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Altruism in Groups: An Evolutionary Games Approach

Ilaria Brunetti†, Rachid El-Azouzi⋆ and Eitan Altman†

Abstract—We revisit in this paper the relation between evolution of species and the mathematical tool of evolutionary games, which has been used to model and predict it. We indicate known shortcoming of this model that restricts the capacity of evolutionary games to model groups of individuals that share a common gene or a common fitness function. In this paper we provide a new concept to remedy this shortcoming in the standard evolutionary games in order to cover this kind of behavior. Further, we explore the relationship between this new concept and Nash equilibrium or ESS. We indicate through the study of some example in the biology as Hawk-Dove game, Stag-Hunt Game and Prisoner’s Dilemma, that when taking into account a utility that is common to a group of individuals, the equilibrium structure may change dramatically. We also study the multiple access control in slotted Aloha based wireless networks. We analyze the impact of the altruism behavior on the performance at the equilibrium.

I. INTRODUCTION

Evolutionary games become a central tool for predicting and even design evolution in many fields. Its origins come from biology where it was introduced by [7] to model conflicts among animals. It differs from classical game theory by its focusing on the evolution dynamics of the fraction of members of the population that use a given strategy, and in the notion of Evolutionary Stable Strategy (ESS, [7]) which includes robustness against a deviation of a whole (possibly small) fraction of the population who may wish to deviate. This is in contrast with the standard Nash equilibrium concept that only incorporates robustness against deviation of a single user. It became perhaps the most important mathematical tool for describing and modeling evolution since Darwin. Indeed, on the importance of the ESS for understanding the evolution of species, Dawkins writes in his book "The Selfish Gene" [4]: "we may come to look back on the invention of the ESS concept as one of the most important advances in evolutionary theory since Darwin." He further specifies: "Maynard Smith’s concept of the ESS will enable us, for the first time, to see clearly how a collection of independent selfish entities can come to resemble a single organised whole. In this paper, we identify inherent restrictions on the modeling capacity of classical evolutionary games apply. Recently, the evolutionary game theory has become of increased interest to social scientists [5]. In computer science, evolutionary game theory is appearing, some examples of applications can be found in multiple access protocols [10], multihoming[11] and resources competition in the Internet [12].

The starting point of this theory is a situation of a very large number of local pairwise interactions between pairs of individuals that are randomly matched. In classical evolutionary games (EG), each individual represents a selfish player in a non-cooperative game that it plays with its randomly matched adversary, and in which each player seeks to maximize its utility. The originality of EG is in postulating that this utility is the Darwinian fitness. The fitness should be understood as the relative rate at which the individual will increase. For a given share of behaviors in the population, a behavior of an individual with a higher fitness would thus result in a higher rate of its reproduction. We make the observation that since classical EG associates with an individual both the interactions with other individuals as well as the fitness, then it is restricted to describing populations in which the individual is the one that is responsible for the reproduction and where the choice of its own strategies is completely selfish. In biology, in some species like bees or ants, the one who interacts is not the one who reproduces. This implies that the Darwinian fitness is related to the entire swarm and not to a single bee and thus, standard EG models excludes these species in which the single individual which reproduces is not the one that interacts with other individuals. Furthermore, in many species, we find altruistic behaviors, which may hurt the individual adopting it, favouring instead the group he belongs to. Altruistic behaviors are typical of parents toward their children: they may incubate them, feed them or protect them from predator’s at a high cost for themselves. Another example can be found in flock of birds: when a bird sees a predator it gives an alarm call to warn the rest of the flock, attracting the predators attention to itself. Also the stinging behavior of bees is an altruistic one: it serves to protect the hive, but its lethal for the bee which strives. In human behavior, many phenomena where individuals do care about other’s benefits in their groups or about their intentions can be observed in the real word. It must be admitted that some phenomena require an explanation in terms of genes which pursue their own interest to the disadvantage of the individual. Hence the assumption of selfishness becomes inconsistent with the real behavior of individual in a population.

Founders of classical EG seem to have been well aware of this problem. Indeed, Vincent writes in [15] "Ants seem to completely subordinate any individual objectives for the good of the group. On the other hand, the social foraging
of hyenas demonstrates individual agendas within a tight-knit social group (Hofer and East, 2003). As evolutionary games, one would ascribe strategies and payoffs to the ant colony, while ascribing strategies and payoffs to the individual hyenas of a pack.” In the case of ants, the proposed solution is thus to model the ant colony as a player. Within the CEG paradigm, this would mean that we have to consider interactions between ant colonies. This however does not allow us anymore to model behavior at the level of the individual.

In this work we present a new model for evolutionary games, in which the concept of the agent as a single individual is substituted by that of the agent as a whole group of individuals. This new concept, named Group Equilibrium Stable Strategy (GESS), allow to model competition between individuals in a population in which the whole group shares a common utility. Even if we still consider pairwise interactions among individuals, our perspective is completely different: we suppose that individuals are simple actors of the game and that the utility to be maximized is the one of their group. Our study of evolutionary games under the altruism inside each group, is built around the GESS. We begin by defining the GESS, deriving it in several ways and exploring its major characteristics. The main focus of this paper is to study how this new concept changes the profile of population and to explore the relationship between GESS and Nash equilibrium or ESS. We characterize through the study of many GESS and we show how the evolution and the equilibrium are influenced by the groups’ size as well as by their immediate payoff. We also provide some primary results through an example on multiple access games, in which any local interaction does not lead to same payoff depending on the type of individuals that are competing, and not only the strategy used. In such application, we evaluate the impact of altruism behavior on the performance of the system.

The paper is structured as follows. We first provide in the next section the needed background on evolutionary games. In the section III we then study the new natural concept GESS and the relationship between GESS and ESS or Nash equilibrium. The characterization of the GESS is studied in section IV. Section V provides some numerical illustration through some famous examples in evolutionary games. In section VI we study the multiple access control in slotted Aloha under altruism behavior. The paper closes with a summary in section VII

II. CLASSICAL EVOLUTIONARY GAMES AND ESS

We consider an infinite population of players and we assume that each member of the population has the same set of available pure strategies \( K = \{1, 2, \ldots, n\} \). We suppose that each individual is repeatedly paired off with an other individual randomly selected within the population. A player may use a mixed strategy \( p \in \Delta(K) \) where \( \Delta(K) = \{ p \in \mathbb{R}^n_+ | \sum_{i \in K} p_i = 1 \} \). Here \( p \) is a probability measure over the set of actions \( K \). This is the case where an individual has the capacity to produce a variety in behaviours. Alternatively, the mixed strategy \( p \) can be interpreted as the vector of densities of individuals adopting a certain pure strategy, where \( p_i \) is the fraction of the population using strategy \( i \in K \). However the original formulation of evolutionary game theory were not required to make distinction between population-level and individual-level variability for infinite population [7].

Let now focus on the case of monomorphic populations in which each individual uses a mixed strategy. We define by \( J(p, q) \) the expected payoff for a tagged individual if it uses a mixed action \( p \) when meeting another individual who adopts the mixed action \( q \). This payoff is called "fitness" and actions with larger fitness are expected to propagate faster in a population. If we define a payoff matrix \( A \) and consider \( p \) and \( q \) to be column vectors, then \( J(p, q) = p' A q \) and the payoff function \( J \) is indeed linear in \( p \) and \( q \). A mixed action \( q \) is called a Nash equilibrium if

\[
\forall p \in \Delta(K), \quad J(q, q) \geq J(p, q)
\]

(1)

In evolutionary games the most important concept of equilibrium is the ESS, which was introduced by [7] as a strategy that, if adopted by most members of a population, it is not invadable by mutant strategies in its suitably small neighbourhood. More precisely, we suppose that the whole population uses a strategy \( q \) and that a small fraction \( \epsilon \) of individuals (mutants) adopts another strategy \( p \). Evolutionary forces are expected to select \( q \) against \( p \) if

\[
J(q, \epsilon p + (1 - \epsilon)q) > J(p, \epsilon p + (1 - \epsilon)q)
\]

(2)

The definition of ESS is thus related to a robustness property against deviations by a whole (possibly small) fraction of the population. This is an important difference that distinguishes the equilibrium in populations as seen by biologists and the standard Nash equilibrium often used in economic context, in which robustness is defined against the possible deviation of a single user. Why do we need the stronger type of robustness? Since we deal with large populations, it is likely to expect that from time to time, some group of individuals may deviate. Thus robustness against deviations by a single user is not sufficient to ensure that deviations will not develop and end up being used by a growing portion of the population. By defining the ESS through the following equivalent definition [16, Proposition 2.1] or [6, Theorem 6.4.1, page 63], it’s possible to establish the relationship between ESS and Nash Equilibrium (NE). Strategy \( q \) is an ESS if it satisfies the two conditions:

- Nash equilibrium condition:
  \[
  J(q, q) \geq J(p, q) \quad \forall p \in K.
  \]
  (3)

- Stability condition:
  \[
  J(p, q) = J(q, q) \Rightarrow J(p, p) < J(q, p). \quad \forall p \neq q
  \]
  (4)

The first condition (3) is the condition for a Nash equilibrium. In fact, if condition (3) is satisfied, then the fraction of mutations in the population will tend to decrease (as it has a lower fitness, meaning a lower growth rate). Thus the action
q is then immune to mutations. If it does not but if still the condition (4) holds, then a population using q is "weakly" immune against mutations using p. Indeed, if the mutant’s population grows, then we shall frequently have individuals with action q competing with mutants. In such cases, the condition \( J(p, p) < J(q, p) \) ensures that the growth rate of the original population exceeds that of the mutants. Then an ESS is a refinement of the Nash equilibrium.

### III. NEW NATURAL CONCEPT ON EVOLUTIONARY GAMES

In this section we present a new concept for evolutionary games, in which the idea of the player as a single individual is substituted by that of a player as a whole group of individuals. The interactions are among individuals but the objective function, which is maximized, is that of the group they belong to. We assume that the population is composed of \( N \) groups, \( G_i, i = 1, 2, \ldots, N \), where the normalized size of \( G_i \) is noted by \( \alpha_i \) with \( \sum_{i=1}^{N} \alpha_i = 1 \).

For clarity of presentation, we restrict our analysis to pairwise interactions, where each group of individuals can compete with a member of its own group or another group of a different type. Individuals dispose of a finite set of actions: \( \mathcal{K} = \{a_1, a_2, \ldots, a_M\} \). Let \( p_{ik} \) be the probability that an individual in the group \( G_i \) chooses an action \( a_k \) in \( \mathcal{K} \); we associate to each group \( i \) the vector of probabilities \( p_i = (p_{i1}, p_{i2}, \ldots, p_{iM}) \) where \( \sum_{j=1}^{M} p_{ij} = 1 \). By assuming that each individual can interact with another individual with equal probability, then the expected utility of a group (player) \( i \) is:

\[
U_i(p_i, p_{-i}) = \sum_{j=1}^{N} \alpha_j J(p_i, p_j),
\]

where \( p_{-i} \) is the profile strategy of other groups and \( J(p_i, p_j) \) is the immediate expected reward of an individual player adopting strategy \( p_i \) against an opponent playing \( p_j \).

#### A. Group Equilibrium Stable Strategy

The definition of GESS is related to the robustness property against deviations inside each group. There are two possible equivalent interpretations of an \( \epsilon \)-deviation in this context:

1. A small deviation in the strategy by all members of a group. If the group \( G_i \) plays according to strategy \( q_i \), the \( \epsilon \)-deviation, where \( \epsilon \in (0, 1) \), consists in a shift to the group’s strategy \( p_i = \epsilon q_i + (1-\epsilon)q_i \);

2. The second is a deviation (possibly) large of a small number of individuals in a group \( G_i \), that means that a fraction \( \epsilon \) of individuals in \( G_i \) plays a different strategy \( p_i \).

After an \( \epsilon \)-deviation under both interpretations the profile of the whole population becomes \( \alpha_i \epsilon q_i + \alpha_i(1-\epsilon)q_i + \sum_{j\neq i} \alpha_j q_j \). Then the average payoff of group \( G_i \) after mutation is given by:

\[
U_i(p_i, q_{-i}) = \sum_{j=1}^{N} \alpha_j J(p_i, p_j)
\]

\[
= U_i(q_i, q_{-i}) + \epsilon^2 \alpha_i \Omega(p_i, q_i) + \epsilon \alpha_i (J(p_i, q_i) + J(q_i, q_i) - 2J(q_i, q_i))
\]

\[
+ J(q_i, q_i) - 2J(q_i, q_i) + \sum_{j \neq i} (J(p_j, q_i) - J(q_i, q_i))
\]

where \( \Omega(p_i, q_i) := J(p_i, p_i) - J(q_i, q_i) + J(q_i, q_i) \).

**Definition 1.** A strategy \( q = (q_1, q_2, \ldots, q_N) \) is a GESS if \( \forall i \in \{1, \ldots, N\}, \forall p_i \neq q_i \), there exists some \( \epsilon_p \in (0, 1) \), which may depend on \( p_i \), such that:

\[
U_i(p_i, q_{-i}) < U_i(q_i, q_{-i}),
\]

Hence from equation (7), strategy \( q \) is a GESS if the two following conditions hold:

- \( \forall p_i \in [0, 1]^M \)

\[
F_i(p_i, q_i) := \alpha_i \Omega(p_i, q_i) - U_i(p_i, q_{-i}) + U_i(q_i, q_{-i}) \geq 0
\]

- \( \exists \epsilon_p \neq q_i \) such that:

\[
F_i(p_i, q_i) = 0 \Rightarrow \Omega(p_i, q_i) < 0 \]

**Remark 1.** The condition (9) can be rewritten as

\[
U_i(q_i, q_{-i}) > U_i(p_i, q_{-i})
\]

which is exactly the definition of the strict Nash equilibrium of the game composed by \( N \) groups in which each group maximises its own utility.

#### B. GESS and standard ESS

Here we analyse the relationship between the standard ESS and our new concept GESS.

**Proposition 1.** Consider games whose immediate expected reward is symmetric, i.e. \( J(p, q) = J(q, p) \). Then any ESS is a GESS.

**Proof:** Let \( q = (q_1, \ldots, q) \) be an ESS. Combining the symmetry of the payoff function and equation (8), we get:

\[
F_i(p_i, q) = -\left(\alpha_i (J(p_i, p_i) + J(q, p_i) - 2J(q, q)) + \sum_{j \neq i} \alpha_j (J(p_j, p_i) - J(q_j, q_i))\right)
\]

\[
= -2\alpha_i(J(p_i, p_i) - J(q_i, q_i)) - \sum_{j \neq i} \alpha_j(J(p_j, q_i) - J(q_j, q_i))
\]

\[
= -(1 + \alpha_i)(J(p_i, q_i) - J(q_i, q_i)) \geq 0
\]

where the second equality follows from the symmetry of the payoff function \( J \) and the last inequality follows form the fact that \( q \) is an ESS and satisfies (3). This implies that \( q \) satisfies the first condition of GESS (8). Now assume that \( F_i(p_i, q) = 0 \) for some \( p_i \neq q_i \), previous equations imply that \( J(p_i, q_i) = J(q_i, q_i) \). Thus the second condition (9) becomes \( \Omega(p_i, q_i) = J(p_i, q_i) - J(q_i, q_i) < 0 \) which coincide with the second condition of ESS (4). This completes the proof.
C. Nash equilibrium and GESS

In the classical evolutionary games, the ESS is a refinement of a Nash equilibrium and we can see that all ESSs are Nash equilibria but not all Nash equilibria are ESSs. In order to characterize this relationship in our context, let us define the game between groups: There are $N$ players in which each player has a finite set of pure strategies $K = \{1, 2, ..., m\}$. We define by $U_i(q_i, q_{-i})$ the utility of player $i$ when using mixed strategy $q_i$ against a population of players using $q_{-i} = (q_1, q_{-1}, q_{-2}, ..., q_N)$.

**Definition 2.** A strategy $q = (q_1, q_2, ..., q_N)$ is a Nash Equilibrium if for every other mixed strategy $p_i \neq q_i$. If it holds for strict inequality, then $q$ is a strict Nash equilibrium.

From the definition of the strict Nash equilibrium, it is easy to show that any strict Nash equilibrium is a GESS defined in equation (7). But in our context, we address several questions on the relationship between the GESS, ESS and the Nash equilibrium defined in (10). For simplicity of presentation, we restrict to the case of two-strategies games. Before studying them, we introduce here some definitions that are needed in the sequel.

**Definition 3.**
- A fully mixed strategy $q$ is a strategy such that all actions of each group have to receive a positive probability, i.e., $0 < q_i < 1 \forall(i, j) \in I \times K$.
- A mixer (pure) group $i$ is the group that uses a mixed (pure) strategy $0 < q_i < 1$ (resp. $q_i \in \{0, 1\}$).
- An equilibrium with mixed and non mixed strategies is an equilibrium in which there is at least one pure group and a mixer group.

IV. ANALYSIS OF $N$-GROUPS GAMES WITH TWO STRATEGIES

We will discuss here $N$-groups games with two strategies. The two possible pure strategies are $A$ and $B$ and the pairwise interactions payoff matrix is given by:

$$
P = \begin{pmatrix}
A & B \\
A & B
\end{pmatrix}
$$

where $P_{ij}, i, j = A, B$ is the payoff of the first (row) individual if it plays strategy $i$ against the second (column) individual playing strategy $j$. We assume that both individuals are the same and hence payoffs of the column player are given by the transposed of $P$. According to the definition of GESS, $q$ is a GESS if it satisfies the conditions (8)-(9), which can be rewritten as:

- $\forall q_i \in [0, 1], i = 1, ..., N$:
  $$F_i(p_i, q) = (q_i - p_i)(\alpha_i(J(q_i, 1) - J(q_i, 0)) + \sum_{j=1}^{N} \alpha_j(J(1, q_j) - J(0, q_j))) \geq 0$$
  $$\tag{11}$$
- If $F(p_i, q) = 0$ for some $p_i \neq q_i$, then:
  $$\Delta = (p_i - q_i)^2 \Delta < 0 \implies \Delta < 0 \tag{12}$$

A. Characterisation of fully mixed GESS

In this section we are interested in characterising the full mixed GESS $q$. According to (11), a full mixed equilibrium $q = (q_1, ..., q_N)$ is a GESS if it satisfies the condition (12) where the equality must holds for all $p_i \in [0, 1]$. This yields to the following equation: $\forall i = 1, ..., N$,

$$\alpha_i(J(q_i, 1) - J(q_i, 0)) + \sum_{j=1}^{N} \alpha_j(J(1, q_j) - J(0, q_j)) = 0$$

which can be rewritten as

$$\alpha_i \Delta q_i + b - d + \alpha_i(c - d) + \sum_{j=1}^{N} \alpha_j g_j = 0$$

This leads to the following expression of the mixed GESS:

$$q_i^* = \frac{d - b + (1 + N)\alpha_i - 1(N - c)}{(N + 1)\alpha_i \Delta} \tag{13}$$

**Proposition 2.** If $\Delta < 0$ and $0 < q_i^* < 1, i = 1, ..., N$, then there exists a unique fully mixed GESS equilibrium given by (13).

We note that the fully mixed GESS is a strict Nash equilibrium since the condition (12) is equivalent to the definition of the strict Nash equilibrium (see remark 1) under the condition $F(p_i, q_1, ..., q_N) = 0, \forall p_i \in [0, 1]$.

B. Characterisation of strong GESS

We call a strong GESS an equilibrium that satisfies the strict inequality (11) for all groups. Similarly to the fully mixed GESS, we present here the condition for the existence of a strong GESS. Without loss of generality, we assume that a pure strong GESS can be represented by $n_A$, where $n_A \in \{1, ..., N\}$ denotes that the $n_A$ first groups use $A$ pure strategy and remaining $N - n_A$ groups chose strategy $B$. For example $n_A = N$ (resp. $n_A = 0$) means that all groups choose pure strategy $A$ (resp. $B$).

**Proposition 3.** If $a \neq c$ or $b \neq d$, then every $N$-player game with two strategies has a GESS. There are the following possibilities for the strong GESS:

i. If $a - c > \max_i(\alpha_i) \cdot (b - a)$ then $n_A = N$ is a strong GESS;

ii. If $b - d < \min_i(\alpha_i) \cdot (d - c)$ then $n_A = 0$ is a strong GESS;

iii. Let $H(n_a) := \sum_{j=1}^{n_A} \alpha_j(a - c) + \sum_{j=n_A+1}^{N} \alpha_j(b - d)$. If $\alpha_i(d - c) < H(n_a) > \alpha_i(a - b)$ then $n_A$ is a strong GESS.

Proof: In order to prove that a strategy $n_A = N$ is a GESS, we have to impose the strict inequality, i.e.: $\forall p_i \neq 1$ for $i \in \{1, ..., n_A\}$ and $\forall p_i \neq 0$ for $i \in \{n_A + 1, ..., N\}$

$$F_i(p_i, 1_{n_A}, 0_{N - n_A}) > 0$$
We show here conditions of the existence only for \( n_A = N \) since the others one straightforward follow from the symmetry of the players in the game.

Consider now that \( (n_A = N) \) is a strong GESS. The inequality (11) becomes: \( \forall p_i \neq 1, \forall i \)

\[
(p_i - 1) \left( \alpha_i (a-b) + \sum_{j=1}^{N} \alpha_j (a-c) \right) = (p_i - 1) \left( \alpha_i (a-b) + a-c \right) < 0,
\]

Since \( p_i < 1 \), one has \( \alpha_i (b-a) < a-c \forall i \). This completes the proof of (i). To show conditions of the other strong GESSs, we follow the lines of the proof of (i).

C. Characterisation of weak GESS

We call a weak GESS an equilibrium in which at least one group uses a strategy that satisfies the condition (12) with equality. Here we distinguish two types of equilibrium: the equilibrium with no mixer group and the equilibrium with mixer and no mixer groups. Conditions for the equilibrium with no mixed strategy are given by Proposition 3 with at least one group satisfying it with equality and \( \Delta < 0 \). In this section we focus only on the equilibrium with mixer and no mixer groups. Without loss of generality, we assume that an equilibrium with mixed and non mixed strategies, can be represented by \((n_A, n_B, q)\) where \( n_A \) denotes that group \( i \) for \( i = 1..n_A \) (resp. \( i = n_A + 1,..., n_A + n_B \)) uses strategy \( A \) (resp. \( B \)). The remaining groups \( N - n_A - n_B \) are mixers in which \( q_i \) is the probability to choose the strategy \( B \) by group \( i \).

**Proposition 4.** Let either \( a \neq c \) or \( b \neq d \) and \( \Delta < 0 \). \((n_A, n_B, q)\) is a weak GESS if:

\[
\begin{align*}
\alpha_i \Delta + d - b + \alpha_i (c-d) + \Delta (\alpha_{n_A} + y) & \geq 0, \quad i = 1,..,n_A \\
\sum_{i=1}^{n_A} \alpha_i + (d-c) & + \Delta (\alpha_{n_A} + y) & \leq 0, \quad i = n_A + 1,..,n_B \\
q_i & = \frac{d + b(\alpha(c - d) + \Delta \alpha)}{\Delta}, \quad i = n_A + n_B + 1,..,N
\end{align*}
\]

where \( y = \frac{(N-n_A-n_B)(d-b-\sum_{j=1}^{n_A} \alpha_j) + (d-c) \sum_{j=n_A+n_B+1}^{N} \alpha_j}{\Delta(N-n_A-n_B+1)} \).

**Proof:** See technical report.

V. SOME EXAMPLES

In this section we analyze a number of examples with two players and two strategies.

A. Hawk and Dove Game

One of the most studied examples in EG theory is the Hawk-Dove game, first introduced by Maynard Smith and Price in "The Logic of Animal Conflict". In this game two animals compete for a resource of a fixed value. Each animal follows one of two strategies, Hawk or Dove, where Hawk corresponds to an aggressive behavior, Dove to a non-aggressive one. If two Hawks meet, they fight and one of them gets the resource while the other is injured, with equal probability. A Hawk always wins against a Dove, whereas if two Doves meet they equally share the resource. The payoff matrix of the game is the following:

\[
\begin{pmatrix}
H & D \\
D & \frac{1}{2}(V-C) & V \\
\end{pmatrix}
\]

where \( C \) represent the cost of the fight, and \( V \) is the benefit the player get from the resource. We suppose that \( C > V \).

In standard GT, this example belongs to the anti-coordination class, whose games always have two strict, pure strategy NEs and one non-strict, mixed strategy NE. In this case the two pure equilibria are \((H, D)\) and \((D, H)\), and the mixed-one is given by: \( q^* = \frac{V}{C} \). The latter is the only ESS: even if the two pure NE are strict, being asymmetric they can’t be ESSs. We set \( V = 2 \) and \( C = 3 \) and study this game in groups framework, considering two groups of size \( \alpha \) and \( 1-\alpha \).

We obtain that the GESSs and the strict NE always coincides. In particular we observe that:

- for \( 0 < \alpha < 0.25 \) the game has one strong GESS \((H, D)\) and a weak GESS \((H, q_2)\);
- for \( 0.25 < \alpha < 0.37 \) one weak GESS \((H, q_2)\);
- for \( 0.37 < \alpha < 0.5 \) one weak GESS, \((q_1', q_2')\).

The size of groups has a strong impact on the behavior of players: in the first interval of \( \alpha \)-values we remark that the GESS is not unique; when increasing the size of the first group, and thus decreasing that of the second one, the probability that the second group plays aggressively against the pure aggressive strategy of the first one increases until we get into the third interval, where both players are mixers. Here we can clearly observe that the equilibrium \( q_1^* \) is decreasing in \( \alpha \); as we supposed that an individual can interact with members of its own group, when increasing \( \alpha \), the probability of meeting an individual in the same group increases and it leads to a less aggressive behavior.

![fig1](image1.png)

**Fig. 1.** The global level of aggressiveness in the two-groups population for the different GESSs, as a function of \( \alpha \)

B. Stag Hunt Game

We now consider a well-known example which belongs to the coordination class, the Stag Hunt game. The story behind has been described by J-J. Rousseau: two individuals go out
on a hunt; if they cooperate they can hunt a stag; otherwise, hunting alone, a hunter can only get e hare, so collaboration is rewarding for players. It represents a conflict between social and safely cooperation. The payoff matrix is the following:

\[
\begin{pmatrix}
S & H \\
S & (a & b) \\
H & (c & d)
\end{pmatrix}
\]

where \( S \) and \( H \) stand respectively for Stag and Hare and \( a > c \geq d > b \). Coordination games have two strict, pure strategy NEs and one non-strict, mixed strategy NE, respectively the risk dominant equilibrium \((H, H)\), the payoff dominant one \((S, S)\) and the mixed symmetric one with 

\[
q_1^* = q_2^* = \frac{d-b}{a-b-c+d}
\]

We set \( a = 2 \), \( b = 0 \), \( c = 1 \), \( d = 1 \) and we look for the equilibria of the two groups game as a function of \( \alpha \). We find that the strict GESSs and the strict NEs don’t coincide, as we found strategies, which are strict GESSs but not NEs. The two-groups Stag-Hunt Game only have the pure-pure strict NE \((S, S)\), for all values of \( \alpha \), whereas, for the GESSs we find that:

- for \( 0 < \alpha < 0.5 \) the game admits two pure-pure strong GESSs: \((S, S)\) and \((H, H)\);
- for \( 0.25 < \alpha < 0.5 \) the game admits three pure-pure strong GESSs: \((S, S)\) and \((H, H)\) and \((S, H)\);
- the game doesn’t admitting any strict mixed NE.

### C. Prisoner’s Dilemma

We consider another classical example in game theory, the Prisoner’s Dilemma, which belongs to a third kind of games, the pure dominance class.

Two criminals are arrested and separately interrogated: they can either accuse the other, either remain silent. If both of them accuse the other (defect), they will be both imprisoned for 2 years. If only one accuse the other, the accused is imprisoned for 3 years while the other is free. If both remain silent (cooperate), each of them will serve one year in jail.

The payoff matrix is the following:

\[
\begin{pmatrix}
C & D \\
C & (a & b) \\
D & (c & d)
\end{pmatrix}
\]

where \( C \) and \( D \) stand respectively for collaboration and defection and where \( c > a > d > b \).

In standard GT, pure dominance class games admit a unique pure, strict and symmetric NE, which also is the unique ESS; in the Prisoner’s Dilemma it’s \((D, D)\).

We set \( a = 2 \), \( b = 0 \), \( c = 3 \), \( d = 1 \) and we study the two-groups corresponding game. As in the previous example, we find strict GESSs which are not strict NEs. In particular, we have that:

- \((C, C)\) is always a GESS and a strict NE for all values of \( \alpha \);
- \((D, D)\) is a GESS for all values of \( \alpha \) but it is never a strict NE;
- \((C, D)\) (symmetrically \((D, C)\)) is always a GESS and a strict NE for \( 0.5 < \alpha < 1 \) (symmetrically \( 0 < \alpha < 0.5 \));
- the game doesn’t admit mixed GESSs;

The two groups game thus admits three pure GESSs and two pure strict NEs for all values of \( \alpha \).

### VI. Multiple Access Control

In this section we briefly introduce a refinement of our model, which will be further developed in future works. We modify the group utility function defined in \((5)\) by supposing that the immediate payoff matrix differs if the interacting individuals belong to the same group or to two different ones.

The utility function of a group \( i \) playing \( q_i \) against a population profile \( q_{-i} \) can be written as follows:

\[
U(q_i, q_{-i}) = \alpha_i K(q_i, q_i) + \sum_{j \neq i} \alpha_{ij} J(q_i, q_j),
\]

where \( K(p, q) \) indicates the immediate expected payoff of an individual playing \( p \) against a member of its own group using \( q \), whereas \( J(p, q) \) is the immediate expected payoff associated to interactions among individuals of different groups.

We present a particular application of this model in Aloha system in which a large population of mobiles interfere with
each other through many local pairwise interactions. We assume that the population is decomposed into $N$ groups $G_i$, $i = 1, 2, ..., N$ of normalized size $\alpha_i$ with $\sum_{i=1}^{N} \alpha_i = 1$. Mobiles are randomly placed over a plane and matched through pairwise interactions where each mobile decides either to transmit ($T$) or to not transmit ($S$) a packet to a receiver when they are within transmission range of each other. Interferences occur as in the Aloha protocol: if more than one transmission within a distance $s$ satisfied: cause interference to

is a GESS if

$q_i$ where

$0 < \delta < 1$ is the cost of transmission. The definition of $P_i$ implies that when two mobiles of the same group $i$ interact, any successful transmission is equally rewarding for the group $i$. The resulting expected payoff functions of a mobile playing $q_i$ against a member belonging to its own group and to a different one, using respectively the same strategy $q_i$ and a different one $q_j$ are the following:

$$K(q_i, q_j) = \mu[q_i\gamma(1 - \delta) + (1 - \gamma)(1 - \delta)(1 - q_i) - 2\delta q_i] + (1 - \gamma)(1 - \delta)(1 - q_i)q_j]$$

$$J(q_i, q_j) = \mu q_i[\gamma(1 - \delta) + (1 - \gamma)(1 - \delta)(1 - q_j) - \delta q_i]$$

The expected throughput of group $i$ is then given by:

$$U(q_i, q_{-i}) = \mu q_i[1 - \delta + (1 - \gamma)\alpha_i(1 - \delta - q_j) - \sum_{j=1}^{N} \alpha_j q_j]$$

By following the same analysis as in section III, the strategy $q_i$ is a GESS if $\forall i, 1, \ldots, N$ the two following conditions are satisfied:

1) $F_i(q_i, q_j) = (q_i - p_i)[1 - \delta + (1 - \gamma)(\alpha_i(1 - \delta - 2q_i) - \sum_{j=1}^{N} \alpha_j q_j)] \geq 0 \ \forall p_i,$

2) If $F_i(q_i, q_j) = 0$ for some $p_i \neq q_j$, then $(p_i - q_j)^2(1 - \gamma)\alpha_i > 0 \ \forall p_i \neq q_j.$

We observe that the inequality $(p_i - q_j)^2(1 - \gamma)\alpha_i > 0$ holds for all values of the parameters which implies that the second condition is always satisfied and that the first condition is sufficient to guarantee the existence of a GESS. In the following proposition we give a characterization of the GESSs of the presented MAC game. Without loss of generality, we reorder the groups so that $\alpha_1 \leq \alpha_2 \ldots \leq \alpha_N$.

**Proposition 5.** We find that:

- The pure symmetric strategy $(S, \ldots, S)$ is never a GESS.
- If a fixed group $G_i$ adopts pure strategy $T$, then at the equilibrium, all smaller groups transmit. If the bigger group $G_N$ use strategy $T$ at the equilibrium, then $\gamma > \gamma_i$.
- If a fixed group $G_i$ adopts pure strategy $S$, then at the equilibrium, all smaller groups also use $S$.
- If a fixed group $G_i$ adopts an equilibrium mixed strategy $q_i \in [0, 1]$, then if $q_i > \frac{1 - \delta}{2\gamma}$, at the equilibrium all smaller groups use pure strategy $T$, whereas if $q_i < \frac{1 - \delta}{2\gamma}$, smaller groups play $S$.
- The game admits a unique fully mixed GESS $q^* = (q_1^*, \ldots, q_N^*)$, given by:

$$q_i^* = \frac{(1 - \delta) + (1 - \gamma)(1 - N)\alpha_i}{2(N + 2)(1 - \gamma)\alpha_i},$$

under the condition: $\gamma < \gamma_{\text{std}}$.

The thresholds $\gamma$ and $\gamma_{\text{std}}$ are defined as follows:

$$\gamma \equiv \min_{\alpha_i} \frac{\alpha_i(N + 2)(1 - \delta) - (1 - \delta)}{\alpha_i(N + 2)(1 - \delta) + (1 + \delta)},$$

$$\gamma_{\text{std}} \equiv \max_{\alpha_i} \left(1 - \frac{1}{\alpha_i(\delta + 1) + 1}\right).$$

**Proof:** The proof is left to the technical report.

As an example, we consider a two groups MAC game, in which we fix a low value of the cost of transmission, $\delta = 0.2$, and groups’ sizes $\alpha_1 = \alpha = 0.4$, $\alpha_2 = 1 - \alpha = 0.6$, and we vary the value of the parameter $\gamma$. The game admits three different equilibria, depending on $\gamma$: a fully mixed GESS $q_1^M$, a pure GESS $q_1^p$, and a pure-mixed one $q_1^pM$. In figure 4 we consider the fully mixed and the pure GESS. We have that for $\gamma < 0.3$ the game admits a GESS $q_1^* = (q_1^*, q_2^*)$, whose components are plotted. Then, for $\gamma = \gamma_{\text{std}} > 0.53$, $q_1^* = (T, T)$. We also plot the value of $q_{\text{std}}^*$ := $\min(1, 1 - \Delta)$. We note that fully mixed equilibrium strategies adopted by the two groups, $q_1^*$, $q_2^*$, are both lower than $q_{\text{std}}^*$.

In figure 5 we plotted the value of the second component of the pure-mixed GESS of the game: $(T, q_T^*)$, which exists only for $0 \leq \gamma < 0.4$, and we compare it to $q_{\text{std}}^*$. We note that, for the second group the probability of transmitting is always lower than in the standar game, whereas.

Let $p_S(p)$ be the probability of a successful transmission in a population under profile $p$. For $N = 2$, we have:

$$p_S(p) = \mu[\gamma(\alpha_1 p_1 + (1 - \alpha)p_2) + (1 - \gamma)(2\alpha^2 p_1 (1 - p_1) + (1 - \alpha)(1 - p_2) p_1 + (1 - p_1) p_2 + 2(1 - \alpha)^2 p_2 (1 - p_2))].$$
In this work we presented a new concept of Evolutionarily Stable Strategy in a group-players framework, the GESS, exploring its relation with the Nash equilibrium and with the standard ESS. By analyzing some examples of two players and two strategies games, we observed how the presence of groups impacts the behavior of individuals and changes the structure of the equilibria. We then introduce a slightly different situation by redefining the utility of a group, in such a way to consider different utilities for interactions among members of the same group or of a different one. Through a MAC example we showed how the presence of groups can favor cooperative behaviors. There are still many issues open for future studies. We are now studying the replicator dynamics in the group-players population in order to investigate the relation between our GESS and the rest point of such dynamics. At the theoretical level, we want to further deepen the study of the utility of a group, considering selfish and altruistic components.

VII. CONCLUSIONS

In this work we presented a new concept of Evolutionarily Stable Strategy in a group-players framework, the GESS, exploring its relation with the Nash equilibrium and with the standard ESS. By analyzing some examples of two players and two strategies games, we observed how the presence of groups impacts the behavior of individuals and changes the structure of the equilibria. We then introduce a slightly different situation by redefining the utility of a group, in such a way to consider different utilities for interactions among members of the same group or of a different one. Through a MAC example we showed how the presence of groups can favor cooperative behaviors. There are still many issues open for future studies. We are now studying the replicator dynamics in the group-players population in order to investigate the relation between our GESS and the rest point of such dynamics. At the theoretical level, we want to further deepen the study of the utility of a group, considering selfish and altruistic components.

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