

Iterative Compression for Solving Hard Network Problems

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Parameterized Approach to Hard Problems

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Example

VERTEX COVER can be solved in time $O(1.2852^k + k|V|)$.
 k : size of the vertex cover

Techniques to Show Fixed-Parameter Tractability

Established Techniques

- ▶ Kernelizations
- ▶ Depth-bounded search trees
- ▶ Dynamic Programming
- ▶ Tree Decompositions
- ▶ ...

Recent Approach

- ▶ Iterative compression

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Inductive approach: Compute a solution for a problem instance using the information provided by a solution for a smaller instance.

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Example for this talk

FEEDBACK VERTEX SET IN TOURNAMENTS.

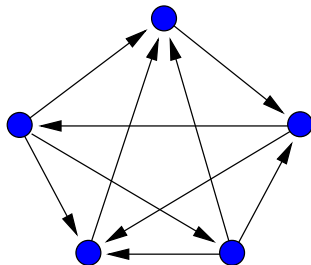
[DOM, GUO, HÜFFNER, NIEDERMEIER, AND TRUSS, CIAC 2006]

FEEDBACK VERTEX SET IN TOURNAMENTS

Definition (FEEDBACK VERTEX SET IN TOURNAMENTS)

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Output: Is there a subset $X \subseteq V$ of at most k vertices such that $G[V \setminus X]$ has no cycles?

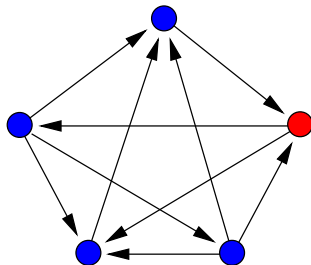


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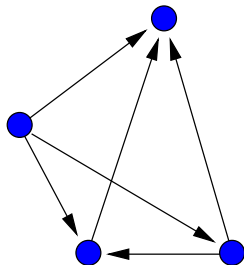


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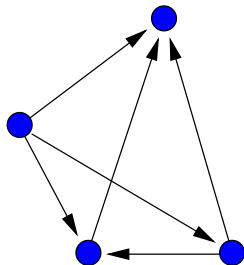


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FEEDBACK VERTEX SET IN TOURNAMENTS is NP-complete

[SPECKENMEYER, WG 1989]

Iterative Compression Framework

Idea

Use a *compression routine* iteratively: Given a solution of size $k + 1$, compute a solution of size k .

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Iterative Compression Framework for Feedback Vertex Set in Tournaments

1. Start with an empty graph G' and an empty solution X .
2. For each vertex v in G :
 - 2.1 Add v to G' and to X .
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 - 2.3 If $|X| > k$, then return “NO”.

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Invariant during the loop:

X is a solution of size at most k for the current graph G' .

Compression Routine (1)

Task

Given a solution X of size $k + 1$, compute a solution X' of size k .

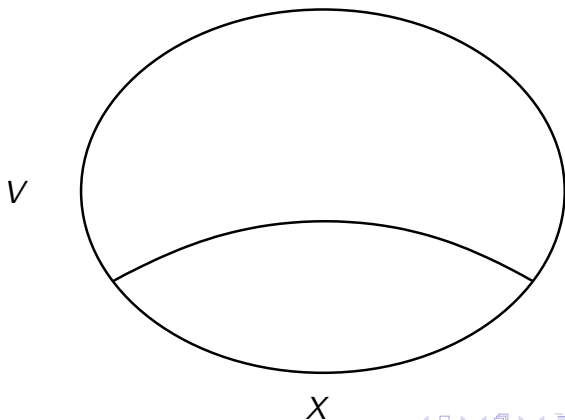
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Approach

Try all $2^{|X|}$ partitions of X into a part to exchange (S) and a part to keep ($X \setminus S$).



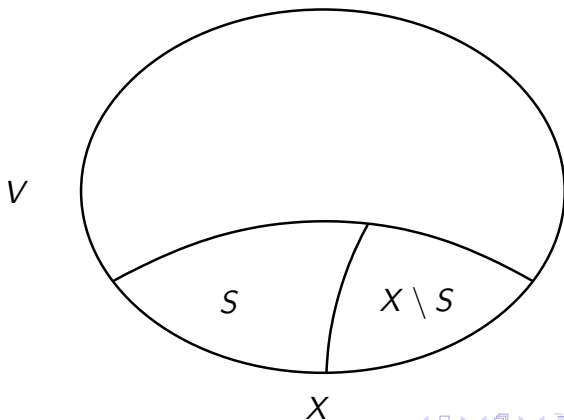
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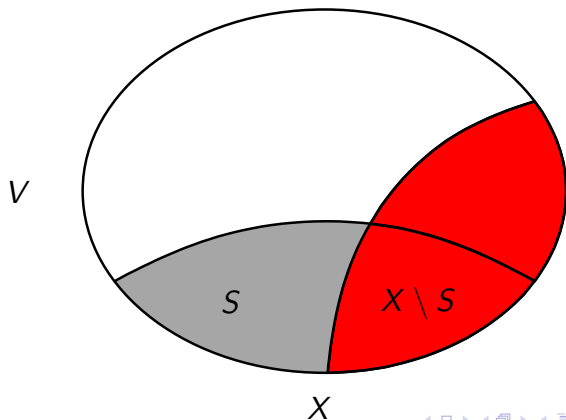
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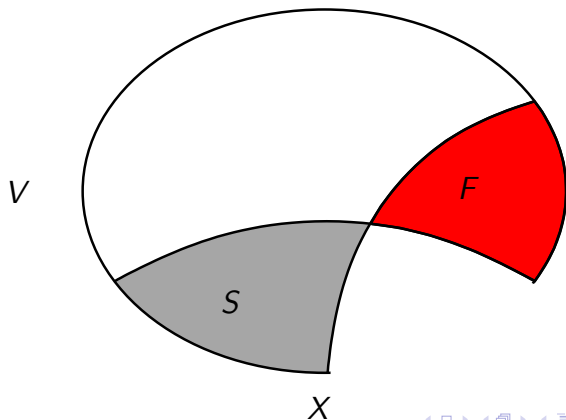
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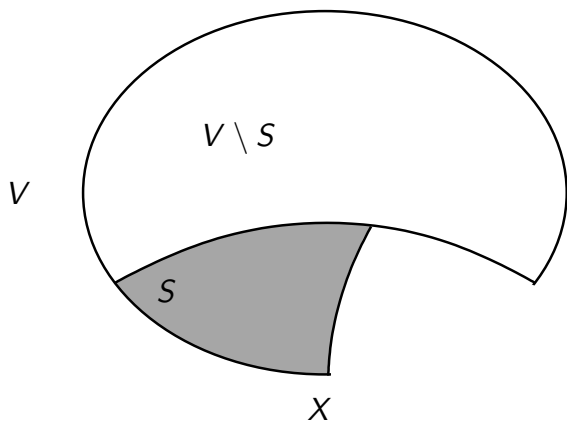
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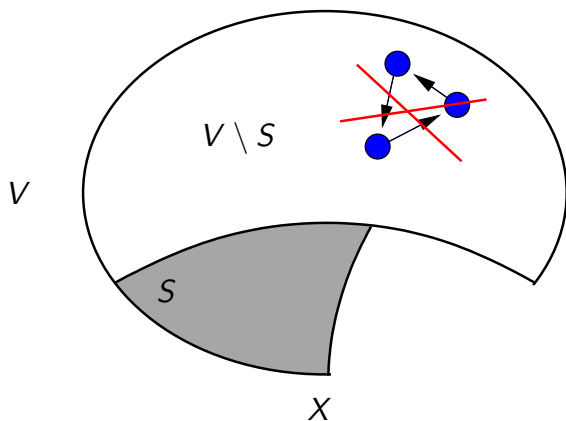
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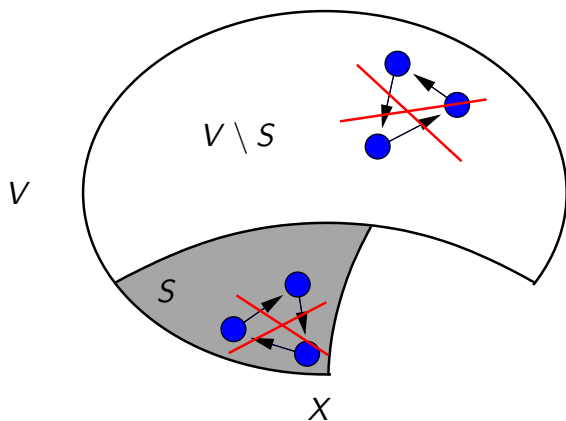
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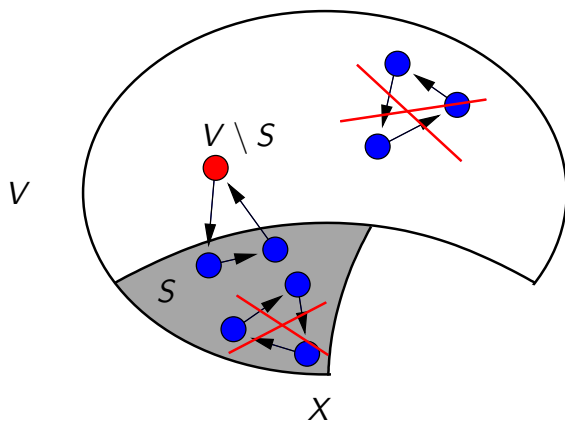
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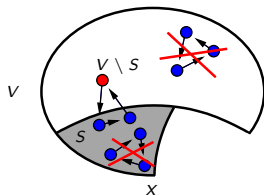
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(Lemma: A tournament contains a cycle iff it contains a triangle.)

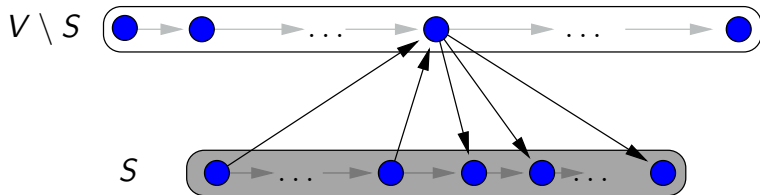
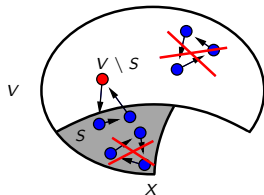
Topological Sorts

- ▶ $G[S]$ is cycle-free \Rightarrow
 S has topological sort $s_1, \dots, s_{|S|}$.
- ▶ $G[V \setminus S]$ is cycle-free \Rightarrow
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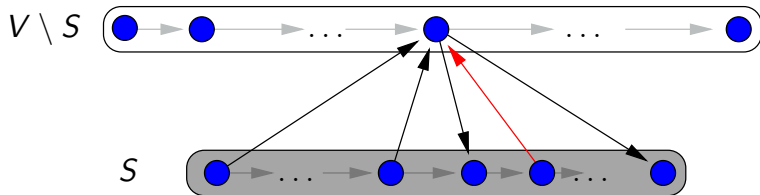
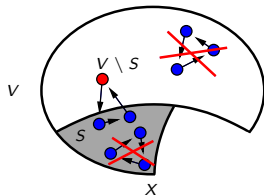


Goal: Insert a maximum subset of $V \setminus S$ into $s_1, \dots, s_{|S|}$.

Observation: Every vertex v of $V \setminus S$ has a “natural position” $p(v)$ relative to the vertices of S .

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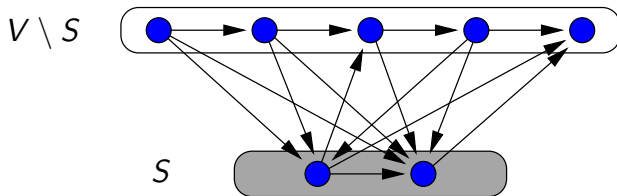
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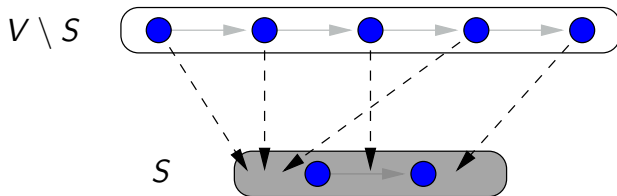
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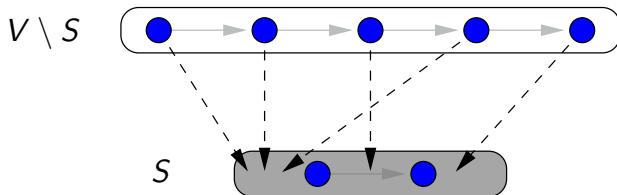
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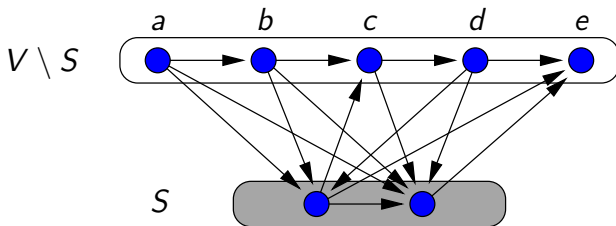
The vertices $(V \setminus S) \setminus F$ must be sortable in such a way that

- ▶ the topological sort of $V \setminus S$ is preserved, and
- ▶ the sort of $V \setminus S$ by “natural position” p is preserved.

⇒ Search for the longest common subsequence of both sorts.

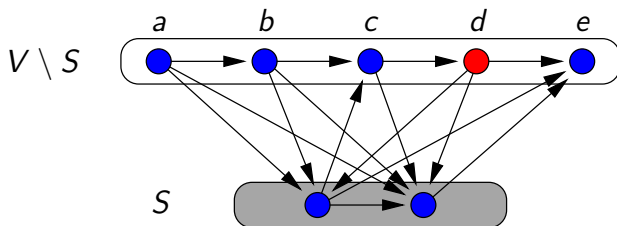
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- ▶ $V \setminus S$ sorted topologically: $abcde$
- ▶ $V \setminus S$ sorted by p : $abdce$
- ▶ A longest common subsequence is $abce$



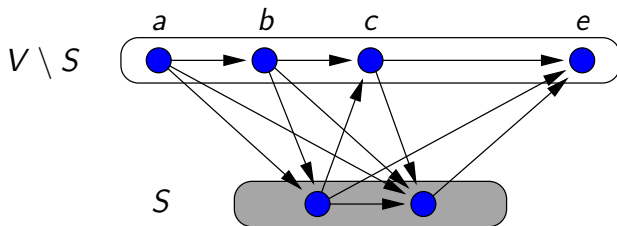
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Overall Running Time

Time consumption:

- ▶ n iterations of the compression routine;
- ▶ 2^{k+1} partitions per iteration;
- ▶ time $O(n \cdot k)$ for destroying triangles, time $O(n \log n)$ for sorting vertices and finding the longest common subsequence.

Running time for solving FEEDBACK VERTEX SET IN
TOURNAMENTS:

$$O(2^k \cdot n^2(\log n + k))$$

Applications of Iterative Compression

- ▶ GRAPH BIPARTIZATION $O(3^k kmn)$
[REED, SMITH, AND VETTA, Operations Research Letters 32, 2004]
- ▶ EDGE BIPARTIZATION $O(2^k m^2)$
[GUO, GRAMM, HÜFFNER, NIEDERMEIER, AND WERNICKE, WADS 2005]
- ▶ FEEDBACK VERTEX SET $O(c^k m)$
[DEHNE, FELLOWS, LANGSTON, ROSAMOND, AND STEVENS, COCOON 2005]
[GUO, GRAMM, HÜFFNER, NIEDERMEIER, AND WERNICKE, WADS 2005]
- ▶ FEEDBACK VERTEX SET IN TOURNAMENTS
 $O(2^k n^2(\lg n + k))$
[DOM, GUO, HÜFFNER, NIEDERMEIER, AND TRUSS, CIAC 2006]
- ▶ CHORDAL DELETION
[MARX, WG 2006]
- ▶ Implementation of GRAPH BIPARTIZATION $O(3^k mn)$
[HÜFFNER, WEA 2005]

Discussion

Advantages

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Future Work

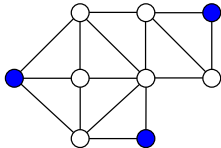
- ▶ Characterize problems amenable to solution compression.
- ▶ How can iterative compression be combined with other techniques?
- ▶ Apply to maximization problems.

A Candidate Problem for Future Research

MINIMUM VERTEX MULTIWAY CUT

Input: A graph $G = (V, E)$, a set of terminals $T \subseteq V$, and an integer $k \geq 0$.

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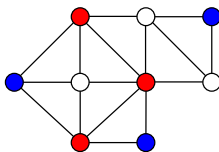


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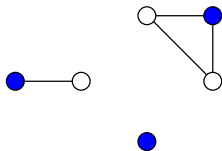


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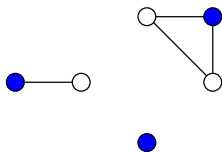


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Known Results

- ▶ NP-complete [CUNNINGHAM, DIMACS Series in DM and TCS, vol. 5, 1991]
- ▶ Fixed-parameter tractable $O(c^{k^3} \cdot \text{poly}(n))$ [MARX, IWPEC 2004]

Thank you!