# NON-CONSTANT SUM RED-AND-BLACK GAMES WITH BET-DEPENDENT WIN PROBABILITY FUNCTION

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#### Abstract

In this paper we investigate a class of N-person non-constant sum redand-black games with bet-dependent win probability functions.

We assume that N players and a gambling house are engaged in a game played in stages, where the player's probability of winning at each stage is a function f of the ratio of his bet to the sum of all players' bets. However, at each stage of the game there is a positive probability that all players lose and the gambling house wins their bets.

We prove that if the win probability function is super-additive and it satisfies  $f(s)f(t) \leq f(st)$ , then a bold strategy is optimal for all players.

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### 1. Introduction

One of the most interesting examples in Dubins and Savage [2] is the famous red-and-black gambling problem. In the discrete version this gambling problem can be described as follows. A player has an initial nonnegative integer fortune x and wants to reach a goal of g units by betting at each stage of the game an integral amount s not greater than his current fortune. If his initial stake is  $s \leq x$ , his next fortune will be x + s with probability w (with 0 < w < 1), and x - s with probability  $\overline{w} = 1 - w$ . And so on until the goal g is reached or the player goes broke. The problem for the player is to decide how many units of his current fortune should be staked at each stage in order to maximize his probability of reaching the goal g.

Dubins and Savage [2] showed that the solution of this problem depends on the value of the probability w. In particular, they showed that in the *subfair* case when  $w \leq 1/2$ , an optimal strategy is bold play which corresponds to always staking the entire fortune or just what is needed to reach the target, whichever is smaller. In the *superfair case* (i.e.  $w \geq 1/2$ ) Ross [6] proved that it is optimal for the player to play timidly, that is always to stake 1 unit of his current fortune when he is between 1 and g - 1 and 0 otherwise. A gambling

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theoretical proof of Ross's result can be found in Maitra and Sudderth [3].

Pontiggia [5] introduced two-person red-and-black games, in which two players hold an integral amount of money and they both aim to win the other player's fortune. Chen and Hsiau [1] extended the results in Pontiggia [5] to two-person red-and-black games with bet-dependent win probability functions. At the present time we are not aware of any non-trivial results establishing Nash Equilibria for N-person red-and-black games.

This paper considers a N-person non-constant sum red-and-black game, in which we have N players and a gambling house, that at each stage of the game has a positive probability of winning the players' bets.

Section 2 introduces the notation and the terminology we will use in the following sections and sets up the formulation of the non-constant sum redand-black game.

In Section 3, we show that if the win probability function is super-additive (i.e.  $\sum_{i=1}^{n} f(s_i) \leq f(\sum_{i=1}^{n} s_i)$ ) and it satisfies the condition  $f(s)f(t) \leq f(st)$ , then it is optimal for all players to bet boldly. Several examples are given as application of this result.

## 2. Preliminaries

We imagine there are N players (with  $N \ge 2$ ) who are engaged in a game played in stages, where they all aim to reach a goal fortune g by betting at each stage an integral amount not greater than their current fortune. However the gambling house may win some of the players' money.

The initial fortunes of the players form a vector  $(x_1, \ldots, x_j, \ldots, x_N)$ , where  $x_j$  denotes the positive integer fortune held by player j, for  $j = 1, \ldots, N$ . Hence, at the initial state, in the system we have a total amount of money equal to  $M = \sum_{j=1}^{N} x_j$ . We assume that  $g \leq M < 2g$ . During the game the amount of money held by the players can become less than M if at some stage of the game all of the players lose and the money goes to the gambling house.

The state space S of this game is defined as the set of all possible fortunes held by the N players:

$$S = \{(x_1, \ldots, x_N) : 0 \le x_j \le g \text{ for } j = 1, \ldots, N\}.$$

This N-dimensional state space is discrete and it includes several absorbing states. In fact, after one of the players reaches the goal g, or the total amount of money held by the players becomes less than g, the game does not move to any other state.

At each stage, each player bids an integral amount of money, which must be less than or equal to the fortune that he holds. We assume each player chooses his action without any knowledge about the actions chosen by the other players. In fact, all the games we consider are noncooperative games. Suppose that player j bids an amount of money  $a_j$ , where his action set is

$$A_j(x_1, \dots, x_N) = \begin{cases} \{1, \dots, x_j\} & \text{if } x_j \in \{1, \dots, g-1\} \\ \{0\} & \text{if } x_j \in \{0, g\} \end{cases}$$

In our game we require each player to bid at least one unit of money if he has a positive fortune less than g; i.e.  $1 \le a_j \le x_j$  if  $x_j > 0$ , for any  $j = 1, \ldots, N$ . If we allow all of them to bid 0, then for this game there is a trivial, uninteresting Nash Equilibrium, in which all players bid 0 forever, and the game stays in the same state perpetually.

The payoff function for each player is given by an indicator function, which takes value 1 if the player reaches the goal g, and 0 otherwise. Notice that in this game the total amount of money in the system can stay the same as the original amount M or eventually decrease if at some stage of the game the gambling house wins the players' bets. Hence, only one of the players can reach the goal, or if the sum of the players' fortunes becomes less than g none of them will be able to reach it. The sum of the payoffs associated to each set of strategies is not constant (i.e. it can be 0 or 1), so the games described here are non-constant sum stochastic games. Notice that at each stage of the game the players have a positive probability of losing at least one unit of money.

#### 3. Non-constant sum N-person red and black game

Suppose that at stage m each player j has  $1 \le x_j^m \le g-1$  units of money, and he bids an amount  $1 \le a_j^m \le x_j^m$ , for  $j = 1, \ldots, N$ . The law of motion for the game at stage m is defined by:

$$(x_{1}^{m+1},\ldots,x_{N}^{m+1}) = \begin{cases} (x_{1}^{m}-a_{1},\ldots,x_{N}^{m}-a_{N}) & w.p. & 1-\sum_{j=1}^{N} f\left(\frac{a_{j}}{\sum_{i=1}^{N}a_{i}}\right), \\ (x_{1}^{m}+\sum_{i=2}^{N}a_{i},\ldots,x_{N}^{m}-a_{N}) & w.p. & f\left(\frac{a_{1}}{\sum_{i=1}^{N}a_{i}}\right), \\ \dots & \dots & \dots \\ (x_{1}^{m}-a_{1},\ldots,x_{j}^{m}+\sum_{k\neq j}a_{k},\ldots,x_{N}^{m}-a_{N}) & w.p. & f\left(\frac{a_{j}}{\sum_{i=1}^{N}a_{i}}\right), \\ \dots & \dots & \dots \\ (x_{1}^{m}-a_{1},\ldots,x_{N}^{m}+\sum_{i=1}^{N-1}a_{i}) & w.p. & f\left(\frac{a_{N}}{\sum_{i=1}^{N}a_{i}}\right). \end{cases}$$

where the function  $f : [0,1] \to [0,1]$  is an increasing, continuous nonzero function with f(0) = 0 and  $f(s) \leq s$ , which represents the win probability

function of the game.

We notice that if  $\sum_{j=1}^{N} x_j^m < g$  at some stage m, then the probability of reaching the goal g is equal to zero for all the players, hence the game stops. Also after one of the players reaches the goal g, he stops staking money, hence the game stops, and the probability for the other players to reach their goal becomes zero. In conclusion, in our games it is impossible for more than one player to reach the goal because of our assumption that the total amount of money M is smaller than 2g.

Since  $f(s) \leq s$ , we see that for any  $j = 1, \ldots, N$ 

$$\begin{split} E[x_{j}^{m+1}|x_{j}^{m}] &= (x_{j}^{m} - a_{j}^{m}) \left[ 1 - \sum_{k=1}^{N} f\left(\frac{a_{k}^{m}}{\sum_{i=1}^{N} a_{i}^{m}}\right) \right] + (x_{j}^{m} + \sum_{k \neq j} a_{k}^{m}) f\left(\frac{a_{j}^{m}}{\sum_{i=1}^{N} a_{i}^{m}}\right) \\ &+ (x_{j}^{m} - a_{j}^{m}) \sum_{k \neq j} f\left(\frac{a_{k}^{m}}{\sum_{i=1}^{N} a_{i}^{m}}\right) = \\ &= (x_{j}^{m} - a_{j}^{m}) \left[ 1 - f\left(\frac{a_{j}^{m}}{\sum_{i=1}^{N} a_{i}^{m}}\right) \right] + (x_{j}^{m} + \sum_{k \neq j} a_{k}^{m}) f\left(\frac{a_{j}^{m}}{\sum_{i=1}^{N} a_{i}^{m}}\right) = \\ &= (x_{j}^{m} - a_{j}^{m}) + f\left(\frac{a_{j}^{m}}{\sum_{i=1}^{N} a_{i}^{m}}\right) \sum_{i=1}^{N} a_{i}^{m} \\ &\leq x_{j}^{m} \end{split}$$

and  $E[x_j^{m+1}|x_j^m] = x_j^m$  for  $x_j^m = 0$  or g. Therefore the processes  $\{x_j^m\}$ , for  $j = 1, \ldots, N$ , of the fortunes of the N players are supermartingales. This means that the game is subfair for all the players.

For a non-constant sum N-person red-and-black game with  $1 \le x_j \le g-1$ and  $g \le \sum_{j=1}^N x_j < 2g$ , for  $j = 1, \ldots, N$ , we can prove the following result.

**Theorem 3.1.** In a non-constant sum N-person red-and-black game, assume that the win probability function f is super-additive (i.e.  $\sum_i f(s_i) \leq f(\sum_i s_i)$ ) and it satisfies the following condition:  $f(s)f(t) \leq f(st)$ . A Nash Equilibrium is for all players to play a bold strategy, i.e.  $a_j(x_1, \ldots, x_N) = x_j$  for all  $j = 1, \ldots, N$  and all  $(x_1, \ldots, x_N) \in S$ .

*Proof.* Suppose all players adopt a bold strategy and set  $Q_j(x_1, \ldots, x_N) = P[\text{player } j \text{ reaches his goal } g \text{ when the game starts at } (x_1, \ldots, x_N)]$ . The corresponding law of motion at stage m of the game is given by:

Non-constant sum red-and-black

$$(x_{1}^{m+1}, \dots, x_{N}^{m+1}) = \begin{cases} (0, \dots, 0) & w.p. \quad 1 - \sum_{j=1}^{N} f\left(\frac{x_{j}}{\sum_{i=1}^{N} x_{i}}\right), \\ (\sum_{i=1}^{N} x_{i}, \dots, 0) & w.p. \quad f\left(\frac{x_{1}}{\sum_{i=1}^{N} x_{i}}\right), \\ \dots & \dots & \dots \\ (0, \dots, \sum_{i=1}^{N} x_{i}) & w.p. \quad f\left(\frac{x_{N}}{\sum_{i=1}^{N} x_{i}}\right). \end{cases}$$

$$(2)$$

Therefore the expected return to player j when all N players use a bold strategy is:

$$Q_j(x_1, \dots, x_j, \dots, x_N) = f\left(\frac{x_j}{\sum_{i=1}^N x_i}\right) Q_j(0, \dots, \sum_{i=1}^N x_i, \dots, 0) =$$
$$= f\left(\frac{x_j}{\sum_{i=1}^N x_i}\right)$$

where  $Q_j(0, \ldots, \sum_{i=1}^N x_i, \ldots, 0) = 1$ , since  $\sum_{i=1}^N x_i \ge g$ . In order to prove that for player j a bold strategy is optimal when all other players play boldly, it suffices to show that  $Q_j(.)$  is excessive (see Theorem 3.10 Maitra and Sudderth [3]). Hence, we need to show that the expected return

Maitra and Sudderth [3]). Hence, we need to show that the expected return that player j gets by initially staking an amount  $1 \le a_j \le x_j$  and then playing boldly for the rest of the game, is always less than or equal to the expected return he gets playing boldly from the beginning.

Suppose now that the first stage of the game player j stakes an amount  $1 \le a_j \le x_j$ . His expected return in this case is given by:

$$\sigma_j(x_1, \dots, x_j, \dots, x_N) = f\left(\frac{x_1}{a_j + \sum_{i \neq j} x_i}\right) Q_j(a_j + \sum_{i \neq j} x_i, 0, \dots, x_j - a_j, \dots, 0) + \dots + f\left(\frac{a_j}{a_j + \sum_{i \neq j} x_i}\right) Q_j(0, \dots, \sum_{i=1}^N x_i, \dots, 0) + \dots + f\left(\frac{x_N}{a_j + \sum_{i \neq j} x_i}\right) Q_j(0, \dots, x_j - a_j, \dots, a_j + \sum_{i \neq j} x_i) + \dots$$

$$+ \left[1 - f\left(\frac{a_j}{a_j + \sum_{i \neq j} x_i}\right) - \sum_{i \neq j} f\left(\frac{x_i}{a_j + \sum_{i \neq j} x_i}\right)\right] Q_j(0, \dots, x_j - a_j, \dots, 0)$$
  
where

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$$Q_j(0, \dots, \sum_{i=1}^N x_i, \dots, 0) = 1, \quad \text{since} \quad \sum_{i=1}^N x_i \ge g,$$
  
 $Q_j(0, \dots, x_j - a_j, \dots, 0) = 0, \quad \text{since} \quad x_j - a_j < g,$ 

$$Q_j(a_j + \sum_{i \neq j} x_i, 0, \dots, x_j - a_j, \dots, 0) = \dots = Q_j(0, \dots, x_j - a_j, \dots, a_j + \sum_{i \neq j} x_i) = f\left(\frac{x_j - a_j}{\sum_{i=1}^N x_i}\right).$$

Therefore we have

$$\sigma_j(x_1, \dots, x_N) = f\left(\frac{a_j}{a_j + \sum_{i \neq j} x_i}\right) + f\left(\frac{x_j - a_j}{\sum_{i=1}^N x_i}\right) \sum_{k \neq j} f\left(\frac{x_k}{a_j + \sum_{i \neq j} x_i}\right).$$

In order to prove that  $Q_j(.)$  is excessive we need to show that the following inequality holds:

$$Q_j(x_1,\ldots,x_N) \ge \sigma_j(x_1,\ldots,x_N),\tag{3}$$

which is equivalent to:

$$f\left(\frac{x_j}{\sum_{i=1}^N x_i}\right) \ge f\left(\frac{a_j}{a_j + \sum_{i \neq j} x_i}\right) + f\left(\frac{x_j - a_j}{\sum_{i=1}^N x_i}\right) \sum_{k \neq j} f\left(\frac{x_k}{a_j + \sum_{i \neq j} x_i}\right).$$

In order to show that inequality (3) holds we need to decompose the quantity  $\frac{x_j}{\sum_{i=1}^N x_i}$  as follows.

$$\begin{aligned} \frac{x_j}{\sum_{i=1}^N x_i} &= \frac{x_j(a_j + \sum_{k \neq j} x_k)}{\sum_{i=1}^N x_i(a_j + \sum_{k \neq j} x_k)} = \\ &= \frac{x_j a_j + x_j \sum_{k \neq j} x_k}{\sum_{i=1}^N x_i(a_j + \sum_{k \neq j} x_k)} = \\ &= \frac{a_j(\sum_{i=1}^N x_i - \sum_{k \neq j} x_k) + x_j \sum_{k \neq j} x_k}{\sum_{i=1}^N x_i(a_j + \sum_{k \neq j} x_k)} = \\ &= \frac{a_j \sum_{i=1}^N x_i + (x_j - a_j) \sum_{k \neq j} x_k}{\sum_{i=1}^N x_i(a_j + \sum_{k \neq j} x_k)} = \\ &= \frac{a_j}{a_j + \sum_{k \neq j} x_k} + \frac{(x_j - a_j) \sum_{k \neq j} x_k}{(a_j + \sum_{k \neq j} x_k) \sum_{i=1}^N x_i} = \\ &= \frac{a_j}{a_j + \sum_{k \neq j} x_k} + \frac{x_j - a_j}{\sum_{i=1}^N x_i} \left( \sum_{k \neq j} \frac{x_k}{a_j + \sum_{k \neq j} x_k} \right) \right) \end{aligned}$$

Suppose the win probability function f is super-additive, i.e.  $\sum_i f(s_i) \leq f(\sum_i s_i)$ , and it satisfies the condition  $f(s)f(t) \leq f(st)$ , for any  $s, t \in [0, 1]$ , hence we can prove the following result.

$$\begin{aligned} f\left(\frac{x_j}{\sum_{i=1}^N x_i}\right) &\geq f\left(\frac{a_j}{a_j + \sum_{k \neq j} x_k}\right) + f\left(\frac{x_j - a_j}{\sum_{i=1}^N x_i} \sum_{k \neq j} \frac{x_k}{(a_j + \sum_{k \neq j} x_k)}\right) \\ &\geq f\left(\frac{a_j}{a_j + \sum_{k \neq j} x_k}\right) + f\left(\frac{x_j - a_j}{\sum_{i=1}^N x_i}\right) f\left(\sum_{k \neq j} \frac{x_k}{(a_j + \sum_{k \neq j} x_k)}\right) \\ &\geq f\left(\frac{a_j}{a_j + \sum_{k \neq j} x_k}\right) + f\left(\frac{x_j - a_j}{\sum_{i=1}^N x_i}\right) \sum_{k \neq j} f\left(\frac{x_k}{a_j + \sum_{k \neq j} x_k}\right) \end{aligned}$$

In conclusion inequality (3) holds for any  $1 \le a_j \le x_j$ . Therefore we proved that  $Q_j(.)$  is excessive for player j and that the profile (Bold, ..., Bold) is a Nash equilibrium for this class of games.

**Remarks 3.1.** The formulation of this game come from the idea that f can be considered as a penalizing function imposed to the players. For example we can imagine a situation where there is a state agency or the same gambling house that has the power to control the players by penalizing their probability of winning.

**Example 3.1.** We consider the win probability function f(s) = ws with 0 < w < 1, introduced in Pontiggia [5] for the weighted red-and-black game. It is clear that  $f(s) \leq s$  for all  $s \in (0, 1)$  and that this function is super-additive, f(s) + f(t) = ws + wt = w(s + t) = f(s + t). Also it is easy to show that this function satisfies the condition  $f(s)f(t) = (ws)(wt) = w^2st \leq wst = f(st)$ . Hence it follows from Theorem 3.1 that (Bold, ..., Bold) is a Nash Equilibrium for a non-constant sum N-person weighted red-and-black game.

**Example 3.2.** We consider the win probability function  $f(s) = s^p$  for some  $p \ge 1$ , introduced in Chen and Hsiau [1]. It is easy to show that  $f(s) \le s$  and that the function is super-additive, since  $f(s)+f(t) = s^p+t^p \le (s+t)^p = f(s+t)$ . Also the second condition is satisfied since  $f(s)f(t) = s^pt^p = (st)^p = f(st)$ . Hence, it follows from Theorem 3.1 that (Bold, ..., Bold) is a Nash Equilibrium for a non-constant sum N-person red-and-black game with this win probability function.

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