

Ordering by weighted number of wins gives a good ranking for weighted tournaments

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We consider the following simple algorithm for feedback arc set problem in weighted tournaments — order the vertices by their weighted indegrees. We show that this algorithm has an approximation guarantee of 5 if the weights satisfy *probability constraints* (for any pair of vertices u and v , $w_{uv} + w_{vu} = 1$). Special cases of feedback arc set problem in such weighted tournaments include feedback arc set problem in unweighted tournaments and rank aggregation. To complement the upper bound, for any constant $\epsilon > 0$, we exhibit an infinite family of (unweighted) tournaments for which the above algorithm (*irrespective* of how ties are broken) has an approximation ratio of $5 - \epsilon$.

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1. INTRODUCTION

Consider a sports tournament where all n players play each other and after all the $\binom{n}{2}$ games are completed, one would like to rank the players with as few inconsistencies as possible. By an inconsistency we mean a higher ranked player actually lost to a lower ranked player. A natural way to generate such a ranking is to rank according to the number of wins (with ties broken in some manner). We show that this natural heuristic has a provably good performance guarantee.

A weighted tournament with *probability constraints* is a complete directed graph $T = (V, E, w)$ where $w_{(\cdot, \cdot)}$ is the weight function on ordered pairs of vertices such that for any $u, v \in V$ with $u \neq v$, $w_{uv} + w_{vu} = 1$ and $w_{uv}, w_{vu} \geq 0$. We will use the term tournament to refer to a weighted tournament with probability constraints. An *unweighted* tournament is a special case of weighted tournaments with probability constraints, where the weights of the edges are either 0 or 1. The *minimum feedback arc set* in T is the smallest weight set $E' \subseteq E$ such that $(V, E \setminus E')$ is acyclic. Alternatively, a minimum feedback arc set can be described by an ordering $\sigma : V \rightarrow \{0, 1, \dots, n-1\}$ which minimizes the weight of *back edges* induced by σ where a back edge $(u, v) \in E$ satisfies $\sigma(u) > \sigma(v)$.

The feedback arc set problem in general directed graphs can be approximated to within $O(\log n \log \log n)$ [Even et al. 1998; Seymour 1995] and is APX-Hard [Karp 1972; Dinur and Safra 2002]. The complementary problem of the maximum acyclic subgraph problem¹ can be approximated to within $2/(1 + \Omega(1/\sqrt{\Delta}))$ (where Δ is the maximum degree) [Berger and Shor 1990; Hassin and Rubinfeld 1977] and is APX-Hard [Papadimitriou and Yannakakis 1991]. The feedback arc set problem in tournaments (shortened to FAS-TOURNAMENT for the rest of the paper) was conjectured to be NP-hard for a long time [Bang-Jensen and Thomassen 1992]. The conjecture was recently proved in [Alon 2006]. Ailon, Charikar and Newman showed that the problem is NP-hard under randomized reductions [Ailon et al. 2005]. Alon derandomized their construction [Alon 2006]. The NP-hardness result was also obtained independently by Conitzer [Conitzer 2006]. The work of Ailon, Charikar and Newman ([Ailon et al. 2005]) also analyzes the following simple randomized 3-approximation algorithm for *unweighted* FAS-TOURNAMENT. Their algorithm first picks a random vertex p to be the “pivot” vertex. All the vertices which are connected to p with an out-edge are placed to the “left” of p and the vertices which are connected to p through an in-edge are placed to the “right”. Then, the algorithm recurses on the two tournaments induced by the vertices placed on either side of p .

Weighted FAS-TOURNAMENT is defined on a weighted tournament T where the weights satisfy probability constraints. Ailon et al. show that running their algorithm for FAS-TOURNAMENT on the unweighted tournament that is the weighted majority² of T yields a 5 approximation for weighted FAS-TOURNAMENT when the weights satisfy probability constraints.

¹The maximum acyclic subgraph of a directed graph $G = (V, E)$ is the largest cardinality subset $E' \subseteq E$ such that the graph (V, E') is acyclic.

²The weighted majority of a weighted tournament T is defined as follows. For any pairs of vertices u and v , orient the edge between u and v in the direction which has the larger weight (breaking ties arbitrarily).

There is a much simpler algorithm for both weighted and unweighted FAS-TOURNAMENT than ones considered by Ailon et al. — order the vertices in increasing order of their (weighted) indegrees where ties are broken arbitrarily. We analyze this algorithm (which we call INCR-INDEG in this paper) and show that it has an approximation guarantee of 5 for both unweighted FAS-TOURNAMENT and weighted FAS-TOURNAMENT when weights satisfy probability constraints.

We also study the problem of RANK-AGGREGATION. In this problem, given n candidates and k permutations of the candidates $\{\pi_1, \pi_2, \dots, \pi_k\}$, we need to find the *Kemeny optimal* ranking, that is, a ranking π such that $\kappa(\pi, \pi_1, \pi_2, \dots, \pi_k) = \sum_{i=1}^k \mathcal{K}(\pi, \pi_i)$ is minimized, where $\mathcal{K}(\pi_i, \pi_j)$ denotes the number of pairs of candidates that are ranked differently by π_i and π_j . The problem is NP-hard even when the number of lists is only four [Bartholdi et al. 1989; Dwork et al. 2001]. There is a simple deterministic 2-approximation for this problem — pick the best of the input rankings. Ailon et al. reduce RANK-AGGREGATION to weighted FAS-TOURNAMENT with probability constraints [Ailon et al. 2005]. This reduction implies that INCR-INDEG is a 5-approximation for RANK-AGGREGATION. Interestingly, INCR-INDEG in the RANK-AGGREGATION setting is exactly the same as the Borda’s method [Borda 1781], which was proposed in the late 18th century. Thus, our results show that the Borda’s method is a factor 5 approximation of the Kemeny optimal ranking. Fagin et al. also show that the ranking produced by Borda’s method is within a constant factor of the Kemeny optimal ranking [Fagin et al. 2005].

To complement our upper bound result, for any $\epsilon > 0$, we exhibit an infinite family of (unweighted) tournaments for which INCR-INDEG has an approximation ratio of $5 - \epsilon$ *irrespective* of how ties are broken. This result shows that our analysis is tight. This is somewhat surprising as it shows that the “dumbest” way of breaking ties is as good as the best tie breaking mechanism at least from the viewpoint of approximation guarantees.

Independent of our work, Van Zuylen [van Zuylen 2005] has designed a deterministic algorithm for weighted FAS-TOURNAMENT with an approximation guarantee of 3 (when the weights satisfy probability constraints). Her algorithm derandomizes the algorithm of Ailon et al. mentioned earlier in this section (the pivot is chosen based upon the solution to the relaxation of a well known LP for FAS-TOURNAMENT).

We now review some work that has appeared subsequent to this work. Van Zuylen, Hegde, Jain and Williamson [van Zuylen et al. 2007] present a combinatorial deterministic algorithm for weighted FAS-TOURNAMENT with an approximation factor of 4. Their algorithm first removes all directed triangles in the graph and then uses a recursive algorithm due to Chudnovsky, Seymour and Sullivan [Chudnovsky et al. 2007]. Recently, Kenyon-Mathieu and Schudy [Kenyon-Mathieu and Schudy 2007] have designed a polynomial time approximation scheme (PTAS) for the weighted FAS-TOURNAMENT. In fact, the algorithm works for more general weighted tournaments than our setting: for any given $b \in (0, 1]$ and for any pair of vertices $i \neq j$, the weights of the edges between the vertices satisfy $w_{ij} + w_{ji} \in [b, 1]$ (the running time is doubly exponential in $1/b$). Note that in our setting $b = 1$. The algorithm is an ingenious combination of some steps of local improvement along

with applications of the PTAS of [Arora et al. 1996] for the maximum acyclic subgraph problem. Balcan et al. [Balcan et al. 2007] show that in a machine learning setting, the “regret” of ranking is within a factor two of that of binary classification. Their reduction uses the ranking-by-number-of-wins algorithm on a suitably defined tournament.

While algorithms for (weighted) FAS-TOURNAMENT with better approximation guarantees are known (and were known when we first presented our results), the algorithm we analyze is arguably the simplest one. As a bonus, our analysis also gives a provable guarantee for the classical Borda’s method for rank aggregation.

The rest of the paper is organized as follows. We introduce some notation and some known facts in Section 2. We analyze the algorithm INCR-INDEG in Section 3. In Section 4, we present an infinite family of tournaments for which INCR-INDEG has an approximation factor of $5 - \epsilon$, for any $\epsilon > 0$. Finally, we conclude with some open questions in Section 5.

2. PRELIMINARIES

We first fix some notation. For any real x (note that x can be negative), $\lfloor x \rfloor$ is defined to be the largest integer n such that $n \leq x$. Similarly, $\lceil x \rceil$ is defined to be the smallest integer n such that $x \leq n$. For any positive integer m we will use $[m]$ to denote the set $\{0, 1, \dots, m-1\}$. Also for any pairs of integers $a < b$, we will use $[a, b]$ to denote the set $\{a, a+1, \dots, b\}$. We will also use (a, b) to denote the set $\{a+1, \dots, b\}$. The vertex set of the input tournament is assumed to be $[n]$. For any edge (u, v) in T , its weight is given by $w_{uv} \geq 0$. For the rest of the paper, the weights are assumed to satisfy *probability constraints*, that is, $w_{uv} + w_{vu} = 1$ for all $u, v \in [n]$. For any vertex $v \in [n]$, $\text{In}(v)$ denotes the (weighted) indegree of v , that is,

$$\text{In}(v) = \sum_{u \in [n] \setminus \{v\}} w_{uv}.$$

We will use $\sigma : [n] \rightarrow [n]$ as a generic permutation. \mathcal{O} will denote the permutation returned by the optimal algorithm for FAS-TOURNAMENT on input T while \mathcal{A} will denote the permutation returned by INCR-INDEG. Given any permutation σ , B_σ denotes the sum of the weight of back edges induced by σ on T , that is,

$$B_\sigma = \sum_{u, v \in [n]: \sigma(u) > \sigma(v)} w_{uv}.$$

We now recall two well known notions of distances between permutations. Given permutations $\sigma, \rho : [n] \rightarrow [n]$; the *Spearman’s footrule distance* between the two permutations is given by

$$\mathcal{F}(\sigma, \rho) = \sum_{v \in [n]} |\sigma(v) - \rho(v)|$$

and the *Kendall-Tau distance* is given by

$$\mathcal{K}(\sigma, \rho) = \frac{1}{2} \cdot \sum_{u, v \in [n]} \mathbf{1}_{(\sigma(u) - \sigma(v)) \cdot (\rho(u) - \rho(v)) < 0}$$

where $\mathbf{1}_{(\cdot)}$ is the indicator function. In other words, $\mathcal{K}(\sigma, \rho)$ is the number of (ordered) pairs which are ordered differently by σ and ρ . The following relationship was shown by Diaconis and Graham.

THEOREM 2.1 [DIACONIS AND GRAHAM 1977]. *Let $n \geq 1$ be an integer and $\sigma, \rho : [n] \rightarrow [n]$ be permutations. Then the following is true:*

$$\mathcal{K}(\sigma, \rho) \leq \mathcal{F}(\sigma, \rho) \leq 2\mathcal{K}(\sigma, \rho).$$

For RANK-AGGREGATION, we will need the notion of *Kemeny distance*. Given the input lists $\{\pi_1, \pi_2, \dots, \pi_k\}$ and an aggregate permutation σ , the Kemeny distance is defined as

$$\kappa(\sigma, \pi_1, \pi_2, \dots, \pi_k) = \sum_{i=1}^k \mathcal{K}(\sigma, \pi_i).$$

We now present the reduction of RANK-AGGREGATION to weighted FAS-TOURNAMENT with probability constraints from [Ailon et al. 2005]. Let $\{\pi_1, \dots, \pi_k\}$ be a RANK-AGGREGATION instance on the set of candidates $[n]$. The equivalent weighted FAS-TOURNAMENT instance is a weighted tournament on $[n]$ such that for any pair of vertices i and j , w_{ij} is the fraction of input permutations which rank i before j . Note that by construction the weights satisfy probability constraints. Finally, sorting the vertices by their weighted indegrees (on the weighted tournament constructed above), is the well known Borda's method [Borda 1781].

3. THE ALGORITHM FOR FAS-TOURNAMENT

We will analyze INCR-INDEG in this section. Recall that INCR-INDEG orders the vertices by their weighted indegrees with ties broken arbitrarily.

Our main result is the following theorem.

THEOREM 3.1. *INCR-INDEG is a 5-approximation for weighted FAS-TOURNAMENT, that is, $B_{\mathcal{A}} \leq 5B_{\mathcal{O}}$.*

The reduction from FAS-TOURNAMENT to RANK-AGGREGATION outlined in Section 2 implies the following corollaries to Theorem 3.1.

COROLLARY 3.2. *INCR-INDEG is a 5-approximation for RANK-AGGREGATION.*

COROLLARY 3.3. *If σ is the Kemeny optimal rank aggregation for input rankings $\pi_1, \pi_2, \dots, \pi_k$ and σ' is an output of the Borda's method for the same input, then*

$$\kappa(\sigma', \pi_1, \pi_2, \dots, \pi_k) \leq 5\kappa(\sigma, \pi_1, \pi_2, \dots, \pi_k).$$

We will prove Theorem 3.1 through the following sequence of lemmas.

LEMMA 3.4. *For any permutation $\sigma : [n] \rightarrow [n]$,*

$$2B_{\sigma} \geq \sum_{v \in [n]} |\sigma(v) - \text{In}(v)|.$$

Note that Lemma 3.4 generalizes the upper bound in Theorem 2.1 to a relationship between similar distance measures between a permutation and a weighted

tournament with probability constraints. Indeed, if the input tournament is unweighted acyclic tournament (with vertices arranged in order of indegrees), then $\text{In}(\cdot)$ is just the identity permutation ι . Further, note that in this case, Lemma 3.4 states that $2\mathcal{K}(\sigma, \iota) \geq \mathcal{F}(\sigma, \iota)$.³

LEMMA 3.5. *For any permutation $\sigma : [n] \rightarrow [n]$,*

$$\sum_{v \in [n]} |\sigma(v) - \text{In}(v)| \geq \sum_{v \in [n]} |\mathcal{A}(v) - \text{In}(v)|.$$

LEMMA 3.6. *For any two permutations $\sigma, \rho : [n] \rightarrow [n]$,*

$$\mathcal{F}(\sigma, \rho) \geq |B_\rho - B_\sigma|.$$

We first show how the above lemmas prove our main result.

Proof of Theorem 3.1: Consider the following sequence of inequalities.

$$\begin{aligned} 4B_{\mathcal{O}} &\geq \sum_{v \in [n]} |\mathcal{O}(v) - \text{In}(v)| + \sum_{v \in [n]} |\mathcal{O}(v) - \text{In}(v)| \\ &\geq \sum_{v \in [n]} |\mathcal{O}(v) - \text{In}(v)| + \sum_{v \in [n]} |\mathcal{A}(v) - \text{In}(v)| \\ &= \sum_{v \in [n]} (|\mathcal{O}(v) - \text{In}(v)| + |\mathcal{A}(v) - \text{In}(v)|) \\ &\geq \mathcal{F}(\mathcal{O}, \mathcal{A}) \\ &\geq B_{\mathcal{A}} - B_{\mathcal{O}}. \end{aligned}$$

The first, second and last inequalities follow from Lemmas 3.4, 3.5 and 3.6 respectively (with $\sigma = \mathcal{O}$ and $\rho = \mathcal{A}$). The third inequality is the triangle inequality while the equality just follows from rearrangement of the terms. Thus, we have $B_{\mathcal{A}} \leq 5B_{\mathcal{O}}$ which proves the theorem. \square

In the rest of this section, we will prove Lemmas 3.4-3.6.

Proof of Lemma 3.4: Consider any arbitrary $v \in [n]$. Let $W_L^-(v)$ be the sum of weights of edges from vertices to the “left” of v (according to σ) to v ; $W_L^+(v)$ be the sum of weights of edges from v to vertices which are to the left of v ; and $W_R^-(v)$ be the sum of weights of edges from vertices which are to the right of v to v . More formally,

$$W_L^-(v) = \sum_{u: \sigma(u) < \sigma(v)} w_{uv},$$

$$W_L^+(v) = \sum_{u: \sigma(u) < \sigma(v)} w_{vu},$$

$$W_R^-(v) = \sum_{u: \sigma(u) > \sigma(v)} w_{uv}.$$

³Theorem 2.1 is stated in terms for two general permutations σ and ρ . However, can one assume w.l.o.g. that $\rho = \iota$ as $\mathcal{F}(\sigma, \rho) = \mathcal{F}(\sigma \circ \rho^{-1}, \iota)$ and $\mathcal{K}(\sigma, \rho) = \mathcal{K}(\sigma \circ \rho^{-1}, \iota)$, where \circ is the composition operator and $\rho \circ \rho^{-1} = \iota$.

By definition, we have

$$W_L^-(v) + W_R^-(v) = \text{In}(v). \quad (1)$$

The following identity follows from definitions and the fact that weights satisfy probability constraints.

$$W_L^+(v) + W_L^-(v) = \sigma(v). \quad (2)$$

Now, $2B_\sigma = \sum_{v \in [n]} (W_L^+(v) + W_R^-(v))$ as each back edge is counted twice in the sum. To complete the proof, we claim that for any $v \in [n]$, $W_L^+(v) + W_R^-(v) \geq |\sigma(v) - \text{In}(v)|$.

Indeed from (1) and (2),

$$\begin{aligned} W_L^+(v) + W_R^-(v) &= \sigma(v) + \text{In}(v) - 2W_L^-(v) \\ &= |\sigma(v) - \text{In}(v)| + 2(\min\{\sigma(v), \text{In}(v)\} - W_L^-(v)) \\ &\geq |\sigma(v) - \text{In}(v)|. \end{aligned}$$

The last inequality again follows from (1) and (2) and the fact that $W_L^+(v), W_R^-(v) \geq 0$. \square

Lemma 3.5 is a restatement of the fact that for any real numbers $a_1 \leq a_2 \leq \dots \leq a_n$, the permutation $\sigma : [n] \rightarrow [n]$ which minimizes the quantity $\sum_{i=1}^n |a_i - \sigma(i)|$ is the identity. For the sake of completeness, we present a proof.

Proof of Lemma 3.5 If σ sorts the vertices in $[n]$ according to their indegrees then the statement of the lemma holds trivially.

So now consider the case when there exists $i \in [n]$ such that $\text{In}(u) > \text{In}(v)$ where $u = \sigma^{-1}(i)$ and $v = \sigma^{-1}(i+1)$. Construct a new ordering σ' that is same as σ except u and v are swapped: $\sigma'(w) = \sigma(w)$ if $w \notin \{u, v\}$ and $\sigma'(u) = i+1$, $\sigma'(v) = i$. We next show that $\sum_{v \in [n]} |\sigma(v) - \text{In}(v)| \geq \sum_{v \in [n]} |\sigma'(v) - \text{In}(v)|$: the rest of the proof is a simple induction. By the construction of σ' ,

$$\begin{aligned} &\sum_{v \in [n]} (|\sigma(v) - \text{In}(v)| - |\sigma'(v) - \text{In}(v)|) \\ &= |i - \text{In}(u)| + |i+1 - \text{In}(v)| - |i - \text{In}(v)| - |i+1 - \text{In}(u)| \\ &= 2(\min\{i, \text{In}(v)\} - \min\{i, \text{In}(u)\}) + 2(\min\{i+1, \text{In}(u)\} - \min\{i+1, \text{In}(v)\}). \end{aligned}$$

The last equality follows from the identity $|x - y| = x + y - 2\min\{x, y\}$. Finally we verify that the last sum is always non-negative. There are three cases. If $i \geq \text{In}(u)$ then the first term is $2(\text{In}(v) - \text{In}(u))$ while the second term is $2(\text{In}(u) - \text{In}(v))$. If $\text{In}(v) \geq i$ then the first term is 0 while the second term is $2\max\{i+1 - \text{In}(v), 0\}$. Finally if $\text{In}(u) > i > \text{In}(v)$ then the first term is $2(\text{In}(v) - i)$ while the second term is $2(i+1 - \text{In}(v))$. \square

Proof of Lemma 3.6: Consider the set of edges which are back edges in σ but are not back edges in ρ . Denote this set by $\mathcal{B}_{\sigma \setminus \rho}$. Also consider the set of edges which are back edges in ρ but are not back edges in σ . Denote this set by $\mathcal{B}_{\rho \setminus \sigma}$.

Note that

$$\sum_{(u,v) \in \mathcal{B}_{\sigma \setminus \rho}} w_{uv} + \sum_{(u,v) \in \mathcal{B}_{\rho \setminus \sigma}} w_{uv} \geq |B_{\rho} - B_{\sigma}|. \quad (3)$$

The crucial observation is that if an edge $(u, v) \in \mathcal{B}_{\sigma \setminus \rho}$ then $(v, u) \in \mathcal{B}_{\rho \setminus \sigma}$. This along with the fact that the weights satisfy probability constraints imply that $\mathcal{K}(\sigma, \rho) = \sum_{(u,v) \in \mathcal{B}_{\sigma \setminus \rho}} w_{uv} + \sum_{(u,v) \in \mathcal{B}_{\rho \setminus \sigma}} w_{uv}$ which by (3) implies $\mathcal{K}(\sigma, \rho) \geq |B_{\rho} - B_{\sigma}|$. Theorem 2.1 completes the proof. \square

4. A LOWER BOUND FOR INCR-INDEG

We will prove the following theorem in this section.

THEOREM 4.1. *For every constant $\epsilon > 0$, there exists an infinite family of (unweighted) tournaments \mathcal{T}_{ϵ} such that arranging the vertices of any tournament in \mathcal{T}_{ϵ} according to their indegrees, irrespective of how ties are broken, results in at least $5 - \epsilon$ times as many back edges as the optimal ordering.*

Note that the above result implies the analysis in Section 3 is tight, even if one modified INCR-INDEG to break ties in some “intelligent” way.

For any tournament T , we will use $\mathcal{I}(T)$ to denote the ordering according to indegrees which induces the least number of back edges (that is, ties are broken “optimally”). Also let $\mathcal{O}(T)$ denote the optimal ordering. For the rest of this section we use tournaments to refer to unweighted tournaments.

We will use two parameters, x and n , in this section. For any $n \geq 5$ and $x \geq 4$ such that x is a perfect square, we will construct a tournament $T_{x,n}$ such that

$$\lim_{x,n \rightarrow \infty} \frac{B_{\mathcal{I}(T_{x,n})}}{B_{\mathcal{O}(T_{x,n})}} = 5, \quad (4)$$

which will prove Theorem 4.1.

In the rest of this section, we will describe the construction of $T_{x,n}$ and show that (4) holds. Fix $n \geq 5$ and $x \geq 4$ such that x is a perfect square. $T_{x,n}$ will have $n(2x + 1)$ vertices. We will partition the vertices into n blocks of $2x + 1$ vertices each. The i^{th} block for any $i = 1, 2, \dots, n$, will be denoted by b^i . Further, for every $j = 0, 1, \dots, 2x$; the j^{th} node in b^i will be denoted by b_j^i . The node b_x^i is the *middle node* of b^i and the sets of nodes $\{b_0^i, b_1^i, \dots, b_{x-1}^i\}$ and $\{b_{x+1}^i, b_{x+2}^i, \dots, b_{2x}^i\}$ are the *left half* and *right half* of b^i respectively. Let ϕ denote the ordering $b_0^1, \dots, b_{2x}^1, b_0^2, \dots, b_{2x}^2, \dots, b_0^n, \dots, b_{2x}^n$. Finally, unless mentioned otherwise, an edge will be called backward or forward in $T_{x,n}$ with respect to ϕ .

The basic idea behind the construction is as follows. In $T_{x,n}$, the sub-tournaments spanned by each b^i would have no back edges if nodes in b^i are arranged according to ϕ . However, back edges from b^{i+1} and b^{i+2} will force $\mathcal{I}(T_{x,n})$ to order the vertices in b^i (more or less) in the reverse order of ϕ . This will result in $\mathcal{I}(T_{x,n})$ inducing many more back edges than the optimal.

We will describe the construction of $T_{x,n}$ by starting with a tournament on the vertex set $\cup_{i=1}^n b^i$ such that all the edges are forward edges according to ϕ . Then we will reverse the direction of some edges between b^i and b^{i+1} (which we call **Type**

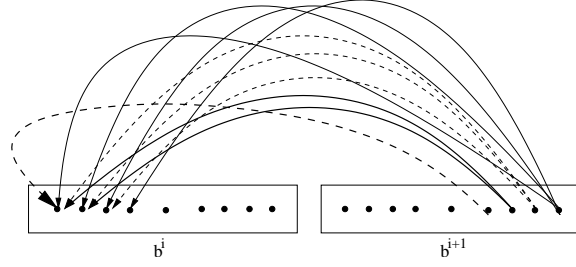


Fig. 1. The **Type I** edges between b^i and b^{i+1} ($1 \leq i < n$) when $x = 4$.

I edges) and some edges between b^i and b^{i+2} (which we call **Type II** edges) to get our final $T_{x,n}$. We now formally define these edges.

Assume we start with a set of edges E on $V = \cup_{i=1}^n b^i$ such that for any $u, v \in V$, $(u, v) \in E$ if and only if $\phi(u) < \phi(v)$. We first describe the **Type I** edges. For every i ($1 \leq i < n$), the last vertex of b^{i+1} has a **Type I** edge to every vertex in the left half of b^i . The second last vertex of b^{i+1} has a **Type I** edge to all but the last vertex in the left half of b^i and so on. More formally, for every $i = 1, 2, \dots, n-1$,

$$\begin{aligned} & \text{for } (j = x+1, x+2, \dots, 2x) \\ & \quad \text{for } (k = 0, 1, \dots, j-x-1) \\ & \quad \quad E \leftarrow (E \setminus \{(b_k^i, b_j^{i+1})\}) \cup \{(b_j^{i+1}, b_k^i)\} \end{aligned}$$

See Figure 1 for an example when $x = 4$.

We turn to the **Type II** edges. First, partition the left and the right half of every b^i into \sqrt{x} consecutive *minigroups* of \sqrt{x} vertices each. A minigroup is *connected* to another if for every $0 \leq \ell \leq \sqrt{x} - 1$ there is an edge from the ℓ th vertex in the first minigroup to the ℓ th vertex in the second minigroup. For any i ($1 \leq i < n-1$), **Type II** edges are introduced to connect the last minigroup in the right half of b^{i+2} to all the minigroups in the left half of b^i . The second last minigroup in the right half of b^{i+2} is connected to all but the last minigroup in the left half of b^i and so on. More formally, for every $i = 1, 2, \dots, n-2$,

$$\begin{aligned} & \text{for } (k = 0, 1, \dots, \sqrt{x}-1) \\ & \quad \text{for } (r = 0, 1, \dots, k) \\ & \quad \quad \text{for } (\ell = 0, 1, \dots, \sqrt{x}-1) \\ & \quad \quad \quad E \leftarrow (E \setminus \{(b_{r\sqrt{x}+\ell}^i, b_{x+k\sqrt{x}+\ell+1}^{i+2})\}) \cup \{(b_{x+k\sqrt{x}+\ell+1}^{i+2}, b_{r\sqrt{x}+\ell}^i)\} \end{aligned}$$

See Figure 2 for an example of **Type II** edges for the case when $x = 4$.

The tournament defined by the vertices V and edges E is the required tournament $T_{x,n}$. We will now estimate the indegrees of the vertices in $T_{x,n}$. Consider an i such that $2 < i < n-1$. Before **Type I** and **Type II** edges were introduced, $T_{x,n}$ was an acyclic graph. Thus, the indegree of the vertex b_j^i (where $0 \leq j \leq 2x$) was the number of vertices connected to it which is $(i-1)(2x+1) + j$. When **Type I** edges were introduced, the indegree of the last vertex in b^i decreased by x (as there was now an edge from it to every vertex in the left half of b^{i-1}) while the indegree of

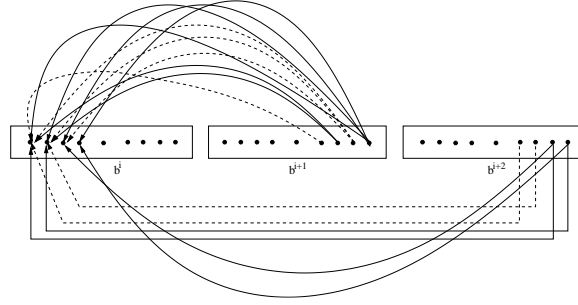


Fig. 2. Type I edges between b^i and b^{i+1} and Type II edges between b^i and b^{i+2} ($1 \leq i < n-1$) when $x = 4$. Type II edges are the ones which are not present in Figure 1.

the first vertex in b^i increased by x (as there was now an edge from every vertex in the right half of b^{i+1} to b_0^i). Similarly, the indegree of the second last vertex decreased by $x-1$ while the indegree of the second vertex increased by $x-1$ and so on. Thus, after all Type I edges were introduced, the degree of b_j^i ($0 \leq j \leq 2x$) was $(i-1)(2x+1)+x$. When Type II edges were introduced, the indegree of every vertex in the last minigroup in the right half of b^i decreased by \sqrt{x} (as the last minigroup is now connected to every minigroup in the left half of b^{i-2}) while the indegree of every vertex in the first minigroup in the left half of b^i increased by \sqrt{x} (as every minigroup in the right half of b^{i+2} is now connected to the first minigroup in the left half of b^i). Similarly, the indegree of every vertex in the second last minigroup in the right half of b^i decreased by $\sqrt{x}-1$ while the indegree of every vertex in the second minigroup in the left half of b^i increased by $\sqrt{x}-1$ and so on. In particular, the indegree of b_x^i did not change. Thus, for $0 \leq j < x$, $b_j^i = (i-1)(2x+1) + x + \left\lceil \frac{x-j}{\sqrt{x}} \right\rceil$ and for $x < j \leq 2x$, $b_j^i = (i-1)(2x+1) + x - \left\lfloor \frac{j-x}{\sqrt{x}} \right\rfloor = (i-1)(2x+1) + x + \left\lfloor \frac{x-j}{\sqrt{x}} \right\rfloor$.

Taking care of the boundary cases we have the following expressions for the indegrees.

$$\text{In}(b_j^1) = \begin{cases} x + \left\lceil \frac{x-j}{\sqrt{x}} \right\rceil, & j \in [0, x] \\ j, & j \in (x, 2x] \end{cases} \quad (5)$$

$$\text{In}(b_j^2) = \begin{cases} 3x + 1 + \left\lceil \frac{x-j}{\sqrt{x}} \right\rceil, & j \in [0, x] \\ 3x + 1, & j \in (x, 2x] \end{cases} \quad (6)$$

For $i = 3, \dots, n-2$ and for $j = 0, 1, \dots, 2x$;

$$\text{In}(b_j^i) = \begin{cases} (i-1)(2x+1) + x + \left\lceil \frac{x-j}{\sqrt{x}} \right\rceil, & j \in [0, x] \\ (i-1)(2x+1) + x + \left\lfloor \frac{x-j}{\sqrt{x}} \right\rfloor, & j \in (x, 2x] \end{cases} \quad (7)$$

$$\text{In}(b_j^{n-1}) = \begin{cases} (n-2)(2x+1) + x, & j \in [0, x] \\ (n-2)(2x+1) + x + \left\lfloor \frac{x-j}{\sqrt{x}} \right\rfloor, & j \in (x, 2x] \end{cases} \quad (8)$$

$$\text{In}(b_j^n) = \begin{cases} (n-1)(2x+1) + j, & j \in [0, x] \\ (n-1)(2x+1) + x + \left\lfloor \frac{x-j}{\sqrt{x}} \right\rfloor, & j \in (x, 2x] \end{cases} \quad (9)$$

We first upper bound the number of back edges in the optimal ordering.

LEMMA 4.2. *The number of back edges in the optimal ordering of $T_{x,n}$ is at most $\frac{x^2n}{2} + o(x^2n)$.*

PROOF. To prove the lemma, we show that $B_\phi \leq x^2n/2 + o(x^2n)$. Note that the only back edges in $T_{x,n}$ according to ϕ are the **Type I** and **Type II** edges. By definition, the number of **Type I** edges is

$$(n-1)(x + x-1 + \dots + 1) = \frac{x(x+1)}{2}(n-1) < \frac{x^2n}{2} + \frac{xn}{2} = \frac{x^2n}{2} + o(x^2n)$$

and the number of **Type II** edges is

$$(n-2)(\sqrt{x} + \sqrt{x}-1 + \dots + 1) = \frac{\sqrt{x}(\sqrt{x}+1)}{2}(n-2) < \frac{xn}{2} + \frac{\sqrt{xn}}{2} = o(x^2n).$$

The proof is complete. \square

We now lower bound the number of back edges induced by $\mathcal{I}(T_{x,n})$.

LEMMA 4.3. *The number of back edges induced by $\mathcal{I}(T_{x,n})$ on $T_{x,n}$ is at least $\frac{5x^2n}{2} - o(x^2n)$.*

Note that Lemmas 4.2 and 4.3 prove equation (4) and thus, Theorem 4.1. We end this section by proving Lemma 4.3.

Proof of Lemma 4.3: We first claim that for any $i < i'$, every node in b^i is placed before every node of $b^{i'}$ by $\mathcal{I}(T_{x,n})$. In other words, all **Type I** and **Type II** edges are back edges (between vertices in the different blocks) according to $\mathcal{I}(T_{x,y})$. The proof of lemma 4.2 shows that this number is at least $x^2n/2 - o(x^2n)$. For any i , let max_i and min_i be the maximum and minimum indegrees of all vertices in b^i . To prove the claim, we will show that

$$\text{for all } i = 1, 2, \dots, n-1; \quad max_i < min_{i+1}. \quad (10)$$

Indeed from Equations (5)-(9), we have the following values for max_i and min_i :

$$max_i = \begin{cases} 2x, & i = 1 \\ (2i-1)x + \sqrt{x} + (i-1), & i = 2, \dots, n-2 \\ (2n-3)x + (n-2), & i = n-1 \end{cases}$$

$$min_i = \begin{cases} 3x+1, & i = 2 \\ (2i-1)x - \sqrt{x} + (i-1), & i = 3, \dots, n-1 \\ (2n-2)x + n-1, & i = n \end{cases}$$

An inspection of the values shows that (10) holds.

Thus, we have counted all the back edges between vertices of b^i and $b^{i'}$ for $i \neq i'$. We now need to count the number of back edges between vertices in the same b^i . Counting conservatively, we assume that there are no such back edges for $i \in \{1, 2, n-1, n\}$. Fix an i such that $2 < i < n-1$. We claim that the number of back edges between vertices in b^i is at least

$$x(2x+1) - x(\sqrt{x}-1). \quad (11)$$

To see this divide the left half of b^i into \sqrt{x} minigroups— $l_0, l_1, \dots, l_{\sqrt{x}-1}$. In particular, l_k consists of the vertices $b_{k\sqrt{x}}^i, b_{k\sqrt{x}+1}^i, \dots, b_{(k+1)\sqrt{x}-1}^i$. Similarly the right half of b^i is divided into \sqrt{x} minigroups— (for left to right) $r_0, r_1, \dots, r_{\sqrt{x}-1}$. Observe from (7) that for any $k = 0, 1, \dots, \sqrt{x} - 1$; the degree of a vertex in l_k and r_k is $(i-1)(2x+1) + x + \sqrt{x} - k$ and $(i-1)(2x+1) + x - k - 1$ respectively. Thus, $\mathcal{I}(T_{x,n})$ will have to arrange vertices in the order $r_{\sqrt{x}-1}, r_{\sqrt{x}-2}, \dots, r_0$, followed by the middle node b_x^i , followed by vertices in the order $l_{\sqrt{x}-1}, l_{\sqrt{x}-2}, \dots, l_0$. Again counting conservatively, we assume that there are no back edges in induced tournaments over any minigroup l_k or r_k (where $k = 0, 1, \dots, \sqrt{x} - 1$). However, note that every remaining edge in $T_{x,n}^i$, the induced tournament over b^i , is a back edge. There are a total of $\binom{2x+1}{2} = x(2x+1)$ edges in $T_{x,n}^i$ while the induced tournaments over any l_k or r_k has $\binom{\sqrt{x}}{2}$ many edges and there are $2\sqrt{x}$ such minigroups. This implies that the number of back edges in $T_{x,n}^i$ is at least $x(2x+1) - 2\sqrt{x} \cdot \sqrt{x}(\sqrt{x}-1)/2$ as claimed in (11).

Recalling that there are $n-4$ choices for i , the number of back edges within some block totaled over all the $n-4$ blocks is

$$(x(2x+1) - x(\sqrt{x}-1))(n-4) \geq 2x^2n - o(x^2n).$$

Adding the estimates of the number of back edges between different b^i 's and number of back edges within the same b^i completes the proof. \square

5. CONCLUSIONS AND OPEN PROBLEMS

In this paper, we analyzed a very simple heuristic for the FAS-TOURNAMENT problem (and a weighted generalization) and showed that it has an approximation factor of 5. Recently Kenyon-Mathieu and Schudy have designed a PTAS for the problem. One of the key ingredients of their algorithm is a sequence of local improvements steps. It is an open question whether such local improvements when applied to our algorithm can give an approximation factor better than 5.

Another interesting problem is to get a tighter bound on how good Borda's method is for approximating the optimal ranking for rank aggregation. Note that the example in Section 4 is not a valid example for rank aggregation (the weights do not satisfy triangle inequality).

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