

Best Strategy for Each Team in The Regular Season to Win Champion in The Knockout Tournament

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Abstract

In ‘J. Schwenk.(2018) [3] What is the Correct Way to Seed a Knockout Tournament? Retrieved from The American Mathematical Monthly’, Schwenk identified a surprising weakness in the standard method of seeding a single elimination (or knockout) tournament. In particular, he showed that for a certain probability model for the outcomes of games it can be the case that the top seeded team would be less likely to win the tournament than the second seeded team. This raises the possibility that in certain situations it might be advantageous for a team to intentionally lose a game in an attempt to get a more optimal (though possibly lower) seed in the tournament. We examine this question in the context of a four team league which consists of a round robin “regular season” followed by a single elimination tournament with seedings determined by the results from the regular season [4]. Using the same probability model as Schwenk we show that there are situations where it is indeed optimal for a team to intentionally lose. Moreover, we show how a team can make the decision as to whether or not it should intentionally lose. We did two detailed analysis. One is for the situation where other teams always try to win every game. The other is for the situation where other teams are smart enough, namely they can also lose some games intentionally if necessary. The analysis involves computations in both probability and (multi-player) game theory.

1 Introduction

In contemporary society, sport competitions such as NBA, NCAA basketball, baseball are more and more prevalent and attracting. In most of these competitions, every team in the knockout tournament

has to play head-to-head matches to eliminate the rival and finally tries best to win the tournament. Whether the knockout tournament is fair and what strategy each team has under the knockout tournament is sparking argue between fans every day. In this article, we use the single elimination tournament model created by J. Schwenk.(2018) [3]. We assume that there are four teams in the playoff: a_1 , a_2 , a_3 , and a_4 . Each of them has a weight, v_1 , v_2 , v_3 , v_4 , respectively, which shows the strength of a team. The larger weight, the stronger the team is. Suppose that $v_1 \geq v_2 \geq v_3 \geq v_4$, here a_1 is the best team in the knockout tournament and we will give the best strategy for it. Let the probability team v_i beats v_j be $p_{ij} = \frac{v_i}{v_i+v_j}$. The schedule of the knockout tournament is in figure 1. In the first round, the first seed plays against the fourth seed. The second seed plays against the third seed. In the second round, the winner between the first seed and the fourth seed plays against the winner between the second seed and the third seed.

According to the model created by J. Schwenk [3], he assumes that teams are ordered “correctly” (i.e. no team is stronger than another higher-seed team), and shows that for a certain probability model for the outcomes of games it can be the case that the top-seeded team would be less likely to win the tournament than the second-seeded team. He provides an example of this phenomenon in an eight-team tournament. In his eight-team tournament setting, there are eight teams in the playoff, let the i th seed be team a_i . Schwenk claims that if a_1 and a_2 has the same weight, a_3 , a_4 , a_5 has the same weight, and a_6 , a_7 , a_8 has the same weight, team a_1 has lower probability to win the champion than team a_2 . In this situation, both a_1 and a_2 are the strongest team, and the second seed has the largest probability to win the champion. This raises the possibility that in certain situations it might be advantageous for a team to intentionally lose a game in an attempt to get a more optimal (though possibly lower) seed in the tournament. It is worth noting that Schwenk’s example does not work in the four-team tournament. If teams are ordered “correctly” in the four-team tournament, theorem 2.1 shows that the first seed has the largest probability to win the champion. However, in our problem setting, we do not assume that teams are ordered “correctly”. Take an example, let $v_1 = 1$, $v_2 = 0.8$, $v_3 = 0.6$, $v_4 = 0.5$, then we have $p_{12} = \frac{4}{7}$, $p_{13} = \frac{5}{8}$, $p_{14} = \frac{2}{3}$, $p_{23} = \frac{5}{9}$, $p_{24} = \frac{3}{5}$, and $p_{34} = \frac{6}{11}$. Assume that a_1 is the first seed, a_2 is the second seed, a_4 is the third seed, and a_3 is the fourth seed. In this case, we have the probability for team a_1 to win the champion is $p_{13}(p_{24}p_{12} + (1 - p_{24})p_{14}) = \frac{5}{8}(\frac{3}{5} \cdot \frac{4}{7} + \frac{2}{5} \cdot \frac{2}{3}) \approx 0.38$. However, if a_1 enters the tournament as the second seed, and a_2 becomes the first seed, a_4 is still the third seed, and a_3 is the fourth seed, in this case, we have the probability for team a_1 to win the champion is $p_{14}(p_{23}p_{12} + (1 - p_{23})p_{13}) = \frac{2}{3}(\frac{5}{9} \cdot \frac{4}{7} + \frac{4}{9} \cdot \frac{5}{8}) \approx 0.40 > 0.38$. This example shows that without the assumption that teams are ordered “correctly”, there exists certain situation where a team might intentionally lose a game during the regular season to become a lower seed to enter the tournament. If all other teams try to become higher-seeded team, it is not hard for us to make a strategy which we can enter the tournament with the proper seed which has the highest probability to win the champion. However, it is more interesting that if all other teams also lose some games intentionally to maximize the probability to win the tournament. In this situation, we are faced with a game theory problem. We will show how to find the best strategy in this situation and whether every team can maximize the probability to win the tournament respectively.

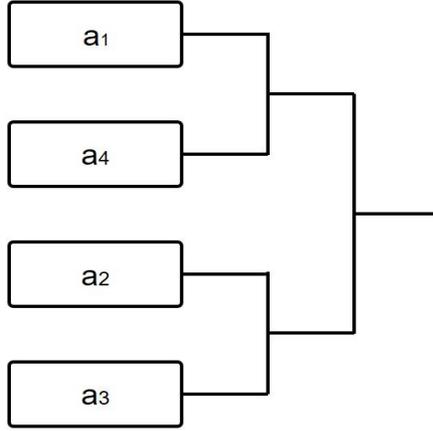


Figure 1: The schedule of the tournament model

1.1 Main Contributions

We build a regular season model with four teams and conduct a detailed analysis of the best team’s strategy, which can help the best team to decide whether to win or lose intentionally in every week in the regular season. It is noteworthy that this model is based on assumption that other teams try their best to win every match, or we can say other teams are not smart enough. We called this Four Teams Regular Season with Not Smart Enough Rivals Model (FRNS). However, in the real world, every professional team is smart enough. Every team has intelligent people to make decisions for them. Thus, we build another model, which assumes that all teams are smart enough and are able to make the most correct decision for them. We call this Four Teams Regular Season with Smart Enough Rivals Model (FRS). More importantly, FRS can get strategy for every team in the regular season.

In game theory, we define pure strategy as a strategy which determines the move a player will make for any situation they could face. We define mixed strategy as an assignment of probability to each pure strategy. In the FRNS model, undoubtedly, the strategy we give to the best team in each week depends on other teams’ weights and the current performance of each team. We define the π_i as the action variable for team a_i . That is $\pi_i = \alpha$ represents team a_i tries to win with probability α and loses intentionally with probability $1 - \alpha$, where $\alpha \in [0, 1]$. We define Π as the collection of vectors $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$, i.e. $\Pi = [0, 1]^4$. We also define $P_\pi^{(i)}(m, W, V)$ as the probability for team a_i to win the champion under action vector $\pi \in \Pi$, team weight vector V and performance vector W in week m . The team weight vector $V = [v_1, v_2, v_3, v_4]$ which contains team weights for all team. The performance vector $W = [W_1, W_2, W_3, W_4]$, where W_i represents the number of winning games before week m for team a_i . Thus, our objective is to find $\sup_{\pi=(\alpha,1,1,1)} P_\pi^{(1)}(m, W, V)$ for every m, W and V . We claim that: under

the FRNS model, for all m, W and possible V , $\operatorname{argmax}_{\pi=(\alpha,1,1,1)} P_{\pi}^{(1)}(m, W, V) \in \{(0, 1, 1, 1), (1, 1, 1, 1)\}$, which shows that the best strategy for the best team in every week is pure strategy. Intuitively, the reason is that other teams' strategies are fixed, i.e. $\pi_2, \pi_3, \pi_4 = 1$.

However, in the FRS model, the action of every team is not fixed due to all teams are smart enough. Our objective is to find $\sup_{\pi \in \Pi} P_{\pi}^{(i)}(m, W, V)$ for every m, W, V and $i \in \{1, 2, 3, 4\}$. To our surprise, the final result shows that the best strategy of every team is not mixed strategy, but pure strategy for almost all possible team weight. Every team can have mixed strategy only in a special case where team a_1 is as strong as a_2 , at the same time team a_3 is as strong as a_4 .

Theorem 1.1. *Under the FRS model, for all $m, W, \pi = (\pi_1, \pi_2, \pi_3, \pi_4)$ satisfies that for each $i \in \{1, 2, 3, 4\}$, $P_{\pi}^{(i)}(m, W, V) \geq P_{\pi'}^{(i)}(m, W, V)$ for $\forall \pi'$ s.t. $\pi'_j = \pi_j, j \neq i$ is a mixed strategy, i.e. $\pi \in (0, 1)^4$ if and only if $v_1 = v_2, v_3 = v_4$. Otherwise, π is a pure strategy, i.e. $\pi \in \{0, 1\}^4$.*

2 Analysis for the FRNS Model

2.1 Description and Assumption of FRNS Model

In this section, we want to analyze the 'regular season' to get a strategy for the best team to decide whether to try to win or lose intentionally in each game. Our logic is that first to analyze the last game in the 'regular season', second to analyze the last two games, third to analyze the last three games, and so on. Before doing the analysis, we will first introduce our FRNS model.

The FRNS model is that we have four teams, a_1, a_2, a_3, a_4 . Their weight is $V = [v_1, v_2, v_3, v_4]$. Assume that $v_1 \geq v_2 \geq v_3 \geq v_4$, so we will help team a_1 to get a strategy $\pi_1 \in [0, 1]$. The winning probability in a single game between a_i and a_j is $p_{ij} = \frac{v_i}{v_i + v_j}$ [3]. We suppose that except a_1 , other teams will try their best to win for every match, i.e. $\pi_2 = \pi_3 = \pi_4 = 1$. Define the subset $(\alpha, 1, 1, 1) = \hat{\Pi} \subseteq \Pi$, where $\alpha \in [0, 1]$, then π_1 is such that $(\pi_1, 1, 1, 1) = \operatorname{argmax}_{\pi \in \hat{\Pi}} P_{\pi}^{(1)}(m, W, V)$ such that for any week m , performance vector W , and possible V . In addition, we build a particular schedule for last three weeks (see figure 2).

Since there are four teams, there are $4! = 24$ seedings in the knockout tournament. However, there exists three types of knockout tournaments, we ignore the exact rank of each team and we only care that which two teams will have a battle in the first week (see figure 3). Here we need another assumption:

Assumptions. *If the number of winning games of two teams is the same, then they will flip a fair coin to decide their seeding in the knockout tournament.*

For example, if finally a_1 and a_2 win 2 games, a_3 and a_4 win 1 game, then in this situation, a_1 and a_2 will have same probability, 50%, to be the first and the second seed. a_3 and a_4 will have same probability, 50%, to be the third and the fourth seed. If finally a_1 wins 3 games and all of a_2, a_3 and a_4 win 1 game, then, a_1 is the first seed and a_2, a_3 and a_4 will have same probability, 33.33%, to be the second, third, and the fourth seed.

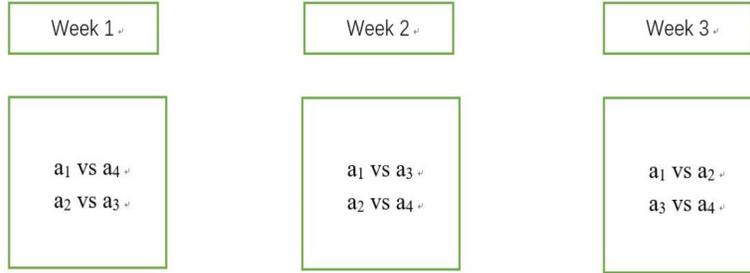


Figure 2: A particular schedule for last three weeks

An interesting idea is that we find the probability for a_1 to win the champion in tournament A is always larger than the one in tournament B and tournament C , no matter what vector V is. Suppose that the probability for a_1 to win the champion in tournament A, B, C is T_A, T_B, T_C , respectively.

Theorem 2.1. *For any weight vector V , subject to $v_1 \geq v_2 \geq v_3 \geq v_4$, we have*

$$T_A \geq T_B \geq T_C \quad (1)$$

The proof of theorem 2.1 can be found in appendix A.1.

We have introduced that our logic is to first analyze the strategy in last week. In the last week, there exists fifteen different W vectors: $[2, 2, 0, 0], [2, 1, 0, 1], [2, 1, 1, 0], [2, 0, 1, 1], [1, 1, 1, 1], [1, 2, 0, 1], [1, 2, 1, 0], [1, 1, 2, 0], [1, 0, 2, 1], [1, 1, 0, 2], [1, 0, 1, 2], [0, 1, 1, 2], [0, 0, 2, 2], [0, 1, 2, 1], [0, 2, 1, 1]$. In the next part, we will pick one specific W vector to do analysis as an example.

2.2 Example $W = [2, 2, 0, 0]$

If at the beginning of last week, the performance vector W is $[2, 2, 0, 0]$, we first consider the game a_3 vs a_4 . There are two possible results: a_3 wins or a_4 wins. If a_3 wins, W will become $[2, 2, 1, 0]$. If a_4 wins, W will become $[2, 2, 0, 1]$. Then we will analyze the match a_1 vs a_2 . If a_1 tries to win, two situations may happen: a_1 wins and a_1 loses. However, if a_1 wants to lose intentionally, the only possible result is a_1 loses. Assume that a_3 defeats a_4 and $W = [2, 2, 1, 0]$, if a_1 defeats a_2 , W will become $[3, 2, 1, 0]$, which leads to tournament A . If a_2 defeats a_1 , W will become $[2, 3, 1, 0]$, which leads to tournament B .

Now we can calculate the probability for a_1 to win the tournament based on different strategy π_1 in last match. Recall that a_3 defeats a_4 with probability p_{34} , and if it happens, W will become $[2, 2, 1, 0]$. a_4 defeats a_3 with probability p_{43} , and if it happens, W will become $[2, 2, 0, 1]$.

$$W = \begin{cases} [2, 2, 1, 0], & \text{with probability } p_{34} \\ [2, 2, 0, 1], & \text{with probability } p_{43} \end{cases} \quad (2)$$

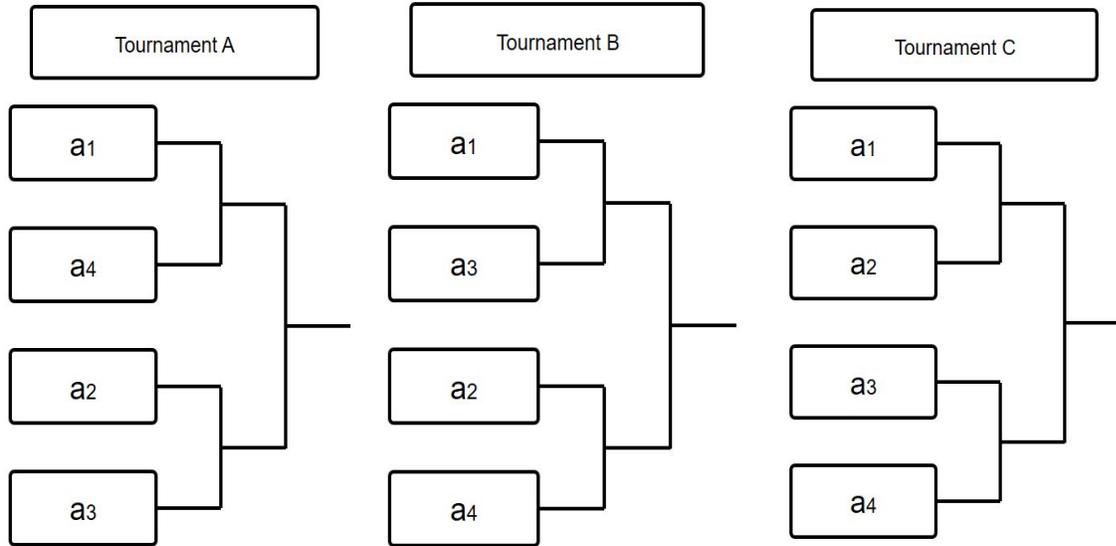


Figure 3: three kind of knockout tournaments

If a_1 tries to win the last match, the analysis is in table 1. Thus, the total probability for a_1 to win the champion if a_1 tries to win in the last match is $P_{\pi=(1,1,1,1)}^{(1)}(3, [2, 2, 0, 0], V) = p_{12}p_{34}T_A + p_{12}p_{43}T_B + p_{21}p_{34}T_B + p_{21}p_{43}T_A$. If a_1 loses intentionally, then the analysis is in table 2. Thus, the total probability for a_1 to win the champion if a_1 loses intentionally in the last match is $P_{\pi=(0,1,1,1)}^{(1)}(3, [2, 2, 0, 0], V) = p_{34}T_B + p_{43}T_A$. Under particular weight vector $V = [v_1, v_2, v_3, v_4]$, we compare $P_{\pi=(1,1,1,1)}^{(1)}(3, [2, 2, 0, 0], V)$ and $P_{\pi=(0,1,1,1)}^{(1)}(3, [2, 2, 0, 0], V)$, the larger one represents the strategy for a_1 .

Theorem 2.2. *Under $W = [2, 2, 0, 0]$ at the beginning of the last week, a_1 should always try to win no matter what weight vector V is.*

a_1	a_3	Win (W)	Vector	Tournament	Probability for team a_1 to win champion
Win	Win	[3, 2, 1, 0]		A	$p_{12}p_{34}T_A$
Win	Lose	[3, 2, 0, 1]		B	$p_{12}p_{43}T_B$
Lose	Win	[2, 3, 1, 0]		B	$p_{21}p_{34}T_B$
Lose	Lose	[2, 3, 0, 1]		A	$p_{21}p_{43}T_A$

Table 1: analysis table if a_1 tries to win

a_1	a_3	Win Vector (W)	Tournament	Probability for team a_1 to win champion
Lose	Win	[2, 3, 1, 0]	B	$p_{34}T_B$
Lose	Lose	[2, 3, 0, 1]	A	$p_{43}T_A$

Table 2: analysis table if a_1 loses intentionally

The proof of theorem 2.2 can be found in appendix A.2. Next, we will show the analytical results for all W .

2.3 Analytical Results

We assume that $v_1 = 1$, v_2, v_3, v_4 are in $(0, 1)$, and $v_2 \geq v_3 \geq v_4$. We let v_2 be the x axis, v_3 be the y axis, v_4 be the z axis. The region in the plots is the set $\{V = (1, v_2, v_3, v_4) | P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) \geq 0, v_2 \geq v_3 \geq v_4\}$ for given W . Here are the results.

For $W = [2, 2, 0, 0], [2, 1, 1, 0], [1, 1, 2, 0], [1, 1, 1, 1], [1, 0, 2, 1], [0, 2, 1, 1], [0, 1, 2, 1], [0, 0, 2, 2]$, since we have completed the analysis of the situation where $W = [2, 2, 0, 0]$ by theorem 2.2, the analysis of rest seven situations of W is similar to the $W = [2, 2, 0, 0]$ example. After some calculation, we can know that for every V , $P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) \leq 0$. Thus, it is always sensible for a_1 to try to win the last match if these eight situations happen.

For $W = [2, 1, 0, 1], [2, 0, 1, 1], [1, 1, 0, 2], [1, 0, 1, 2]$, by the similar method, we find that $P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) \geq 0$ holds for every V . We can conclude that the region in figure 4 is the whole region, i.e. $\{V = (1, v_2, v_3, v_4) | v_2 \geq v_3 \geq v_4\}$. Thus, it is always sensible for a_1 to lose the last match intentionally if these four situations happen.

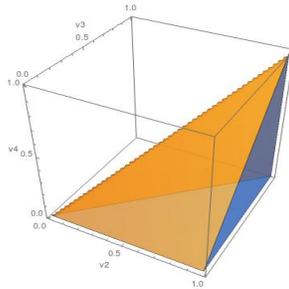


Figure 4: the 3d region plot for the difference of two values in $[2, 1, 0, 1], [2, 0, 1, 1], [1, 1, 0, 2], [1, 0, 1, 2]$

For $[1, 2, 0, 1]$ and $[0, 1, 1, 2]$, similarly, we can find that the strategy for a_1 to decide whether to try

to win or lose intentionally in the last match depends on the team weight vector $V = (1, v_2, v_3, v_4)$. Different team weight leads to different strategy.

Theorem 2.3. *Under a specific team weight vector $V = (1, v_2, v_3, v_4)$, $W = [1, 2, 0, 1]$ or $[0, 1, 1, 2]$, then a_1 should try to win if and only if the following inequality holds. Otherwise, a_1 should lose intentionally.*

$$3v_2v_3v_4^2 + 2v_2v_4^3 + 6v_2^2v_3^2v_4 + v_2v_3^2v_4 + v_3^3v_4 + 2v_3v_4^3 + 10v_3^2v_4^2 - 3v_2^2v_3^3 - v_2^2v_3v_4^2 - 2v_2^2v_4^3 - v_2v_3^4 - 3v_2v_3^3v_4 - 2v_2v_3^2v_4^2 - v_3^4v_4 - 6v_3^3v_4^2 - 6v_3^2v_4^3 \leq 0$$

The proof of theorem 2.3 can be found in appendix A.3. The plot of this region is figure 5, which is made by Mathematica.

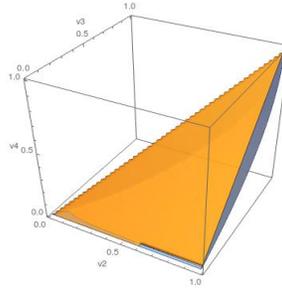


Figure 5: the 3d region plot for the difference of two values in $[1, 2, 0, 1]$ and $[0, 1, 1, 2]$

For $[1, 2, 1, 0]$, we can find that the strategy for a_1 to decide whether to try to win or lose intentionally in the last match depends on the team weight vector $V = (1, v_2, v_3, v_4)$. Different team weight leads to different strategy. Next theorem will show the relationship between the strategy and vector V :

Theorem 2.4. *Under a specific team weight vector $V = (1, v_2, v_3, v_4)$, $W = [1, 2, 1, 0]$, then a_1 should try to win if and only if the following inequality holds. Otherwise, a_1 should lose intentionally.*

$$3v_2^2v_3^2v_4 + v_2v_3^4 + v_2v_3^3v_4 + v_3^3v_4^2 + 4v_3^3v_4 + 4v_3^2v_4^2 + 5v_2v_3v_4 - 4v_2^2v_3v_4^2 - 5v_2^2v_4^3 - 4v_3^4v_4 - 2v_2v_3v_4^2 - v_3^2v_4 - v_3v_4^2 \leq 0$$

The proof of theorem 2.4 can be found in appendix A.4. The plot of this region is in figure 6, which is made by Mathematica.

2.4 Analysis for the Second Week

After completing the analysis of the last week, we want to analyze for the second week. Our idea is that given a winning vector W^* at the beginning of the second week and the team weight vector V , if team a_1 tries to win, then four situations may happen on the second week: a_1 defeats a_3 , a_2 defeats a_4

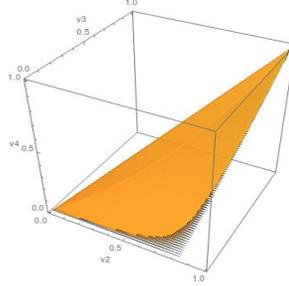


Figure 6: the 3d region plot for the difference of two values in $[1, 2, 1, 0]$

with probability $p_{13}p_{24}$; a_1 defeats a_3 , a_4 defeats a_2 with probability $p_{13}p_{42}$; a_3 defeats a_1 , a_2 defeats a_4 with probability $p_{31}p_{24}$; a_3 defeats a_1 , a_4 defeats a_2 with probability $p_{31}p_{42}$. Assume these four situations bring W^* to W_1, W_2, W_3, W_4 respectively, since we have completed the analysis of the last week, then we can get

$$\begin{aligned} P_{\pi=(1,1,1,1)}^{(1)}(2, W^*, V) &= p_{13}p_{24} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, W_1, V), P_{\pi=(0,1,1,1)}^{(1)}(3, W_1, V)\} \\ &\quad + p_{13}p_{42} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, W_2, V), P_{\pi=(0,1,1,1)}^{(1)}(3, W_2, V)\} \\ &\quad + p_{31}p_{24} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, W_3, V), P_{\pi=(0,1,1,1)}^{(1)}(3, W_3, V)\} \\ &\quad + p_{31}p_{42} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, W_4, V), P_{\pi=(0,1,1,1)}^{(1)}(3, W_4, V)\} \end{aligned}$$

If a_1 loses intentionally in the second week, then two situations may happen on the second week: a_3 defeats a_1 , a_2 defeats a_4 with probability p_{24} ; a_3 defeats a_1 , a_4 defeats a_2 with probability p_{42} . These two situations bring W^* to W_3, W_4 respectively, then we can get

$$\begin{aligned} P_{\pi=(0,1,1,1)}^{(1)}(2, W^*, V) &= p_{24} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, W_3, V), P_{\pi=(0,1,1,1)}^{(1)}(3, W_3, V)\} \\ &\quad + p_{42} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, W_4, V), P_{\pi=(0,1,1,1)}^{(1)}(3, W_4, V)\} \end{aligned}$$

Theorem 2.5. *Suppose the winning vector is W^* at the beginning of the second week, then if*

$$P_{\pi=(1,1,1,1)}^{(1)}(2, W^*, V) \geq P_{\pi=(0,1,1,1)}^{(1)}(2, W^*, V) \tag{3}$$

a_1 should try to win in the second week. Otherwise, a_1 should lose intentionally.

Next, we take $W^* = [1, 1, 0, 0]$ as an example.

2.4.1 Example: $W^* = [1, 1, 0, 0]$

Suppose that after the first week, $W^* = [1, 1, 0, 0]$, then if a_1 tries to win in the second week, after the second week, W^* may become the following four vectors:

$$\begin{cases} [2,2,0,0], & \text{with probability } p_{13}p_{24} \\ [2,1,0,1], & \text{with probability } p_{13}p_{42} \\ [1,2,1,0], & \text{with probability } p_{31}p_{24} \\ [1,1,1,1], & \text{with probability } p_{31}p_{42} \end{cases} \quad (4)$$

Since we have already found the best strategy under $W_1 = [2, 2, 0, 0]$, $W_2 = [2, 1, 0, 1]$, $W_3 = [1, 2, 1, 0]$, $W_4 = [1, 1, 1, 1]$, recall that $P_{\pi=(1,1,1,1)}^1(2, W^*, V)$ is the probability for a_1 to win the champion under $W^* = [1, 1, 0, 0]$ at the beginning of the second week if a_1 tries to win in the second week, then by theorem 2.5 , we can get

$$\begin{aligned} P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V) &= p_{13}p_{24} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, [2, 2, 0, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(3, [2, 2, 0, 0], V)\} \\ &+ p_{13}p_{42} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, [2, 1, 0, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(3, [2, 1, 0, 1], V)\} \\ &+ p_{31}p_{24} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, [1, 2, 1, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(3, [1, 2, 1, 0], V)\} \\ &+ p_{31}p_{42} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, [1, 1, 1, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(3, [1, 1, 1, 1], V)\} \end{aligned}$$

Similarly, recall that $P_{\pi=(0,1,1,1)}^1(2, W^*, V)$ is the probability for a_1 to win the champion under $W^* = [1, 1, 0, 0]$ at the beginning of the second week if a_1 loses intentionally in the second week, note the W^* may become the following two vectors:

$$\begin{cases} [1,2,1,0], & \text{with probability } p_{24} \\ [1,1,1,1], & \text{with probability } p_{42} \end{cases} \quad (5)$$

Then, $P_{\pi=(0,1,1,1)}^{(1)}(2, W^*, V)$ can be calculated by theorem 2.5.

$$\begin{aligned} P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V) &= p_{24} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, [1, 2, 1, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(3, [1, 2, 1, 0], V)\} \\ &+ p_{42} \max\{P_{\pi=(1,1,1,1)}^{(1)}(3, [1, 1, 1, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(3, [1, 1, 1, 1], V)\} \end{aligned}$$

Now, it is natural to compare the difference between $P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V)$ and $P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V)$ to get the best strategy in the second week. We will show the result in the following subsection.

2.4.2 Analytical Results

We assume that $v_1 = 1$, v_2, v_3, v_4 are in $(0, 1)$, and $v_2 \geq v_3 \geq v_4$. We let v_2 be the x axis, v_3 be the y axis, v_4 be the z axis. It is worth mentioning that in this section, we only provide the plots by Mathematica for the set $\{V = (v_2, v_3, v_4) | P_{\pi=(0,1,1,1)}(2, W, V) - P_{\pi=(1,1,1,1)}(2, W, V) \geq 0\}$ for given W instead of giving an analytical formula for the region, because the analytical formula is consisted of several extremely complicated polynomial, which is hard to write it out explicitly.

For $W = [1, 1, 0, 0]$, the result is in figure 7. The region in the plot is the set

$$\{V = (v_2, v_3, v_4) | P_{\pi=(0,1,1,1)}(2, W, V) - P_{\pi=(1,1,1,1)}(2, W, V) \geq 0\} \text{ where } W = [1, 1, 0, 0].$$

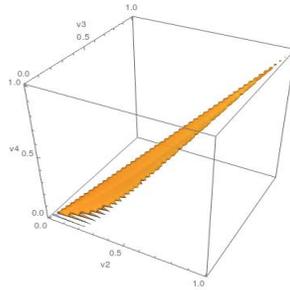


Figure 7: values of v_2, v_3, v_4 such that $P_{\pi=(0,1,1,1)}(2, W, V) - P_{\pi=(1,1,1,1)}(2, W, V) \geq 0$ under $W = [1, 1, 0, 0]$

We can also get the region plot by Mathematica under $W = [1, 0, 1, 0], [0, 1, 0, 1], [0, 0, 1, 1]$, the results are in figure 8.

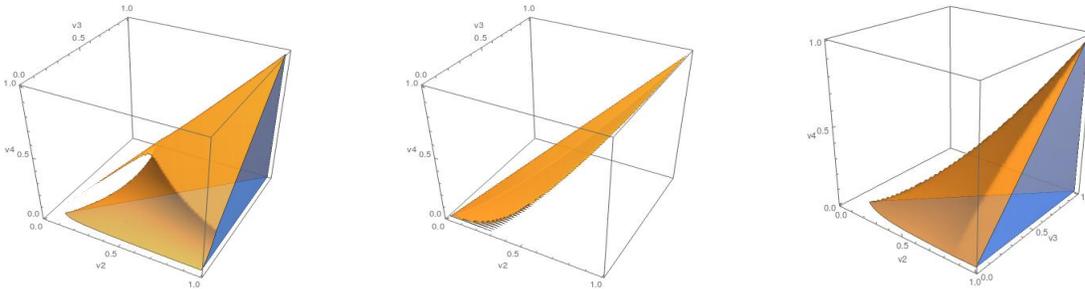


Figure 8: Left: values of v_2, v_3, v_4 such that $P_{\pi=(0,1,1,1)}(2, W, V) - P_{\pi=(1,1,1,1)}(2, W, V) \geq 0$ under $W = [1, 0, 1, 0]$. Middle: values of v_2, v_3, v_4 such that $P_{\pi=(0,1,1,1)}(2, W, V) - P_{\pi=(1,1,1,1)}(2, W, V) \geq 0$ under $W = [0, 1, 0, 1]$. Right: values of v_2, v_3, v_4 such that $P_{\pi=(0,1,1,1)}(2, W, V) - P_{\pi=(1,1,1,1)}(2, W, V) \geq 0$ under $W = [0, 0, 1, 1]$

Thus, a_1 can decide to try to win or lose intentionally in the second week from the figure 7, 8. Now we move to analyze the strategy for a_1 in the first week!

2.5 Analysis for the First Week

After completing the analysis of the second week, we want to analyze for the first week. In the first week, if a_1 tries to win, winning vector W may go to these four situations:

$$\left\{ \begin{array}{l} [1,1,0,0], \text{ with probability } p_{14}p_{23} \\ [1,0,1,0], \text{ with probability } p_{14}p_{32} \\ [0,1,0,1], \text{ with probability } p_{41}p_{23} \\ [0,0,1,1], \text{ with probability } p_{41}p_{32} \end{array} \right. \quad (6)$$

Since we have already found the best strategy under $[1, 1, 0, 0]$, $[1, 0, 1, 0]$, $[0, 1, 0, 1]$, $[0, 0, 1, 1]$, recall that $P_{\pi=(1,1,1,1)}^1(1, [0, 0, 0, 0], V)$ is the probability for a_1 to win the champion at the beginning of the first week if a_1 tries to win, similar to the analysis for the second week, we can get:

$$\begin{aligned} P_{\pi=(1,1,1,1)}^1(1, [0, 0, 0, 0], V) &= p_{14}p_{23} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V)\} \\ &+ p_{14}p_{32} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V)\} \\ &+ p_{41}p_{23} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V)\} \\ &+ p_{41}p_{32} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V)\} \end{aligned}$$

If a_1 loses intentionally, winning vector W may go to these two situations:

$$\left\{ \begin{array}{l} [0,1,0,1], \text{ with probability } p_{23} \\ [0,0,1,1], \text{ with probability } p_{32} \end{array} \right. \quad (7)$$

Then, $P_{\pi=(0,1,1,1)}^1(1, [0, 0, 0, 0], V)$ can be written as

$$\begin{aligned} P_{\pi=(0,1,1,1)}^1(1, [0, 0, 0, 0], V) &= p_{23} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V)\} \\ &+ p_{32} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V)\} \end{aligned}$$

Next we can compare the difference and due to the huge amount of calculation, the Mathematica cannot provide a nice 3d region plot, instead, we use Python to make the 3d scatter plot, which can also provide some information of the rough shape of the region. We let v_2 be the x axis, v_3 be the y axis, v_4 be the z axis. The points in the plots is the set of point $\{V = (v_2, v_3, v_4) | P_{\pi=(0,1,1,1)}^1(1, [0, 0, 0, 0], V) - P_{\pi=(1,1,1,1)}^1(1, [0, 0, 0, 0], V) \geq 0\}$ for given W . Here is the result.

Thus, a_1 can investigate the team weight of a_2 , a_3 , a_4 at the beginning of the first week. If their team weight is in figure 9, a_1 should lose intentionally in the first week. Otherwise, a_1 should try to win. Next, we will provide a specific example to see how to make decision for a_1 in the first week given the weight v_2 , v_3 , and v_4 .

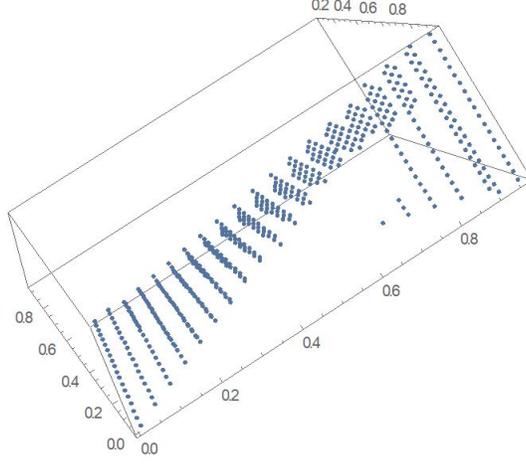


Figure 9: the list points of $[v_2, v_3, v_4]$ when it is better for a_1 to lose intentionally in week 1

2.5.1 Example: $v_2 = 0.5, v_3 = v_4 = 0.1$

In this part, we shows an example where team a_1 should lose intentionally in the first week. We pick a point, $(0.5, 0.1, 0.1)$, which is in figure 9. We will calculate a_1 's probability of winning the tournament conditional on winning in the first week, and conditional on losing intentionally in the first week respectively.

Let $V = (1, 0.5, 0.1, 0.1)$, we have

$$\begin{aligned}
P_{\pi=(1,1,1,1)}^1(1, [0, 0, 0, 0], V) &= p_{14}p_{23} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V)\} \\
&+ p_{14}p_{32} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V)\} \\
&+ p_{41}p_{23} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V)\} \\
&+ p_{41}p_{32} \max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V)\}
\end{aligned}$$

We have introduced how to calculate $P_{\pi=(1,1,1,1)}^{(1)}(2, W^*, V)$ in section 2.4. Under $V = (1, 0.5, 0.1, 0.1)$, we have

$$\max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V)\} = P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 1, 0, 0], V) \approx 0.604,$$

$$\max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V)\} = P_{\pi=(1,1,1,1)}^{(1)}(2, [1, 0, 1, 0], V) \approx 0.614,$$

$$\max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V)\} = P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 1, 0, 1], V) \approx 0.620, \quad (8)$$

$$\max\{P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V), P_{\pi=(0,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V)\} = P_{\pi=(1,1,1,1)}^{(1)}(2, [0, 0, 1, 1], V) \approx 0.603. \quad (9)$$

Then, we have a_1 's probability of winning the tournament conditional on winning in the first week is

$$P_{\pi=(1,1,1,1)}^1(1, [0, 0, 0, 0], V) \approx 0.604p_{14}p_{23} + 0.614p_{14}p_{32} + 0.620p_{41}p_{23} + 0.603p_{41}p_{32} \approx 0.606. \quad (10)$$

We also have a_1 's probability of winning the tournament conditional on losing intentionally in the first week is

$$P_{\pi=(0,1,1,1)}^1(1, [0, 0, 0, 0], V) \approx 0.620p_{23} + 0.603p_{32} \approx 0.618. \quad (11)$$

Since $0.606 < 0.618$, we know that under $V = (1, 0.5, 0.1, 0.1)$, team a_1 should lose intentionally in the first week to get higher probability to win the champion. Hence, we have fully analyzed this FRNS model and in the next part we will analyze the FRS model.

3 Analysis for the FRS Model

3.1 Description and Assumption of FRS Model

In this part, all assumptions are as same as the model in previous section except for the strategies of the other three teams. Recall that in the FRNS model, we assume that other teams try to win every game, i.e. $\pi_2 = \pi_3 = \pi_4 = 1$. However, in this part, we assume that other three teams are smart enough to lose some games intentionally to maximize their winning probability of the tournament. Hence, the strategy for one particular game may not be pure strategy. Instead, every team may have a mixed strategy for each game. In this model, $\pi_i \in [0, 1]$ for $i \in \{1, 2, 3, 4\}$. Note that the boundary point 0 represents the strategy to lose intentionally and 1 represents the strategy to try to win. If $\pi_i \notin \{0, 1\}$, it means team a_i tries to win with probability π_i and loses intentionally with probability $1 - \pi_i$. We assume that while two teams a_i and a_j are playing a game, if both of them try to win, then a_i has probability $p_{ij} = \frac{v_i}{v_i + v_j}$ to win. If one of them tries to win and the other loses intentionally, then we assume that the one who tries to win will have 100% probability to win this game. If both of them loses intentionally, then the game is decided by flipping a fair coin. In the game theory problem with mixed strategy, we usually have to find the Nash equilibrium. The Nash equilibrium is a concept of game theory where the optimal outcome of a game is one where no player has an incentive to deviate from his

chosen strategy after considering an opponent's choice. Hence, in this model, in week m , given winning vector W and weight vector V , $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$ is a Nash equilibrium if for each $i \in \{1, 2, 3, 4\}$, $P_\pi^{(i)}(m, W, V) \geq P_{\pi'}^{(i)}(m, W, V)$ for $\forall \pi'$ s.t. $\pi'_j = \pi_j, j \neq i$.

3.2 Example $W = [1, 2, 0, 1]$

3.2.1 Analysis for $(\pi_1, \pi_2, \pi_3, \pi_4)$

Similar with the analysis of FRNS model, we still first analyze the last game. Take $W = [1, 2, 0, 1]$ as an example, recall that in our FRNS model, we denote T_A, T_B, T_C as the probability for a_1 to win the champion in tournament A, B, C respectively. In this section, we define T_{ij} as the probability team a_i to win the champion in the tournament j , where $i \in \{1, 2, 3, 4\}, j \in \{A, B, C\}$. By Theorem 2.1, we know that $T_{1A} \geq T_{1B} \geq T_{1C}$ when $v_1 \geq v_2 \geq v_3 \geq v_4$. Since in our FRS model, we not only analyze the strategy of a_1 , but also analyze the strategies of a_2, a_3, a_4 . Next, we will introduce a lemma to show the relationship between T_{ij} .

Lemma 3.1. *If $v_1 \geq v_2 \geq v_3 \geq v_4$, then*

$$T_{1A} \geq T_{1B} \geq T_{1C} \quad (12)$$

$$T_{2B} \geq T_{2A} \geq T_{2C} \quad (13)$$

$$T_{3C} \geq T_{3A} \geq T_{3B} \quad (14)$$

$$T_{4C} \geq T_{4B} \geq T_{4A} \quad (15)$$

The proof of lemma 3.1 can be found in the appendix A.5. We just simply calculate all these variables and find the difference between each of them.

In addition, we assume that in week 3, a_1 vs a_2, a_3 vs a_4 happen simultaneously. Thus, before playing the game, any team cannot know the result of the other game. We introduce two variables: A_{12}, A_{34} , which represent the real probability for a_1 defeats a_2 , and a_3 defeats a_4 . The real probability of A_{12} is the sum of winning probability for a_1 under four situations: a_1 tries to win and a_2 tries to win, a_1 tries to win and a_2 loses intentionally, a_1 loses intentionally and a_2 tries to win, a_1 loses intentionally and a_2 loses intentionally. Recall that in this section, we still assume that if both team lose intentionally, we can flip a fair coin to decide who wins the game. Thus, the formula of A_{12} is:

$$A_{12} = p_{12}\pi_1\pi_2 + \pi_1(1 - \pi_2) + \frac{1}{2}(1 - \pi_1)(1 - \pi_2) \quad (16)$$

Similarly,

$$A_{34} = p_{34}\pi_3\pi_4 + \pi_3(1 - \pi_4) + \frac{1}{2}(1 - \pi_3)(1 - \pi_4) \quad (17)$$

Next, we define the payoff function as the probability to win the champion under given strategy. Define $Q_{ij}, i \in \{1, 2, 3, 4\}, j \in \{a, b, c, d\}$ as the probability for a_i to win the tournament under condition

j , such that Q_{1a}, Q_{2a} , represents the probability for a_1, a_2 to win the champion if both a_1 and a_2 tries to win, respectively. Q_{1b}, Q_{2b} , represents the probability for a_1, a_2 to win the champion if a_1 loses intentionally and a_2 tries to win, respectively. Q_{1c}, Q_{2c} , represents the probability for a_1, a_2 to win the champion if a_1 tries to win and a_2 loses intentionally, respectively. Q_{1d}, Q_{2d} , represents the probability for a_1, a_2 to win the champion if both a_1 and a_2 loses intentionally, respectively. Similarly, we define Q_{3a}, Q_{4a} as the probability for a_3, a_4 to win the champion if both a_3 and a_4 tries to win, respectively. Q_{3b}, Q_{4b} , represents the probability for a_3, a_4 to win the champion if a_3 loses intentionally and a_4 tries to win, respectively. Q_{3c}, Q_{4c} , represents the probability for a_3, a_4 to win the champion if a_3 tries to win and a_4 loses intentionally, respectively. Q_{3d}, Q_{4d} , represents the probability for a_3, a_4 to win the champion if both a_3 and a_4 loses intentionally, respectively.

To calculate the payoff functions, we show some examples. For example, to calculate Q_{1a} , we notice that if a_3 defeats a_4 , then $W = [1, 2, 1, 1]$. If both a_1 and a_2 tries to win and if a_1 wins, then $W = [2, 2, 1, 1]$, recall our flipping coin assumption, a_1 has $\frac{1}{2}(T_{1A}+T_{1B})$ probability to win the champion. If a_2 wins, then $W = [1, 3, 1, 1]$, a_1 has $\frac{1}{3}(T_{1A} + T_{1B} + T_{1C})$ probability to win the champion. Otherwise, if a_4 defeats a_3 , then $W = [1, 2, 0, 2]$, then if a_1 wins, $W = [2, 2, 0, 2]$, a_1 has $\frac{1}{3}(T_{1A} + T_{1B} + T_{1C})$ probability to win the champion. If a_2 wins, $W = [1, 3, 0, 2]$, a_1 will enter tournament A , and has T_{1A} probability to win the champion. Thus,

$$Q_{1a} = A_{34}(p_{12}\frac{1}{2}(T_{1A}+T_{1B})+(1-p_{12})\frac{1}{3}(T_{1A}+T_{1B}+T_{1C}))+ (1-A_{34})(p_{12}\frac{1}{3}((T_{1A}+T_{1B}+T_{1C})+(1-p_{12})T_{1A})) \quad (18)$$

Notice that if $i = 1, 2$, Q_{ij} is a function with parameter $v_1, v_2, v_3, v_4, A_{34}$. Since A_{34} is a function with parameter π_3, π_4 ,

$$Q_{ij} = F_{ij}(v_1, v_2, v_3, v_4, \pi_3, \pi_4) \quad (19)$$

Similarly, if $i = 3, 4$,

$$Q_{ij} = F_{ij}(v_1, v_2, v_3, v_4, \pi_1, \pi_2) \quad (20)$$

Now, we want to introduce our method to calculate $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$. In our algorithm, the main logic is to write the probability for each team to win the champion as a function with variable π and Q_{ij} . By the definition of the Nash equilibrium, if all teams own mixed strategy, then the partial derivatives of probability for team a_i to win the champion with respect to the variable π_i all equal to 0. If not, the teams will have pure strategy. Thus, we calculate the partial derivative of the probability for each team to win the champion respect to the team i 's strategy π_i and get the solutions of π . We will expand the analysis of how to solve π now. Recall that in our main logic, firstly, we have to find the probability of each team to win the champion. Let E_1, E_2, E_3, E_4 be probability of team a_1, a_2, a_3, a_4 to win the tournament, respectively. Recall that $Q_{ij}, i \in \{1, 2, 3, 4\}, j \in \{a, b, c, d\}$ is the probability for a_i to win the champion under condition j . Define $P(j)$ as the probability of event j . Here for team a_i , we have

$$E_i = \sum_{j \in \{a, b, c, d\}} Q_{ij}P(j) \quad (21)$$

a ₁	Try to Win	Try to Lose
a ₂		
Try to Win	Q _{1a} , Q _{2a}	Q _{1b} , Q _{2b}
Try to Lose	Q _{1c} , Q _{2c}	Q _{1d} , Q _{2d}

a ₃	Try to Win	Try to Lose
a ₄		
Try to Win	Q _{3a} , Q _{4a}	Q _{3b} , Q _{4b}
Try to Lose	Q _{3c} , Q _{4c}	Q _{3d} , Q _{4d}

Figure 10: The game theory table for four teams

For example, recall that $Q_{1a}, Q_{1b}, Q_{1c}, Q_{1d}$ represents the probability for a_1 to win the tournament if both a_1 and a_2 tries to win, if a_1 loses intentionally and a_2 tries to win, if a_1 tries to win and a_2 loses intentionally, and if both a_1 and a_2 loses intentionally, respectively. Then by equation 21, we have $E_1 = \pi_1\pi_2Q_{1a} + (1 - \pi_1)\pi_2Q_{1b} + \pi_1(1 - \pi_2)Q_{1c} + (1 - \pi_1)(1 - \pi_2)Q_{1d}$. Similarly, we can write the following equation system.

$$\begin{cases} E_1 = \pi_1\pi_2Q_{1a} + (1 - \pi_1)\pi_2Q_{1b} + \pi_1(1 - \pi_2)Q_{1c} + (1 - \pi_1)(1 - \pi_2)Q_{1d}, \\ E_2 = \pi_1\pi_2Q_{2a} + (1 - \pi_1)\pi_2Q_{2b} + \pi_1(1 - \pi_2)Q_{2c} + (1 - \pi_1)(1 - \pi_2)Q_{2d}, \\ E_3 = \pi_3\pi_4Q_{3a} + (1 - \pi_3)\pi_4Q_{3b} + \pi_3(1 - \pi_4)Q_{3c} + (1 - \pi_3)(1 - \pi_4)Q_{3d}, \\ E_4 = \pi_3\pi_4Q_{4a} + (1 - \pi_3)\pi_4Q_{4b} + \pi_3(1 - \pi_4)Q_{4c} + (1 - \pi_3)(1 - \pi_4)Q_{4d}, \end{cases} \quad (22)$$

Secondly, we want to see how each team's strategy effects their winning probability. We calculate the partial derivative for each probability in equation system 22 with respect to the variable π_i , recall that Q_{1j}, Q_{2j} only depend on π_3 and π_4 , Q_{3j}, Q_{4j} only depend on π_1 and π_2 , then, we can get

$$\begin{cases} \frac{\partial E_1}{\partial \pi_1} = \pi_2(Q_{1a} - Q_{1b} - Q_{1c} + Q_{1d}) + Q_{1c} - Q_{1d}, \\ \frac{\partial E_2}{\partial \pi_2} = \pi_1(Q_{2a} - Q_{2b} - Q_{2c} + Q_{2d}) + Q_{2b} - Q_{2d}, \\ \frac{\partial E_3}{\partial \pi_3} = \pi_4(Q_{3a} - Q_{3b} - Q_{3c} + Q_{3d}) + Q_{3c} - Q_{3d}, \\ \frac{\partial E_4}{\partial \pi_4} = \pi_3(Q_{4a} - Q_{4b} - Q_{4c} + Q_{4d}) + Q_{4b} - Q_{4d}, \end{cases} \quad (23)$$

From equation system 23, we notice that $\frac{\partial E_i}{\partial \pi_i}$ is linearly related to the strategy variable π_j , where a_j is the rival of a_i in the last week. Hence, we can draw the following conclusion:

Theorem 3.1. *If $\frac{\partial E_i}{\partial \pi_i} \neq 0$, then team a_i should make a pure strategy. Moreover, if $\frac{\partial E_i}{\partial \pi_i} > 0$, team a_i should try to win. If $\frac{\partial E_i}{\partial \pi_i} < 0$, team a_i should lose intentionally.*

Proof. Notice that if $\frac{\partial E_i}{\partial \pi_i} > 0$, then $\frac{\partial E_i}{\partial \pi_i}$ is a linear function with positive slope. We get the maximum of E_i if π_i achieves its maximum, which is 1. Similarly, if $\frac{\partial E_i}{\partial \pi_i} < 0$, then we get the maximum of E_i if π_i achieves its minimum, which is 0. \square

Now, we focus on finding the Nash equilibrium. We first apply the Nash's Existence Theorem to show the existence of the Nash equilibrium π .

Theorem 3.2 (Nash's Existence Theorem). *[1] If we allow mixed strategies (where a pure strategy is chosen at random, subject to some fixed probability), then every game with a finite number of players in which each player can choose from finitely many pure strategies has at least one Nash equilibrium.*

In our problem setting, the number of players is finite, each player has two pure strategies: win or lose. Hence, we know that the Nash equilibrium exists in our problem setting. The next proposition shows that the teams can have mixed strategies if and only if all partial derivatives of winning probabilities equals to zero.

Proposition 3.3. *For all m , W , and possible V , $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$ is the Nash equilibrium if and only if*

$$\begin{cases} \frac{\partial E_1}{\partial \pi_1} = 0, \\ \frac{\partial E_2}{\partial \pi_2} = 0, \\ \frac{\partial E_3}{\partial \pi_3} = 0, \\ \frac{\partial E_4}{\partial \pi_4} = 0, \end{cases} \quad (24)$$

Recall that we have mentioned that at this moment, we do not know the value of $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$, hence we solve the equation system in proposition 3.3 without expanding A_{12} and A_{34} , we can get

$$\begin{cases} A_{34} = \frac{4T_{1A} - 2T_{1B} - 2T_{1C}}{5T_{1A} - T_{1B} - 4T_{1C}}, \\ A_{34} = \frac{4T_{2A} - 2T_{2B} - 2T_{2C}}{5T_{2A} - T_{2B} - 4T_{2C}}, \\ A_{12} = \frac{4T_{3A} - 2T_{3C} - 2T_{3B}}{5T_{3A} - T_{3B} - 4T_{3C}}, \\ A_{12} = \frac{4T_{4A} - 2T_{4C} - 2T_{4B}}{5T_{4A} - T_{4B} - 4T_{4C}}, \end{cases} \quad (25)$$

From equation system 25, we know that the Nash equilibrium follows

$$A_{34} = \frac{4T_{1A} - 2T_{1B} - 2T_{1C}}{5T_{1A} - T_{1B} - 4T_{1C}} = \frac{4T_{2A} - 2T_{2B} - 2T_{2C}}{5T_{2A} - T_{2B} - 4T_{2C}} \quad (26)$$

$$A_{12} = \frac{4T_{3A} - 2T_{3C} - 2T_{3B}}{5T_{3A} - T_{3B} - 4T_{3C}} = \frac{4T_{4A} - 2T_{4C} - 2T_{4B}}{5T_{4A} - T_{4B} - 4T_{4C}} \quad (27)$$

We have to solve these two equations. We know that the strategy is related to all team weights. Next, we claim the following theorem:

Theorem 3.4. *The only solution of 26, 27 is that $v_1 = v_2 = 1$, $v_3 = v_4 \leq 1$.*

By theorem 3.4, we know that the only situation where all teams have mixed strategies is when a_2 is as strong as a_1 , a_3 is as strong as a_4 . Next, we will give a proof of theorem 3.4.

Proof. Equation 26 implies that

$$T_{1B}T_{2C} - T_{1B}T_{2A} + T_{1C}T_{2A} + T_{1A}T_{2B} - T_{1A}T_{2C} - T_{1C}T_{2B} = 0 \quad (28)$$

We can write 28 as

$$T_{1A}(T_{2B} - T_{2C}) + T_{1B}(T_{2C} - T_{2A}) + T_{1C}(T_{2A} - T_{2B}) = 0 \quad (29)$$

To solve 29, we have to introduce the Chebyshev's sum inequality [2]

Lemma 3.2 (Chebyshev's sum inequality). *If $a_1 \geq a_2 \geq \dots \geq a_n$, $b_1 \geq b_2 \geq \dots \geq b_n$, then*

$$\frac{1}{n} \sum_{k=1}^n a_k b_k \geq \left(\frac{1}{n} \sum_{k=1}^n a_k \right) \left(\frac{1}{n} \sum_{k=1}^n b_k \right) \quad (30)$$

Recall lemma 3.1, $T_{1A} \geq T_{1B} \geq T_{1C}, T_{2B} \geq T_{2A} \geq T_{2C}$, if $(T_{2C} - T_{2A}) \geq (T_{2A} - T_{2B})$, then by Chebyshev's sum inequality,

$$T_{1B}(T_{2C} - T_{2A}) + T_{1C}(T_{2A} - T_{2B}) \geq \frac{1}{2}(T_{1B} + T_{1C})(T_{2C} - T_{2A} + T_{2A} - T_{2B}) = \frac{1}{2}(T_{1B} + T_{1C})(T_{2C} - T_{2B}) \quad (31)$$

Applying 31 to 29, we get

$$0 \geq T_{1A}(T_{2B} - T_{2C}) + \frac{1}{2}(T_{1B} + T_{1C})(T_{2C} - T_{2B}) = (T_{1A} - \frac{T_{1B} + T_{1C}}{2})(T_{2B} - T_{2C}) \quad (32)$$

This only happens when $T_{1A} = T_{1B} = T_{1C}$. Otherwise,

$$T_{2C} - T_{2A} \leq T_{2A} - T_{2B} \quad (33)$$

We can also write equation 28 as $T_{2B}(T_{1A} - T_{1C}) + T_{2A}(T_{1C} - T_{1B}) + T_{2C}(T_{1B} - T_{1A}) = 0$. By applying Chebyshev's sum inequality, we know that if $T_{1C} - T_{1B} \geq T_{1B} - T_{1A}$, the only possible solution is that $T_{2A} = T_{2B} = T_{2C}$. Otherwise,

$$T_{1C} - T_{1B} \leq T_{1B} - T_{1A} \quad (34)$$

Similarly, we apply the same method to equation 27. We get

$$T_{4A} - T_{4B} \leq T_{4B} - T_{4C} \quad (35)$$

$$T_{3B} - T_{3A} \leq T_{3A} - T_{3C} \quad (36)$$

By solving 33, 34, 35, 36, we get that the only solution is that $v_1 = v_2 = 1$, $v_3 = v_4 \leq 1$. \square

Now, we know that if $v_1 = v_2 = 1$, $v_3 = v_4 \leq 1$, every team has mixed strategy. Next, we want to find what the mixed strategy is. In this situation,

$$T_{1A} = T_{1B} = T_{2A} = T_{2B} \geq T_{1C} = T_{2C} \quad (37)$$

$$T_{3C} = T_{4C} = 1 - T_{1C} \quad (38)$$

$$T_{3A} = T_{3B} = T_{4A} = T_{4B} = 1 - T_{1A} \quad (39)$$

We re-calculate A_{12} and A_{34} by plugging 37, 38, 39 to equation 26, 27, we can get

$$A_{34} = \frac{4T_{1A} - 2T_{1B} - 2T_{1C}}{5T_{1A} - T_{1B} - 4T_{1C}} = \frac{2T_{1A} - 2T_{1C}}{4T_{1A} - 4T_{1C}} = \frac{1}{2} \quad (40)$$

Similarly, $A_{12} = \frac{1}{2}$.

Corollary 3.5. *Under the situation where $v_1 = v_2 = 1$, $v_3 = v_4 \leq 1$, the mixed strategy $(\pi_1, \pi_2, \pi_3, \pi_4)$ satisfies $\pi_1 = \pi_2$, $\pi_3 = \pi_4$*

Proof. By expanding A_{34} ,

$$A_{34} = p_{34}\pi_3\pi_4 + \pi_3(1 - \pi_4) + \frac{1}{2}(1 - \pi_3)(1 - \pi_4) = \pi_3\pi_4 + 2\pi_3 - 2\pi_3\pi_4 + \frac{1}{2} - \pi_3 - \pi_4 + \pi_3\pi_4 = \frac{1}{2} \quad (41)$$

We can get $\pi_3 = \pi_4$, similarly, by expanding A_{12} , we can also get $\pi_1 = \pi_2$. \square

Next, we re-calculate the probability for each team to win the champion, E_i , $i = \{1, 2, 3, 4\}$, under $\pi_1 = \pi_2$, $\pi_3 = \pi_4$.

Corollary 3.6. *Under $\pi_1 = \pi_2$, $\pi_3 = \pi_4$, the probability for each team to win the champion, E_i , does not depend on $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$.*

Proof.

$$\begin{aligned} E_1 &= \pi_1\pi_2Q_{1a} + (1 - \pi_1)\pi_2Q_{1b} + \pi_1(1 - \pi_2)Q_{1c} + (1 - \pi_1)(1 - \pi_2)Q_{1d} \\ &= \pi_1^2(Q_{1a} - Q_{1b} - Q_{1c} + Q_{1d}) + \pi_1(Q_{1b} + Q_{1c} - 2Q_{1d}) + Q_{1d} \end{aligned}$$

Due to $T_{1A} = T_{1B}$, also by equation 40 we know $A_{12} = A_{34} = \frac{1}{2}$, we can get

$$Q_{1a} - Q_{1b} - Q_{1c} + Q_{1d} = 0 \quad (42)$$

$$Q_{1b} + Q_{1c} - 2Q_{1d} = 0 \quad (43)$$

Thus, $E_1 = Q_{1d}$. Similarly, we can get $E_2 = Q_{2d}, E_3 = Q_{3d}, E_4 = Q_{4d}$. This proves that E_i is independent with the strategy variable π . \square

Hence, we can draw the following conclusion:

Theorem 3.7. Conclusion

If $v_1 = v_2 = 1$, $v_3 = v_4 \leq 1$, then all $\pi_1 = \pi_2$, $\pi_3 = \pi_4$ are Nash equilibriums. Otherwise, all teams should use pure strategy.

Now, we want to raise a question: How to compute the Nash equilibrium if $\pi_1 \neq \pi_2$ or $\pi_3 \neq \pi_4$? From theorem 3.7, we know that all teams have pure strategy if $\pi_1 \neq \pi_2$ or $\pi_3 \neq \pi_4$, i.e. $\pi \in \{0, 1\}^4$. Since $\text{card}(\{0, 1\}^4) = 2^4 = 16$, we can plug all these 16 possible π to the verification process: given a vector π , we can calculate all Q_{ij} and draw the game theory table. Then, we can check whether the game theory problem has dominant strategy or not. By theorem 3.2, we know that there exists some solutions among the 16 possible π .

If we are the coach of a sport team, to determine whether to try to win the next game or to lose intentionally, we can use our algorithm to get the answer. We first approximate the team weights v_1 , v_2 , v_3 , v_4 by previous data. If the win vector before the next game is W , by applying our algorithm, we can get the strategy $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)$ and the probability of winning the champion of each team E_1, E_2, E_3, E_4 .

4 Future Works

One important question is that whether theorem 3.7 holds when there are eight teams. A big difference between the four-team model and eight-team model is that under the four-team model, lemma 3.1 holds for any $v_1 \geq v_2 \geq v_3 \geq v_4$. However, under the eight-team model, according to the model created by J. Schwenk [3], he showed that under some specific situations, the second seed has larger probability than the first seed to win the tournament, which implies that we cannot generalize lemma 3.1 for the eight-team model. Hence, one of the future work is to find the Nash equilibrium under the eight-team model. One possible method is following the definition of Nash equilibrium, proposition 3.3, to solve the equation system. However, without lemma 3.1, we may not draw the conclusion that all teams should use pure strategy for most cases.

References

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Appendices

A Proofs for FRNS Model

A.1 Proof of Theorem 2.1

Proof.

$$\begin{aligned}
 T_A &= \frac{v_1}{v_1 + v_4} \left(\frac{v_1}{v_1 + v_2} \frac{v_2}{v_2 + v_3} + \frac{v_1}{v_1 + v_3} \frac{v_3}{v_2 + v_3} \right) = \frac{v_1^2 [v_2(v_1 + v_3) + v_3(v_1 + v_2)]}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)(v_2 + v_3)} \\
 T_B &= \frac{v_1}{v_1 + v_3} \left(\frac{v_1}{v_1 + v_2} \frac{v_2}{v_2 + v_4} + \frac{v_1}{v_1 + v_4} \frac{v_4}{v_2 + v_4} \right) = \frac{v_1^2 [v_2(v_1 + v_4) + v_3(v_2 + v_4)]}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)(v_2 + v_4)} \\
 T_C &= \frac{v_1}{v_1 + v_2} \left(\frac{v_1}{v_1 + v_3} \frac{v_3}{v_3 + v_4} + \frac{v_1}{v_1 + v_4} \frac{v_4}{v_3 + v_4} \right) = \frac{v_1^2 [v_3(v_1 + v_4) + v_4(v_1 + v_3)]}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)(v_3 + v_4)}
 \end{aligned}$$

Compare P_A and P_B , we compute

$$\begin{aligned}
 T_A - T_B &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) \left(\frac{v_2(v_1 + v_3) + v_3(v_1 + v_2)}{v_2 + v_3} - \frac{v_2(v_1 + v_4) + v_3(v_2 + v_4)}{v_2 + v_4} \right) \\
 &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (v_1v_2 + 2v_2v_3 + v_1v_3)(v_2 + v_4) - (v_1v_2 + v_2v_3 + v_2v_4 + v_3v_4)(v_2 + v_3) \\
 &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_2v_3^2 - 2v_3^2v_4) = \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_3^2(v_2 - v_4))
 \end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_A - T_B \geq 0$

Then, compare T_B and T_C , we compute

$$\begin{aligned}
 T_B - T_C &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) \left(\frac{v_2(v_1 + v_4) + v_3(v_2 + v_4)}{v_2 + v_4} - \frac{v_3(v_1 + v_4) + v_4(v_1 + v_3)}{v_3 + v_4} \right) \\
 &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) ((v_1v_2 + v_2v_4 + v_2v_3 + v_3v_4)(v_3 + v_4) - (v_1v_3 + v_1v_4 + 2v_3v_4)(v_2 + v_4)) \\
 &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) ((v_3^2v_4 - v_3v_4^2) + (v_2v_3^2 - v_2v_3v_4) + (v_1v_2v_4 - v_1v_3v_4)) \\
 &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (v_3v_4(v_3 - v_4) + v_2v_3(v_2 - v_4) + v_1v_4(v_2 - v_3))
 \end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_B - T_C \geq 0$

Thus, $T_A \geq T_B \geq T_C$ for any v_1, v_2, v_3, v_4 , subject to $v_1 \geq v_2 \geq v_3 \geq v_4$. \square

A.2 Proof of Theorem 2.2

Proof.

$$\begin{aligned}
P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) &= (p_{12}p_{34}T_A + p_{12}p_{43}T_B + p_{21}p_{34}T_B + p_{21}p_{43}T_A) \\
&\quad - (p_{34}T_A + p_{43}T_B) \\
&= T_A p_{12}(2p_{34} - 1) + T_B p_{12}(1 - 2p_{34}) \\
&= (T_A - T_B)p_{12}(2p_{34} - 1)
\end{aligned}$$

$\because T_A \geq T_B, p_{34} \geq \frac{1}{2}$, i.e. $2p_{34} - 1 \geq 0$, $\therefore P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) \geq P_{\pi=(0,1,1,1)}^{(1)}(3, W, V)$ for all $V = [v_1, v_2, v_3, v_4]$.

This implies that it is always sensible for a_1 to try to win if the performance vector W of first two week is $[2, 2, 0, 0]$. \square

A.3 Proof of Theorem 2.3

Proof. Under $W = [1, 2, 0, 1]$ or $[0, 1, 1, 2]$, we have

$$\begin{aligned}
P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) &= (p_{21} - 1)p_{34}T_A + (p_{21} - 1)p_{43}\frac{1}{3}(T_A + T_B + T_C) \\
&\quad + p_{12}p_{43}\frac{1}{3}(T_A + T_B + T_C) + p_{21}p_{43}\frac{1}{2}(T_A + T_B) \\
&= p_{12} \left(-\frac{2}{3}T_A + \frac{1}{3}T_B + \frac{1}{3}T_C + p_{34}\left(\frac{5}{6}T_A - \frac{1}{6}T_B - \frac{2}{3}T_C\right) \right)
\end{aligned}$$

Since we want to find $V = (v_1, v_2, v_3, v_4)$ such that $P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) \geq 0$, which is equal to solve for $p_{34} \geq \frac{4T_A - 2T_B - 2T_C}{5T_A - T_B - 4T_C}$. Hence, we want to solve the following inequality:

$$\frac{v_3}{v_3 + v_4} \geq \frac{4T_A - 2T_B - 2T_C}{5T_A - T_B - 4T_C} \tag{44}$$

By expanding the right side of the inequality, we can get

$$\begin{aligned}
&3v_2v_3v_4^2 + 2v_2v_4^3 + 6v_2^2v_3^2v_4 + v_2v_3^2v_4 + v_3^3v_4 + 2v_3v_4^3 + 10v_3^2v_4^2 - 3v_2^2v_3^3 \\
&\quad - v_2^2v_3v_4^2 - 2v_2^2v_4^3 - v_2v_3^4 - 3v_2v_3^3v_4 - 2v_2v_3^2v_4^2 - v_3^4v_4 - 6v_3^3v_4^2 - 6v_3^2v_4^3 \leq 0
\end{aligned}$$

which leads to theorem 2.3. \square

A.4 Proof of Theorem 2.4

Proof. Under $W = [1, 2, 1, 0]$, we have

$$\begin{aligned} P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) &= (p_{21} - 1)p_{34}T_B + (p_{21} - 1)p_{43}\frac{1}{3}(T_A + T_B + T_C) \\ &\quad + p_{12}p_{34}\frac{1}{2}(T_A + T_B) + p_{12}p_{43}\frac{1}{3}(T_A + T_B + T_C) \\ &= p_{12}\left(\frac{1}{6}T_A + \frac{1}{6}T_B - \frac{1}{3}T_C + p_{34}\left(\frac{1}{6}T_A - \frac{5}{6}T_B + \frac{2}{3}T_C\right)\right) \end{aligned}$$

Since we want to find $V = (v_1, v_2, v_3, v_4)$ such that $P_{\pi=(1,1,1,1)}^{(1)}(3, W, V) - P_{\pi=(0,1,1,1)}^{(1)}(3, W, V) \geq 0$, which is equal to solve for $p_{34} \geq \frac{T_A + T_B - 2T_C}{-T_A + 5T_B - 4T_C}$. Hence, we want to solve the following inequality:

$$\frac{v_3}{v_3 + v_4} \geq \frac{T_A + T_B - 2T_C}{-T_A + 5T_B - 4T_C} \quad (45)$$

By expanding the right side of the inequality, we can get

$$\begin{aligned} 3v_2^2v_3^2v_4 + v_2v_3^4 + v_2v_3^3v_4 + v_3^3v_4^2 + 4v_3^3v_4 + 4v_3^2v_4^2 + 5v_2v_3v_4 \\ - 4v_2^2v_3v_4^2 - 5v_2^2v_4^3 - 4v_3^4v_4 - 2v_2v_3v_4^2 - v_3^2v_4 - v_3v_4^2 \leq 0 \end{aligned}$$

which leads to theorem 2.4. □

A.5 Proof of Lemma 3.1

Proof. By theorem 2.1, we know that if $v_1 \geq v_2 \geq v_3 \geq v_4$, then

$$T_{1A} \geq T_{1B} \geq T_{1C} \quad (46)$$

Similar to theorem 2.1, we calculate T_{2A} , T_{2B} , T_{2C} ,

$$\begin{aligned} T_{2A} &= \frac{v_2}{v_2 + v_3} \left(\frac{v_2}{v_1 + v_2} \frac{v_1}{v_1 + v_4} + \frac{v_2}{v_2 + v_4} \frac{v_4}{v_1 + v_4} \right) = \frac{v_2^2[v_1(v_2 + v_4) + v_4(v_1 + v_2)]}{(v_1 + v_2)(v_1 + v_4)(v_2 + v_3)(v_2 + v_4)} \\ T_{2B} &= \frac{v_2}{v_2 + v_4} \left(\frac{v_2}{v_1 + v_2} \frac{v_1}{v_1 + v_3} + \frac{v_2}{v_2 + v_3} \frac{v_3}{v_1 + v_3} \right) = \frac{v_2^2[v_1(v_2 + v_3) + v_3(v_1 + v_2)]}{(v_1 + v_2)(v_1 + v_3)(v_2 + v_3)(v_2 + v_4)} \\ T_{2C} &= \frac{v_2}{v_1 + v_2} \left(\frac{v_2}{v_2 + v_3} \frac{v_3}{v_3 + v_4} + \frac{v_2}{v_2 + v_4} \frac{v_4}{v_3 + v_4} \right) = \frac{v_2^2[v_3(v_2 + v_4) + v_4(v_2 + v_3)]}{(v_1 + v_2)(v_3 + v_4)(v_2 + v_3)(v_2 + v_4)} \end{aligned}$$

Compare T_{2B} and T_{2A} , we compute

$$\begin{aligned}
T_{2B} - T_{2A} &= \left(\frac{v_2^2}{(v_1 + v_2)(v_2 + v_3)(v_2 + v_4)} \right) \left(\frac{v_1(v_2 + v_3) + v_3(v_1 + v_2)}{v_1 + v_3} - \frac{v_1(v_2 + v_4) + v_4(v_1 + v_2)}{v_1 + v_4} \right) \\
&= \left(\frac{v_2^2}{(v_1 + v_2)(v_2 + v_3)(v_2 + v_4)} \right) (v_1v_2 + 2v_1v_3 + v_2v_3)(v_1 + v_4) - (v_1v_2 + 2v_1v_4 + v_2v_4)(v_1 + v_3) \\
&= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_1^2v_3 - 2v_1^2v_4) = \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_1^2(v_3 - v_4))
\end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_{2B} - T_{2A} \geq 0$

Then, compute $T_{2A} - T_{2C}$, we can get

$$\begin{aligned}
T_{2A} - T_{2C} &= \left(\frac{v_2^2}{(v_1 + v_2)(v_2 + v_3)(v_2 + v_4)} \right) \left(\frac{v_1(v_2 + v_4) + v_4(v_1 + v_2)}{v_1 + v_4} - \frac{v_3(v_2 + v_4) + v_4(v_2 + v_3)}{v_3 + v_4} \right) \\
&= \left(\frac{v_2^2}{(v_1 + v_2)(v_2 + v_3)(v_2 + v_4)} \right) (v_1v_2 + 2v_1v_4 + v_2v_4)(v_3 + v_4) - (v_2v_3 + v_2v_4 + 2v_3v_4)(v_1 + v_4) \\
&= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_1v_4^2 - 2v_3v_4^2) = \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_4^2(v_1 - v_3))
\end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_{2A} - T_{2C} \geq 0$

Hence, we can draw the conclusion that

$$T_{2B} \geq T_{2A} \geq T_{2C} \quad (47)$$

Next, we calculate T_{3A}, T_{3B}, T_{3C} ,

$$\begin{aligned}
T_{3A} &= \frac{v_3}{v_2 + v_3} \left(\frac{v_3}{v_1 + v_3} \frac{v_1}{v_1 + v_4} + \frac{v_3}{v_3 + v_4} \frac{v_4}{v_1 + v_4} \right) = \frac{v_3^2[v_1(v_3 + v_4) + v_4(v_1 + v_3)]}{(v_1 + v_3)(v_1 + v_4)(v_2 + v_3)(v_3 + v_4)} \\
T_{3B} &= \frac{v_3}{v_1 + v_3} \left(\frac{v_3}{v_2 + v_3} \frac{v_2}{v_2 + v_4} + \frac{v_3}{v_3 + v_4} \frac{v_4}{v_2 + v_4} \right) = \frac{v_3^2[v_2(v_3 + v_4) + v_4(v_2 + v_3)]}{(v_1 + v_3)(v_2 + v_3)(v_2 + v_4)(v_3 + v_4)} \\
T_{3C} &= \frac{v_3}{v_3 + v_4} \left(\frac{v_3}{v_1 + v_3} \frac{v_1}{v_1 + v_2} + \frac{v_3}{v_2 + v_3} \frac{v_2}{v_1 + v_2} \right) = \frac{v_3^2[v_1(v_2 + v_3) + v_2(v_1 + v_3)]}{(v_1 + v_2)(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)}
\end{aligned}$$

Compare T_{3C} and T_{3A} , we compute

$$\begin{aligned}
T_{3C} - T_{3A} &= \left(\frac{v_3^2}{(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)} \right) \left(\frac{v_1(v_2 + v_3) + v_2(v_1 + v_3)}{v_1 + v_2} - \frac{v_1(v_3 + v_4) + v_4(v_1 + v_3)}{v_1 + v_4} \right) \\
&= \left(\frac{v_3^2}{(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)} \right) (2v_1v_2 + v_1v_3 + v_2v_3)(v_1 + v_4) - (v_1v_3 + 2v_1v_4 + v_3v_4)(v_1 + v_2) \\
&= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_1^2v_2 - 2v_1^2v_4) = \left(\frac{v_3^2}{(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)} \right) (2v_1^2(v_2 - v_4))
\end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_{3C} - T_{3A} \geq 0$

Then, we compare T_{3A} and T_{3B} ,

$$\begin{aligned} T_{3A} - T_{3B} &= \left(\frac{v_3^2}{(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)} \right) \left(\frac{v_1(v_3 + v_4) + v_4(v_1 + v_3)}{v_1 + v_4} - \frac{v_2(v_3 + v_4) + v_4(v_2 + v_3)}{v_2 + v_4} \right) \\ &= \left(\frac{v_3^2}{(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)} \right) ((v_1v_3 + 2v_1v_4 + v_3v_4)(v_2 + v_4) - (v_2v_3 + 2v_2v_4 + v_3v_4)(v_1 + v_4)) \\ &= \left(\frac{v_1^2}{(v_1 + v_4)(v_1 + v_2)(v_1 + v_3)} \right) (2v_1v_4^2 - 2v_2v_4^2) = \left(\frac{v_3^2}{(v_1 + v_3)(v_2 + v_3)(v_3 + v_4)} \right) (2v_4^2(v_1 - v_2)) \end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_{3A} - T_{3B} \geq 0$

Hence, we can draw the conclusion that

$$T_{3C} \geq T_{3A} \geq T_{3B} \quad (48)$$

Finally, we calculate T_{4A}, T_{4B}, T_{4C} ,

$$\begin{aligned} T_{4A} &= \frac{v_4}{v_1 + v_4} \left(\frac{v_4}{v_2 + v_4} \frac{v_2}{v_2 + v_3} + \frac{v_4}{v_3 + v_4} \frac{v_3}{v_2 + v_3} \right) = \frac{v_4^2[v_2(v_3 + v_4) + v_3(v_2 + v_4)]}{(v_1 + v_4)(v_2 + v_4)(v_2 + v_3)(v_3 + v_4)} \\ T_{4B} &= \frac{v_4}{v_2 + v_4} \left(\frac{v_4}{v_1 + v_4} \frac{v_1}{v_1 + v_3} + \frac{v_4}{v_3 + v_4} \frac{v_3}{v_1 + v_3} \right) = \frac{v_4^2[v_1(v_3 + v_4) + v_3(v_1 + v_4)]}{(v_1 + v_4)(v_1 + v_3)(v_2 + v_4)(v_3 + v_4)} \\ T_{4C} &= \frac{v_4}{v_3 + v_4} \left(\frac{v_4}{v_1 + v_4} \frac{v_1}{v_1 + v_2} + \frac{v_4}{v_2 + v_4} \frac{v_2}{v_1 + v_2} \right) = \frac{v_4^2[v_1(v_2 + v_4) + v_2(v_1 + v_4)]}{(v_1 + v_2)(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \end{aligned}$$

Compare T_{4C} and T_{4B} , we compute

$$\begin{aligned} T_{4C} - T_{4B} &= \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) \left(\frac{v_1(v_2 + v_4) + v_2(v_1 + v_4)}{v_1 + v_2} - \frac{v_1(v_3 + v_4) + v_3(v_1 + v_4)}{v_1 + v_3} \right) \\ &= \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) (2v_1v_2 + v_1v_4 + v_2v_4)(v_1 + v_3) - (2v_1v_3 + v_1v_4 + v_3v_4)(v_1 + v_2) \\ &= \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) (2v_1^2v_2 - 2v_1^2v_3) = \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) (2v_1^2(v_2 - v_3)) \end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_{4C} - T_{4B} \geq 0$

Then, we compare T_{4B} and T_{4A} ,

$$\begin{aligned} T_{4B} - T_{4A} &= \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) \left(\frac{v_1(v_3 + v_4) + v_3(v_1 + v_4)}{v_1 + v_3} - \frac{v_2(v_3 + v_4) + v_3(v_2 + v_4)}{v_2 + v_3} \right) \\ &= \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) (2v_1v_3 + v_1v_4 + v_3v_4)(v_2 + v_3) - (2v_2v_3 + v_2v_4 + v_3v_4)(v_1 + v_3) \\ &= \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) (2v_1v_3^2 - 2v_2v_3^2) = \left(\frac{v_4^2}{(v_1 + v_4)(v_2 + v_4)(v_3 + v_4)} \right) (2v_3^2(v_1 - v_2)) \end{aligned}$$

$\because v_1 \geq v_2 \geq v_3 \geq v_4, \therefore T_{4B} - T_{4A} \geq 0$

Hence, we can draw the conclusion that

$$T_{4C} \geq T_{4B} \geq T_{4A} \tag{49}$$

□