

Integer Linear Programming

Input: An integer linear program with k variables.

Parameter: k

Question: Does this ILP have a solution?

This Problem is fixed parameter tractable.

The running time is $f(k)n^{O(1)}$, but the $f(k)$ are painfully large.

Proof: very involved...

Feedback Vertex Set

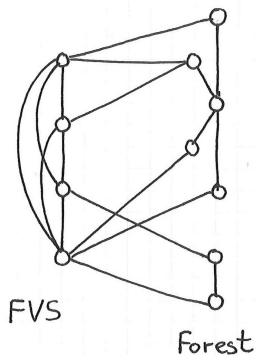
Input: A graph G and a number k

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Question: Are there $\leq k$ nodes whose removal makes G acyclic?

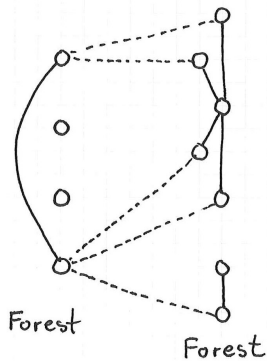
Is FVS fixed parameter tractable?

Iterative Compression



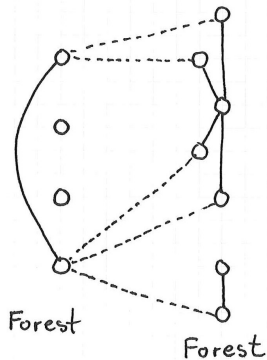
Assume we already know a FVS of size k .

Does this help to find a FVS of size $k - 1$?



Step 1: Find a subset of the FVS to keep

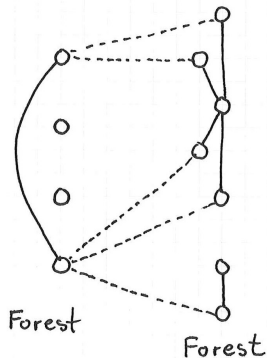
Plan: Add vertices from the forest to this FVS



Step 2: Apply reduction rules

Contract components of the FVS into one vertex

Contract degree-2 vertices in the forest



Step 3: Branching algorithm

If a node in the forest has two neighbors in the FVS:

- a) put it into the FVS
- b) contract its neighborhood

Running time

Size of the branching tree 2^k

Total size of all branching trees:

$$\sum_{j=0}^k \binom{k}{j} 2^k = 4^k$$

Total running time $4^k n^{O(1)}$

Overview

Introduction

Parameterized Algorithms

Further Techniques

Parameterized Complexity Theory

Depth-First Search Trees

Input: A graph G and a number k

Parameter: k

Question: Is there a path of length k in G ?

Construct a **depth-first search tree**.

What is the helpful property of a DFS tree?

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A Simple Theorem

Theorem

Let G be a graph and k a number.

Then it takes only polynomial time to find one of these:

- 1. A cycle of length at least k*
- 2. A tree decomposition of treewidth at most k*

Proof

$k + 1$ cops slowly traverse the DFS tree

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Long Paths

The theorem allows us to find paths of length k easily:

1. If we find a cycle longer than k , there obviously is a path of length k as well
2. Otherwise we use the tree decomposition and Courcelle's theorem:

$$\exists x_1 \dots \exists x_{k+1} (\text{inc}(x_1, x_2) \wedge \dots \wedge \text{inc}(x_k, x_{k+1}) \wedge x_1 \neq x_2 \dots)$$

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Question: Can we solve
Vertex Cover this way?

A Complicated Theorem

Theorem (Bodlaender)

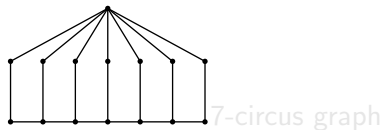
Let G be a graph and k, l some numbers.

It takes $f(k, l)|G|$ steps to find one of these:

1. A subdivision of the $2 \times k$ grid
2. A subdivision of the l -circus graph
3. A tree decomposition of G of treewidth $2(k - 1)^2(l - 1) + 1$.

Proof

Again, using a DFS tree.



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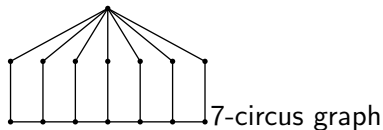
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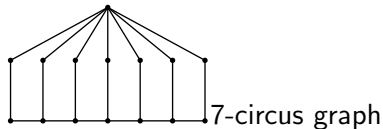
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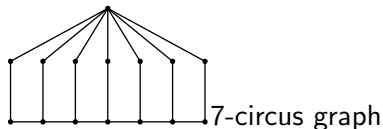
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Proof

Again, using Why did it work for planar graphs?



Application 1

Max-Leaf-Spanning-Tree:

Input: A graph G and a number k

Parameter: k

Question: Does G have a spanning tree with at least k leaves?

Both the $2 \times k$ grid and the k -circus graph contain a tree with k leaves.

That is, Max-Leaf-Spanning-Tree is fixed parameter tractable.

Application 1

Max-Leaf-Spanning-Tree:

Input: A graph G and a number k

Parameter: k
Does the following statement hold?

Question: If a graph contains a tree with k leaves, does it also contain a spanning tree with at least k leaves?

Both the $2 \times k$ grid and the k -circus graph contain a tree with k leaves.

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Application 2

Feedback Vertex Set:

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Proof

Apply Bodlaender's theorem.

1. Small tree decomposition: Courcelle
2. $2 \times 3k$ grid: No
3. $4k$ -circus graph: Remove the tip and check for a FVS of size $k - 1$.

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Overview

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Parameterized Algorithms

Further Techniques

Parameterized Complexity Theory

Parameterized Complexity Theory

Classical complexity theory:

- ▶ Complexity classes P , NP , etc.
- ▶ Languages $L \in P$, $L \subseteq \Sigma^*$
- ▶ Framework insufficient for parameterized problems

Parameterized Complexity Theory

Definition

A **parameterized problem** over the alphabet Σ is a set of pairs (w, k) , where $w \in \Sigma^*$ and $k \in \mathbf{N}$.

It is not allowed that there exists w and $k \neq k'$ with $(w, k) \in L$ and $(w, k') \in L$, if L is a parameterized problem.

The second condition states that k is a function of w .

Parameterized Complexity Theory

We like to state parameterized problems as follows:

Input: A graph G and a number k

Parameter: k

Question: Does G contain a clique of size k as a subgraph?

Parameterized Complexity Theory

The parameter can be some arbitrary number, if it can be **easily computed** from the input.

Input: A graph G and a number k

Parameter: The diameter of G

Question: Does G contain a clique of size k as a subgraph?

Here it is easy to compute $(G, \Delta(G))$ from G in order to get formally a parameterized problem.

Parameterized Complexity Theory

One goal of complexity theory is to categorize problems into **easy** and **hard** ones.

For this purpose P and NP are best known.

Others are:

- ▶ NC and L
- ▶ AC^0 and NC^1
- ▶ $EXPTIME$ and $EXPSPACE$
- ▶ etc. etc.

Parameterized Complexity Theory

In parameterized complexity theory the easy problems can be found in the class *FPT*.

Definition

The class *FPT* contains all parameterized problems that are fixed parameter tractable.

Formally: $L \in \text{FPT}$, if there is an algorithm solving $(w, k) \in L$ in at most $f(k)|w|^c$ steps, where c is a constant and $f: \mathbf{N} \rightarrow \mathbf{N}$ an arbitrary function.

Parameterized Complexity Theory

A fundamental concept in complexity theory are **reductions**.

Important example: **polynomial time many-one reductions**:

$g: \Sigma^* \rightarrow \Sigma^*$ reduces the problem L_1 to L_2 , if

1. $w \in L_1 \iff g(w) \in L_2$.
2. $g(w)$ can be computed in $|w|^{O(1)}$ steps.

Important property: If L_1 can be reduced to L_2 and $L_1 \notin P$, then $L_2 \notin P$.



“I can’t find an efficient algorithm, but neither can all these famous people.”

Important property: If L_1 can be reduced to L_2 and $L_1 \notin P$, then $L_2 \notin P$.

Parameterized Complexity Theory

Question: Is this reduction useful for parameterized problems?

1. $w \in L_1 \iff g(w) \in L_2$.
2. $g(w)$ can be computed in $|w|^{O(1)}$ steps.

Does the corresponding property hold: If L_1 can be reduced to L_2 and $L_1 \notin FPT$, then $L_2 \notin FPT$.

Parameterized Complexity Theory

That corresponding property does not hold:

We can map (w, k) to $(w, |w|)$!

If we reduce a problem **to itself** like this, we have $f(|w|)|w|^c$ steps instead of $f(|w|)|w|^c$ steps to compute a solution.

It that way we can solve **every computable** problem.

A polynomial time reduction is **not fine grained enough**.

Parameterized Reductions

Definition

A parameterized problem $L_1 \subseteq \Sigma^*$ can be reduced to $L_2 \subseteq \Gamma^*$ by a **parameterized reduction** if

- ▶ $r, s: \mathbf{N} \rightarrow \mathbf{N}$ are computable functions,
- ▶ there is a function $g: \Sigma^* \times \mathbf{N} \rightarrow \Gamma^*$, $(w, k) \mapsto (w', k')$, that can be computed in $r(k)|w|^{O(1)}$ steps and $k' = s(k)$,
- ▶ $(w, k) \in L_1$ if and only if $g(w, k) \in L_2$.

Parameterized Reductions

Theorem

If $L_1 \notin FPT$ and there is a parameterized reduction from L_1 to L_2 , then $L_2 \notin FPT$.

Proof

Assume $L_2 \in FPT$. We can compute $(w', k') = g(w, k)$ in $r(k)|w|^c$ steps such that $k' = s(k)$ and $|w'| \leq r(k)|w|^c$.

Then test whether $(w', k') \in L_2$ taking $f'(k')|w'|^d \leq f'(s(k))r(k)^d|w|^{cd}$ steps.

Because $(w, k) \in L_1 \iff (w', k') \in L_2$, we answered whether $(w, k) \in L_1$ holds and therefore $L_1 \in FPT$.