

# Separable and Low-Rank Continuous Games

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**Abstract**—Separable games are a structured subclass of continuous games whose payoffs take a sum-of-products form; the zero-sum case has been studied in earlier work. Included in this subclass are all finite games and polynomial games. Separable games provide a unified framework for analyzing and generating results about the structural properties of low rank games. This work extends previous results on low-rank finite games by allowing for multiple players and a broader class of payoff functions. We also discuss computation of exact and approximate equilibria in separable games. We tie these results together with alternative characterizations of separability which show that separable games are the largest class of continuous games to which low-rank arguments apply.

## I. INTRODUCTION

There has been much interest recently in the problem of computing mixed strategy Nash equilibria of games. The hardness of this problem for cases as simple as two-player finite games has been established by Chen and Deng [2], and this suggests the need to search for classes of games which are computationally tractable. Many such classes have been identified, but in this paper we will consider a generalization of the low-rank games studied by Lipton et al. [6]. Here the computational simplification arises because it can be proven that in an equilibrium, each player chooses a “simple strategy” without loss of generality, where simple refers to the fact that the players randomize among a small number of pure strategies with positive probability.

In games with infinite strategy sets, additional complications arise in computing equilibria. Without compactness of the strategy spaces and continuity of the payoff functions, an equilibrium may fail to exist at all. Even when these assumptions are made, as in so-called continuous games, the strategies played in equilibrium may be arbitrarily complicated. Specifically, two-player zero-sum games with compact intervals for strategy sets and rational payoff functions can be constructed in which the unique equilibrium strategies are the Cantor measure or worse, see Karlin [5] for an example. To compute equilibria of continuous games, we therefore require some additional structure which will ensure that an equilibrium exists which can be succinctly represented.

There are several possible structures which could be imposed. In this paper we study a separable structure in which the payoffs to each player can be written as a weighted sum of products of functions in each player’s strategy separately, e.g. as polynomials. Zero-sum games

possessing such a separable structure have been studied by Karlin [5]. Efficient algorithms for computing equilibria of zero-sum games with polynomial payoffs have been given by Parrilo [7]. Since continuous functions can be uniformly approximated by polynomials, a large class of games can be exactly or approximately represented as separable games.

Another advantage of the separable structure is that for games of this type, the complexity of the payoffs gives an immediate bound on the complexity of the strategies played in equilibrium. If the payoff functions are viewed as matrices whose indices range over compact metric spaces rather than merely finite sets, then the separable form of the payoffs can be viewed as a “finite rank” condition. Since a generic continuous function has “infinite rank,” separable games are a natural generalization of games with low-rank payoff matrices. In this more general setting we extend the results of [6] on low-rank two-player finite games, allowing for an arbitrary finite number of players as well as infinite strategy sets and a broader class of payoff functions.

The rest of this paper is organized as follows. In Section II, we present the basic definitions and theorems as well as introduce a running example. Then in Section III we define the rank of a continuous game, present a theorem demonstrating the relevance of this definition, and show how to bound the rank for arbitrary separable games and compute it exactly for finite and polynomial games. We also give a characterization theorem which shows that separable games form the largest class of continuous games to which low-rank arguments apply. In Section IV we discuss computation of equilibria and approximate equilibria. We close with conclusions and directions for future work.

## II. BASIC THEORY

The theorems stated in this background section are either known or are slight generalizations of known results about separable games. However, we present them in a setting which will allow us to extend them in later sections.

Some notational conventions used throughout are that subscripts refer to players, while superscripts are reserved for other indices, rather than exponents. If  $S_j$  are sets for  $j = 1, \dots, n$  then  $S = \prod_{j=1}^n S_j$  and  $S_{-i} = \prod_{j \neq i} S_j$ . The  $n$ -tuple  $s$  and the  $(n-1)$ -tuple  $s_{-i}$  are formed from the points  $s_j$  similarly. We use the symbols  $\text{span } S$ ,  $\text{aff } S$ ,  $\text{conv } S$ , and  $\bar{S}$  to denote the span, affine hull, convex hull, and closure of the set  $S$ , respectively.

*Definition 2.1:* An  $n$ -player **continuous game** is defined by  $n$  **pure strategy spaces**  $C_i$  assumed to be nonempty compact metric spaces and  $n$  **utility** or **payoff functions**  $u_i : C \rightarrow \mathbb{R}$  assumed to be continuous. Throughout,  $\Delta_i$  will

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denote the space of Borel probability measures  $\sigma_i$  over  $C_i$ , referred to as **mixed strategies**, and the  $u_i$  will be extended from  $C$  to  $\Delta$  by expectation, defining

$$u_i(\sigma) = \int_C u_i(s) d\sigma.$$

*Definition 2.2:* An  $n$ -player **separable game** is an  $n$ -player continuous game with utility functions  $u_i : C \rightarrow \mathbb{R}$  taking the form

$$u_i(s) = \sum_{j_1=1}^{m_1} \cdots \sum_{j_n=1}^{m_n} a_i^{j_1 \cdots j_n} f_1^{j_1}(s_1) \cdots f_n^{j_n}(s_n), \quad (1)$$

where  $a_i^{j_1 \cdots j_n} \in \mathbb{R}$  and the  $f_i^j : C_i \rightarrow \mathbb{R}$  are continuous. In the special case when the  $C_i$  are subsets of  $\mathbb{R}$  and the  $f_i^j$  are polynomials in  $s_i$ , the game is called a **polynomial game**.

When it is convenient to do so, and always for polynomial games, we will begin the summations in (1) at  $j_i = 0$ . For polynomial games we can then use the convention that  $f_i^j(s_i) = s_i^j$ , where the superscript on the right hand side denotes an exponent rather than an index.

*Example 1:* Consider a two player game with  $C_1 = C_2 = [-1, 1] \subset \mathbb{R}$ . Letting  $x$  and  $y$  denote the pure strategies of players 1 and 2, respectively, we define the utility functions

$$\begin{aligned} u_1(x, y) &= 2xy + 3y^3 - 2x^3 - x - 3x^2y^2, \\ u_2(x, y) &= 2x^2y^2 - 4y^3 - x^2 + 4y + x^2y. \end{aligned} \quad (2)$$

This is a polynomial game, and we will return to it periodically to apply the results presented.

Let  $V_i$  denote the space of all finite-valued signed measures (henceforth simply called measures) on  $C_i$ , which can be identified with the dual of the Banach space  $C(C_i)$  of all continuous real-valued functions on  $C_i$  endowed with the sup norm. Throughout, we will assume  $V_i$  is endowed with the weak\* topology. This is the weakest topology such that whenever  $f : C_i \rightarrow \mathbb{R}$  is continuous,  $\sigma \mapsto \int f d\sigma$  is a continuous linear functional on  $V_i$ .

We can extend the utility functions of a continuous game to all of  $V$  in the same way they are extended from  $C$  to  $\Delta$ , yielding a multilinear functional on  $V$ . For a fixed separable game we can extend the  $f_i^j$  from  $C_i$  to  $V_i$  similarly, yielding the so-called **generalized moment** functionals, so (1) holds with  $s$  replaced by  $\sigma$ . In the case of polynomial games when  $f_i^j(s_i) = s_i^j$ , the generalized moment functionals are just the classical moment functionals. We will abuse notation and identify the elements of  $C_i$  with the atomic measures in  $\Delta_i$ , so  $C_i \subseteq \Delta_i \subset V_i$ . Note that  $\text{conv } C_i$  and  $\text{span } C_i$  are the set of all finitely-supported probability measures and the set of all finitely-supported finite signed measures, respectively. The following are standard results (see Ch. 2 of Parthasarathy [8] for the proofs).

*Proposition 2.3:*

- The sets  $C_i$  and  $\Delta_i$  are compact.
- The closures of  $\text{conv } C_i$  and  $\text{span } C_i$  are  $\Delta_i$  and  $V_i$ , respectively.

The simplifications which occur in separable games as opposed to general continuous games can be expressed in

terms of the following three notions of equivalence between two measures:

*Definition 2.4:* Two measures  $\sigma_i, \tau_i \in V_i$  are

- **moment equivalent** if  $f_i^j(\sigma_i) = f_i^j(\tau_i)$  for all  $j$  (representation-dependent and only defined for separable games).
- **payoff equivalent** if  $u_j(\sigma_i, s_{-i}) = u_j(\tau_i, s_{-i})$  for all  $j$  and all  $s_{-i} \in C_{-i}$ .
- **almost payoff equivalent** if  $u_j(\sigma_i, s_{-i}) = u_j(\tau_i, s_{-i})$  for all  $j \neq i$  and all  $s_{-i} \in C_{-i}$ .

Note that in separable games moment equivalence implies payoff equivalence and in all continuous games payoff equivalence implies almost payoff equivalence. Since the  $f_i^j$  and  $u_j$  are linear and multilinear functionals on  $V_i$  and  $V$ , respectively, these equivalence relations can be expressed in terms of (potentially infinitely many) linear constraints on  $\sigma_i - \tau_i$ .

*Definition 2.5:* Let

- $W_i = \{\text{measures moment equivalent to } 0\}$
- $X_i = \{\text{measures payoff equivalent to } 0\}$
- $Y_i = \{\text{measures almost payoff equivalent to } 0\}$

where 0 denotes the zero measure in  $V_i$ .

Then  $W_i \subseteq X_i \subseteq Y_i$ , and  $\sigma_i - \tau_i \in X_i$  if and only if  $\sigma_i$  is payoff equivalent to  $\tau_i$ , etc. Furthermore, the subspaces  $X_i$  and  $Y_i$  are representation-independent and well-defined for any continuous game, separable or not. Note that these subspaces are given by the intersection of the kernels of (potentially infinitely many) continuous linear functionals, hence they are closed.

We will analyze separable games by considering the quotients of  $V_i$  by these subspaces, i.e.  $V_i \text{ mod these three equivalence relations}$ . To avoid defining excessively many symbols let  $\Delta_i/W_i$  denote the image of  $\Delta_i$  in  $V_i/W_i$  and so forth. The following theorem presents the most fundamental result about separable games.

*Theorem 2.6:* In a separable game every mixed strategy  $\sigma_i$  is moment equivalent to a finitely-supported mixed strategy  $\tau_i$  with  $|\text{supp}(\tau_i)| \leq m_i + 1$ . Moreover, if  $\sigma_i$  is countably-supported  $\tau_i$  can be chosen with  $\text{supp}(\tau_i) \subset \text{supp}(\sigma_i)$ .

*Proof:* Note that the map

$$f_i : \sigma_i \mapsto (f_i^1(\sigma_i), \dots, f_i^{m_i}(\sigma_i))$$

is linear and weak\* continuous, with kernel  $W_i$ . Thus  $V_i/W_i$  has dimension at most  $m_i$ . Then

$$\begin{aligned} f_i(\Delta_i) &= f_i(\overline{\text{conv } C_i}) \subseteq \overline{f_i(\text{conv } C_i)} \\ &= \overline{\text{conv } f_i(C_i)} = \text{conv } f_i(C_i). \end{aligned} \quad (3)$$

The first three steps follow from Proposition 2.3, continuity of  $f_i$ , and linearity of  $f_i$ , respectively. The final equality follows from the fact that  $\text{conv } f_i(C_i)$  is compact, being the convex hull of a compact subset of a finite-dimensional space. The reverse inclusion is obvious, so we have  $f_i(\Delta_i) = \text{conv } f_i(C_i) = \overline{f_i(\text{conv } C_i)}$ . Thus any mixed strategy is moment equivalent to a finitely-supported mixed strategy, and applying Carathéodory's theorem to the set  $\text{conv } f_i(C_i) \subset \mathbb{R}^{m_i}$  yields the uniform bound. Since a countable convex

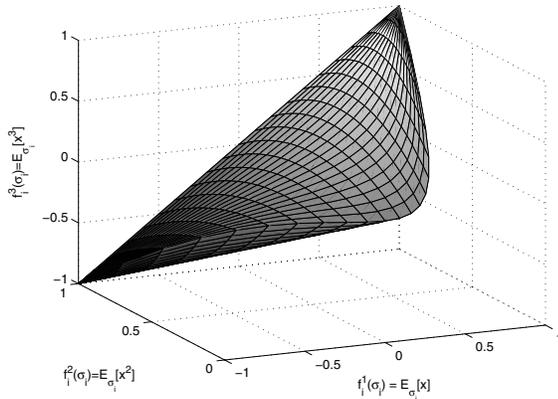


Fig. 1. The space  $f_i(\Delta_i) \cong \Delta_i/W_i$  of possible moments for either player under the payoffs given in (2) due to a measure  $\sigma_i$  on  $[-1, 1]$ . The zeroth moment, which is identically unity, has been omitted.

combination of points in a bounded subset of  $\mathbb{R}^{m_i}$  can always be written as a finite convex combination of at most  $m_i + 1$  of those points, the final claim follows. ■

This theorem can also be proven by a separating hyperplane argument as applied to zero-sum games by Karlin [5]. Combining this theorem with Glicksberg's result [3] that every continuous game has an equilibrium yields:

*Corollary 2.7:* Any separable game has a Nash equilibrium in which player  $i$  mixes among at most  $m_i + 1$  pure strategies.

*Example 1 (cont'd):* Apply the standard definition of the  $f_i^j$  to the polynomial game with payoffs given in (2). The set of moments  $\Delta_i/W_i \cong f_i(\Delta_i)$  as described in Theorem 2.6 is shown in Figure 1. In this case it is the same for both players.

For each player the range of the indices in (1) is  $0 \leq j_i \leq 3$ , so by Corollary 2.7, this game has an equilibrium in which each player mixes among at most  $4 + 1 = 5$  pure strategies. To produce this bound, we have not used any information about the payoffs except for the degree of the polynomials. However, there is extra structure here to be exploited. For example,  $u_2$  depends on the expected value  $E_{\sigma_1}[x^2]$ , but not on  $E_{\sigma_1}[x]$  or  $E_{\sigma_1}[x^3]$ . In particular, player 2 is indifferent between the two strategies  $\pm x$  of player 1 for all  $x$ , insofar as this choice does not affect his payoff (though it does affect what strategy profiles are equilibria). In the following section, we will show how to give improved bounds on the number of strategies played in equilibrium which take these simplifications into account in a systematic manner.

### III. THE RANK OF A CONTINUOUS GAME

By comparing the two-player case of (1) with the singular value decomposition for matrices, separable games can be thought of as games of "finite rank." Now we will define rank precisely, use it to give a bound on the cardinality of the support of equilibrium strategies, and show how to compute the rank of finite and polynomial games. This will generalize the results by Lipton et al. [6] on low-rank two-

player finite games to multiplayer finite and polynomial games, simultaneously improving these results slightly. The primary tool will be the notion of almost payoff equivalence from Definition 2.4. In what follows, the dimension of a set will refer to the dimension of its affine hull.

*Definition 3.1:* The **rank** of a continuous game is defined to be  $\rho = (\rho_1, \dots, \rho_n)$  where  $\rho_i = \dim \Delta_i/Y_i$ . A game is said to have **finite rank** if  $\rho_i < \infty$  for all  $i$ .

Since  $W_i \subseteq Y_i$  and  $V_i/W_i$  is finite-dimensional for any separable game, it is clear that separability implies finite rank. In Section ?? we show that the converse is also true. Using the rank of a game, Corollary 2.7 can now be improved as follows:

*Theorem 3.2:* Given an equilibrium  $\sigma$  of a separable game with rank  $\rho$ , there exists an equilibrium  $\tau$  such that each player  $i$  mixes among at most  $\rho_i + 1$  pure strategies and  $u_i(\sigma) = u_i(\tau)$ .

*Proof:* By Theorem 2.6, we can assume without loss of generality that each player's mixed strategy  $\sigma_i$  is finitely supported. Fix  $i$ , let  $\psi_i : V_i \rightarrow V_i/Y_i$  denote the canonical projection transformation and let  $\sigma_i = \sum_j \lambda^j s_i^j$  be a finite convex combination of pure strategies. By linearity of  $\psi_i$  we have

$$\psi_i(\sigma_i) = \sum_j \lambda^j \psi_i(s_i^j).$$

Carathéodory's theorem states that (renumbering the  $s_i^j$  and adding some zero terms if necessary) we can write

$$\psi_i(\sigma_i) = \sum_{j=0}^{\rho_i} \mu^j \psi_i(s_i^j),$$

a convex combination but perhaps with fewer terms. Let  $\tau_i = \sum_{j=0}^{\rho_i} \mu^j s_i^j$ . Then  $\psi_i(\sigma_i) = \psi_i(\tau_i)$ . Since  $\sigma$  was a Nash equilibrium, and  $\sigma_i$  is almost payoff equivalent to  $\tau_i$ ,  $\sigma_j$  is a best response to  $(\tau_i, \sigma_{-i,j})$  for all  $j \neq i$ . On the other hand  $\sigma_i$  was a mixture among best responses to the mixed strategy profile  $\sigma_{-i}$ , so the same is true of  $\tau_i$ , making it a best response to  $\sigma_{-i}$ . Thus  $(\tau_i, \sigma_{-i})$  is a Nash equilibrium. Repeating the above steps for each player in turn completes the construction of  $\tau$ . ■

If a submatrix is formed from a matrix by "sampling," i.e. selecting a subset of the rows and columns, the rank of the submatrix is bounded by the rank of the original matrix. The same is true of continuous games.

*Proposition 3.3:* Let  $(\{C_i\}, \{u_i\})$  be a continuous game with rank  $\rho$  and  $\tilde{C}_i$  be a nonempty compact subset of  $C_i$  for each  $i$ , with  $\tilde{u}_i = u_i|_{\tilde{C}_i}$ . Then the game  $(\{\tilde{C}_i\}, \{\tilde{u}_i\})$  satisfies  $\tilde{\rho}_i \leq \rho_i$  for all  $i$ .

*Proof:* Since  $\tilde{\Delta}_i \subseteq \Delta_i$  and  $Y_i \cap \tilde{V}_i \subseteq \tilde{Y}_i$ ,

$$\begin{aligned} \tilde{\rho}_i &= \dim \tilde{\Delta}_i / \tilde{Y}_i \leq \dim \tilde{\Delta}_i / (Y_i \cap \tilde{V}_i) \\ &= \dim \tilde{\Delta}_i / Y_i \leq \dim \Delta_i / Y_i = \rho_i. \end{aligned}$$

*Definition 3.4:* The game  $(\{\tilde{C}_i\}, \{\tilde{u}_i\})$  in Proposition 3.3 is called a **sampled game** or a **sampled version** of  $(\{C_i\}, \{u_i\})$ . ■

Note that if we take  $\tilde{C}_i$  to be finite for each  $i$ , then the sampled game is a finite game. If the original game is separable and hence has finite rank, then Proposition 3.3 yields a uniform bound on the complexity of finite games which can arise from this game by sampling. This fact is applied to the problem of computing approximate equilibria in Section IV below. Note that while the proof of Proposition 3.3 is trivial, there exist other kinds of bounds on the cardinality of the support of equilibria (e.g. for special classes of polynomial games as studied by Karlin [5]) which do not share this sampling property.

With the significance of rank made clear by Theorem 3.2 and Proposition 3.3, we will now present a bound on the rank for arbitrary separable games which is tight in the case of finite and polynomial games. For consistency while deriving these results we will use the convention that the summations in (1) begin at zero, even for non-polynomial games. For finite games we will assume each player's strategy set is a set of the form  $C_i = \{0, 1, \dots, m_i\}$  and let  $f_i^{j_i}(s_i) = \delta(j_i - s_i)$  where  $\delta$  is the Kronecker delta function. We will assume that polynomial games are written so that  $f_i^j(s_i) = s_i^j$ . This way the coefficients in (1) are just the payoffs for finite games and the coefficients of the polynomials for polynomial games.

Despite the fact that  $V_i$  may be infinite-dimensional, the problem of computing the rank of a separable game can be reduced to computing the dimension of certain finite-dimensional convex sets. To show this, we first require several preliminary definitions and a lemma. Let  $f_i : V_i \rightarrow M_i$ , where  $M_i = \mathbb{R}^{m_i+1} \supseteq V_i/W_i$ , be the generalized moment function defined by

$$f_i(\sigma_i) = (f_i^0(\sigma_i), \dots, f_i^{m_i}(\sigma_i)).$$

We write  $x_i = (x_i^0, \dots, x_i^{m_i})$  for a typical element of  $M_i$  and we define multilinear functionals  $v_i : M \rightarrow \mathbb{R}$  by

$$v_i(x) = \sum_{j_1=0}^{m_1} \dots \sum_{j_n=0}^{m_n} a_i^{j_1 \dots j_n} x_1^{j_1} \dots x_n^{j_n}, \quad (4)$$

so  $u_i(\sigma) = v_i(f_1(\sigma_1), \dots, f_n(\sigma_n))$ . Let  $e_i^0, \dots, e_i^{m_i}$  denote the standard unit vectors in  $M_i$ .

*Lemma 3.5:* Consider the linear functionals on  $V_i$  of the form  $v_l(f_i(\cdot), x_{-i})$  for  $1 \leq l, i \leq n$  where  $l \neq i$  and  $x_{-i} \in M_{-i}$ . Then

$$Y_i \supseteq \bigcap_{\substack{x_k \in \{e_k^0, \dots, e_k^{m_k}\} \\ k, l \neq i}} \ker v_l(f_i(\cdot), x_{-i}) \quad (5)$$

with equality holding for all  $i$  if  $\text{span } f_i(C_i) = M_i$  for all  $i$ .

*Proof:* The definition of  $Y_i$  can be written as

$$Y_i = \bigcap_{\substack{s_k \in C_k \\ k, l \neq i}} \ker u_l(\cdot, s_{-i}).$$

Letting  $x_k = f_k(s_k)$  and applying (4), we obtain

$$Y_i = \bigcap_{\substack{x_k \in f_k(C_k) \\ k, l \neq i}} \ker v_l(f_i(\cdot), x_{-i}). \quad (6)$$

Without changing the set on the right hand side, we can take the intersection over the kernels of a larger class of linear functionals, those which are linear combinations of the functionals considered in (6). Using the multilinearity of the  $v_l$  yields

$$Y_i = \bigcap_{\substack{x_k \in \text{span } f_k(C_k) \\ k, l \neq i}} \ker v_l(f_i(\cdot), x_{-i}). \quad (7)$$

If we replace  $\text{span } f_k(C_k)$  by  $M_k$  we will be taking the intersection over a larger collection of sets, so

$$Y_i \supseteq \bigcap_{\substack{x_k \in M_k \\ k, l \neq i}} \ker v_l(f_i(\cdot), x_{-i})$$

with equality if  $\text{span } f_k(C_k) = M_k$  for all  $k \neq i$ . Reversing the procedure used to pass from (6) to (7), we may replace the collection of linear functionals on the right hand side by a spanning set. Letting  $x_k$  range over the standard unit vectors in  $M_k$  for all  $k \neq i$  yields such a spanning set (by the multilinearity of the  $v_l$ ), which proves the lemma. ■

The linear functional  $v_l(f_i(\cdot), x_{-i})$  appearing in (5) can be written as the composition  $v_l(\cdot, x_{-i}) \circ f_i$ . Since  $x_{-i}$  is of the form

$$x_{-i} = (e_1^{j_1}, \dots, e_{i-1}^{j_{i-1}}, e_{i+1}^{j_{i+1}}, \dots, e_n^{j_n}),$$

the matrix for the linear functional  $v_l(\cdot, x_{-i})$  is the row vector consisting of the  $m_i + 1$  coefficients  $a_l^{j_1 \dots j_n}$  as  $j_i$  ranges from 0 to  $m_i$ .

*Definition 3.6:* Let  $S_i$  denote the matrix composed of all such row vectors as  $j_k$  ranges from 0 to  $m_k$  and  $k, l \neq i$ ; the order of these vectors is irrelevant. The matrix  $S_i$  has  $(n-1)\prod_{k \neq i} (m_k+1)$  rows and  $m_i+1$  columns. For a set  $X \subseteq M_i$  we will write  $S_i \cdot X$  to denote the image set  $\{S_i x \mid x \in X\}$ .

*Theorem 3.7:* The rank is bounded by

$$\rho_i \leq \dim [S_i \cdot f_i(C_i)]$$

with equality for all  $i$  if  $\text{span } f_i(C_i) = M_i$  for all  $i$ . In particular, equality holds for arbitrary finite games and for polynomial games which satisfy  $|C_i| > m_i$ .

*Proof:* Given  $S_i$ , we can rephrase Lemma 3.5 as:  $Y_i \supseteq \ker [S_i \cdot f_i]$  with equality for all  $i$  when  $\text{span } f_i(C_i) = M_i$  for all  $i$ . Using the definition of rank we have

$$\begin{aligned} \rho_i &= \dim \Delta_i / Y_i \leq \dim \Delta_i / (\ker [S_i \cdot f_i]) \\ &= \dim [S_i \cdot f_i(\Delta_i)] = \dim [S_i \cdot \text{conv } f_i(C_i)] \\ &= \dim [S_i \cdot f_i(C_i)] \end{aligned}$$

with equality for all  $i$  if  $\text{span } f_i(C_i) = M_i$  for all  $i$ . The third equality follows from (3) and the final equality follows from the definition of the dimension of a set as the dimension of its affine hull.

In the case of a finite game, we have

$$\begin{aligned} \text{aff } f_i(C_i) &= \text{aff } f_i(\{0, \dots, m_i\}) = \text{aff } \{e_i^0, \dots, e_i^{m_i}\} \\ &= \left\{ (x_i^0, \dots, x_i^{m_i}) \in \mathbb{R}^{m_i+1} \mid \sum_{k=0}^{m_i} x_i^k = 1 \right\}. \quad (8) \end{aligned}$$

In the case of a polynomial game

$$\begin{aligned} \text{aff } f_i(C_i) &= \text{aff}\{(1, x, x^2, \dots, x^{m_i}) | x \in C_i\} \\ &= \text{aff}\{e_i^0, e_i^0 + e_i^1, e_i^0 + e_i^2, \dots, e_i^0 + e_i^{m_i}\} \quad (9) \\ &= \{(x_i^0, \dots, x_i^{m_i}) \in \mathbb{R}^{m_i+1} | x_i^0 = 1\}, \end{aligned}$$

where we assume that  $|C_i| > m_i$  to get the penultimate equality, which is a standard result. Note that in both the finite and polynomial cases we have shown that  $\text{aff } f_i(C_i)$  is a hyperplane which does not pass through the origin, so  $\text{span } f_i(C_i) = M_i$ . ■

Let  $F_i$  be the matrix with one fewer column than  $S_i$  whose  $j^{\text{th}}$  column is the  $j^{\text{th}}$  column of  $S_i$  minus the  $0^{\text{th}}$  column of  $S_i$  for  $1 \leq j \leq m_i$ . Let  $P_i$  be  $S_i$  with the  $0^{\text{th}}$  column removed. In view of (8) and (9) we have proven the following.

*Proposition 3.8:* The rank can be computed or bounded as follows. For any:

- Separable game,  $\rho_i \leq \text{rank } S_i$ .
- Finite game,  $\rho_i = \text{rank } F_i \geq \text{rank } S_i - 1$ .
- Polynomial game with  $|C_i| > m_i$  for all  $i$ ,  $\rho_i = \text{rank } P_i \geq \text{rank } S_i - 1$ .

Now consider a two-player finite game having payoff matrices  $U_1$  and  $U_2$  where player 1 chooses rows and player 2 chooses columns. Then  $S_1 = U_2'$  and  $S_2 = U_1$ , so the general separable game rank bound in Proposition 3.8 applied to Theorem 3.2 shows that the game has an equilibrium in which player 1 mixes among at most  $\text{rank } U_2 + 1$  pure strategies and player 2 mixes among at most  $\text{rank } U_1 + 1$  pure strategies, as proven by Lipton et al. [6]. The finite game bound proven in Proposition 3.8 yields strictly better results in some cases (by at most one), and applies also to the multiplayer case which was left open in [6].

*Example 1 (cont'd):* We can apply Proposition 3.8 and Theorem 3.2 to the example with payoffs given by (2). Using the procedure described above, we produce the matrices

$$S_1 = \begin{bmatrix} 0 & 0 & -1 & 0 \\ 4 & 0 & 1 & 0 \\ 0 & 0 & 2 & 0 \\ -4 & 0 & 0 & 0 \end{bmatrix}, S_2 = \begin{bmatrix} 0 & 0 & 0 & 3 \\ -1 & 2 & 0 & 0 \\ 0 & 0 & -3 & 0 \\ -2 & 0 & 0 & 0 \end{bmatrix}$$

and

$$P_1 = \begin{bmatrix} 0 & -1 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix}, P_2 = \begin{bmatrix} 0 & 0 & 3 \\ 2 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

Thus the rank of the example game is

$$\rho = (\text{rank } P_1, \text{rank } P_2) = (1, 3),$$

so in fact there exists an equilibrium in which player 1 mixes among at most 2 pure strategies and player 2 mixes among at most 4 pure strategies.

We close this section with two alternative characterizations of separable games, using the concept of finite rank as defined above.

*Theorem 3.9:* For a continuous game, the following are equivalent:

- 1) The game is separable.

- 2) The game has finite rank.

- 3) For each player  $i$ , every countably supported mixed strategy  $\sigma_i$  is almost payoff equivalent to a finitely supported mixed strategy  $\tau_i$  with  $\text{supp}(\tau_i) \subset \text{supp}(\sigma_i)$ .

*Proof:* See the extended version [9] for a proof and an example of a nonseparable continuous game for which every (not necessarily countably supported) mixed strategy is payoff equivalent to a pure strategy, which shows that the property  $\text{supp}(\tau_i) \subset \text{supp}(\sigma_i)$  cannot be removed from condition 3, even if a uniform bound on the cardinality of  $\text{supp}(\tau_i)$  is assumed and almost payoff equivalence is strengthened to payoff equivalence. ■

#### IV. COMPUTATION OF EQUILIBRIA AND APPROXIMATE EQUILIBRIA

In this section, we present an algorithm for computing approximate equilibria of separable games with infinite strategy sets which follows directly from the results on the rank of games given in Section III. First, we consider briefly the problem of computing exact equilibria.

The moments of an equilibrium can in principle be computed using the following generalization of the equilibrium formulation given by Başar and Olsder [1]:

$$\begin{aligned} \max \quad & \sum_{i=1}^n [v_i(x) - p_i] \\ \text{s.t.} \quad & x_i \in \Delta_i/W_i \text{ for all } i \\ & v_i(f_i(s_i), x_{-i}) \leq p_i \text{ for all } i, \text{ all } s_i \in C_i \end{aligned}$$

where  $x_i$  are the moments,  $f_i$  is the moment function, and  $v_i$  is the payoff function on the moment spaces as defined in Section III. Each player also has an auxiliary variable  $p_i$ . The optimum objective value of this problem is zero and is attained exactly at the Nash equilibria. To compute equilibria by this method, we require an explicit description of the spaces of moments  $\Delta_i/W_i$ . We also require a method for computing the payoff to player  $i$  if he plays a best response to an  $m_{-i}$ -tuple of moments for the other players.

While it seems doubtful that such descriptions could be found for arbitrary  $f_i^j$ , they do exist for two-player polynomial games in which the pure strategy sets are intervals. In this case they can be given in terms of linear matrix inequalities as in Parrilo's treatment of the zero-sum case [7]. This yields a problem with multiaffine objective and linear matrix inequality constraints.

*Example 1 (cont'd):* Directly solving this nonconvex problem with MATLAB's `fmincon` has proven error-prone, as there appear to be many local minima which are not global. However, we were able to compute the equilibrium measures

$$\begin{aligned} \sigma_1 &= 0.5532\delta(x+1) + 0.4468\delta(x-0.1149), \\ \sigma_2 &= \delta(y-0.7166) \end{aligned}$$

(i.e. player 1 plays the pure strategy  $x = -1$  with probability 0.5532 and so on) for the payoffs in (2) by this method.

The difficulties in computing equilibria by general nonconvex optimization techniques suggest the need for more specialized systematic methods. As a step toward this, we present an algorithm for computing approximate equilibria

of separable games. There are several possible definitions of approximate equilibrium, but here we will use:

**Definition 4.1:** A mixed strategy profile  $\sigma \in \Delta$  is an  $\epsilon$ -**equilibrium** ( $\epsilon \geq 0$ ) if

$$u_i(s_i, \sigma_{-i}) \leq u_i(\sigma) + \epsilon$$

for all  $s_i \in C_i$  and  $i = 1, \dots, n$ . If  $\epsilon = 0$  then  $\sigma$  is called a **Nash equilibrium**.

Throughout the rest of this section, we will consider a separable game for which  $C_i$  is a compact interval for each  $i$ , and for which the utility functions satisfy a Lipschitz condition

$$|u_i(s_i, s_{-i}) - u_i(s'_i, s_{-i})| \leq L_i |s_i - s'_i|$$

for some  $L_i \geq 0$ , all  $s_{-i} \in C_{-i}$  and all  $i$ . Clearly this is equivalent to requiring the same inequality for all  $\sigma_{-i} \in \Delta_{-i}$ . We must also be able to compute the values of  $u_i$  efficiently, so for example  $u_i$  could be a polynomial with rational coefficients. These assumptions could be relaxed, say by making  $C_i$  a box in  $\mathbb{R}^n$  or by requiring a Hölder condition in place of a Lipschitz condition, but for the sake of clarity we do not do so here.

Fix  $\epsilon > 0$ . Divide the interval  $C_i$  into equal subintervals of length no more than  $2\frac{\epsilon}{L_i}$ ; at most  $\lceil \frac{1}{2\epsilon} l(C_i) L_i \rceil$  such intervals are required, where  $l(C_i)$  denotes the length of the interval  $C_i$ . Let  $\tilde{C}_i$  be the set of center points of these intervals and let  $\tilde{u}_i$  be the corresponding sampled payoffs. Suppose  $\sigma$  is a Nash equilibrium of the sampled game. Choose any  $s_i \in C_i$  and let  $s'_i$  be an element of  $\tilde{C}_i$  closest to  $s_i$ , so  $|s_i - s'_i| \leq \frac{\epsilon}{L_i}$ . Then

$$\begin{aligned} & u_i(s_i, \sigma_{-i}) - u_i(\sigma) \\ & \leq u_i(s_i, \sigma_{-i}) - u_i(s'_i, \sigma_{-i}) + u_i(s'_i, \sigma_{-i}) - u_i(\sigma) \\ & \leq |u_i(s_i, \sigma_{-i}) - u_i(s'_i, \sigma_{-i})| + \tilde{u}_i(s'_i, \sigma_{-i}) - \tilde{u}_i(\sigma) \\ & \leq L_i \frac{\epsilon}{L_i} + 0 = \epsilon \end{aligned}$$

so  $\sigma$  is automatically an  $\epsilon$ -equilibrium of the original separable game. Thus it will suffice to compute a Nash equilibrium of the finite sampled game.

To do so, first compute or bound the rank  $\rho$  of the original separable game as in Proposition 3.8. By Theorem 3.2 and Proposition 3.3, the sampled game has a Nash equilibrium in which player  $i$  mixes among at most  $\rho_i + 1$  pure strategies, independent of how large  $|\tilde{C}_i|$  is. The number of possible choices of at most  $\rho_i + 1$  pure strategies from  $\tilde{C}_i$  is

$$\sum_{k=1}^{\rho_i+1} \binom{|\tilde{C}_i|}{k} \leq \binom{|\tilde{C}_i| + \rho_i}{1 + \rho_i},$$

which is a polynomial in  $|\tilde{C}_i|$  since  $\rho_i$  is fixed.

This leaves the step of checking whether there exists an equilibrium  $\sigma$  for a given choice of  $\text{supp}(\sigma_i) \subseteq \tilde{C}_i$  with  $|\text{supp}(\sigma_i)| \leq \rho_i + 1$  for each  $i$ , and if so, computing such an equilibrium. If the game has two players, the set of such equilibria for given supports is described by  $O(|\tilde{C}_1| + |\tilde{C}_2|)$  linear equalities and inequalities (with more than two players these would become nonlinear), and hence an equilibrium

or certificate of nonexistence of an equilibrium for a given support can be found in time polynomial in  $|\tilde{C}_1| + |\tilde{C}_2| \propto \frac{1}{\epsilon}$ . Since the number of possible supports to be checked is also polynomial in  $\frac{1}{\epsilon}$ , we have the following result.

**Proposition 4.2:** For  $\epsilon > 0$ , an  $\epsilon$ -equilibrium of a two-player separable game satisfying the conditions of this section can be found in time polynomial in  $\frac{1}{\epsilon}$  and exponential in  $\rho$ .

While an algorithm which is polynomial in  $\rho$  would be preferable, the complexity results for finite games, e.g. those proven by Chen and Deng [2], suggest that this is most likely impossible.

## V. CONCLUSIONS AND FUTURE WORK

We have shown that separable games provide a natural setting for the study of games with payoffs satisfying a low-rank condition. This level of abstraction allows the low-rank results of Lipton et al. [6] to be extended to the multiplayer and polynomial cases. Since the rank of a separable game gives a bound on the cardinality of the supports of equilibria for any sampled version of that separable game, approximate equilibria can be computed in time polynomial in  $\frac{1}{\epsilon}$  by discretizing the strategy spaces and applying standard computational techniques for low-rank games.

Other types of low-rank conditions have been studied for finite games, for example Kannan and Theobald have considered the condition that the sum of the payoff matrices be low-rank [4]. It is likely that that the discretization techniques used here can be applied in an analogous way to yield results about computing approximate equilibria of continuous games when the sum of the payoffs of the players is a separable function.

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