

FPT Approximation Schemes for Shift Bribery

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Motivation

What is the “best” way to cope with NP-hard problems?

Some opinions:

Robert: FPA \succ Rand \succ Avg \succ Heur

Jiehua: Rand \succ FPA \succ Avg \succ Heur

(heretical) André: Heur \succ Avg \succ Rand \succ FPA

FPA fixed-parameter approximation

Rand randomized algorithms

Avg average-case analysis

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- ▶ A set of candidates $C = \{c_1, \dots, c_m\}$.
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Copeland $^\alpha$ Head-to-head contest between each two candidates.

Winner: 1 point, loser: 0 points, tie: α points each.

Shift Bribery

Setting:

An external agent may influence voter to shift some preferred candidate forward in the voters' preference list.

[Elkind, Faliszewski, and Slinko. Swap Bribery. SAGT '09]

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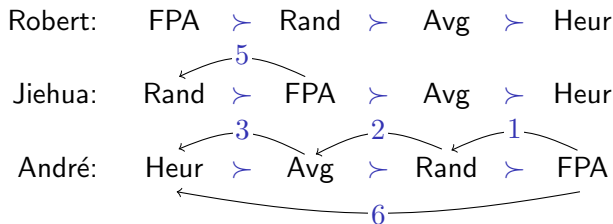
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- ▶ **each voter has a different price depending on the “shift length”**

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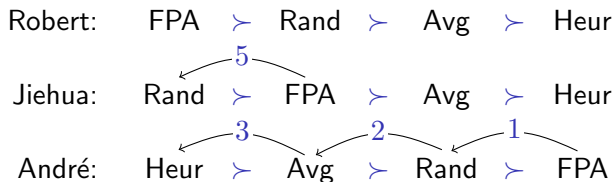


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Question: Is there a shift action $\vec{s} = (s_1, \dots, s_n) \in \mathbb{N}^n$ with $\Pi(\vec{s}) \leq B$ such that p is a winner according to rule \mathcal{R} ?



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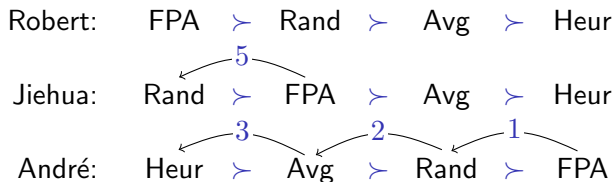
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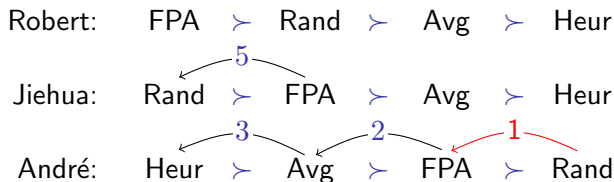
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What is known about the Shift Bribery?

[Elkind, Faliszewski, and Slinko. Swap Bribery. SAGT '09]:

- ▶ Polynomial-time algorithm for t -Approval voting rule
- ▶ NP-hard for Borda, Copeland, and other voting rules
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[Schlotter, Faliszewski, and Elkind. Campaign management under approval-driven voting rules. AAI '11]:

- ▶ Polynomial-time algorithm for Bucklin and Fallback voting rule
- ▶ Problem variant: **Support Bribery** (briber extends incomplete votes)

Our Results

[Bredereck, Chen, Faliszewski, N., and Niedermeier. Prices matter for the parameterized complexity of shift bribery. AAAI '14]:

- ▶ There is a factor- $(1 + \varepsilon)$ approximation algorithm solving \mathcal{R} **Shift Bribery** in time $O^*(\lceil n/\varepsilon + 1 \rceil^n)$ for $\mathcal{R} \in \{\text{Copeland, Borda}\}$.

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Update: \mathcal{R} **Shift Bribery** for $\mathcal{R} \in \{\text{Copeland, Borda}\}$ is fixed-parameter tractable with respect to the number m of candidates.

[Bredereck, Faliszewski, Niedermeier, Skowron, and Talmon. Elections with Few Candidates: Prices, Weights, and Covering Problems. ADT' 15]

Further Results

		\mathcal{R} Shift Bribery				
parameter	\mathcal{R}	unit	convex	all	sort	0/1

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#voters (n)	Borda Copeland ^{α}					

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#voters (n)	Borda	FPT-AS				FPT
	Copeland $^\alpha$	FPT-AS / W[1]-hard				FPT

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Algorithm outline:

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Observation: Above algorithm works for any voting rule where winner determination is fixed-parameter tractable with respect to n .

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Rescaling:

- ▶ Guess maximum cost π_{\max} spend at a single voter.
- ▶ Set $K = \varepsilon \cdot \pi_{\max}/n$ and

$$\text{for each } v_i \in V, j \in \{1, \dots, m\} : \pi'_i(j) = \begin{cases} \left\lceil \frac{\pi_i(j)}{K} \right\rceil & \text{if } \pi_i(j) \leq \pi_{\max} \\ \infty & \text{otherwise.} \end{cases}$$

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Running time: Step 1: $O((nm)^2)$

Step 2: $O((\lceil n/\varepsilon \rceil + 1)^n \cdot \text{running time for winner determination})$

In total: $O^*((\lceil n/\varepsilon \rceil + 1)^n)$ for Copeland and Borda voting rule

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Approximation factor: algorithm solution: \vec{s}' optimal solution: \vec{s}
(A solution is a vector $\in \{0, \dots, m\}^n$ storing the number of shifts for each voter.)

cost of \vec{s}' : $\Pi(\vec{s}')$

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$$\Pi(\vec{s}') \leq K \cdot \Pi'(\vec{s}') \leq K \cdot \Pi'(\vec{s}) \leq \Pi(\vec{s}) + K \cdot n$$

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$$\begin{aligned} \Pi(\vec{s}') &\leq K \cdot \Pi'(\vec{s}') \leq K \cdot \Pi'(\vec{s}) \leq \Pi(\vec{s}) + K \cdot n \\ &= \Pi(\vec{s}) + \varepsilon \cdot \pi_{\max} \leq \Pi(\vec{s}) + \varepsilon \cdot \Pi(\vec{s}) = (1 + \varepsilon)\Pi(\vec{s}) \end{aligned}$$

Fixed-Parameter Approximation Scheme I: Summary

We saw: There is a factor- $(1 + \varepsilon)$ approximation algorithm solving \mathcal{R} **Shift Bribery** in time $O^*(\lceil n/\varepsilon + 1 \rceil^n)$ for $\mathcal{R} \in \{\text{Copeland, Borda}\}$.

Recall:

- ▶ Copeland **Shift Bribery** parameterized by the number n of voters is $W[1]$ -hard for unit prices.
- ▶ There is a polynomial-time factor- m approximation algorithm for Copeland **Shift Bribery**. [Elkind and Faliszewski. *Approximation algorithms for campaign management*. WINE '10]

Open:

Is there a polynomial-time factor- $\log(m)$ (or better) approximation for Copeland **Shift Bribery**?

Fixed-Parameter Approximation Scheme II: Overview

\mathcal{R} Shift Bribery

Input: An election $E = (C, V)$, a specific candidate $p \in C$, and a price function list $\Pi = (\pi_1, \dots, \pi_n)$.

Taks: Find a minimum-cost shift action that makes p a winner according to rule \mathcal{R} ?

Parameter $m := |C|$.

Goal: $(1 + \varepsilon)$ approximation in $f(m, \varepsilon) \cdot \text{poly}$

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First Observations:

- ▶ at most $m!$ different preference orders
 \rightsquigarrow at most $m!$ “blocks” of voters
- ▶ The prices give a sorting of the voters in one block.
 \rightsquigarrow For each two voters of one block, one never shifts p farther in the preference order of the more expensive voter.

Fixed-Parameter Approximation Scheme II: Idea

Brief idea of the algorithm:

Step 1: Spent most of the budget optimally in $O^*(f(m, \varepsilon))$ time:
(exact part of the algorithm)

Step 2: Spent the remaining budget uniformly.
(approximation part of the algorithm)

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Theorem

There is a factor- $(1 + 2\varepsilon + \varepsilon^2)$ approximation algorithm solving \mathcal{R} **Shift Bribery** for sortable prices in time $O^*(M^{M \cdot \lceil \ln(M/\varepsilon) \rceil + 1})$ (where $M = m \cdot m!$) for $\mathcal{R} \in \{\text{Copeland, Borda}\}$.

Fixed-Parameter Approximation Scheme II: Remarks

- ▶ Algorithm relies on two observations:
 - ▶ At most $m!$ different preference orders (voter blocks).
 - ▶ Prices induce sorting on voters in one voter block.
- ▶ FPT algorithm with respect to parameter m uses the same two observations to construct a mixed integer programming.

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Open questions:

- ▶ What is the parameterized complexity in the unrestricted price functions?
(Known: **Shift Bribery** is in XP wrt. m)
- ▶ Is there a fixed-parameter approximation scheme for unrestricted price functions?

Conclusion

Mostly only polynomial-time approximation algorithms or only fixed-parameter algorithms for voting problems.

- ~> Just few fixed-parameter approximation results for voting problems.
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Thank you!