

# Fixed-size Minimax for Committee Elections: Approximation and Local Search Heuristics

COMSOC '06

6 December 2006

**Rob LeGrand**

Washington University in St. Louis

legrand@cse.wustl.edu

**Evangelos Markakis**

University of Toronto

vangelis@cs.toronto.edu

**Aranyak Mehta**

IBM Almaden Research Center

mehtaa@us.ibm.com

# Electing a committee from approval ballots

$n = 5$  candidates

$m = 6$  ballots

11110

00011

approves of  
candidates  
4 and 5

01111

00111

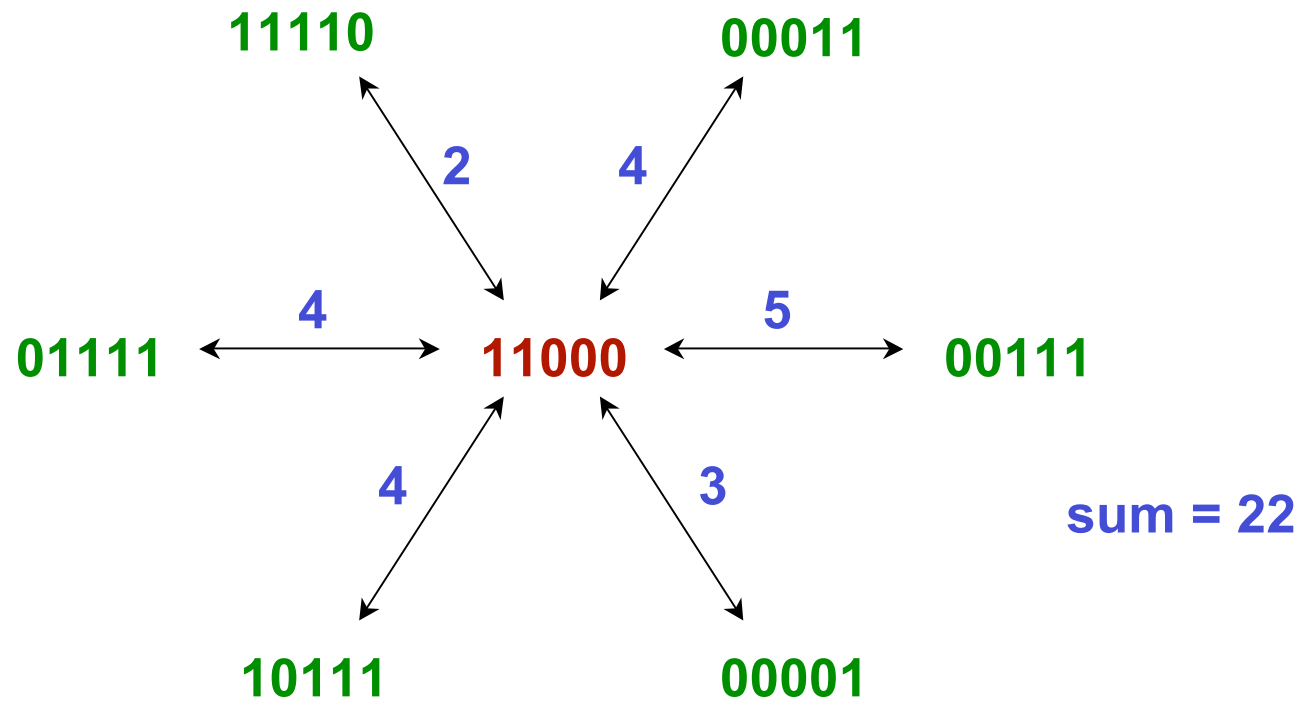
10111

00001

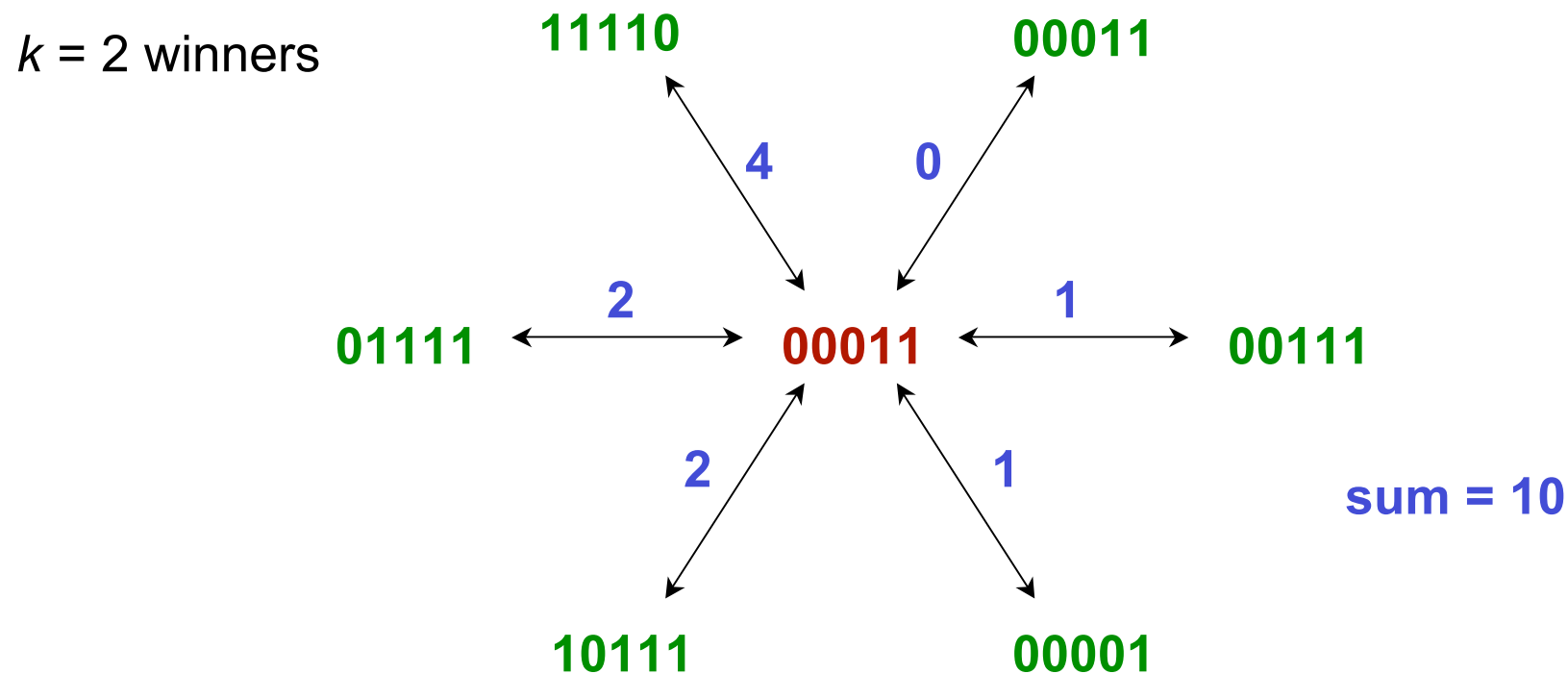
- What's the best committee of size  $k = 2$ ?

# Sum of Hamming distances

$k = 2$  winners



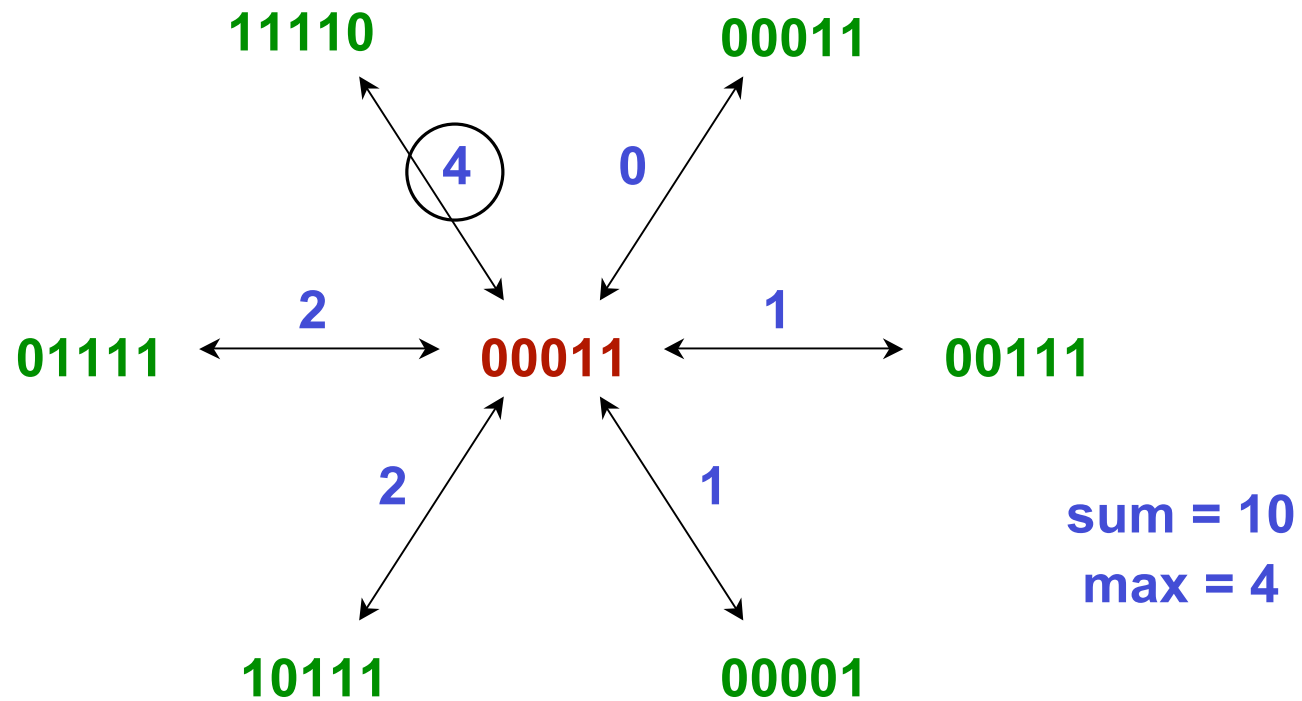
## Fixed-size minisum



- Minisum elects winner set with smallest sumscore
- Easy to compute (pick candidates with most approvals)

# Maximum Hamming distance

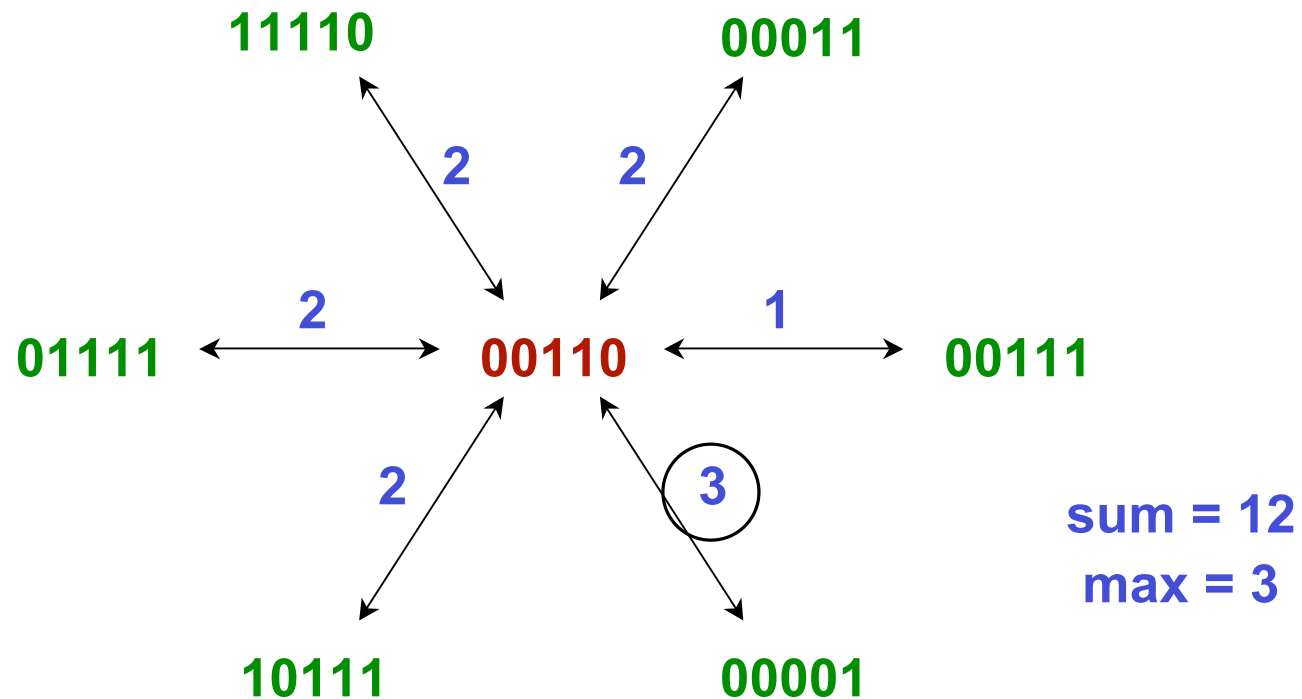
$k = 2$  winners



# Fixed-size minimax

[Brams, Kilgour & Sanver, '04]

$k = 2$  winners



- Minimax elects winner set with smallest maxscore
- Harder to compute?

# Complexity

Endogenous minimax = EM = BSM(0, $n$ )	Bounded-size minimax = BSM( $k_1$ , $k_2$ )	Fixed-size minimax = FSM( $k$ ) = BSM( $k$ , $k$ )
<b>NP-hard</b>  [Frances & Litman, '97]	<b>NP-hard</b>  (generalization of EM)	?

# Complexity

Endogenous minimax = EM = BSM(0, $n$ )	Bounded-size minimax = BSM( $k_1$ , $k_2$ )	Fixed-size minimax = FSM( $k$ ) = BSM( $k$ , $k$ )
<b>NP-hard</b>  [Frances & Litman, '97]	<b>NP-hard</b>  (generalization of EM)	<b>NP-hard</b>  (this paper)



# Approximability

Endogenous minimax = EM = BSM(0, $n$ )	Bounded-size minimax = BSM( $k_1$ , $k_2$ )	Fixed-size minimax = FSM( $k$ ) = BSM( $k$ , $k$ )
has a PTAS* [Li, Ma & Wang, '99]	no known PTAS; no known constant- factor approx.	no known PTAS; no known constant- factor approx.

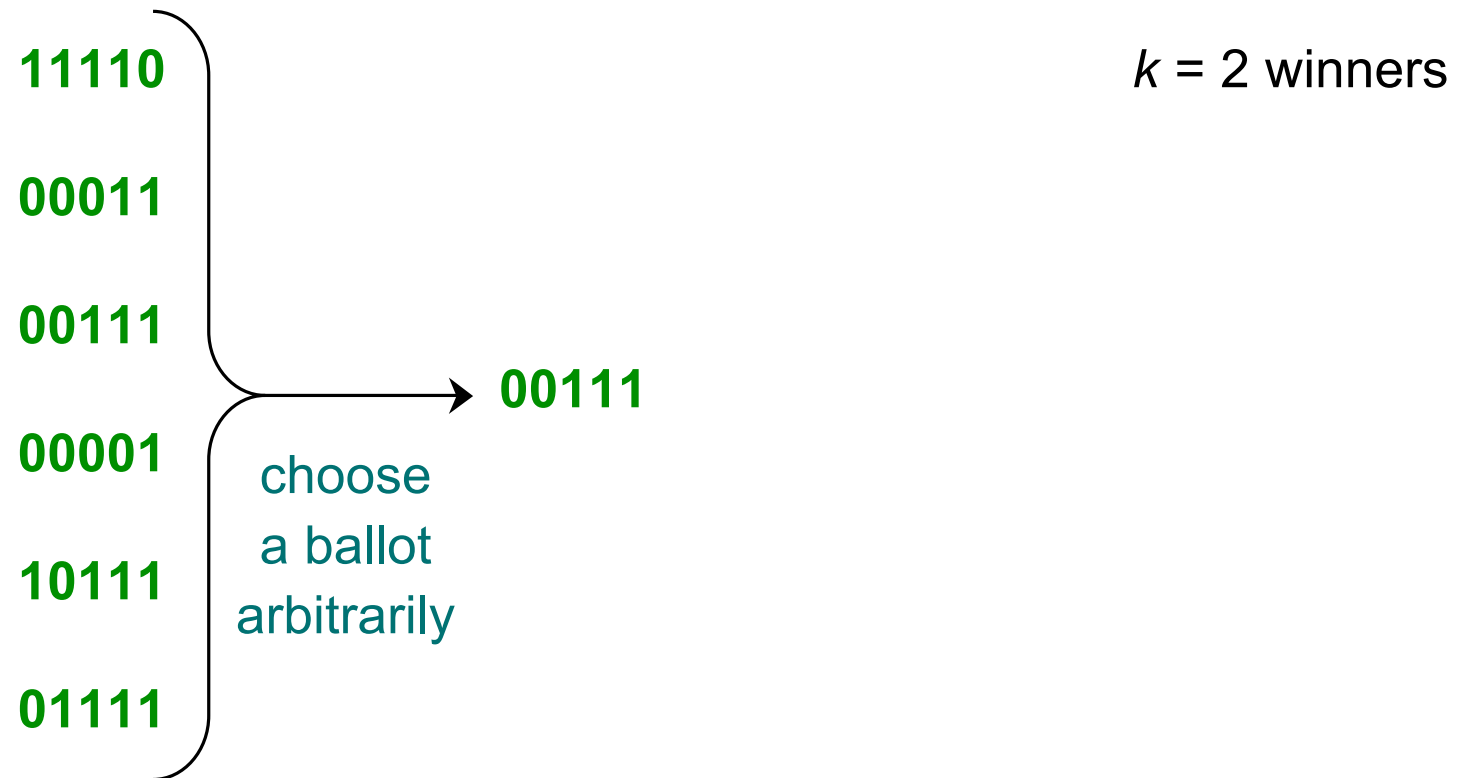
\* Polynomial-Time Approximation Scheme: algorithm with approx. ratio  $1 + \epsilon$  that runs in time polynomial in the input and exponential in  $1/\epsilon$

# Approximability

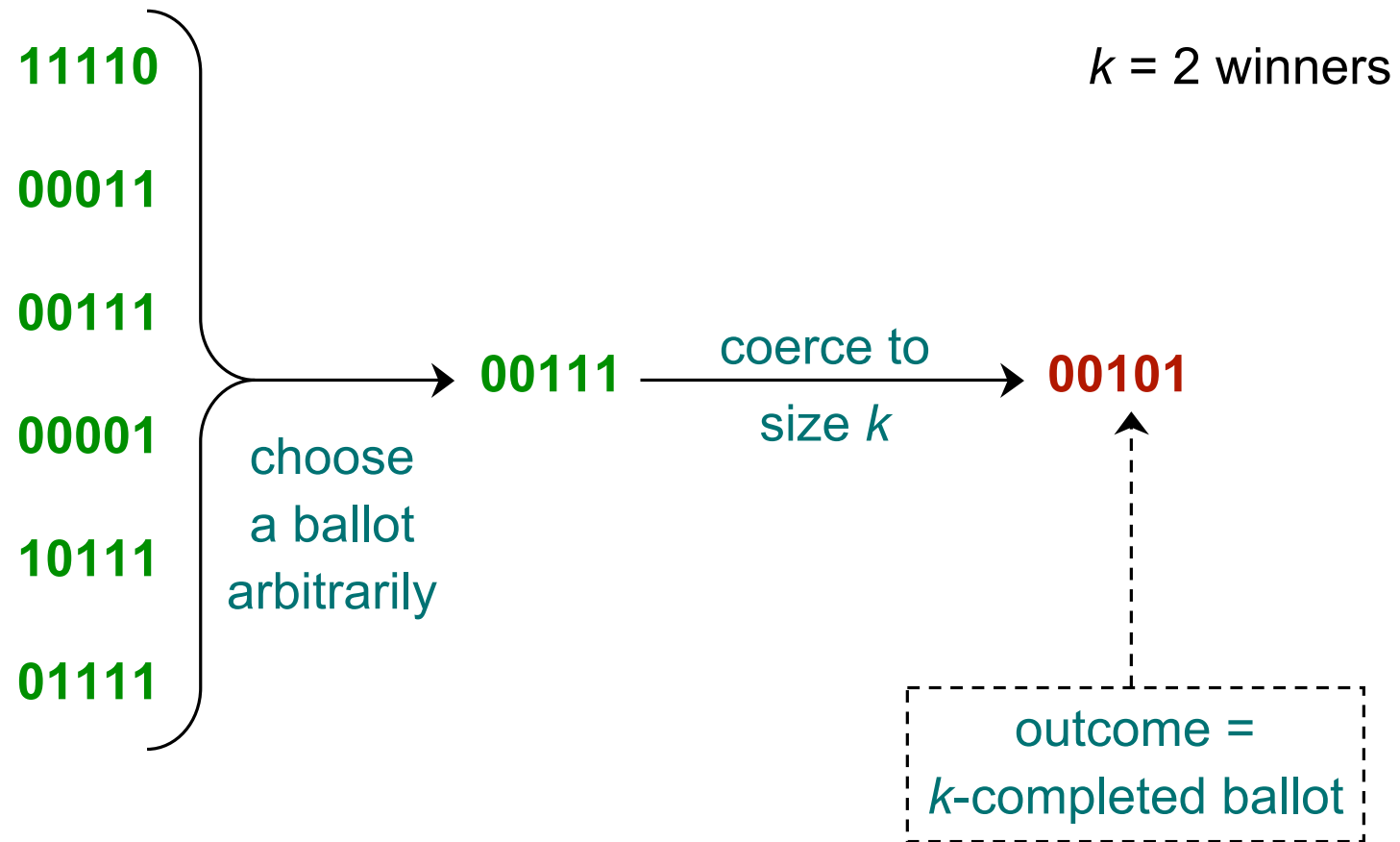
Endogenous minimax = EM = BSM(0, $n$ )	Bounded-size minimax = BSM( $k_1$ , $k_2$ )	Fixed-size minimax = FSM( $k$ ) = BSM( $k$ , $k$ )
has a PTAS* [Li, Ma & Wang, '99]	no known PTAS; has a 3-approx. (this paper)	no known PTAS; has a 3-approx. (this paper)

\* Polynomial-Time Approximation Scheme: algorithm with approx. ratio  $1 + \varepsilon$  that runs in time polynomial in the input and exponential in  $1/\varepsilon$

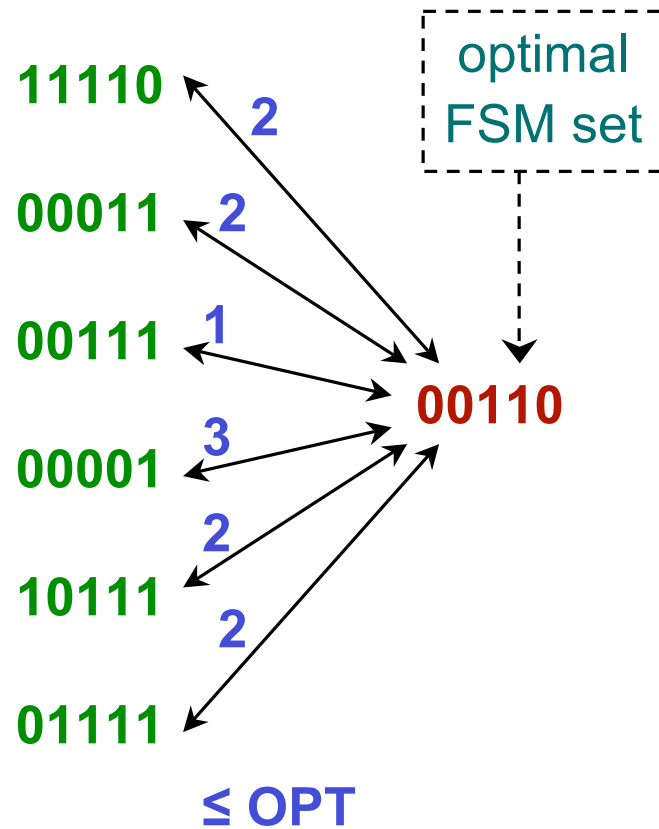
# Approximating FSM



# Approximating FSM

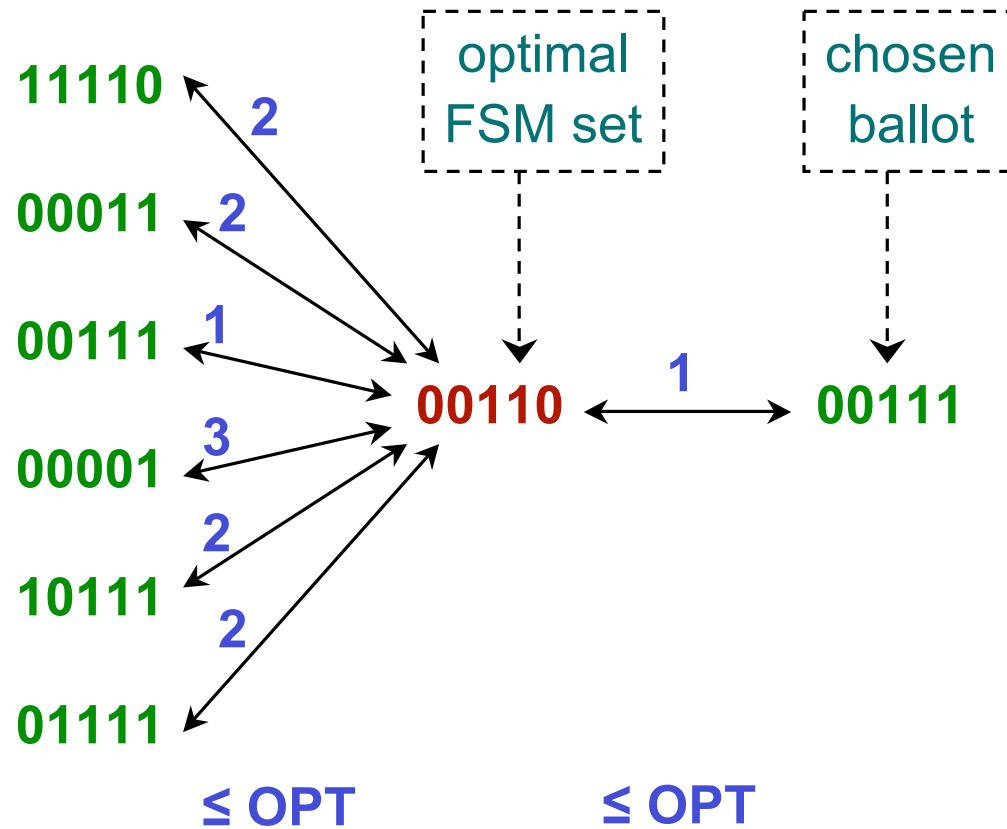


## Approximation ratio $\leq 3$



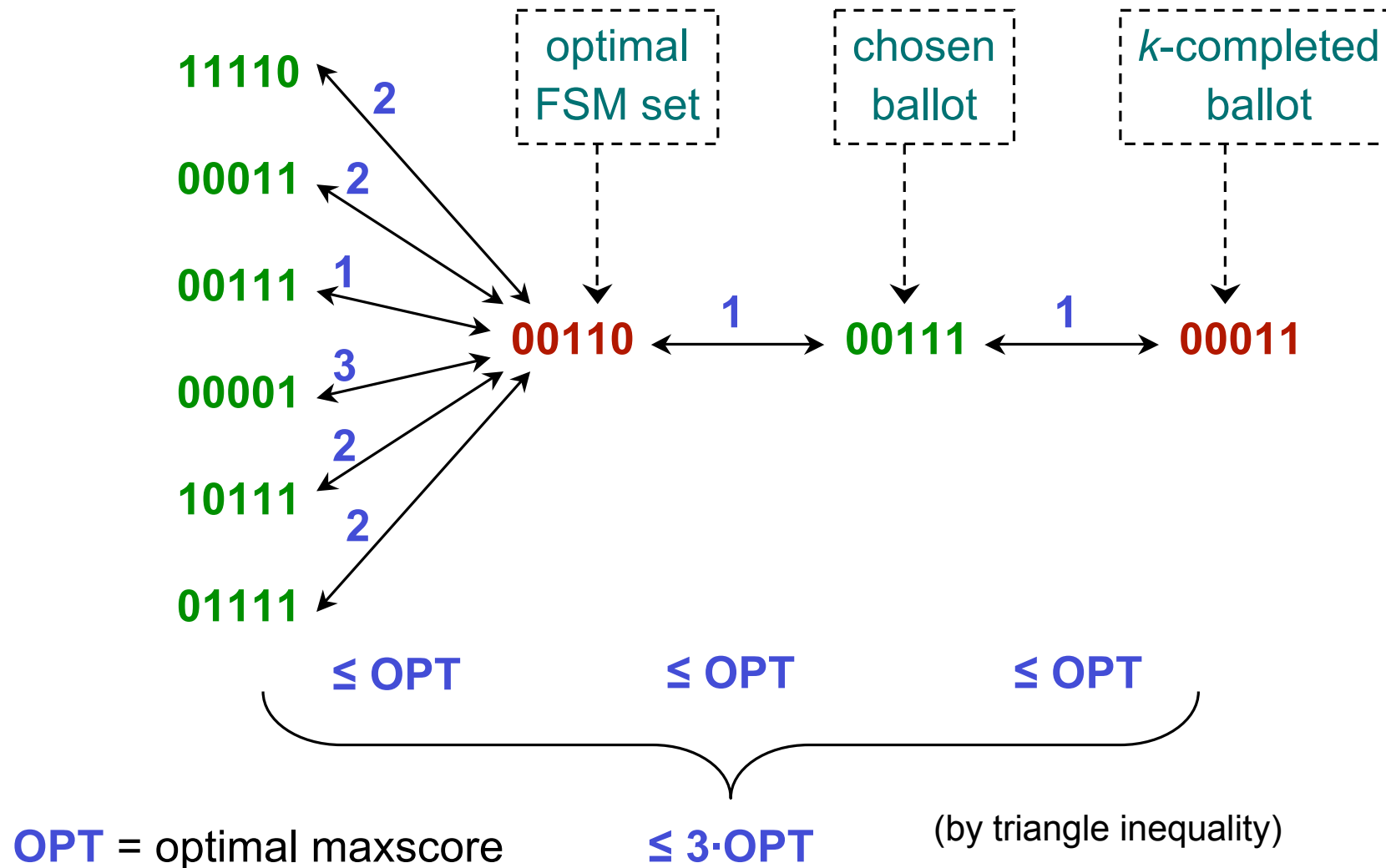
**OPT** = optimal maxscore

# Approximation ratio $\leq 3$



**OPT** = optimal maxscore

# Approximation ratio $\leq 3$



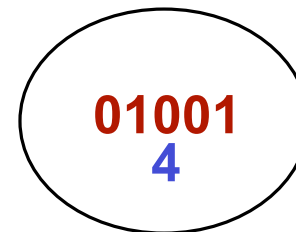
## Better in practice?

- So far, we can guarantee a winner set no more than 3 times as bad as the optimal.
  - Nice in theory . . .
- How can we do better in practice?
  - Try local search



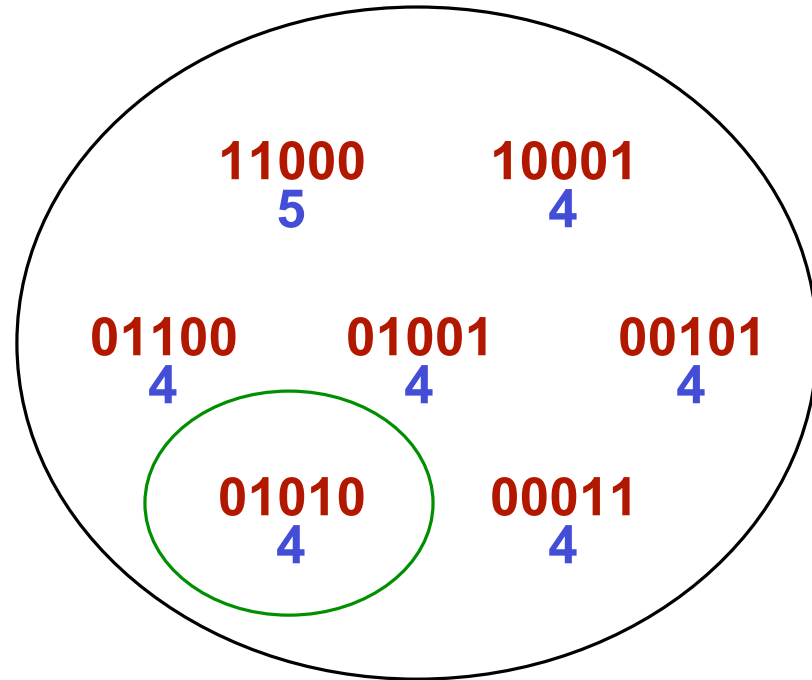
# Local search approach for FSM

1. Start with some  $c \in \{0,1\}^n$   
of weight  $k$



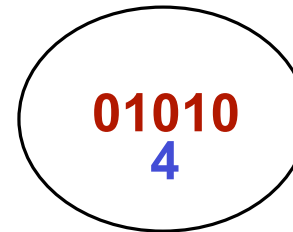
## Local search approach for FSM

1. Start with some  $c \in \{0,1\}^n$  of weight  $k$
2. In  $c$ , swap up to  $r$  0-bits with 1-bits in such a way that minimizes the maxscore of the result



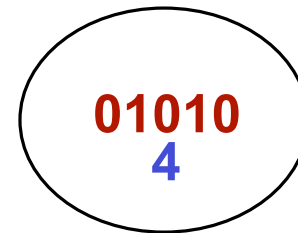
## Local search approach for FSM

1. Start with some  $c \in \{0,1\}^n$  of weight  $k$
2. In  $c$ , swap up to  $r$  0-bits with 1-bits in such a way that minimizes the maxscore of the result



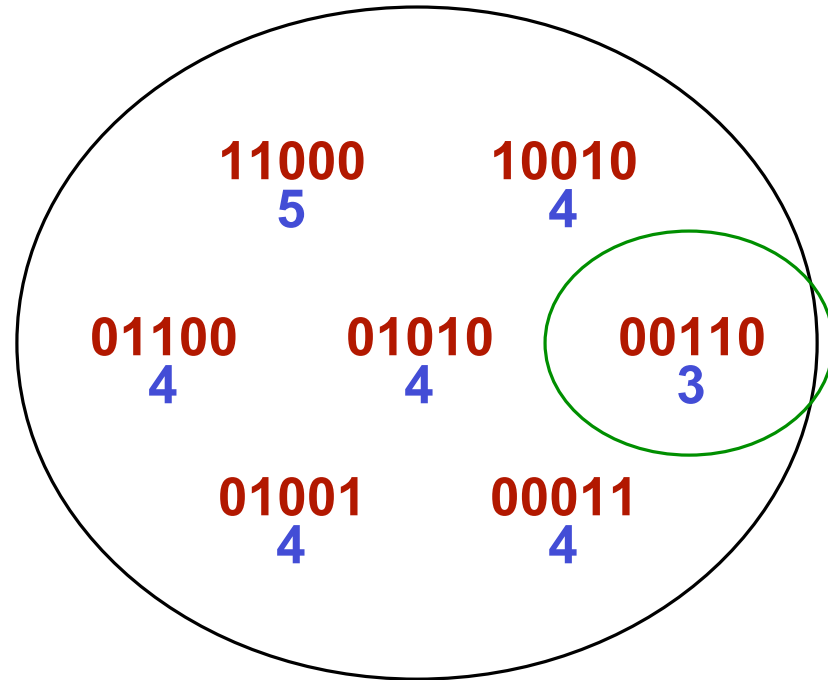
## Local search approach for FSM

1. Start with some  $c \in \{0,1\}^n$  of weight  $k$
2. In  $c$ , swap up to  $r$  0-bits with 1-bits in such a way that minimizes the maxscore of the result



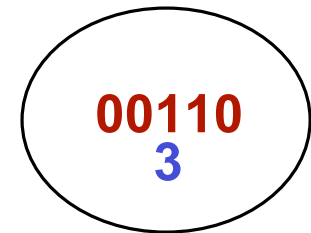
## Local search approach for FSM

1. Start with some  $c \in \{0,1\}^n$  of weight  $k$
2. In  $c$ , swap up to  $r$  0-bits with 1-bits in such a way that minimizes the maxscore of the result
3. Repeat step 2 until  $\text{maxscore}(c)$  is unchanged  $n$  times
4. Take  $c$  as the solution



## Local search approach for FSM

1. Start with some  $c \in \{0,1\}^n$  of weight  $k$
2. In  $c$ , swap up to  $r$  0-bits with 1-bits in such a way that minimizes the maxscore of the result
3. Repeat step 2 until  $\text{maxscore}(c)$  is unchanged  $n$  times
4. Take  $c$  as the solution



## Specific FSM heuristics

- Two parameters:
  - where to start vector  $c$ :
    1. a fixed-size-minisum solution
    2. a  $k$ -completion of a ballot (3-approx.)
    3. a random set of  $k$  candidates
    4. a  $k$ -completion of a ballot with highest maxscore
  - radius of neighborhood  $r$ : 1 and 2

## Heuristic evaluation

- Real-world ballots from GTS 2003 council election
- Found exact minimax solution
- Ran each heuristic 5000 times
- Compared exact minimax solution with heuristics to find realized approximation ratios
  - example:  $15/14 = 1.0714$ 
    - maxscore of solution found = 15
    - maxscore of exact solution = 14
- We also performed experiments using ballots generated according to random distributions (see paper)



## Average approx. ratios found

	radius = 1	radius = 2
fixed-size minimax	1.0012	1.0000
3-approx.	1.0017	1.0000
random set	1.0057	1.0000
highest- maxscore	1.0059	1.0000

performance on GTS '03 election data

$n = 24$  candidates,  $k = 12$  winners,  $m = 161$  ballots

## Largest approx. ratios found

	radius = 1	radius = 2
fixed-size minimax	1.0714	1.0000
3-approx.	1.0714	1.0000
random set	1.0714	1.0000
highest- maxscore	1.0714	1.0000

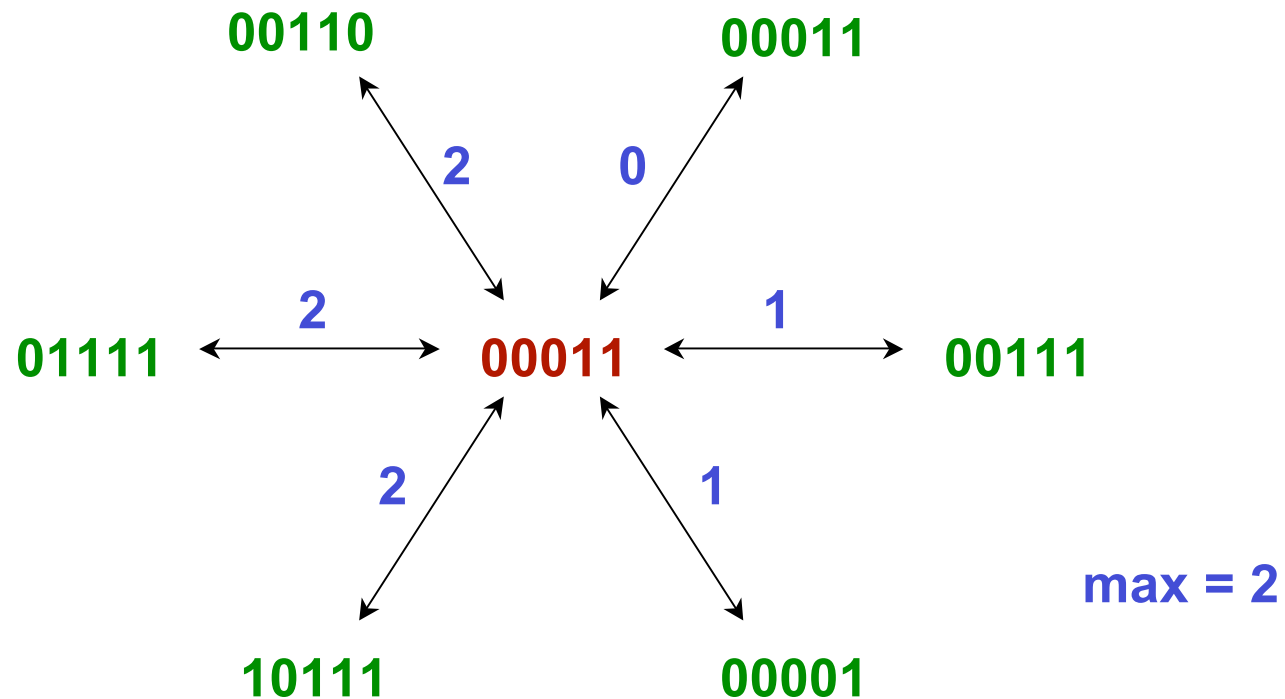
performance on GTS '03 election data

$n = 24$  candidates,  $k = 12$  winners,  $m = 161$  ballots

## Conclusions from all experiments

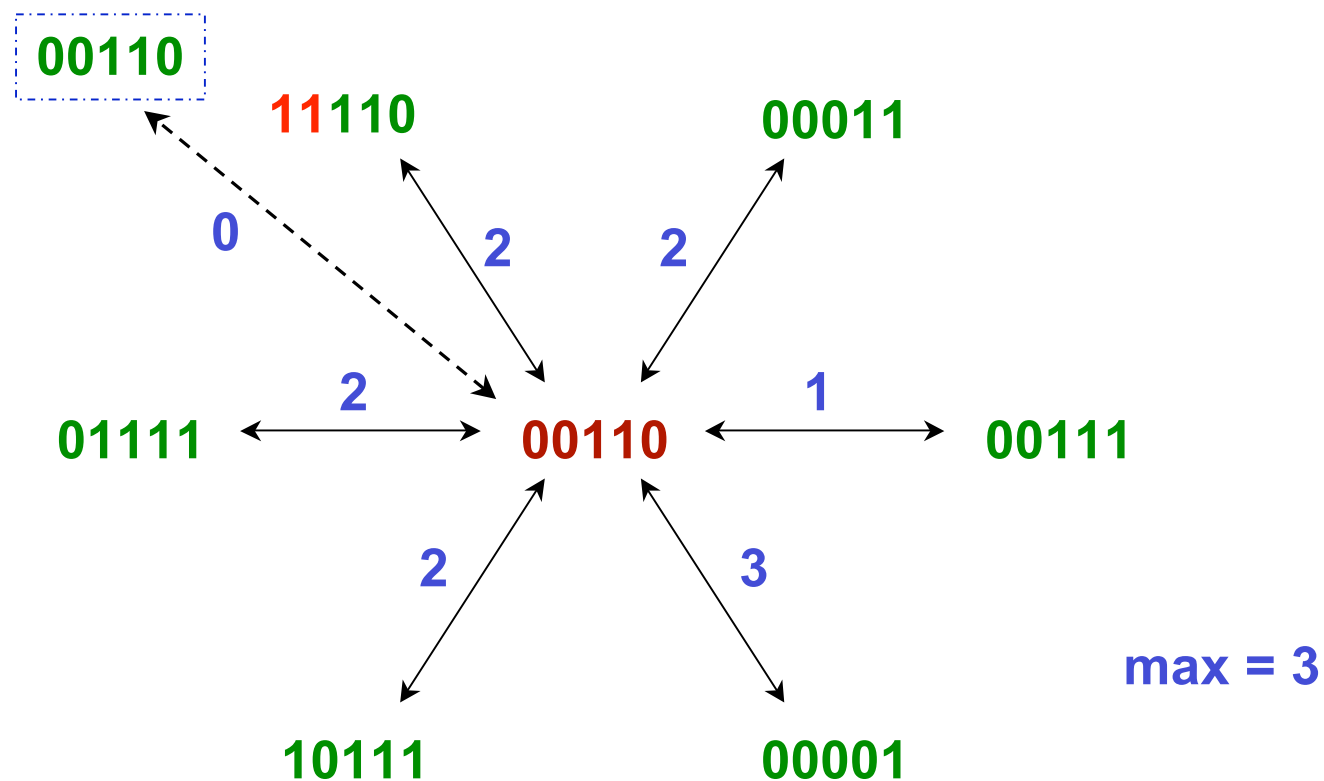
- All heuristics perform near-optimally
  - highest ratio found: 1.2
  - highest average ratio  $< 1.04$
- When radius is larger, performance improves and running time increases
- The fixed-size-minisum starting point performs best overall (with our 3-approx. a close second)

# Manipulating FSM



- Voters are sincere
- Another optimal solution: **00101**

# Manipulating FSM



- A voter manipulates and realizes ideal outcome

## Nonmanipulable “FSM”?

Electing a set found using our 3-approximation for FSM gives a nonmanipulable procedure:

- For the voters whose ballots are *not* chosen, voting insincerely cannot affect the outcome
- For the voter whose ballot *is* chosen, the outcome will be one of the sets of size  $k$  closest to the voter's wishes

## Conclusions

- BSM and FSM are NP-hard
- Both can be approximated with ratio 3
- Polynomial-time local search heuristics perform well in practice
  - some retain ratio-3 guarantee
- Exact FSM can be manipulated
- Our 3-approximation for FSM is nonmanipulable

## Future work

- Investigate weighted version of minimax [Brams, Kilgour & Sanver, '06]
- What is the best approximation ratio for FSM achievable in polynomial time? (Is there a PTAS?)
- What is the nonmanipulable FSM approximation algorithm with the best ratio?

**Thanks!**