

## Approximability of Manipulating Elections

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### Abstract

In this paper, we set up a framework to study approximation of manipulation, control, and bribery in elections. We show existence of approximation algorithms (even fully polynomial time approximation schemes) as well as obtain inapproximability results.

### Introduction

Elections are an essential mechanism that each democratic society uses to make joint decisions. They are also important tools within computer science. For example, (Dwork *et al.* 2001) show how to build a meta search-engine via conducting elections between other search engines; Ephrati and Rosenschein (1993) use voting to solve certain planning problems; and in the context of multiagent systems, elections and voting are naturally used to obtain the joint decisions of agent societies.

Unfortunately, a famous result of Gibbard and Satterthwaite states that for any reasonable election system (with at least 3 candidates) there exist scenarios where at least some voters have an incentive to vote strategically, i.e., to vote not according to their true preferences but in a way that yields a result more desirable for them. Similarly, the result of an election can be skewed by an external agent who bribes some of the voters to change their votes or even by the authority organizing the election, via, e.g., encouraging or discouraging particular candidates from participating, or via arranging voting districts in a certain way.

The possibility that the result of the election can be skewed via strategic voting, bribery, or procedural control is very disconcerting. In the early 90s, Bartholdi, Orlin, Tovey, and Trick (1989; 1991; 1992) suggested a brilliant way to circumvent this issue. They observed that since all voters are computationally-bounded entities, even if various forms of manipulating elections are possible in principle,<sup>1</sup> they constitute a real threat only if it is computationally easy, for a given election system, to determine the appropriate actions that affect the result (i.e., for the case of strategic voting

to determine how the manipulators should vote; for the case of bribery determine who to bribe and how, etc.). To measure the computational difficulty of manipulation and control, Bartholdi, Orlin, Tovey, and Trick used the complexity-theoretic notion of NP-hardness.

The ideas of Bartholdi, Orlin, Tovey and Trick did not receive that much attention until a few years ago, when it became apparent that elections and voting are important tools in the context of multiagent systems, and that software agents are capable of much more systematic attempts at manipulating elections than, e.g., humans. Thus, in recent years, many papers focused on the computational analysis of voting rules with respect to manipulation, e.g., (Conitzer, Sandholm, & Lang 2007; Elkind & Lipmaa 2005; Procaccia & Rosenschein 2006; Procaccia, Rosenschein, & Zohar 2007; Hemaspaandra & Hemaspaandra 2007), bribery (Faliszewski, Hemaspaandra, & Hemaspaandra 2006; Faliszewski *et al.* 2007; Faliszewski 2008), and control (Hemaspaandra, Hemaspaandra, & Rothe 2007; Faliszewski *et al.* 2007; Procaccia, Rosenschein, & Zohar 2007). Most of these papers focus on obtaining polynomial-time algorithms and NP-hardness results for various forms of affecting the result of elections. However, NP-hardness gives only worst-case complexity guarantees, and it might very well be the case that even though, say, manipulation in a given election system is NP-hard, finding effective manipulations is often easy. Recently, Conitzer and Sandholm (2006), and Procaccia and Rosenschein (2006) looked into these issues and they provide examples of voting rules and distributions of votes for which this is the case. We will refer to the approach presented, among others, in these two papers as *frequency of hardness approach*.

In this paper we refine the study of the complexity of manipulating elections via analysis of approximability of NP-hard election-manipulation problems. An important contribution of this paper is a natural, uniform objective function that can be used to measure the effectiveness of a particular manipulation, bribery, or control attempt. Thus we set up a general framework for studying approximation for these problems.<sup>2</sup> Our function is particularly interesting for

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<sup>1</sup>In the literature the technical term *manipulation* means strategic voting. In this section by *manipulating* we mean the general notion of affecting the result of an election.

<sup>2</sup>To be technically correct, our approach is limited to voting rules that assign numerical scores to the candidates. This is the case for most standard voting rules.

the case of manipulation, where defining a natural objective function is not straightforward.

We show existence of approximation algorithms (even fully polynomial-time approximation schemes) for manipulation for a large subclass of voting rules known as scoring protocols (for bounded numbers of candidates) as well as obtain inapproximability results regarding several prominent families of scoring protocols (e.g., veto and  $k$ -approval for unbounded number of candidates). We prove NP-hardness of bribery for Borda count and inapproximability of bribery for Borda count for the case where each voter has a price for changing its vote. To the best of our knowledge, our NP-hardness result for Borda is the first hardness result for a problem of affecting the result of unweighted Borda count elections via modifying the voters.

**Related work** Several previous papers studied approximation of various manipulation and bribery problems but each of them used objective functions specifically tailored to their tasks. In particular, Faliszewski (2008) studied approximability of the total cost of a bribery for plurality and approval voting. Zuckerman, Procaccia and Rosenschein (2008), among other things, studied approximability of the minimum number of unweighted manipulators needed to change the result of an election for several voting systems, including Borda count.

We also contrast our approach and results with the frequency of hardness approach. The existence of an approximation algorithm (in particular, the existence of a fully polynomial-time approximation scheme) for a given election problem is much stronger evidence that this problem is practically easy than a frequency of hardness result stating that the problem is easy often, according to some distribution. The reason for this is that a polynomial-time approximation algorithm guarantees to find a near-optimal answer for *every* input instance. If our problem is frequently easy it might still be the case that the instances that we encounter in practice happen to be the “rare” difficult ones. On the other hand, inapproximability is a worst case notion. If a problem is in general inapproximable, it might still be the case that most of its instances are easy (are easily approximable). Nonetheless, inapproximability of a given NP-hard election problem is stronger evidence of its computational hardness than NP-hardness alone.

In this paper we focus on manipulation and bribery rather than on control. We mention, however, that Brelsford (2007) studied several control problems from the point of view of approximation.

## Preliminaries

**Elections** An election  $E$  is a pair  $(C, V)$ , where  $C$  is a finite set of candidates and  $V$  a finite multiset of strict linear orders on  $C$ . An order  $v \in V$  is called a vote and represents the preference of a voter over the candidates. The winner of an election  $E$  depends on the underlying election system. In this paper we consider only election systems that are represented by scoring protocols and families of scoring protocols. A scoring protocol is a vector  $(\alpha_1, \dots, \alpha_m)$  of natural numbers with  $\alpha_1 \geq \dots \geq \alpha_m$ . Using this protocol the win-

ner of an election  $E$  with  $m$  candidates can be determined as follows: Every candidate  $c$  gets  $\alpha_i$  points for every vote that ranks  $c$  in the  $i$ th place and  $\text{score}_E(c)$  is the sum of all points  $c$  gets in this way. In the end  $c$  is a winner if no other candidate has a higher score. We also allow the votes in  $E$  to have weights, in this case each vote with weight  $w \in \mathbb{N}$  is counted as  $w$  identical votes.

Let  $(S_i)_{i \geq 1}$  be a family of scoring protocols such that  $S_i$  is a scoring protocol of length  $i$ . We represent by  $(S_i)_{i \geq 1}$  the election system that uses  $S_m$  to determine the winner of an election with  $m$  candidates.

Borda count is the election system using  $((i-1, i-2, \dots, 0))_{i \geq 1}$ , and veto is the election system using  $((1, 1, \dots, 1, 0))_{i \geq 1}$ .

**Approximating Elections** In this paper we study approximation algorithms for manipulation and bribery. In both problems our goal is to ensure that a specified candidate is a winner but in manipulation we attempt to reach this goal via, in essence, adding a certain number of votes, whereas in bribery we do so via changing up to a given number of votes. (Note that sometimes manipulation is defined as allowing to change a *specified* set of voters. Our version allows to state our results in an easier notation—it is easy to see that these notions describe the exact same issue.) We also study the manipulation problem where the voters additionally have weights, and the bribery problem where the votes have prices which the briber has to pay in order to change the vote.

We require our approximation algorithms to produce “solutions” to their respective instances. A solution is a strategy specifying which actions to perform, i.e., what votes to add for the case of manipulation and which votes to change (and how to change them) for the case of bribery.

In our model we assume that we know all the votes that are supposed to be cast in the election. In reality, however, we are often limited to only having a guess regarding these votes. Thus we are interested in finding a strategy that benefits the specified candidate as much as possible so that this candidate has a good change of becoming a winner even if the guess is a little off.

In the setting of scoring protocols (or any other score-based election system), a natural way to measure the performance of a candidate  $p$  in an election  $E$  (written as  $\text{perf}^E(p)$ ) is  $\text{score}_E(p) - \max\{\text{score}_E(c) \mid c \in C \setminus \{p\}\}$ , the difference between the score of  $p$  and that of  $p$ ’s strongest competitor.  $\text{perf}^E(p)$  tells us “how much”  $p$  wins the election or “how close”  $p$  is to winning it. Obviously,  $p$  wins the election  $E$  if and only if  $\text{perf}^E(p)$  is nonnegative.

A natural measure of the effectiveness of a manipulating action  $s$  within election  $E$  is the increase of performance of the favorite candidate  $p$  obtained by applying this action.

**Definition 1**  $\beta(E, s) = \text{perf}^{s(E)}(p) - \text{perf}^E(p)$ , where  $s(E)$  denotes the election resulting from applying action  $s$  to  $E$ .

We now define our optimization problems (which we will prefix with the election system under consideration):

**\$-bribery-max** The input  $I$  consists of an election  $E$ , for each existing vote a natural number defining its price, a

preferred candidate  $p$ , and a natural number  $k$  representing the budget available to the briber. Solutions consist of a set of votes in the election  $E$  such that the sum of their prices does not exceed the budget  $k$ , and new votes to replace them with. The goal is to find a solution  $s$  maximizing  $\beta(E, s)$ .

**weighted-manipulation-max** Here, the input  $I$  consists of an election  $E$  where each vote is accompanied by its weight, the preferred candidate  $p$ , and a list of weights (of the manipulators). A solution consists of a vote for each manipulator. Again, the goal is to find a solution  $s$  such that  $\beta(E, s)$  is maximal.

Note that if we could compute the maximum value of  $\beta$  for a given optimization problem then, naturally, we could solve the corresponding decision problem. This means that if the corresponding decision problem is NP-hard (as is often the case for manipulation and bribery) then we cannot hope to compute the optimal value of  $\beta$  exactly. However, since  $\beta$  is defined as an increase in  $p$ 's performance and thus its maximum value is positive in most settings, we can attempt to use natural techniques to compute it approximately.

**Approximation Algorithms and Elections** The quality of an approximation algorithm is usually measured by comparing the solutions it computes to the optimal ones. For our optimization problems, an instance  $I$  contains the election itself, the preferred candidate, and additional parameters limiting the possible strategies for affecting the result of the election (i.e., the available budget in  $\$$ -bribery and the weights of the manipulators in manipulation). For such an instance  $I$  containing the election  $E$ , we define an *optimal solution* to be a solution  $s$  that achieves the maximal possible value of  $\beta(E, s)$  among *all* legal solutions. We define  $\text{OPT}(I)$  as  $\beta(E, s)$  for such an optimal solution  $s$ .

Given an instance  $I$ , an approximation algorithm  $\mathcal{A}$  is required to produce a legal solution  $\mathcal{A}(I)$ , that is, a solution that respects the constraints specified in  $I$ . For a positive real constant  $c$ , we say that  $\mathcal{A}$  is a *factor  $c$  approximation algorithm*, if for each instance  $I$  containing the election  $E$ , we have that  $\beta(E, \mathcal{A}(I)) \geq \frac{1}{c} \text{OPT}(I)$ . Such an algorithm guarantees that the effectiveness of the solution obtained from applying  $\mathcal{A}$  to the instance is at least  $\frac{1}{c}$  of the effectiveness that the optimal solution achieves.

We say that an algorithm  $\mathcal{A}$  is a *polynomial-time approximation scheme* if for each input  $(I, \varepsilon)$ , where  $\varepsilon$  is a rational value between 0 and 1 and where  $I$  contains an election  $E$ , it holds that: (a)  $\mathcal{A}$  produces a solution  $s = \mathcal{A}(I, \varepsilon)$  such that  $\beta(E, s) \geq (1 - \varepsilon) \cdot \text{OPT}(I)$ , and (b) for each fixed value of  $\varepsilon$ ,  $\mathcal{A}$  runs in time polynomial in  $|I|$ . If, in fact,  $\mathcal{A}$  runs in time polynomial in both  $|I|$  and  $\frac{1}{\varepsilon}$  then  $\mathcal{A}$  is a *fully polynomial-time approximation scheme* (FPTAS). In the current paper we only consider maximization problems, analogous definitions can be given for minimization problems as well.

## Manipulation in Scoring Protocols

Hemaspaandra and Hemaspaandra (2007) showed that for each scoring protocol  $\alpha = (\alpha_1, \dots, \alpha_m)$  such that it is not the case that  $\alpha_2 = \dots = \alpha_m$  the problem  $\alpha$ -weighted-manipulation is NP-complete (see also (Conitzer, Sand-

holm, & Lang 2007; Procaccia & Rosenschein 2006)). In this section we show that, nonetheless, weighted manipulation is easy for a large class of scoring protocols in practice. We do so via showing FPTASes for the scoring protocols in this class.

Let  $\alpha$  be a scoring protocol. An instance  $I$  of  $\alpha$ -weighted-manipulation-max is a tuple  $(E, w, p)$  where  $E = (C, V)$  is an election with candidate set  $C$  and weighted nonmanipulative voters  $V$ ,  $w = (w_1, \dots, w_n)$  is a sequence of weights of the manipulative voters, and  $p \in C$  is our preferred candidate. Our goal is to maximize the performance of  $p$ . That is, our goal is to find a solution  $sol$  such that  $\beta(E, sol) = \text{OPT}(I)$ .

**Theorem 2** *Let  $\alpha = (\alpha_0, \dots, \alpha_m)$  be a scoring protocol such that  $\alpha_0 > \alpha_1$ . There is an algorithm  $\mathcal{A}$  that given a rational number  $\varepsilon$ ,  $0 < \varepsilon < 1$ , and an instance  $I = (E, w, p)$  of  $\alpha$ -weighted-manipulation-max computes, in polynomial time in  $|I|$  and  $\frac{1}{\varepsilon}$ , a solution  $sol$  such that  $\beta(E, sol) \geq (1 - \varepsilon) \text{OPT}(I)$ .*

Before we prove this theorem, we need to build some infrastructure. Let  $\alpha = (\alpha_0, \dots, \alpha_m)$  be a scoring protocol where  $\alpha_0 > \alpha_1$  and let  $C = \{p, c_1, \dots, c_m\}$  be a set of candidates.  $p$  is our preferred candidate whose performance we want to maximize. We implicitly assume that we have a set  $V$  of nonmanipulative voters, however in this discussion the only incarnation of the nonmanipulative voters is through the sequence  $s$  below.

We let  $w = (w_1, \dots, w_n)$  be the sequence of weights of the manipulators. Naturally, to maximize  $p$ 's performance, each manipulator ranks  $p$  first. The complexity of  $\alpha$ -weighted-manipulation-max comes from the difficulty in arranging the remainder of the manipulators' votes in such a way as to minimize the score of  $p$ 's most dangerous competitor.

By  $\mathcal{E}(C, w)$  we mean the set of all elections over the candidate set  $C$  with voter set containing exactly voters with weights  $w_1, \dots, w_n$ . Let  $s = (s_1, \dots, s_m)$  be a sequence of nonnegative integers. Intuitively, the sequence  $s$  gives the scores that candidates  $c_1$  through  $c_m$  receive from the nonmanipulative voters. By  $S_\alpha(E, s)$  we mean  $\max_{i \in \{1, \dots, m\}} \{\text{score}_E(c_i) + s_i\}$  and by  $T_\alpha(w, s)$  we mean  $\min_{E \in \mathcal{E}(C, w)} S_\alpha(E, s)$ . Function  $T_\alpha(w, s)$  measures the smallest possible top score of a candidate from  $\{c_1, \dots, c_m\}$  after the manipulators cast their votes. For each scoring protocol  $\alpha$  there is an FPTAS for  $T_\alpha$ .

**Lemma 3** *Let  $\alpha = (\alpha_0, \dots, \alpha_m)$  be a scoring protocol and let  $C = \{p, c_1, \dots, c_m\}$ . There is an algorithm  $\mathcal{T}$  that given a rational number  $\varepsilon$ ,  $0 < \varepsilon < 1$ , a sequence  $s = (s_1, \dots, s_m)$  of nonnegative integers and a sequence of manipulators weights  $w = (w_1, \dots, w_n)$  computes an election  $E \in \mathcal{E}(C, w)$  such that  $S_\alpha(E, s) \leq (1 + \varepsilon) T_\alpha(w, s)$ . Algorithm  $\mathcal{T}$  runs in polynomial time in  $n, m$ , and  $\frac{1}{\varepsilon}$ .*

We skip the proof. However, we mention that it follows via a standard application of the scaling technique for developing FPTASes. With Lemma 3 at hand we can prove Theorem 2

*Proof.* Our input is  $I = (E, w, p)$ , where  $E = (C, V)$  is an election with candidate set  $C = \{p, c_1, \dots, c_m\}$  and set  $V$

of nonmanipulative voters,  $w = (w_1, \dots, w_n)$  is a sequence of manipulators' weights, and  $p$  is our preferred candidate. Our goal is to find a solution  $sol$  (a collection of votes for the manipulators to cast) that maximizes  $\beta(E, sol)$ .

Let  $W = \sum_{i=1}^n w_i$  and let  $w_{\max} = \max\{w_1, \dots, w_n\}$ . For each  $i$  in  $\{1, \dots, m\}$  let  $s_i = \text{score}_E(c_i)$ . We assume that the candidates  $c_1, \dots, c_m$  are listed in such an order that  $s_1 \geq s_2 \geq \dots \geq s_m$ . Since  $\alpha_0 > \alpha_1$ , in every optimal solution each manipulator ranks  $p$  first and so  $\text{OPT}(I) = W\alpha_0 - (T_\alpha(w, s) - s_1)$ . It would seem that computing approximately  $T_\alpha(w, s)$  should be enough to get a good approximation of  $\text{OPT}(I)$ , but unfortunately  $T_\alpha(w, s)$  can be much bigger than  $\text{OPT}(I)$ . We have to, in some sense, reduce its value first.

Note that we can disregard all candidates  $c_j$  such that  $s_1 - s_j > \alpha_1 W$ . If there are  $k$  such candidates then the manipulators may simply rank them on the first  $k$  positions after  $p$ . For the sake of simplicity, we assume that there are no such candidates.

Let  $s' = (s_1 - s_m, \dots, s_m - s_m)$ . It is easy to see that  $\text{OPT}(I) = W\alpha_0 - (T_\alpha(w, s) - s_1) = W\alpha_0 - (T_\alpha(w, s') - s'_1)$ . Additionally, via the above paragraph, we have that for each  $s'_i$  it holds that  $s'_i \leq \alpha_1 W$ . However, this means that  $T_\alpha(w, s') \leq 2\alpha_1 W$ . This is so because at worst the candidate whose score is the value of  $T_\alpha(w, s')$  gets  $\alpha_1 W$  points from  $s'$  and another  $\alpha_1 W$  points from the manipulators.

Using algorithm  $\mathcal{T}$  from Lemma 3, we fill-in the manipulators' votes to form an election  $E' \in \mathcal{E}(C, w)$  such that all voters in  $E'$  rank  $p$  first and  $T_\alpha(w, s') \leq S_\alpha(s', E') \leq (1 + \varepsilon')T_\alpha(w, s')$ , where  $\varepsilon' = \frac{1}{2\alpha_1}\varepsilon$ . (Recall that in our setting  $\alpha_1$  is a constant.) Votes obtained in this way are the solution  $sol$  that our algorithm produces and we have  $\beta(E, sol) = W\alpha_0 - (S_\alpha(s', E') - s'_1)$ . Note that

$$\begin{aligned} \text{OPT}(I) &= W\alpha_0 - (T_\alpha(w, s') - s'_1) \\ &\geq W\alpha_0 - (S_\alpha(s', E') - s'_1) \\ &\geq W\alpha_0 - ((1 + \varepsilon')T_\alpha(w, s') - s'_1) \\ &= W\alpha_0 - (T_\alpha(w, s') - s'_1) - \varepsilon'T_\alpha(w, s') \\ &= \text{OPT}(I) - \varepsilon'T_\alpha(w, s'). \end{aligned}$$

Since  $\text{OPT}(I) \geq W$  (this is a consequence of the fact that  $\alpha_0 > \alpha_1$ ),  $T_\alpha(w, s') \leq 2\alpha_1 W$ , and  $\varepsilon' = \frac{1}{2\alpha_1}\varepsilon$ , via the above calculations,  $\text{OPT}(I) \geq W\alpha_0 - (S_\alpha(s', E') - s'_1) \geq (1 - \varepsilon)\text{OPT}(I)$  and thus  $\text{OPT}(I) \geq \beta(E, sol) \geq (1 - \varepsilon)\text{OPT}(I)$ . This completes the proof.  $\square$

Interestingly, we can use Theorem 2 to obtain results similar to those of Zuckerman, Procaccia, and Rosenschein, but for the case of scoring protocols  $\alpha = (\alpha_0, \dots, \alpha_m)$  such that  $\alpha_0 > \alpha_1$ . Note that Theorem 4 below says that there is a separate algorithm for each scoring protocol of the given form.

**Theorem 4** *Let  $\varepsilon$  be a rational number,  $0 < \varepsilon < 1$ , and let  $\alpha = (\alpha_0, \dots, \alpha_m)$  be a scoring protocol such that  $\alpha_0 > \alpha_1$ . There is an algorithm that given an instance  $I = (E, w, p)$  of  $\alpha$ -weighted-manipulation, where  $w = (w_1, \dots, w_n)$  is the sequence of manipulators' weights, has the property that if there is a successful manipulation for instance  $I$ , it finds,*

*in polynomial time in  $|I|$  and  $\frac{1}{\varepsilon}$ , a successful manipulation for instance  $I' = (E, (w_1, \dots, w_n, w_{n+1}), p)$ , where  $w_{n+1} = \lceil \varepsilon \max\{w_1, \dots, w_n\} \rceil$ .*

We omit the easy proof. (The idea is to simply find a good enough approximation and then add a single voter, with appropriate weight, that ranks  $p$  first.)

Theorem 2 notwithstanding, we now show that for the case of an unbounded number of candidates there are no FPTASes for veto-weighted-manipulation-max and for  $k$ -approval-weighted-manipulation-max, unless  $P \neq NP$ .

**Theorem 5** *If  $P \neq NP$ , there is no FPTAS for veto-weighted-manipulation-max.*

It is well known that if a unary version of an NP-complete number problem is still NP-complete, i.e., if a version of an NP-complete problem where each number is encoded in unary is still NP-complete, then such a problem cannot have an FPTAS unless  $P \neq NP$ , see, e.g., (Garey & Johnson 1979). Thus, Theorem 5 follows directly from the following theorem.

**Theorem 6** *unary-veto-weighted-manipulation is NP-complete.*

*Proof.* We will reduce from the NP-complete problem Unary-3-Partition (Garey & Johnson 1979): Given a multi-set  $A$  of  $3m$  positive integers in unary and an integer bound  $B$  in unary such that for each  $a \in A$ ,  $B/4 < a < B/2$  and such that  $\sum_{a \in A} a = mB$ , does there exist a partition of  $A$  into  $m$  subsets  $A_1, \dots, A_m$  such that  $\sum_{a \in A_i} a = B$  for all  $i$ ? (Note that  $\|A_i\| = 3$  for all  $i$ ; hence the problem's name.)

Our reduction works as follows. The election consists of one voter of weight  $B$  with preference  $c_1 > c_2 > \dots > c_m > p$  and the manipulators have weights  $a_1, \dots, a_{3m}$ .

We claim that there is a successful partition of  $A$  if and only if  $p$  can be made a winner in our constructed election. First suppose that there exists a partition of  $A$  into  $m$  subsets  $A_1, \dots, A_m$  such that  $\sum_{a \in A_i} a = B$ . Let the manipulators corresponding to  $A_i$  veto candidate  $c_i$ . Note that every candidate  $c_i$  receives exactly  $B$  vetoes. In the resulting election,  $\text{score}(p) = mB$  ( $p$  is never vetoed), and  $\text{score}(c_i) = B + mB - B = mB$ , and so  $p$  is a winner of the election. For the converse, suppose the manipulators vote in such a way that  $p$  is a winner of the election. Without loss of generality, we may assume that  $p$  is never vetoed, and so  $\text{score}(p) = mB$ . In order for  $p$  to be a winner,  $\text{score}(c_i)$  can be at most  $mB$ .  $\sum_{c_i} \text{score}(c_i) = m^2 B$ , and so this can only happen if  $\text{score}(c_i) = mB$  for all  $c_i$ . It follows that each  $c_i$  receives exactly  $B$  vetoes. Let  $A_i$  consist of the multi-set of the weights of the voters that veto  $c_i$ . Then  $A_1, \dots, A_m$  is a partition such that  $\sum_{a \in A_i} a = B$  for all  $i$ .  $\square$

The same approach can be used to show NP-hardness (and thus non-existence of FPTAS unless  $P = NP$ ) for unary manipulation for many other scoring protocols with an unbounded number of candidates. In particular,

**Theorem 7** *Let  $k \geq 2$ . If  $P \neq NP$ , there is no FPTAS for  $k$ -veto-weighted-manipulation-max,  $k$ -approval-weighted-manipulation-max, and generalized versions of*

$k$ -approval-weighted-manipulation-max where, as in  $k$ -approval, voters give only points to the first  $k$  candidates, but any  $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_k > 0$  is allowed.

## The Bribery Problem for Borda Count

In this section, we prove hardness results for the Borda count election system.

**NP-hardness of the decision version** We start by showing that Borda-bribery is NP-complete. In Borda-bribery we are given an election  $E$ , a distinguished candidate  $p$ , and a nonnegative integer  $k$ , and we ask if it is possible to ensure that  $p$  is a winner of  $E$  via modifying at most  $k$  votes. Note that our result regards the simplest variant of bribery where each voter has unit weight and unit price. Hardness of more involved variants (i.e., ones including prices or weights or both) follows naturally.

Our proof works via a reduction from a specifically crafted restriction of the set cover problem.

*Problem:* 34-XC  
*Input:* Set  $S$ ,  $\|S\| = n$ , sets  $T_1, \dots, T_{\frac{3}{4}n}$ , where  $\|T_i\| = 4$  for each  $T_i$  and where each  $s \in S$  is in exactly 3 sets  $T_i$ .  
*Question:* Is there a set  $I \subseteq \{1, \dots, \frac{3}{4}n\}$  such that for  $i, j \in I, i \neq j, T_i \cap T_j = \emptyset$  and  $\cup_{i \in I} T_i = S$ ?

Note that, by definition, each correct solution  $I$  has exactly  $\frac{1}{4}n$  elements. One can show, via a routine proof, that 34-XC is NP-complete.

**Theorem 8** *Borda-bribery is NP-complete.*

*Proof.* For a set  $A$  of candidates, writing  $A$  in a vote means  $A$  in some arbitrary, but fixed, order.  $\overleftarrow{A}$  denotes the candidates of  $A$  in reverse order.

Let  $S = \{s_1, \dots, s_n\}$ ,  $T_1, \dots, T_{\frac{3}{4}n}$  be an instance of 34-XC. Let  $k_1 = n^4$  and  $k_2 = n^4 - 8n^3 - 4n^2$  (without loss of generality, assume that  $n$  is large enough for  $k_2$  to be positive). Our candidate set is  $\{p\} \cup S \cup P_1 \cup P_2$ , where  $P_1$  and  $P_2$  are sets of padding candidates such that  $\|P_1\| = k_1$  and  $\|P_2\| = k_2$ . We set  $P = P_1 \cup P_2$ . The voter set is defined as follows. For each set  $T_i$ , we introduce a voter who votes as follows:

$$v_i = T_i > P_1 > S \setminus T_i > P_2 > p.$$

We also introduce  $m$  votes of the form  $p > S > P$  and  $m$  votes of the form  $\overleftarrow{S} > p > \overleftarrow{P}$ . By increasing  $m$  we increase the point differences between pairs of candidates where one candidate comes from  $S \cup \{p\}$  and the other from  $P$ , without at the same time affecting the point differences between pairs of candidates where both candidates come from  $S \cup \{p\}$  or both come from  $P$ . In particular, we can choose  $m$  large enough such that with bribing at most  $\frac{1}{4}n$  voters, the padding candidates cannot be made to win the election. We choose such a value for  $m$ , and hence the briber only has to ensure that the candidate  $p$  has at least as many points as each candidate in  $S$  in order for  $p$  to win the election. This allows us to establish a direct correspondence between bribery attempts bribing at most  $\frac{1}{4}n$  voters and the 34-XC instance, by showing that a bribery is successful if and only if the bribed votes

of the form  $v_i$  correspond to a set cover (and the new votes are set in a reasonable way, i.e., they rank  $p$  first, then the padding candidates, and then the candidates in  $S$ ). Details are left out due to space restrictions.  $\square$

**Nonapproximability of Bribery** We now show that there are no efficient approximation algorithms for  $\$$ -bribery-max. The following result does not only show that there is no polynomial-time approximation algorithm for the problem that achieves an approximation rate of a constant factor, it also excludes a polynomial relationship between results that can be achieved efficiently and the optimal solution.

**Theorem 9** *For every polynomial  $q$  there is no polynomial-time approximation algorithm  $\mathcal{A}$  for Borda- $\$$ -bribery-max such that for all instances  $I$  containing the election  $E$ ,  $\mathcal{A}$  computes a solution  $s$  such that  $q(\beta(E, s)) \geq \text{OPT}(I)$ , unless  $\text{P} = \text{NP}$ .*

Note that this result is significantly stronger than just excluding a constant-ratio approximation algorithm: It also shows that, for example, there is no polynomial-time approximation algorithm  $\mathcal{A}$  that guarantees to produce a solution  $\mathcal{A}(I)$  for every instance  $I$  containing the election  $E$  such that  $\beta(E, \mathcal{A}(I))$  is at least  $(\text{OPT}(I))^{1/c}$  for every constant  $c$ .

*Proof.* Let  $q$  be a polynomial and assume  $\mathcal{A}$  is a polynomial time approximation algorithm for  $\$$ -bribery-max, such that for all instances  $I$  containing the election  $E$ :  $q(\beta(E, \mathcal{A}(I))) \geq \text{OPT}(I)$ . We show that we can use  $\mathcal{A}$  to decide 34-XC in polynomial time. Note that the construction is similar but easier than the one in the proof of Theorem 8.

Choose  $d, n_0 \in \mathbb{N}$  such that  $q(k) \leq k^d$  for all  $k \geq n_0$ . Let  $S = \{s_1, \dots, s_n\}$ ,  $T_1, \dots, T_{\frac{3}{4}n}$  be an instance of 34-XC. Without loss of generality assume that  $n \geq n_0$ . Let  $m$  be a natural number such that  $m > n^{2d} + \frac{3}{4}n^2 - \frac{15}{4}n + 11$  and let  $C = S \cup \{p, c_1, \dots, c_m\}$  be a set of candidates, where  $p$  is our preferred candidate. Let  $V = \{v_1, \dots, v_{\frac{3}{4}n}\}$  be a set of votes with

$$v_i = p > T_i > c_1 > \dots > c_m > S \setminus T_i$$

for every  $i \in \{1, \dots, \frac{3}{4}n\}$ , and let  $W = \{w_1, w'_1, \dots, w_l, w'_l\}$  be a set of votes with

$$w_i = p > s_1 > \dots > s_n > c_1 > \dots > c_m$$

and

$$w'_i = s_n > \dots > s_1 > p > c_m > \dots > c_1$$

for every  $i \in \{1, \dots, l\}$ . We set the price of each vote in  $V$  to 1 and the price of each vote in  $W$  to  $\frac{1}{4}n + 1$ . The effect of  $W$  is that it leaves the relative scores of  $p$  and the candidates in  $S$  invariant, while increasing them relatively to the scores of the padding candidates  $c_1, \dots, c_m$ . We introduce enough of these votes such that for every possible bribery, the candidates in  $S$  will always have more points than the padding candidates. Clearly a polynomial number of these votes suffices. The algorithm  $\mathcal{A}$  cannot change these votes, since their cost exceeds the budget  $\frac{1}{4}n$ .

Let  $E$  be the election with candidates  $C$  and votes  $V \cup W$ . We apply  $\mathcal{A}$  on the instance  $I = (E, \frac{1}{4}n, p)$  and show that  $q(\beta(E, \mathcal{A}(I))) > n^{2d}$  if and only if  $S, T_1, \dots, T_{\frac{3}{4}n}$  is a yes-instance of 34-XC. This shows that we can use  $\mathcal{A}$  to decide the problem 34-XC, which can only happen if  $P = NP$ .

First note that  $\text{score}_E(p) = \frac{3}{4}n(n+m) + l(m + \frac{n}{2})$  and for each  $s \in S$ :  $3(n+m-4) + l(m + \frac{n}{2}) \leq \text{score}_E(s) \leq 3(n+m-1) + (\frac{3}{4}n-3)(n-5) + l(m + \frac{n}{2})$ .

Now let  $S, T_1, \dots, T_{\frac{3}{4}n}$  be a yes-instance of 34-XC and let  $J \subseteq \{1, \dots, \frac{3}{4}n\}$  specify an exact cover of  $S$ . We bribe in the following way: For every  $i \in J$  we replace  $v_i$  by  $v'_i = p > c_1 > \dots > c_m > S$ . Let  $V' = \{v'_i \mid i \in J\} \cup \{v_i \mid i \in \{1, \dots, \frac{3}{4}n\} \setminus J\}$  be the set of votes obtained from  $V$  with this bribe and  $E'$  the election with candidates  $C$  and votes  $V' \cup W$ . Since  $J$  is an exact cover, for every candidate  $s \in S$  we changed exactly one of the votes in  $V$  where  $s$  was in a position among the top five candidates, therefore there are two votes left in  $V$  where  $s$  is in one of the first five positions and in all other votes in  $V$   $s$  is voted among the last  $n$  candidates. Thus  $\text{score}_{E'}(s) \leq 2(n+m-1) + (\frac{3}{4}n-2)(n-1) + l(m + \frac{n}{2})$ . Note that  $\text{score}_{E'}(p) = \text{score}_E(p)$ . It follows  $q(\beta(E, \mathcal{A}(I))) \geq \text{OPT}(I) \geq m - \frac{3}{4}n^2 + \frac{15}{4}n - 11 > n^{2d}$ .

Now assume there is no exact cover for  $S, T_1, \dots, T_{\frac{3}{4}n}$ . Note that  $\mathcal{A}(I)$  changes exactly  $\frac{1}{4}n$  votes from  $V$  and no vote from  $W$ . Let  $E_{\mathcal{A}}$  be the election obtained from  $E$  by applying the bribing strategy  $\mathcal{A}(I)$ . Since there is no exact cover, there is a candidate  $s \in S$  such that for all  $v_i \in V$  with  $s \in T_i$  it holds that  $v_i$  is not changed by  $\mathcal{A}(I)$  and therefore  $v_i$  is a vote in  $E_{\mathcal{A}}$ . That means there are at least three votes in  $E_{\mathcal{A}}$  that rank  $s$  among the first 5 candidates, therefore we get the lower bound  $\text{score}_{E_{\mathcal{A}}}(s) \geq 3(n+m-4) + l(m + \frac{n}{2})$ . By assuming that  $p$  is ranked first in all votes it follows  $\text{score}_{E_{\mathcal{A}}}(p) \leq \frac{3}{4}n(n+m-1) + l(m + \frac{n}{2})$ . Hence  $\beta(E, \mathcal{A}(I)) \leq \frac{3}{4}n^2 - \frac{27}{4}n + 24$ . Without loss of generality assume that  $n$  is large enough to ensure that  $\frac{3}{4}n^2 - \frac{27}{4}n + 24 \leq n^2$ . Then  $q(E, \beta(\mathcal{A}(I))) \leq q(n^2) \leq n^{2d}$ , concluding the proof.  $\square$

Note that the above proof also works for a variant of the (unpriced) bribery problem where voters do not have prices, but only indicators whether they can be bribed or not.

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