# Fixing a Balanced Knockout Tournament 

Haris Aziz and Serge Gaspers and Simon Mackenzie and Nicholas Mattei<br>NICTA and UNSW, Sydney, Australia<br>\{haris.aziz, serge.gaspers, simon.mackenzie, nicholas.mattei\}@nicta.com.au

Paul Stursberg<br>Technische Universität München, Germany<br>paul.stursberg@ma.tum.de

Toby Walsh<br>NICTA and UNSW, Sydney, Australia<br>toby.walsh@nicta.com.au


#### Abstract

Balanced knockout tournaments are one of the most common formats for sports competitions, and are also used in elections and decision-making. We consider the computational problem of finding the optimal draw for a particular player in such a tournament. The problem has generated considerable research within AI in recent years. We prove that checking whether there exists a draw in which a player wins is NP-complete, thereby settling an outstanding open problem. Our main result has a number of interesting implications on related counting and approximation problems. We present a memoization-based algorithm for the problem that is faster than previous approaches. Moreover, we highlight two natural cases that can be solved in polynomial time. All of our results also hold for the more general problem of counting the number of draws in which a given player is the winner.


## Introduction

Balanced knockout tournaments are one of the most widelyused formats for sports competitions (Horen and Riezman 1985; Connolly and Rendleman 2011; Groh et al. 2012). A prominent example is the Wimbledon Men's tennis tournament in which 128 players enter the tournament and the player who wins seven consecutive matches right from the first round to the final wins the tournament. The format is also used in certain elimination style election and decision making schemes and has received considerable interest in AI (Vu, Altman, and Shoham 2009; Vassilevska Williams 2010) as well as economics and operations research (Rosen 1986; Vu et al. 2013; Tullock 1980; Laslier 1997).

Consider the setting in which there is a set of players $N=[n]$ (we use the notation $[n]:=\{1, \ldots, n\}$ ) where $n=2^{c}$ for some integer $c$. Given $N$, an ordered balanced knockout tournament $\mathrm{T}(N, \pi)$ is defined as a balanced binary tree with $n$ leaf nodes where the seeding $\pi$ specifies the labelling of the leaf nodes with respect to $N$. All ordered balanced knockout tournaments that are isomorphic to each other (with respect to the labelling of the leaf nodes) are said to have the same draw. They are represented by a single (unordered) balanced knockout tournament (BKT) $\mathrm{T}(N, \sigma)$

[^0]where $\sigma$ denotes the draw. The set of all draws is denoted by $\Sigma$. Whereas the total number of seedings is $n$ !, the number of draws is $\frac{n!}{2^{n-1}}$ but even this grows very rapidly. For a tournament like Wimbledon, $n=128$ and the number of distinct draws is $2.3 * 10^{177}$. This is significantly more than the number of atoms in the universe, or even a googol.

A BKT $\mathrm{T}(N, \sigma)$ is conducted in the following fashion. Players that correspond to sibling leaf nodes play a match against each other. The winner of the match proceeds up the tree to the next round. The winner of $\mathrm{T}(N, \sigma)$ is the player who reaches the root node. We are given a pairwise comparison matrix $P$ such that $P_{j k} \in[0,1]$ denotes the probability of player $j$ beating player $k$ in a pairwise elimination match and $0 \leq P_{i j}=1-P_{j i} \leq 1$. Given $N, P$ and a draw $\sigma$, each player $i \in N$ has a certain probability $\operatorname{wp}(i, N, P, \sigma)$ of being the winner of $\mathrm{T}(N, \sigma)$. This probability can be computed in time $O\left(n^{2}\right)$ via a recursive formulation (Vu, Altman, and Shoham 2009). We denote by $\operatorname{mwp}(i, N, P):=\max _{\sigma \in \Sigma}(\operatorname{wp}(i, N, P, \sigma))$ the maximum possible winning probability of $i$ in $\mathrm{T}(N, \sigma)$ taken over all draws $\sigma \in \Sigma$.

We consider the Probabilistic Tournament Fixing PRoblem (PTFP) in which the probability of each player beating another player is known and the goal is to find a draw that maximizes the probability of a certain player winning the BKT.

Probabilistic Tournament Fixing Problem (PTFP)
Instance: Player set $N$, pairwise comparison matrix $P$, a distinguished player $i^{*} \in N$, and target probability $q \in[0,1]$.
Question: Does there exist a draw $\sigma$ for the player set $N$ for which the probability of $i^{*}$ winning $\mathrm{T}(N, \sigma)$ is at least $q$ ?

PTFP was proposed by Vu, Altman, and Shoham (2009) and has been studied in numerous papers (see e.g., (Stanton and Vassilevska Williams 2011a; 2011c; 2011b)). It is a well-motivated problem in sports analytics (Silver 2011). Vu, Altman, and Shoham (2009) and Vassilevska Williams (2010) showed that PTFP is NP-hard for various restrictions. These results also imply that
computing the maximum possible winning probability of a given player is NP-hard. Nevertheless, the computational complexity of a particularly natural and interesting problem, the Tournament Fixing Problem (TFP), has remained a major open question. In the TFP, $P$ is deterministic i.e., $P_{j k} \in\{0,1\}$ for all players $j, k \in N$. We say that player $j$ beats player $k$ iff $P_{j k}=1$. The winner of each match is deterministically known beforehand and the question is whether there exists a draw for which a given player can win in the corresponding BKT.

## Tournament Fixing Problem (TFP)

Instance: Player set $N$, deterministic pairwise comparison matrix $P$, and a distinguished player $i^{*} \in N$.
Question: Does there exist a draw $\sigma$ for the player set $N$ for which $i^{*}$ is the winner of $\mathrm{T}(N, \sigma)$ ?

TFP is equivalent to checking whether there exists a seeding $\pi$ for which $i$ is the winner of $\mathrm{T}(N, \pi)$. We note that TFP is a special case of the problem with same name as defined in (Vassilevska Williams 2010). ${ }^{1}$ TFP is also a special case of \#TFP - the problem of counting the number of draws for which a given player is the winner. This count can be used to compute the probability of a player winning in a draw chosen uniformly at random. It can also be considered as the relative strength of the player.

Contributions. We first settle the computational complexity of TFP by showing that it is NP-complete. The problem was explicitly stated as an open problem a number of times (Vu, Altman, and Shoham 2009; Vassilevska Williams 2010; Russell and van Beek 2011; Stanton and Vassilevska Williams 2011b; 2011a; 2011c; Lang et al. 2012; Vu et al. 2013). As a corollary, we show that unless $P=N P$, there exists no polynomial-time approximation algorithm for computing the maximal winning probability of a player. The inapproximability result provides additional motivation for the line of work in which heuristic algorithms have been proposed for PTFP (Vu et al. 2013). Another corollary is that there exists no fully polynomial time randomized approximation scheme (FPRAS) for counting the number of draws for which a player is the winner.

In view of these intractability results, we identify two natural cases for which even \#TFP can be solved in polynomial time. In the first case, the players can be divided into a constant number of player types. It appeals to the scenario where players can be divided into groups based on similar intrinsic ability. In the second case, there is a linear ordering on the ability of players with a constant number of exceptions where a player with lower ability beats a player with higher ability. ${ }^{2}$ Finally, we provide an exact memoization-based al-

[^1]gorithm to solve \#TFP that is faster than known exact approaches to solve the problem: it runs in time $O\left(2.8285^{n}\right)$ and uses space $O\left(1.7548^{n}\right)$. If only polynomial space is available, the running time becomes $4^{n+o(n)}$, and we give a range of possible time-space trade-offs.

Related Work. After the work of Vu, Altman, and Shoham (2009), PTFP and TFP have been studied in a number of AI papers. Vassilevska Williams (2010) identified various sufficient conditions for a player to be a winner of a BKT. In a followup paper, Stanton and Vassilevska Williams (2011c) focused on when weak players can possibly win a BKT. Stanton and Vassilevska Williams (2011b; 2011a) identified conditions in a probabilistic model under which the tournament organizer can fix the tournament with high probability. In (Vu and Shoham 2011), the problem of designing 'fair' draws was considered. Lang et al. (2007) and Lang et al. (2012) examined winner determination in voting trees that need not be balanced.

TFP can also be considered as the problem of checking whether a player is a possible winner in a BKT. Computing possible winner for other voting rules where the information on the preferences is not complete has been studied extensively (Xia and Conitzer 2011; Aziz et al. 2012). Another related problem is checking whether a sports team can still win a round-robin competition when all the matches have not yet been completed (Kern and Paulusma 2004; Gusfield and Martel 2002).

## TFP is NP-complete

In this section, we settle the complexity of TFP. For convenience, we will represent the pairwise comparison matrix $P$ as a graph where an edge from $i$ to $j$ exists iff $i$ beats $j$.

## Theorem 1. TFP is NP-complete.

Proof. We reduce from the NP-hard variant of the 3SAT problem in which every literal appears at most twice. Given such a 3SAT instance $F=(X, C)$ where $X=$ $\left\{x_{1}, \ldots, x_{|X|}\right\}$ is the set of variables and $C$ the set of clauses, we build an instance of TFP where a draw exists such that a distinguished player $m_{1}$ wins iff $F$ is satisfiable. The TFP instance consists of a set of players $N=$ $\{1, \ldots, n\}$ where $n$ is the smallest power of 2 greater than $32 \cdot|X|$. The resulting knock-out tournament will thus consist of $R:=\lceil\log (32 \cdot|X|)\rceil-1 \geq 5$ rounds where the first (lowest) four rounds will be used to store the gadgets while the later rounds will enforce certain outcomes for the gadgets. We can decompose the set of players $N$ as follows

$$
\begin{equation*}
N=M \dot{\cup} S \dot{\cup} G^{\mathrm{X}} \dot{\cup} G^{\mathrm{CG}} \dot{\cup} G^{\mathrm{F}} \tag{1}
\end{equation*}
$$

where players in $G^{\mathrm{X}}$ are used in the choice gadgets that will model the variable assignment, players in $G^{\mathrm{CG}}$ are used in the clause/garbage gadgets that will model the behaviour of the clauses, and players in $G^{\mathrm{F}}$ are used in filler gadgets that will be used to balance the BKT. Players in $S$ are special

[^2]

Figure 1: The spawning process. By the shown draw, the leftmost player can win the BKT.
players that will ensure the connection between choice and clause gadgets. Finally we will show that for $m_{1}$ to win the BKT, all $k:=\frac{n}{16}$ players in $M=\left\{m_{1}, m_{2}, \ldots, m_{k}\right\}$ will have to proceed to the fifth round. We will use a total of $k$ gadgets where each $m_{i}$ will be associated with one particular gadget. This gadget will ensure that $m_{i}$ can proceed to the fifth round. More precisely, we use $|X|$ choice gadgets numbered from 1 to $|X|,|X|$ clause/garbage gadgets (that may each contain multiple clauses) numbered from $|X|+1$ to $2|X|$ and $(k-2|X|)$ filler gadgets numbered from $(2|X|+1)$ to $k$. Using this numbering, we can further partition the above sets as follows:

$$
\begin{equation*}
G^{\mathrm{X}}=\bigcup_{j=1}^{|X|} G_{j} ; \quad G^{\mathrm{CG}}=\bigcup_{j=|X|+1}^{2|X|} G_{j} ; \quad G^{\mathrm{F}}=\bigcup_{j=2|X|+1}^{k} G_{j} \tag{2}
\end{equation*}
$$

Note that the sets $G_{j}$ have 10,13 or 15 elements, depending on whether they are a subset of $G^{\mathrm{X}}, G^{\mathrm{CG}}$ or $G^{\mathrm{F}}$, respectively.

Set $M$. The relation between elements of the set of winners $M$ is recursively defined via a linear ordering of players as follows: We start with player $m_{1}$. At each iteration, every player a spawns a new player $b$ placed directly to his right. In the pairwise comparison graph, each player beats all players to his left in this construction except for the one that spawned him. This recursive construction is repeated until a total of $k$ players are present (see Figure 1).

Lemma 1. There is a draw $\sigma$ such that $m_{1}$ wins the BKT $T(M, \sigma)$.

Global structure. We now describe how the sets from (1) and (2) relate to each other. In many places, we will use the right-left-rule, that is elements from sets with a higher index will beat elements from sets with a lower index. For instance, for all $j>j^{\prime}$ and elements $i \in\left\{m_{j}\right\} \cup G_{j}$, $i^{\prime} \in\left\{m_{j^{\prime}}^{\prime}\right\} \cup G_{j}$ where $\left(i, i^{\prime}\right) \neq\left(m_{j}, m_{j^{\prime}}\right)$ we have that $i$ beats $i^{\prime}$. All members of $S$ beat all other players unless explicitly stated otherwise. The set $S$ can be partitioned into subsets $S_{j}$ corresponding to each variable of the SAT instance, such that $\forall j \in[|X|]: S_{j}=S_{x_{j}} \cup S_{\bar{x}_{j}} \cup\left\{s_{j}^{*}\right\}$ where $\left|S_{x_{j}}\right|=3$ and $\left|S_{\bar{x}_{j}}\right|=3$. We further define the set of particles (these will move between the choice and clause/garbage gadgets) by $S_{p}:=\bigcup_{j \in[|X|]}\left(S_{x_{j}} \dot{\cup} S_{\bar{x}_{j}}\right)$. The members of $S_{p}$


Figure 2: Pairwise comparisons in the $j$-th choice gadget. All arrows not shown in the figure run downwards, horizontal arcs run right to left. Vertices grouped in a component have the same relation with vertices outside the component.
follow the right-left rule between each other, i.e., for $j>j^{\prime}$, elements from $S_{x_{j}}$ and $S_{\bar{x}_{j}}$ beat elements from $S_{x_{j^{\prime}}}$ and $S_{\bar{x}_{j^{\prime}}}$. The players in $\left\{s_{j}^{*} \mid j \in[|X|]\right\}$ follow the right-left rule amongst themselves. For each $j \in[|X|], s_{j}^{*}$ is beaten by all other members of $S_{j}$ and beats all members of $S_{p} \backslash S_{j}$.

Choice gadget. For each $j \in[|X|]$, the $j$-th choice gadget consists of player $m_{j}$, all ten elements of $G_{j}$ and all of $S_{j}$. Note that some elements of $S_{j}$ will appear again in the clause/garbage gadgets. The pairwise comparison graph for these elements (for fixed $j$ ) is shown in Figure 2. The choice gadget is structured in such a way that it is possible for $m_{j}$ to win a subtournament composed of all elements in the gadget except two elements of either $S_{x_{j}}$ or $S_{\bar{x}_{j}}$, as illustrated in Figure 4. We will also show that this is the only way in which $m_{j}$ can reach the fifth round in a tournament won by $m_{1}$.

Clause/garbage gadget. We now describe the internal structure of the clause/garbage gadgets. The $j$-th gadget consists of $m_{j}$ and the 13 elements of $G_{j}$, two of these are denoted $c / g$. The pairwise comparison graph for these players is shown in Figure 3. For each clause $c_{i} \in C$ we will call one of the players denoted $c / g$ associated with clause $c_{i}$, all remaining players $c / g$ are garbage players. All players shown in the figure are beaten by all players in $S$ with the following exceptions: (i) garbage players beat all players from $S_{p}$, (ii) players associated with clause $c_{i}$ beat all players from the set $S_{x_{j}}$ or $S_{\bar{x}_{j}}$ if $x_{j}$ or $\bar{x}_{j}$ occurs in clause $c_{i}$, respectively.

The clause gadget is structured in such a way that it is possible for $m_{j}$ to win a subtournament composed of all elements in the gadget with the addition of a compatible $S_{p}$ element for both clause/garbage players included. We will also show that if we want $m_{1}$ to win the tournament, this is the only way in which $m_{j}$ can reach the fifth round.


Figure 3: Pairwise comparison graph for a clause/garbage gadget. All arrows not shown in the figure run downwards, horizontal arcs run right to left.


Figure 4: The subtournament for choice gadget $j$. In this case, $x_{j}=$ true is selected.

The filler gadget $j$ consists of $m_{j}$ and the players in set $G_{j}$. To create the pairwise comparison graph relating them to one another, we use the same method of spawning new players as for $M$, starting from player $m_{j}$ (see Figure 1).

We note that for all $j \in[k]$, player $m_{j}$ beats exactly four players from the set $G_{j}$.

Lemma 2. For all $j \in[k]$ the following hold:
a) Let $\sigma$ be a draw where $\forall i<j$, player $m_{i}$ reaches the fifth round by winning a 16-player subtournament that contains all players from $G_{i}$ and $m_{j}$ plays against all four players from $G_{j}$ that he beats. Then, the 16-player subtournament by which $m_{j}$ proceeds to the fifth round consists exactly of:
i) if the $j$-th gadget is a choice gadget: $\left\{m_{j}\right\} \cup \dot{\cup} G_{j} \dot{\cup} S_{j}$ where two elements of either $S_{x_{j}} \dot{\cup}\left\{s_{j}^{*}\right\}$ or of $S_{\bar{x}_{j}} \dot{\cup}\left\{s_{j}^{*}\right\}$ are removed,
ii) if the $j$-th gadget is a clauselgarbage gadget: $\left\{m_{j}\right\} \dot{\cup}$ $G_{j}$ and for each of the $c / g$ players in $G_{j}$ one additional element of $S_{p}$ that he can beat,
iii) if the $j$-th gadget is a filler gadget: $\left\{m_{j}\right\} \dot{\cup} G_{j}$.
b) If the 16-player subtournament for the first four rounds in which $m_{j}$ is placed consists exactly of $i$ ), ii), or iii) for the respective type of the $j$-th gadget, then a draw for the subtournament exists by which $m_{j}$ reaches the fifth round.

The proof of Lemma 2 is easy but lengthy and thus omitted. Figure 4 shows the relevant subtournament for a choice gadget. We can now show the following, which will also show that for all $j \in[k]$ the antecedent of Lemma 2a) is satisfied and thus the consequent holds.

Lemma 3. For $m_{1}$ to win the tournament, all players $m_{j} \in$ $M$ must reach the fifth round, winning a 16-player subtournament that contains all players from $G_{j}$ by playing against all four players from $G_{j}$ that they can beat.

We can now finally show the following lemma.

Lemma 4. There exists a draw such that $m_{1}$ wins the tournament if and only if the SAT instance is satisfiable.

Proof. First, note that by Lemma 1 and Lemma 3, $m_{1}$ can win the tournament iff all players from the set $M$ reach the fifth round. As only $|M|$ players can reach the fifth round in total, $m_{1}$ can win the tournament iff each player $m_{j} \in M$ can win a 16-player subtournament that consists exactly of the players stated in Lemma 2. We thus have to show that a draw in which each player $m_{j} \in M$ wins a 16-player subtournament with exactly the players stated in Lemma 2 exists iff the 3SAT instance is satisfiable. We show one direction of the proof and omit the other direction.
$(\Rightarrow)$ Let a draw be given such that all players in $M$ win a 16-player subtournament consisting exactly of the players stated in Lemma 2. First, note that whenever a player $s_{j}^{*}$ is not placed in the $j$-th choice gadget, we get a contradiction as by Lemma 2, this player cannot be placed anywhere else while all players from $M$ still reach the fifth round. The following truth assignment is thus well-defined: $A\left(x_{j}\right)=$ true if $m_{j}$ 's subtournament includes all players from $S_{j}$ except two players from $S_{x_{j}}, A\left(x_{j}\right)=$ false if $m_{j}$ 's subtournament includes all players from $S_{j}$ except two players from $S_{\bar{x}_{j}}$

We now show that $A$ satisfies all clauses. Let $c_{i} \in C$ be a clause and $c / g$ the player associated with that clause. Let $j^{*}$ be the 16 -player subtournament including $c / g$. By the assumption, $m_{j^{*}}$ wins the subtournament, thus Lemma $2 \mathrm{im}-$ plies that it contains a player that $c / g$ beats, i.e. a player from set $S_{x_{j}}$ or $\left(S_{\bar{x}_{j}}\right)$ where $x_{j}$ or $\left(\bar{x}_{j}\right)$ occurs in clause $C_{i}$. By the definition of $A$, that player can only be used in the subtournament if the corresponding literal evaluates to true (as otherwise, the player is used in $m_{j}$ 's subtournament).

This concludes the proof of the theorem.

## Implications of Theorem 1

Theorem 1 simultaneously implies a number of results from the literature, in particular Theorem 1, 2, and 3 in (Vu, Altman, and Shoham 2009) and Theorem 1 in (Vassilevska Williams 2010). It also yields some further corollaries. For PTFP and $\alpha \geq 1$, an algorithm is called an $\alpha$-approximation algorithm for the maximum winning probability if it can find a draw in which the winning probability of the given player $i$ is at least $\operatorname{mwp}(i, N, P) / \alpha$.
Corollary 1. Unless $P=N P$, there exists no polynomialtime algorithm for PTFP that approximates the maximum winning probability of a player within any given factor.

It immediately follows from Theorem 1 that \#TFP is NP-hard. We next highlight that even randomisation is not very helpful for \#TFP. Let $\Gamma$ be a finite alphabet in which we agree to describe our problem instances and solutions. A fully polynomial time randomized approximation scheme (FPRAS) for a function $f: \Gamma^{*} \rightarrow \mathbb{Q}$ is an algorithm that takes input $x \in \Gamma^{*}$ and a parameter $\epsilon \in \mathbb{Q}_{>0}$, and returns with probability at least $\frac{3}{4}$ a number between $f(x) /(1+\epsilon)$ and $(1+\epsilon) f(x)$. Moreover, an FPRAS runs in time polynomial in the size of $x$ and $1 / \epsilon$. RP is the complexity class consisting of problems that can be solved in randomized polyno-
mial time. The statement below follows from Proposition 8 in (Welsh and Merino 2000) and Theorem 1.

## Corollary 2. Unless $N P=R P$, there is no FPRAS for \#TFP.

Hence, unless NP $=R P$, there also does not exist an FPRAS for computing the probability of a player winning a BKT for a draw chosen uniformly at random.

## Tractable Cases

We first examine a natural case in which players are divided into player types: All players of one type beat exactly the same subset of players of other types. The result of matches between players of the same type is irrelevant as we do not care which player in each type wins. In this variant that we call \#TFP-types, the objective is to count the number of draws for which a player of a given type wins. We first adapt the concept of the pairwise comparison matrix. For two player types $i$ and $j$, define $P_{i j}=1$ if an $i$-player wins a match between $i$ and $j$ and $P_{i j}=0$ otherwise. Note that this definition is chosen such that $P_{i i}=1$ for all player types $i$.
Theorem 2. \#TFP-types is polynomial-time solvable if there are a constant number of player types.

Proof. We will use a dynamic programming scheme. All vectorial inequalities assume equal dimensions and are meant component-wise. Let $N_{x}:=\left\{\eta=\left(\eta_{1}, \ldots, \eta_{k}\right) \geq\right.$ $\left.0 \mid \eta_{1}+\cdots+\eta_{k}=2^{x}\right\}$ and denote by $\# \operatorname{TFP}\left(i, x, \eta \in N_{x}\right)$ the number of draws for an $x$-round tournament involving $\eta_{j}$ players of type $j$ for all $j \in[k]$ in which a player of type $i$ wins the tournament. We will assume that the players of one type are not distinguishable. For fixed $i$ and $x$, there are $\binom{k+2^{x}}{k}$ such problems that potentially need to be considered. \#TFP $(i, x, \eta)$ is computed via the following recursion.

$$
\begin{align*}
\# \operatorname{TFP}(i, x, \eta)= & \sum_{\substack{\eta^{\prime} \in N_{x-1} \\
\eta^{\prime} \leq \eta}} \# \operatorname{TFP}\left(i, x-1, \eta^{\prime}\right) \\
& \cdot \sum_{\substack{j \in[k] \\
P_{i j}=1 \\
\eta_{j}^{\prime}<\eta_{j}}} \# \operatorname{TFP}\left(j, x-1, \eta-\eta^{\prime}\right) \tag{3}
\end{align*}
$$

The base cases are given by

$$
\# \operatorname{TFP}(i, 0, \eta)= \begin{cases}1 & \eta_{i}=1 \text { and } \forall j \neq i: \eta_{j}=0  \tag{4}\\ 0 & \text { otherwise }\end{cases}
$$

for all $i \in[k]$. Eq. (3) only uses values of $\# \operatorname{TFP}\left(j, y, \eta^{\prime}\right)$ with $y<x$, thus every problem only needs to be solved once. For constant $k$, solving a problem requires $O\left(n^{k}\right)$ operations and for constant $k$ it holds that $\binom{k+2^{x}}{k} \in O\left(n^{k}\right)$. Thus, $O\left(\log (n) n^{2 k}\right)$ operations are necessary to compute $\# \operatorname{TFP}\left(i, \log (n),\left(n_{1}, \ldots, n_{k}\right)\right)$.

The second tractable case that we identify is:
Theorem 3. \#TFP is polynomial-time solvable if there is a linear ordering of strengths of the players with at most a constant number of pairwise relations flipped.

Proof. Let $b$ be the number of backwards arcs that do not respect the linear ordering. We can show by induction that the problem can be reduced to \#TFP-types with at most $4 b+$ 3 player types. Since the number of player types is constant, the theorem follows from Theorem 2.

## An Exponential Time Algorithm for \#TFP

\#TFP can be trivially solved in time $2^{O(n \log n)}$ via a bruteforce enumeration of all possible draws. In exponential time algorithmics (Fomin and Kratsch 2010), the aim is to design algorithms solving the problem exactly with worstcase running times outperforming the brute-force solution. In this section we give an algorithm for \#TFP running in time $O\left(2.8285^{n}\right)$ using space $O\left(1.7548^{n}\right)$. If only polynomial space is available, the running time becomes $4^{n+o(n)}$, and we give a range of possible time-space trade-offs. The algorithm is based on the recursion formula (3) and memoization used at various levels of the recursion. We use poly $(n)$ to stand for a polynomial function in $n$.
Theorem 4. For every $y, 2 \leq y \leq \log n$, there is an algorithm solving \#TFP in time $T(n)=\operatorname{poly}(n) \cdot \prod_{p=0}^{y-1} \frac{n}{2^{p}} \cdot 2^{n / 2^{p}}$ and space $S(n)=\operatorname{poly}(n) \cdot\left(2^{y}\right)^{n / 2^{y}} \cdot\left(\frac{2^{y}}{2^{y}-1}\right)^{n-n / 2^{y}}$.

Proof Sketch. The algorithm recursively evaluates the formula (3) for the special case where each player type consists of one player, starting from $\# \mathrm{TFP}(i, \log n,(1, \ldots, 1))$. It uses memoization for all recursive calls \#TFP $(\cdot, x, \cdot)$ where $x \leq(\log n)-y$. That is, the algorithm uses a table indexed by players, level, and player type vector. To evaluate $\# \operatorname{TFP}(\cdot, x, \cdot)$ with $x \leq(\log n)-y$, the algorithms first checks in this table whether this recursive call has already been evaluated. Only if the value has not yet been computed, it computes the result recursively, and stores it in the table. Then, it returns the result that is stored in the table. The number of table entries used by memoization is upper bounded by $n$ times the number of subsets of size at most $n / 2^{y}$. Using Stirling's approximation for factorials, the space usage of the algorithm can be upper bounded by $S(n)=\operatorname{poly}(n) \cdot\left(2^{y}\right)^{n / 2^{y}} \cdot\left(\frac{2^{y}}{2^{y}-1}\right)^{n-n / 2^{y}}$. The running time of the algorithm is the time used for the recursion without memoization, poly $(n) \cdot \prod_{p=0}^{y-1} \frac{n}{2^{p}} \cdot 2^{n / 2^{p}}$, plus the time for the part with memoization, which is upper bounded by $S(n) \cdot n \cdot 2^{n / 4}=O\left(2.0868^{n}\right)$. Thus, for any $y, 2 \leq y \leq \log n$, we obtain an algorithm with running time $T(n)=\operatorname{poly}(n) \cdot \Pi_{p=0}^{y-1} \frac{n}{2^{p}} \cdot 2^{n / 2^{p}}$ using space $S(n)=\operatorname{poly}(n) \cdot\left(2^{y}\right)^{n / 2^{y}} \cdot\left(\frac{2^{y}}{2^{y}-1}\right)^{n-n / 2^{y}}$.

Various time and space requirements of the algorithm are reported in Table 1. Using the rule of thumb that for current computing architectures, the space requirements of an algorithm become a bottleneck if they exceed the square root of the time requirements, the analyses for $y=2$ and $y=3$ currently seem the most relevant.

| $y$ | time $T(n)$ | space $S(n)$ |
| :---: | :---: | :---: |
| 2 | $O\left(2.8285^{n}\right)$ | $O\left(1.7548^{n}\right)$ |
| 3 | $O\left(3.3636^{n}\right)$ | $O\left(1.4576^{n}\right)$ |
| 4 | $O\left(3.6681^{n}\right)$ | $O\left(1.2634^{n}\right)$ |
| $\log n$ | $4^{n+o(n)}$ | $\operatorname{poly}(n)$ |

Table 1: Running times and space requirements for the algorithm of Theorem 4 for various time-space trade-offs.

## Conclusions

In this paper we considered problems related to tournament fixing. Although being able to change draws is not always realistic, the computational problems that are considered have been analyzed in post analysis of tournament draws and also shed light on the relative strengths of players. Our main result is that TFP is NP-complete. We discussed a number of implications of the result. We complement the computational hardness result in the paper by presenting algorithms for \#TFP - both for the general case as well as restricted cases. A possible future direction is to propose parametrized algorithms for TFP.

## Acknowledgments

NICTA is funded by the Australian Government as represented by the Department of Broadband, Communications and the Digital Economy and the Australian Research Council through the ICT Centre of Excellence program. Serge Gaspers is the recipient of an Australian Research Council Discovery Early Career Researcher Award (project number DE120101761).

## References

Aziz, H.; Brill, M.; Fischer, F.; Harrenstein, P.; Lang, J.; and Seedig, H. G. 2012. Possible and necessary winners of partial tournaments. In Proceedings of the 11th International Joint Conference on Autonomous Agents and Multi-Agent Systems (AAMAS), 585-592. IFAAMAS.
Connolly, R. A., and Rendleman, R. J. 2011. Tournament qualification, seeding and selection efficiency. Technical Report 2011-96, Tuck School of Business.
Fomin, F. V., and Kratsch, D. 2010. Exact Exponential Algorithms. Springer.
Groh, C.; Moldovanu, B.; Sela, A.; and Sunde, U. 2012. Optimal seedings in elimination tournaments. Economic Theory 49(1):59-80.
Gusfield, D., and Martel, C. 2002. The structure and complexity of sports elimination numbers. Algorithmica 32(1):73-86.
Horen, J., and Riezman, R. 1985. Comparing draws for single elimination tournaments. Operations Research 33(2):249-262.
Kern, W., and Paulusma, D. 2004. The computational complexity of the elimination problem in generalized sports competitions. Discrete Optimization 1(2):205-214.
Lang, J.; Pini, M. S.; Rossi, F.; Venable, K. B.; and Walsh, T. 2007. Winner determination in sequential majority voting. In Proceedings of the 20th International Joint Conference on Artificial Intelligence (IJCAI), 1372-1377.

Lang, J.; Pini, M. S.; Rossi, F.; Salvagnin, D.; Venable, K. B.; and Walsh, T. 2012. Winner determination in voting trees with incomplete preferences and weighted votes. Journal of Autonomous Agents and Multi-Agent Systems 25(1):130-157.
Laslier, J.-F. 1997. Tournament Solutions and Majority Voting. Springer-Verlag.
Rosen, S. 1986. Prizes and incentives in elimination tournaments. The American Economic Review 76(4):701-715.
Russell, T., and van Beek, P. 2011. An empirical study of seeding manipulations and their prevention. In Proceedings of the Twenty-Second International Joint Conference on Artificial Intelligence (IJCAI 2011), 350-356. IJCAI/AAAI.
Silver, N. 2011. When 15 th is better than 8th: The math shows the bracket is backward. On Five-ThirtyEight Blog, http://fivethirtyeight.blogs.nytimes.com/2011/03/15/when-15th-is-better-than-8th-the-math-shows-the-bracket-isbackward/. Last Accessed Feb. 2014.
Stanton, I., and Vassilevska Williams, V. 2011a. Manipulating single-elimination tournaments in the BravermanMossel model. In IJCAI Workshop on Social Choice and Artificial Intelligence.
Stanton, I., and Vassilevska Williams, V. 2011b. Manipulating stochastically generated single-elimination tournaments for nearly all players. In Proceedings of 7th International Workshop on Internet and Network Economics (WINE), 326-337.
Stanton, I., and Vassilevska Williams, V. 2011c. Rigging tournament brackets for weaker players. In Proceedings of the 22nd International Joint Conference on Artificial Intelligence (IJCAI), 357-364.
Tullock, G. 1980. Toward a Theory of the Rent-seeking Society. Texas A\&M University Press.
Vassilevska Williams, V. 2010. Fixing a tournament. In Proceedings of the 24th AAAI Conference on Artificial Intelligence (AAAI), 895-900. AAAI Press.
Vu, T.; Altman, A.; and Shoham, Y. 2009. On the complexity of schedule control problems for knockout tournaments. In Proceedings of the 8th International Joint Conference on Autonomous Agents and Multi-Agent Systems (AAMAS), 225-232.
Vu, T., and Shoham, Y. 2011. Fair seeding in knockout tournaments. ACM Transactions on Intelligent Systems Technology 3(1):1-17.
Vu, T.; Hazon, N.; Altman, A.; Kraus, S.; Shoham, Y.; and Wooldridge, M. 2013. On the complexity of schedule control problems for knock-out tournaments.
Welsh, D. J. A., and Merino, C. 2000. The potts model and the tutte polynomial. Journal of Mathematical Physics 41(3):1127-1152.
Xia, L., and Conitzer, V. 2011. Determining possible and necessary winners under common voting rules given partial orders. Journal of Artificial Intelligence Research 41(2):2567.


[^0]:    Copyright © $\mathfrak{C}$ 2014, Association for the Advancement of Artificial Intelligence (www.aaai.org). All rights reserved.

[^1]:    ${ }^{1}$ In the version of TFP defined in (Vassilevska Williams 2010), there can be additional constraints in which certain matches are disallowed.
    ${ }^{2}$ The condition is quite natural since in many competitions there is a clear-cut ranking of the players according to their skills with only a few pairs of players for which the weaker player can beat the stronger player. For example, as of $15 / 01 / 2014$, Nikolay Davy-

[^2]:    denko was the only tennis player among the men's top 64 who had a winning head-to-head record against Rafael Nadal.

