, 31	61, 16	67, 46	

Table S3: Relative frequencies of first-order beliefs in each game – Individuals

game # 3				game # 13				
<i>(D)</i>				<i>(D)</i>				
	(95.83%)	(4.17%)			(79.17%)	(20.83%)		
(16.67%)	72, 93	31, 46		(95.83%)	94, 23	38, 57		
(83.33%)	84, 52	55, 79		(4.17%)	45, 89	14, 18		
game # 1				game # 12				
<i>(D)</i>				<i>(D)</i>				
	(87.5%)	(12.5%)			(83.33%)	(16.67%)		
(25%)	75, 51	42, 27		(58.33%)	21, 92	87, 43		
(75%)	48, 80	89, 68		(41.67%)	55, 36	16, 12		
game # 7				game # 11				
<i>(D)</i>				<i>(D)</i>				
	(4.17%)	(91.67%)	(4.17%)		(8.33%)	(91.67%)		
(54.17%)	59, 58	46, 83	85, 61	(20.83%)	31, 32	68, 46		
(45.83%)	38, 29	70, 52	37, 23	(4.17%)	72, 43	47, 61		
				(75%)	91, 65	43, 84		
game # 9				game # 16				
<i>(D)</i>				<i>(D)</i>				
	(12.5%)	(87.5%)			(8.33%)	(8.33%)	(83.33%)	
(0%)	28, 37	57, 58		(91.67%)	42, 64	57, 43	80, 39	
(8.33%)	22, 36	60, 84		(8.33%)	28, 27	39, 68	61, 87	
(91.67%)	51, 69	82, 45						
game # 6				game # 15				
<i>(ND)</i>				<i>(ND)</i>				
	(83.33%)	(16.67%)			(83.33%)	(16.67%)		
(4.17%)	53, 86	24, 19		(20.83%)	76, 93	25, 12		
(79.17%)	79, 57	42, 73		(8.33%)	43, 40	74, 62		
(16.67%)	28, 23	71, 50		(70.83%)	94, 16	59, 37		
game # 14				game # 2				
<i>(ND)</i>				<i>(ND)</i>				
	(8.33%)	(0%)	(91.67%)		(4.17%)	(33.33%)	(62.5%)	
(29.17%)	21, 26	52, 73	75, 44	(75%)	42, 45	95, 78	18, 96	
(70.83%)	88, 55	25, 30	59, 81	(25%)	64, 76	14, 27	39, 61	
game # 8				game # 10				
<i>(ND)</i>				<i>(ND)</i>				
	(83.33%)	(16.67%)	(0%)		(87.5%)	(8.33%)	(4.17%)	
(25%)	87, 32	18, 37	63, 76	(75%)	67, 91	95, 64	31, 35	
(75%)	65, 89	96, 63	24, 30	(25%)	89, 49	23, 53	56, 78	
game # 5				game # 4				
<i>(ND)</i>				<i>(ND)</i>				
	(12.5%)	(87.5%)			(16.67%)	(83.33%)		
(20.83%)	72, 59	26, 20		(4.17%)	46, 16	57, 88		
(8.33%)	33, 14	59, 92		(8.33%)	71, 49	28, 24		
(70.83%)	28, 83	85, 61		(87.5%)	42, 82	84, 60		
game # 17				game # 18				
<i>(D)</i>				<i>(D)</i>				
	(8.33%)	(91.67%)			(0%)	(0%)	(12.5%)	(87.5%)
(4.17%)	22, 14	57, 55		(62.5%)	56, 58	38, 29	89, 62	32, 86
(0%)	30, 42	28, 37		(37.5%)	15, 23	43, 31	61, 16	67, 46
(12.5%)	15, 60	61, 88						
(83.33%)	45, 66	82, 31						

Table S4: Relative frequencies of first-order beliefs in each game – Teams

game # 3 (D)				game # 13 (D)				
	(75%)	(25%)			(54.17%)	(45.83%)		
(16.67%)	72, 93	31, 46		(91.67%)	94, 23	38, 57		
(83.33%)	84, 52	55, 79		(8.33%)	45, 89	14, 18		
game # 1 (D)				game # 12 (D)				
	(75%)	(25%)			(100%)	(0%)		
(37.5%)	75, 51	42, 27		(75%)	21, 92	87, 43		
(62.5%)	48, 80	89, 68		(25%)	55, 36	16, 12		
game # 7 (D)				game # 11 (D)				
	(8.33%)	(83.33%)	(8.33%)		(4.17%)	(95.83%)		
(70.83%)	59, 58	46, 83	85, 61	(20.83%)	31, 32	68, 46		
(29.17%)	38, 29	70, 52	37, 23	(0%)	72, 43	47, 61		
				(79.17%)	91, 65	43, 84		
game # 9 (D)				game # 16 (D)				
	(20.83%)	(79.17%)			(50%)	(0%)	(50%)	
(0%)	28, 37	57, 58		(91.67%)	42, 64	57, 43	80, 39	
(16.67%)	22, 36	60, 84		(8.33%)	28, 27	39, 68	61, 87	
(83.33%)	51, 69	82, 45						
game # 6 (ND)				game # 15 (ND)				
	(54.17%)	(45.83%)			(66.67%)	(33.33%)		
(0%)	53, 86	24, 19		(12.5%)	76, 93	25, 12		
(91.67%)	79, 57	42, 73		(8.33%)	43, 40	74, 62		
(8.33%)	28, 23	71, 50		(79.17%)	94, 16	59, 37		
game # 14 (ND)				game # 2 (ND)				
	(0%)	(4.17%)	(95.83%)		(4.17%)	(29.17%)	(66.67%)	
(25%)	21, 26	52, 73	75, 44	(70.83%)	42, 45	95, 78	18, 96	
(75%)	88, 55	25, 30	59, 81	(29.17%)	64, 76	14, 27	39, 61	
game # 8 (ND)				game # 10 (ND)				
	(95.83%)	(4.17%)	(0%)		(100%)	(0%)	(0%)	
(8.33%)	87, 32	18, 37	63, 76	(91.67%)	67, 91	95, 64	31, 35	
(91.67%)	65, 89	96, 63	24, 30	(8.33%)	89, 49	23, 53	56, 78	
game # 5 (ND)				game # 4 (ND)				
	(16.67%)	(83.33%)			(25%)	(75%)		
(12.5%)	72, 59	26, 20		(0%)	46, 16	57, 88		
(8.33%)	33, 14	59, 92		(0%)	71, 49	28, 24		
(79.17%)	28, 83	85, 61		(100%)	42, 82	84, 60		
game # 17 (D)				game # 18 (D)				
	(54.17%)	(45.83%)			(0%)	(0%)	(12.5%)	(87.5%)
(4.17%)	22, 14	57, 55		(62.5%)	56, 58	38, 29	89, 62	32, 86
(0%)	30, 42	28, 37		(37.5%)	15, 23	43, 31	61, 16	67, 46
(16.67%)	15, 60	61, 88						
(79.17%)	45, 66	82, 31						

Table S5: Relative frequencies of second-order beliefs in each game – Individuals

game # 3			(D)		
	(87.5%)	(12.5%)			
(20.83%)	72, 93	31, 46			
(79.17%)	84, 52	55, 79			

game # 13			(D)		
	(70.83%)	(29.17%)			
(91.67%)	94, 23	38, 57			
(8.33%)	45, 89	14, 18			

game # 1			(D)		
	(83.33%)	(16.67%)			
(8.33%)	75, 51	42, 27			
(91.67%)	48, 80	89, 68			

game # 12			(D)		
	(87.5%)	(12.5%)			
(75%)	21, 92	87, 43			
(25%)	55, 36	16, 12			

game # 7			(D)		
	(8.33%)	(83.33%)	(8.33%)		
(100%)	59, 58	46, 83	85, 61		
(0%)	38, 29	70, 52	37, 23		

game # 11			(D)		
	(8.33%)	(91.67%)			
(4.17%)	31, 32	68, 46			
(4.17%)	72, 43	47, 61			
(91.67%)	91, 65	43, 84			

game # 9			(D)		
	(12.5%)	(87.5%)			
(0%)	28, 37	57, 58			
(25%)	22, 36	60, 84			
(75%)	51, 69	82, 45			

game # 16			(D)		
	(8.33%)	(16.67%)	(75%)		
(79.17%)	42, 64	57, 43	80, 39		
(20.83%)	28, 27	39, 68	61, 87		

game # 6			(ND)		
	(87.5%)	(12.5%)			
(20.83%)	53, 86	24, 19			
(79.17%)	79, 57	42, 73			
(0%)	28, 23	71, 50			

game # 15			(ND)		
	(83.33%)	(16.67%)			
(25%)	76, 93	25, 12			
(4.17%)	43, 40	74, 62			
(70.83%)	94, 16	59, 37			

game # 14			(ND)		
	(8.33%)	(16.67%)	(75%)		
(4.17%)	21, 26	52, 73	75, 44		
(95.83%)	88, 55	25, 30	59, 81		

game # 2			(ND)		
	(16.67%)	(20.83%)	(62.5%)		
(95.83%)	42, 45	95, 78	18, 96		
(4.17%)	64, 76	14, 27	39, 61		

game # 8			(ND)		
	(75%)	(8.33%)	(16.67%)		
(0%)	87, 32	18, 37	63, 76		
(100%)	65, 89	96, 63	24, 30		

game # 10			(ND)		
	(79.17%)	(8.33%)	(12.5%)		
(87.5%)	67, 91	95, 64	31, 35		
(12.5%)	89, 49	23, 53	56, 78		

game # 5			(ND)		
	(8.33%)	(91.67%)			
(4.17%)	72, 59	26, 20			
(16.67%)	33, 14	59, 92			
(79.17%)	28, 83	85, 61			

game # 4			(ND)		
	(12.5%)	(87.5%)			
(8.33%)	46, 16	57, 88			
(4.17%)	71, 49	28, 24			
(87.5%)	42, 82	84, 60			

game # 17			(D)		
	(25%)	(75%)			
(4.17%)	22, 14	57, 55			
(0%)	30, 42	28, 37			
(25%)	15, 60	61, 88			
(70.83%)	45, 66	82, 31			

game # 18			(D)		
	(8.33%)	(0%)	(16.67%)	(75%)	
(87.5%)	56, 58	38, 29	89, 62	32, 86	
(12.5%)	15, 23	43, 31	61, 16	67, 46	

Table S6: Relative frequencies of second-order beliefs in each game – Teams

game # 3				game # 13					
		<i>(D)</i>				<i>(D)</i>			
		(91.67%)	(8.33%)			(62.5%)	(37.5%)		
(25%)		72, 93	31, 46	(95.83%)		94, 23	38, 57		
(75%)		84, 52	55, 79	(4.17%)		45, 89	14, 18		
game # 1				game # 12					
		<i>(D)</i>				<i>(D)</i>			
		(83.33%)	(16.67%)			(87.5%)	(12.5%)		
(8.33%)		75, 51	42, 27	(45.83%)		21, 92	87, 43		
(91.67%)		48, 80	89, 68	(54.17%)		55, 36	16, 12		
game # 7				game # 11					
		<i>(D)</i>				<i>(D)</i>			
		(4.17%)	(79.17%)	(16.67%)			(20.83%)	(79.17%)	
(87.5%)		59, 58	46, 83	85, 61	(25%)	31, 32	68, 46		
(12.5%)		38, 29	70, 52	37, 23	(4.17%)	72, 43	47, 61		
(70.83%)					(70.83%)	91, 65	43, 84		
game # 9				game # 16					
		<i>(D)</i>				<i>(D)</i>			
		(20.83%)	(79.17%)			(29.17%)	(4.17%)	(66.67%)	
(0%)		28, 37	57, 58	(87.5%)		42, 64	57, 43	80, 39	
(29.17%)		22, 36	60, 84	(12.5%)		28, 27	39, 68	61, 87	
(70.83%)		51, 69	82, 45						
game # 6				game # 15					
		<i>(ND)</i>				<i>(ND)</i>			
		(87.5%)	(12.5%)			(75%)	(25%)		
(4.17%)		53, 86	24, 19	(25%)		76, 93	25, 12		
(87.5%)		79, 57	42, 73	(0%)		43, 40	74, 62		
(8.33%)		28, 23	71, 50	(75%)		94, 16	59, 37		
game # 14				game # 2					
		<i>(ND)</i>				<i>(ND)</i>			
		(0%)	(4.17%)	(95.83%)			(4.17%)	(20.83%)	(75%)
(12.5%)		21, 26	52, 73	75, 44	(79.17%)	42, 45	95, 78	18, 96	
(87.5%)		88, 55	25, 30	59, 81	(20.83%)	64, 76	14, 27	39, 61	
game # 8				game # 10					
		<i>(ND)</i>				<i>(ND)</i>			
		(91.67%)	(0%)	(8.33%)			(91.67%)	(8.33%)	(0%)
(8.33%)		87, 32	18, 37	63, 76	(87.5%)	67, 91	95, 64	31, 35	
(91.67%)		65, 89	96, 63	24, 30	(12.5%)	89, 49	23, 53	56, 78	
game # 5				game # 4					
		<i>(ND)</i>				<i>(ND)</i>			
		(8.33%)	(91.67%)			(4.17%)	(95.83%)		
(4.17%)		72, 59	26, 20	(4.17%)		46, 16	57, 88		
(4.17%)		33, 14	59, 92	(4.17%)		71, 49	28, 24		
(91.67%)		28, 83	85, 61	(91.67%)		42, 82	84, 60		
game # 17				game # 18					
		<i>(D)</i>				<i>(D)</i>			
		(20.83%)	(79.17%)			(0%)	(0%)	(29.17%)	(70.83%)
(0%)		22, 14	57, 55	(62.5%)		56, 58	38, 29	89, 62	32, 86
(0%)		30, 42	28, 37	(37.5%)		15, 23	43, 31	61, 16	67, 46
(4.17%)		15, 60	61, 88						
(95.83%)		45, 66	82, 31						

Table S7: Relative frequency of correct higher order beliefs (%) – Individuals (row players)

<i>game</i>	<i>First – order beliefs</i>		<i>Second – order beliefs</i>	
	<i>correct</i>	<i>incorrect</i>	<i>correct</i>	<i>incorrect</i>
game #1	62.5	37.5	62.50	37.50
game #2	54.17	45.83	62.50	37.50
game #3	70.83	29.17	75.00	25.00
game #4	62.5	37.5	75.00	25.00
game #5	83.33	16.67	62.50	37.50
game #6	70.83	29.17	66.67	33.33
game #7	70.83	29.17	58.33	41.67
game #8	41.67	58.33	62.50	37.50
game #9	62.50	37.50	70.83	29.17
game #10	75.00	25.00	70.83	29.17
game #11	66.67	33.33	66.67	33.33
game #12	79.17	20.83	50.00	50.00
game #13	45.83	54.17	87.50	12.50
game #14	70.83	29.17	70.83	29.17
game #15	50.00	50.00	62.50	37.50
game #16	37.50	62.50	62.50	37.50
game #17	50.00	50.00	75.00	25.00
game #18	54.17	45.83	58.33	41.67
<i>all games</i>	<i>61.57</i>	<i>38.43</i>	<i>66.67</i>	<i>33.33</i>

*Table S8: Relative frequency of correct higher order beliefs (%) –
Individuals (column players)*

<i>game</i>	<i>First – order beliefs</i>		<i>Second – order beliefs</i>	
	<i>correct</i>	<i>incorrect</i>	<i>correct</i>	<i>incorrect</i>
game #1	58.33	41.67	70.83	29.17
game #2	54.17	45.83	37.50	62.50
game #3	58.33	41.67	79.17	20.83
game #4	66.67	33.33	62.50	37.50
game #5	50.00	50.00	83.33	16.67
game #6	66.67	33.33	83.33	16.67
game #7	45.83	54.17	66.67	33.33
game #8	66.67	33.33	75.00	25.00
game #9	62.50	37.50	75.00	25.00
game #10	79.17	20.83	75.00	25.00
game #11	33.33	66.67	87.50	12.50
game #12	54.17	45.83	79.17	20.83
game #13	83.33	16.67	70.83	29.17
game #14	70.83	29.17	70.83	29.17
game #15	50.00	50.00	70.83	29.17
game #16	70.83	29.17	54.17	45.83
game #17	70.83	29.17	70.83	29.17
game #18	45.83	54.17	58.33	41.67
<i>all games</i>	<i>60.42</i>	<i>39.58</i>	<i>70.60</i>	<i>29.40</i>

Table S9: Relative frequency of correct higher order beliefs (%) – Teams (row players)

<i>game</i>	<i>First – order beliefs</i>		<i>Second – order beliefs</i>	
	<i>correct</i>	<i>incorrect</i>	<i>correct</i>	<i>incorrect</i>
game #1	66.67	33.33	70.83	29.17
game #2	62.50	37.50	66.67	33.33
game #3	45.83	54.17	66.67	33.33
game #4	79.17	20.83	83.33	16.67
game #5	62.50	37.50	70.83	29.17
game #6	62.50	37.50	75.00	25.00
game #7	58.33	41.67	75.00	25.00
game #8	62.50	37.50	87.50	12.50
game #9	45.83	54.17	58.33	41.67
game #10	79.17	20.83	83.33	16.67
game #11	83.33	16.67	62.50	37.50
game #12	70.83	29.17	58.33	41.67
game #13	41.67	58.33	87.50	12.50
game #14	79.17	20.83	66.67	33.33
game #15	37.50	62.50	58.33	41.67
game #16	66.67	33.33	70.83	29.17
game #17	62.50	37.50	83.33	16.67
game #18	70.83	29.17	58.33	41.67
<i>all games</i>	<i>63.19</i>	<i>36.81</i>	<i>71.30</i>	<i>28.70</i>

Table S10: Relative frequency of correct higher order beliefs (%) – Teams (column players)

<i>game</i>	<i>First – order beliefs</i>		<i>Second – order beliefs</i>	
	<i>correct</i>	<i>incorrect</i>	<i>correct</i>	<i>incorrect</i>
game #1	58.33	41.67	75.00	25.00
game #2	54.17	45.83	70.83	29.17
game #3	54.17	45.83	70.83	29.17
game #4	79.17	20.83	79.17	20.83
game #5	54.17	45.83	70.83	29.17
game #6	75.00	25.00	54.17	45.83
game #7	50.00	50.00	70.83	29.17
game #8	58.33	41.67	79.17	20.83
game #9	83.33	16.67	58.33	41.67
game #10	66.67	33.33	83.33	16.67
game #11	50.00	50.00	70.83	29.17
game #12	37.50	62.50	79.17	20.83
game #13	83.33	16.67	29.17	70.83
game #14	66.67	33.33	87.50	12.50
game #15	41.67	58.33	45.83	54.17
game #16	75.00	25.00	45.83	54.17
game #17	70.83	29.17	62.50	37.50
game #18	62.50	37.50	70.83	29.17
<i>all games</i>	62.27	37.73	66.90	33.10

Figure F1: Frequency of Nash strategy in the 18 games (choices, first- and second-order beliefs) – actually observed versus random play

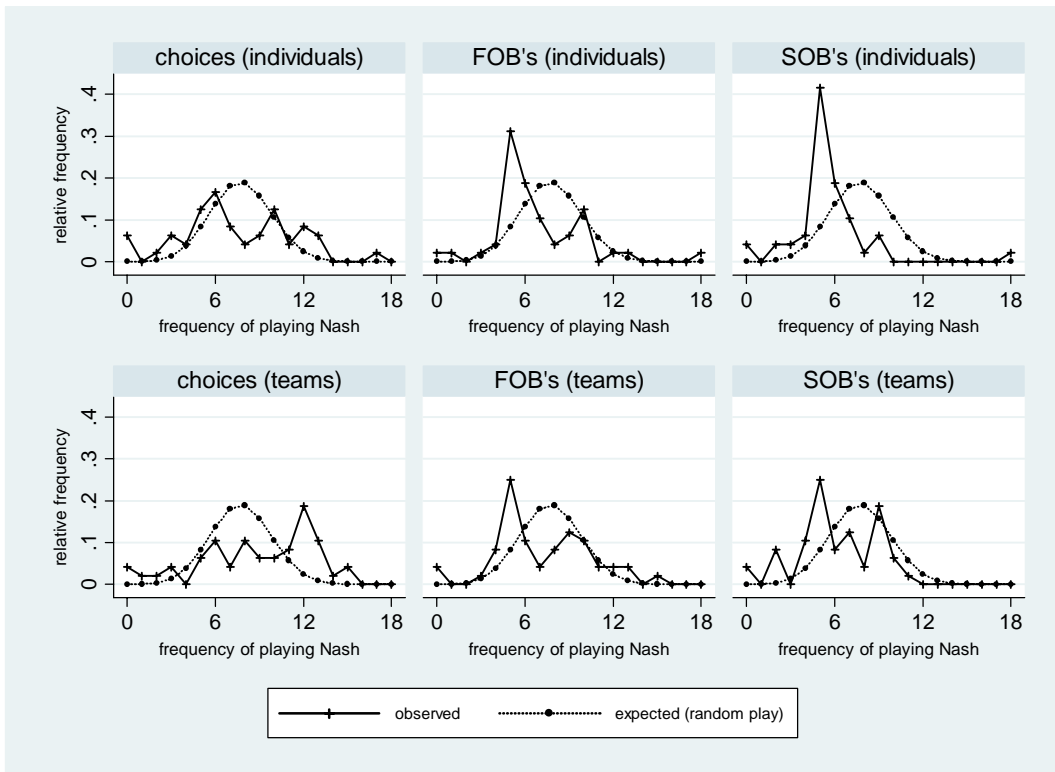


Figure F2: Frequency of consistency in the 18 games (own consistency, expected consistency of opponent) – actually observed versus random play

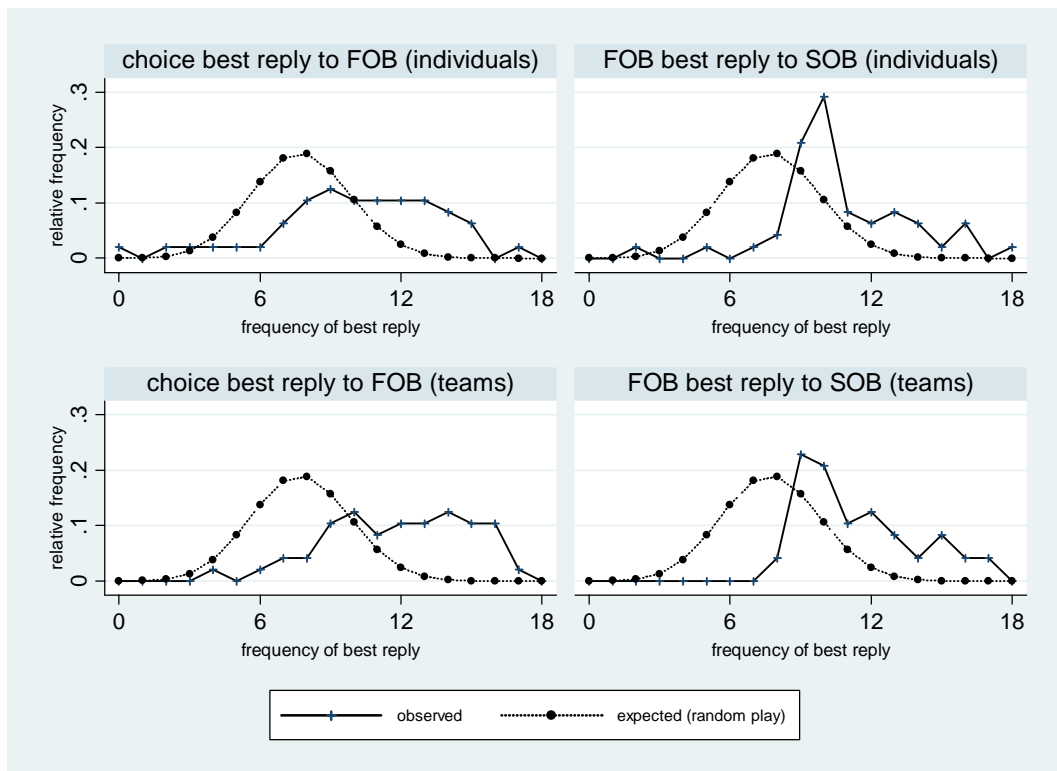


Figure F3: Frequency of Nash-consistency in the 18 games (own consistency, expected consistency of opponent) – actually observed versus random play

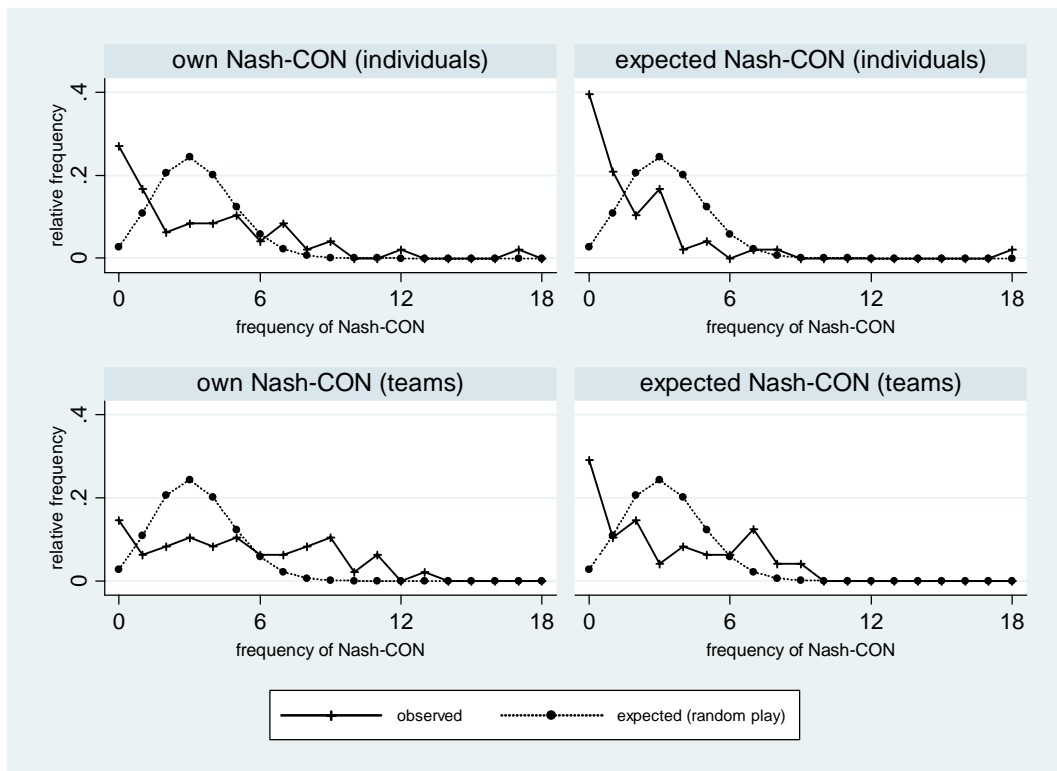


Figure F4: Best responses of different strategic and non-strategic types

game # 3 (D)		[1R, 2R]	
A, P, O, S ^I		D12, L2, E, S ^I	
A	72, 93	31, 46	
D	84, 52	55, 79	

game # 13 (D)		[1R, 2R]	
A, P, O		D12, L2, E, S	
D	94, 23	38, 57	
A	45, 89	14, 18	

game # 1 (D)		[2R, 1R]	
D12, L2, E, S		A, P, O	
	75, 51	42, 27	
	48, 80	89, 68	

game # 12 (D)		[2R, 1R]	
D12, L2, E, S		A, P, O	
	21, 92	87, 43	
	55, 36	16, 12	

game # 7 (D)		[2R, 1R]	
D12, L2, E, S		A, P, O	
	59, 58	46, 83	85, 61
	38, 29	70, 52	37, 23

game # 11 (D)		[2R, 1R]	
D12, L2, E, S		A, P, O	
	31, 32	68, 46	
	72, 43	47, 61	
	91, 65	43, 84	

game # 9 (D)		[1R, 2R]	
D12, L2, E, S		A, P, O	
	28, 37	57, 58	
A	22, 36	60, 84	
D	51, 69	82, 45	

game # 16 (D)		[1R, 2R]	
D12, L2, E, S ^I		P, A, O, S ^I	
D	42, 64	57, 43	80, 39
A	28, 27	39, 68	61, 87

game # 6 (ND)		[3R, 2R]	
A, P, O, S ^I		D12, L2, E, S ^I	
A	53, 86	24, 19	
P, O, D1, L2, S	79, 57	42, 73	
D2, E	28, 23	71, 50	

game # 15 (ND)		[3R, 2R]	
A, P, O, S ^I		D12, L2, E, S ^I	
A	76, 93	25, 12	
D2, E	43, 40	74, 62	
P, O, D1, L2, S	94, 16	59, 37	

game # 14 (ND)		[2R, 3R]	
D12, L2, E, S		A, P, O, D1, D2, S	
	21, 26	52, 73	75, 44
	88, 55	25, 30	59, 81

game # 2 (ND)		[2R, 3R]	
D12, L2, E		A, P, O, S	
	42, 45	95, 78	18, 96
	64, 76	14, 27	39, 61

game # 8 (ND)		[∞R, ∞R]	
L2, E, S		A, P, O, D12, S	
	87, 32	18, 37	63, 76
A, P, O, D12	65, 89	96, 63	24, 30

game # 10 (ND)		[∞R, ∞R]	
L2, E, S		A, P, O, D12	
	67, 91	95, 64	31, 35
	89, 49	23, 53	56, 78

game # 5 (ND)		[∞R, ∞R]	
L2, E, S ^I		A, P, O, D12, S ^I	
E	72, 59	26, 20	
A, P	33, 14	59, 92	
O, D12, L2, S	28, 83	85, 61	

game # 4 (ND)		[∞R, ∞R]	
L2, E, S		A, P, O, D12	
A, P	46, 16	57, 88	
E	71, 49	28, 24	
O, D12, L2, S	42, 82	84, 60	

game # 17 (D)		[1R, 2R]	
D12, L2, E, S		A, P, O	
	22, 14	57, 55	
	30, 42	28, 37	
A	15, 60	61, 88	
D	45, 66	82, 31	

game # 18 (D)		[2R, 1R]	
D12, L2, E, S		A, P, O	
	56, 58	38, 29	89, 62
	15, 23	43, 31	61, 16
			32, 86
			67, 46

A ... altruistic; P ... pessimistic; O ... optimistic; L2, D1, D2 ... strategic types (see section 3.6 in paper); E ... equilibrium; S ... sophisticated (S^I ... sophisticated against individuals; S^T ... sophisticated against teams); D12 = D1 and D2.

Experimental instructions

In the following we present the instructions for the individual treatment as a baseline and indicate parts which were used for the team treatment by underlined letters and parts which were used for the mixed treatment by italic letters.

18 games in which two randomly chosen cabins are interacting with each other

In this experiment you will make decisions in 18 different “matrix-games”. In each „matrix-game“ **two different cabins are interacting** with each other. From a total of eight cabins, in every game two cabins are randomly selected to interact with each other. *Out of those two cabins there is always one cabin with three participants and one cabin with one participant. This means in this experiment interaction takes place between teams of three participants and single players. It is randomly selected whether you are a member of a team or a single player by drawing cards at the beginning of the experiment.*

In each “matrix game” both cabins have **up to four choices**. In most games the quantity of choices is two or three. All choices in your own cabin are indicated with the following signs:

- §
- %
- @
- μ

Choices in the other cabin are indicated with the following signs:

- §§
- %%
- @@
- μμ

The combination of choices of both randomly selected cabins decides about your potential payment in the experiment.

Three types of screens in each game (three decisions in each game)

For each of the 18 games you will be randomly matched with another cabin. Within each game you need to make three different choices on three different screen types. For those three decisions the matching of the cabins remains constant.

Please take care of the fact that the sequence of those three decisions is randomly determined. Because of this you always need to pay attention to the upper row placed above the matrices. On the following three pages you will find examples of the three different screen-types (decisions). Note that the numbers within the matrices are just examples.

Payment

During the experiment you will have to make $18 * 3$ decisions. At the end of the experiment one of the 18 games and one of the corresponding three decisions will be randomly assigned to be relevant for your payment.

This random selection is done by a draw from a deck of cards showing numbers from 1 to 18 (to select one of the 18 games) and from a deck of cards showing numbers 1 to 3 (to select one of the three decisions). You will draw the card by **yourself**. In the following you will find an explanation how much you can earn.

Independently from your decisions you will receive a show up fee of **4 Euros**.

Feedback

Until the end of period 18 you will not get any feedback about the choices of the other cabins. As soon as the game and the screen-type (decision) which will be relevant for payment are selected, you will get detailed information about the corresponding decisions and the payment resulting from these decisions.

Teams in cabins

*In the case that there are three participants in your cabin you need to agree on **one common team decision**. At the end of the experiment **each team member gets the payment as explained in the instructions**.*

Teams in cabins

In this experiment there is a team consisting of three members in each cabin. The team members are required to agree on **one common team decision**. At the end of the experiment **each team member gets the payment as explained in the instructions**.

Screen type A. What is your choice?

Periode 1 von 1 Verbleibende Zeit [sec]: 19

Wie ist Ihre Wahl?

Here you can see the actual period.

Please take care of this first row to identify the screen type and the kind of decision you need to make.

Ihre Gewinne				
	\$\$	%%	@@	μμ
\$	65	52	x	x
%	58	101	x	x
@	x	x	x	x
μ	x	x	x	x

Gewinne des anderen Spielers				
	\$\$	%%	@@	μμ
	51	27	x	x
	80	68	x	x
	x	x	x	x
	x	x	x	x

In this row you can see the options of the other cabin.

In this column you can see your options. In the case that there are only "x-signs" in a certain row, this row cannot be selected.

In this box you can find your potential profits.

In this box you can find the potential profits of the other cabin.

On the left hand side of the screen you can find your potential profit, depending on your own choice (in this example „\$“ oder „%“) and on the choice of the cabin that is matched with you in a certain period (in this example „\$\$“ oder „%%“).

Lets assume that you choose the option „%“, while the other cabin chooses the option „\$\$“, in this particular case you would earn 58 experimental points while the other cabin earns 80 experimental points.

(Please take care that the screen for the other cabin is transposed (looks contrariwise). If you have any questions about this, please ask the instructor of the experiment)

In the case that at the end of the experiment screen type A is randomly selected to be relevant for your payment, your profit (measured in experimental points) will be determined by the combination of your own choice and the choice of the other cabin.

The relevant exchange rate between experimental points and Euros is:

$$1 \text{ experimental point} = 0.30 \text{ Euros}$$

In the example illustrated above this would result in a payoff of 14.4 Euros (plus show up fee) for yourself and 24 Euros (plus show up fee) for the other cabin.

Screen type B. According to your opinion, which option will be chosen by the other cabin?

Periode 1 von 1 Verbleibende Zeit [sec]: 26

Welche Wahl wird der Andere Ihrer Meinung nach treffen?

Please take care of this first row to identify the screen type and the kind of decision you need to make.

Ihre Gewinne

	\$\$	%%	@@	μμ
\$	65	52	x	x
%	58	101	x	x
@	x	x	x	x
μ	x	x	x	x

Gewinne des anderen Spielers

	\$\$	%%	@@	μμ
\$	51	27	x	x
%	80	68	x	x
@	x	x	x	x
μ	x	x	x	x

Here you need to indicate your opinion regarding the other cabin's choice.

In the case of screen type B you need to correctly predict the decision of the other cabin. In order to do that you need to select the choice which you expect from the other cabin in the screen-area as explained above.

In the case that at the end of the experiment screen type B is randomly selected to be relevant for your payment, your profit will be **15 Euros** if your guess was **correct**. To assess whether your guess was correct we will compare your expectation indicated on this screen with the real choice of the other cabin.

Screen type C. What is the choice that the other player expects from you?

Periode 1 von 1
Verbleibende Zeit [sec]: 18

Welche Wahl wird der Andere Ihrer Meinung nach von Ihnen erwarten?

Ihre Gewinne

	\$\$	%%	@@	μμ
\$	65	52	x	x
%	58	101	x	x
@	x	x	x	x
μ	x	x	x	x

Gewinne des anderen Spielers

	\$\$	%%	@@	μμ
\$	51	27	x	x
%	80	68	x	x
@	x	x	x	x
μ	x	x	x	x

Please take care of this first row to identify the screen type and the kind of decision you need to make.

Here you need to indicate your guess about the expectation of the other cabin regarding your own choice.

In the case of screen type C you need to correctly predict the other cabin's expectation regarding your own choice.

On screen type B (which is explained on the previous page) you need to correctly predict the decision of the other cabin. For each game the other cabin needs to make the same prediction which means that the other cabin needs to state an expectation regarding your own choice. On screen type C you need to indicate which choice the other cabin expects from you.

In the case that at the end of the experiment screen type C is randomly selected to be relevant for payment, your profit will be **15 Euros** if your guess regarding the expectation of the other cabin was **correct**.

To assess whether your guess was correct we will compare your decision on this screen (which is your guess regarding the expectation of the other cabin) with the real expectation of the other cabin.

Control questions

Please answer the following questions which are all related to the matrices of the three screen-types explained before.

Screen-type A.

- 1) Let's assume you choose option „§“ and the other cabin chooses option „%%“:
 - a. How many experimental points do you earn? _____
 - b. How many experimental points does the other cabin earn? _____
- 2) Let's assume you choose option „%“ and the other cabin chooses option „§§“:
 - a. How many experimental points do you earn? _____
 - b. How many experimental points does the other cabin earn? _____

Screen-type B.

- 3) Let's assume you expect the other cabin to choose option „%%“ and the other cabin chooses option „§§“.
 - a. How much do you earn (show up fee excluded), if this screen-type (decision) is selected to be relevant for payment? _____
- 4) Let's assume you expect the other cabin to choose option „§§“ and the other cabin chooses option „§§“.
 - a. How much do you earn (show up fee excluded), if this screen-type (decision) is selected to be relevant for payment? _____

Screen-type C.

- 5) Let's assume the other cabin expects you to choose option „%“ and let's assume you select option „%“ on screen-type C.
 - a. How much do you earn (show up fee excluded), if this screen-type (decision) is selected to be relevant for payment? _____
- 6) Let's assume the other cabin expects you to choose option „§“ and let's assume you select option „%“ on screen-type C.
 - a. How much do you earn (show up fee excluded), if this screen-type (decision) is selected to be relevant for payment? _____

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Matthias Sutter, Simon Czermak and Francesco Feri

Strategic sophistication of individuals and teams in experimental normal-form games

Abstract

We present an experiment on strategic thinking and behavior of individuals and teams in one-shot normal-form games. Besides making choices, decision makers have to state their first- and second-order beliefs. We find that teams play the Nash strategy significantly more often, and their choices are more often consistent by being a best reply to first order beliefs. We identify the complexity of a game and the payoffs in equilibrium as determining the likelihood of consistent behavior according to textbook rationality. Using a mixture model, the estimated probability to play strategically is 62% for teams, but only 40% for individuals.

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