

# A Structural View on Parameterizing Problems: Distance from Triviality

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# Parameterization for hard problems

For exact algorithms for NP-hard problems, we probably have to accept exponential runtimes.

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**Approach:** Try to confine the combinatorial explosion to some parameter  $k$ .

## Definition

For some *parameter*  $k$  of a problem, the problem is called *fixed-parameter tractable* with respect to  $k$  if there is an algorithm that solves it in  $f(k) \cdot n^{O(1)}$ .

# Finding Parameters

Usually, many parameters are sensible.

## Example

VERTEX COVER: Given a graph  $G = (V, E)$  and an integer  $k$ , is there  $V' \subseteq V$  with  $|V'| \leq k$  such that each edge has at least one endpoint in  $V'$ ?

- ▶ Parameterization by solution size:  
If the vertex cover has size  $k$ :  
 $O(1.3^k + kn)$  time algorithm

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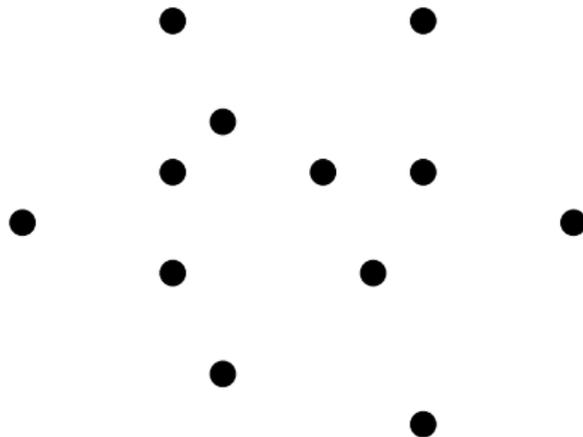
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- ▶ Parameterization by solution size:  
If the vertex cover has size  $k$ :  
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- ▶ Parameterization by structure:  
If treewidth is bounded by  $w$ :  
 $O(2^w \cdot n)$  time algorithm

# 2D-TRAVELING SALESMAN PROBLEM

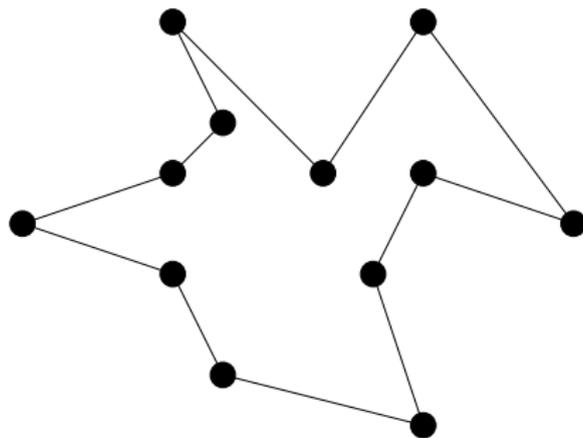
**Given:**  $n$  points from  $\mathbf{R}^2$



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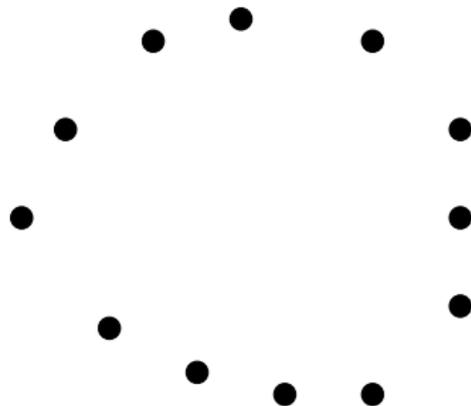
**Given:**  $n$  points from  $\mathbf{R}^2$

**Task:** Find a minimal length tour through all points



# Simple cases of the 2D-TRAVELING SALESMAN PROBLEM

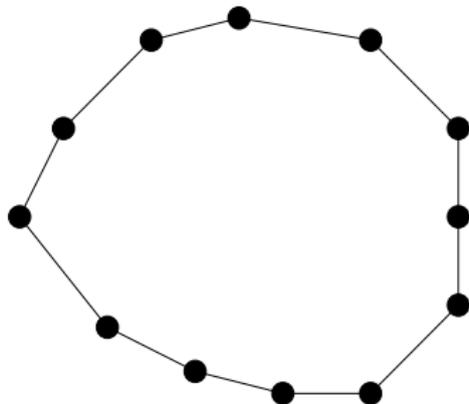
Trivial case: all vertices on the border of a convex region



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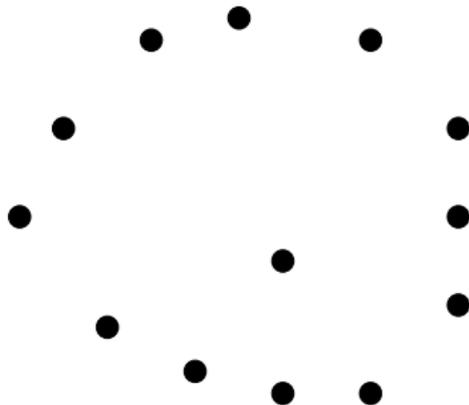
Trivial case: all vertices on the border of a convex region

- ▶ Walk all vertices in clockwise order



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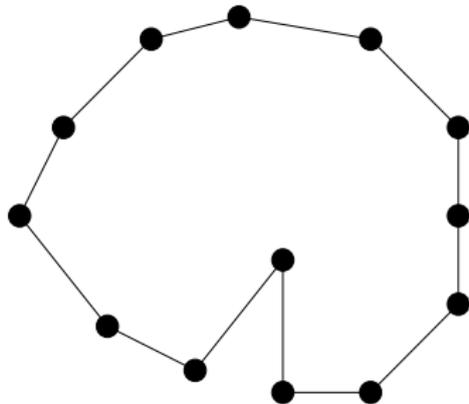
*Nearly* trivial case: one vertex inside the border of a convex region



# Simple cases of the 2D-TRAVELING SALESMAN PROBLEM

*Nearly* trivial case: one vertex inside the border of a convex region

- ▶ Few possibilities; polynomial time



# Parameterized 2D-TRAVELING SALESMAN PROBLEM

## Generalized question:

How fast can we solve 2D-TRAVELING SALESMAN PROBLEM for an instance with  $k$  points inside of the convex hull?

[DEJNEKO, HOFFMANN, OKAMOTO&WOEGINGER, COCOON'04]

## Theorem

2D-TRAVELING SALESMAN PROBLEM *with  $k$  inner points can be solved in  $O(2^k \cdot k^2 \cdot n)$  time.*

# Negative Results for Distance from Triviality Parameterization

GRAPH COLORING [LEIZHEN CAI, DISCRETE APPL. MATH. 2003]

Is there a vertex coloring of a graph with  $c$  colors such that no edge joins vertices of equal colors?

- ▶ NP-complete in general, but polynomial time solvable on split graphs and bipartite graphs

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- ▶ split graph by adding  $k$  vertices? — W[1]-hard
- ▶ bipartite graph by adding  $k$  edges? — NP-c for  $k \geq 3$

# Scheme for Parameterization by Distance from Triviality

Assume that we study a hard problem.

1. Determine efficiently solvable special cases  
(e.g., the restriction to special graph classes)  
—the triviality.
2. Identify useful distance measures from the triviality  
(e.g., the treewidth of a graph)  
—the (structural) parameter.

# Case Studies

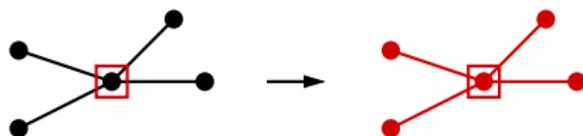
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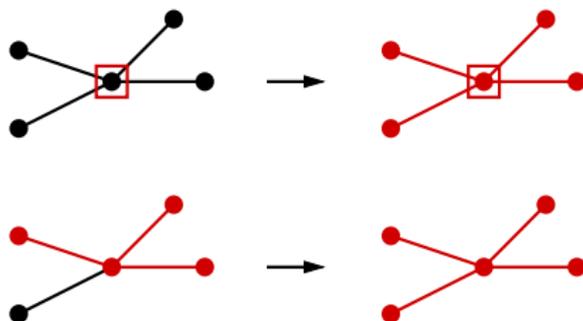
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# POWER DOMINATING SET

- ▶ POWER DOMINATING SET is NP-complete.

[HAYNES, HEDETNIEMI, HEDETNIEMI&HENNING SIAM J. DISCRETE MATH.  
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# POWER DOMINATING SET

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- ▶ **POWER DOMINATING SET** is APX-hard and  $W[1]$ -hard with respect to the number of monitoring devices.  
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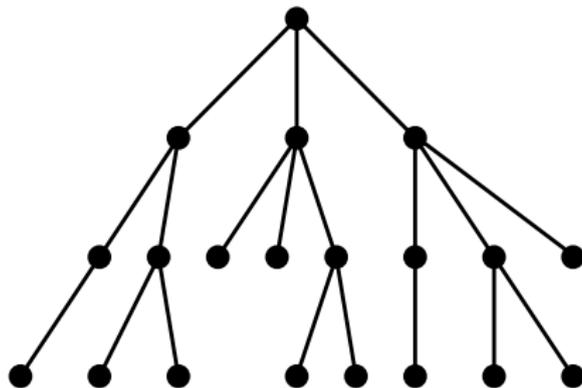
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[KNEIS, MÖLLE, RICHTER&ROSSMANITH 2004]  
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- ▶ There is a linear time algorithm solving POWER DOMINATING SET on trees.

**Triviality:** Trees.

# POWER DOMINATING SET on Trees

Idea for the linear time algorithm:

- ▶ Work layer-wise bottom-up from the leaves.
- ▶ Place a monitoring device in vertices with at least two unobserved children.

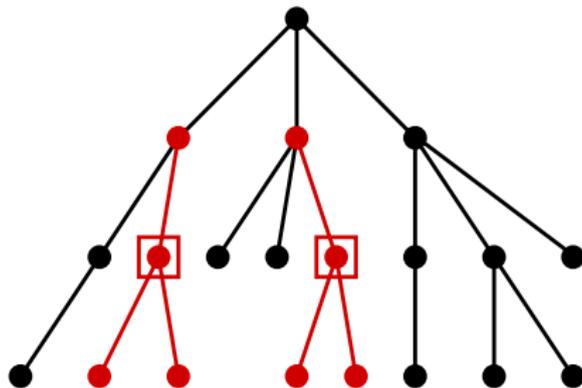




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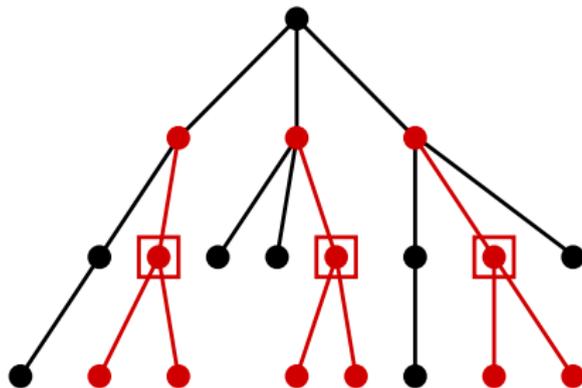
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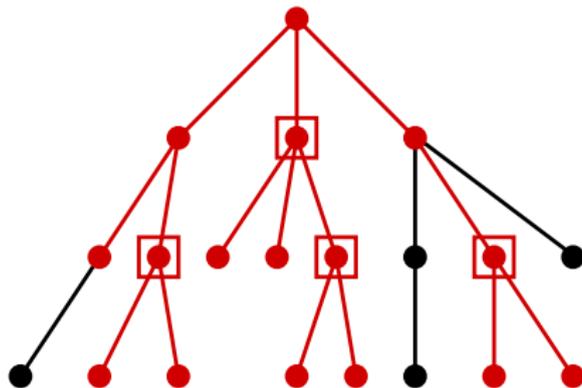




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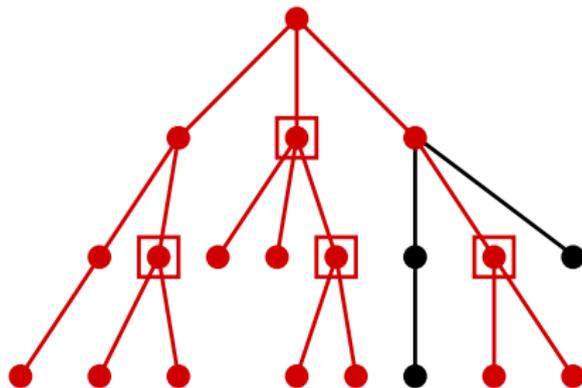
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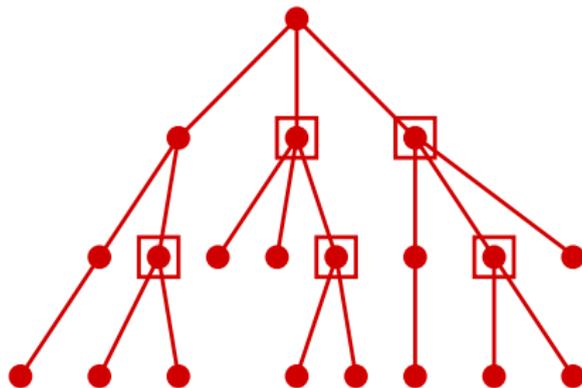
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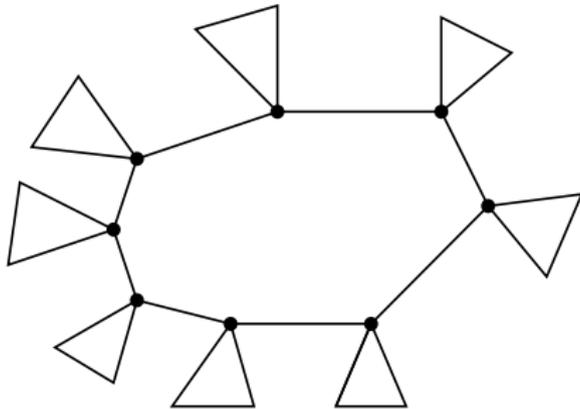
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**Distance from Triviality:** Number of edges added.

First we consider a single added edge.

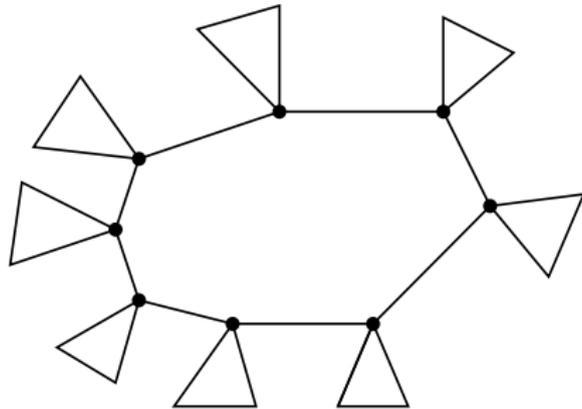


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- ▶ Treat trees with linear time algorithm.
- ▶ We can prune observed edges and singleton vertices.

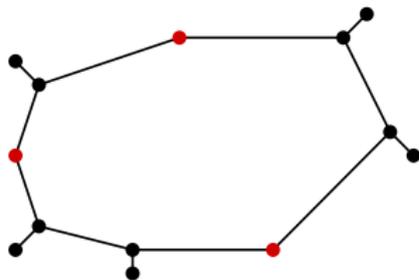


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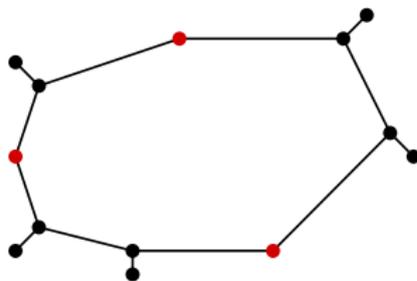


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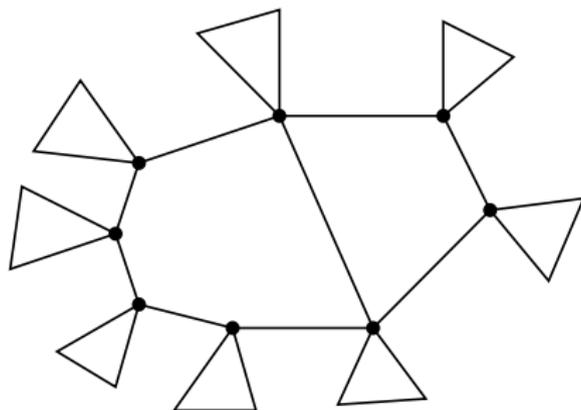
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- ▶ Treat trees with linear time algorithm.
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- ▶ Branch on first vertex for placing a monitoring device, solve the rest in linear time.



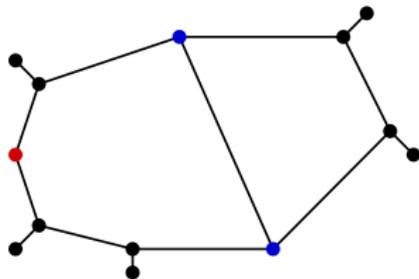
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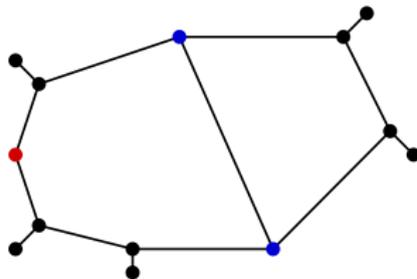


# POWER DOMINATING SET on Almost Trees

- ▶ POWER DOMINATING SET on a tree with  $k$  edges added
- ▶ After treating trees, we additionally have **joints**.

Branch for each joint  $x$ :

- ▶  $x$  gets a monitoring device
- ▶  $x$  does not get a monitoring device
  - ▶ Branch further according to the local effect of  $x$



# POWER DOMINATING SET on Almost Trees

*Observation:* The number of joints is bounded by  $2k$ .  
Therefore, the number of branches depends only on  $k$ , not on  $n$ :

## Theorem

*POWER DOMINATING SET for a graph which originates from a tree with  $k$  edges added is fixed-parameter tractable with respect to  $k$ .*

# CLIQUE

**Input:** *A graph  $G$  and a nonnegative integer  $s$ .*

**Question:** *Does  $G$  contain a clique, that is, a complete subgraph, of size  $s$ ?*

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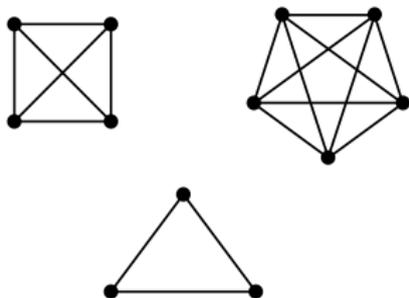
**Question:** *Does  $G$  contain a clique, that is, a complete subgraph, of size  $s$ ?*

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- ▶  $W[1]$ -hard with respect to  $s$

# CLIQUE on Cluster Graphs: Trivial Case

## Definition

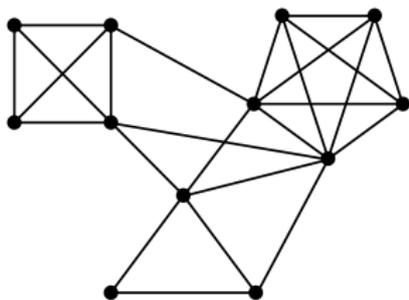
A *cluster graph* is a graph where every connected component is a clique.



**Triviality:** Cluster graphs.

# CLIQUE on Nearly Cluster Graphs

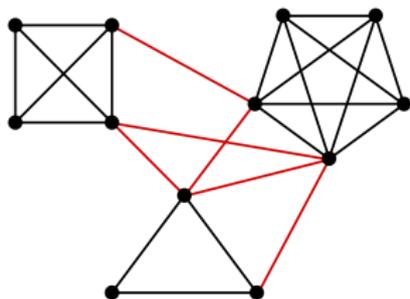
**Distance from Triviality:**  $k$  edges added.



Solving CLIQUE:

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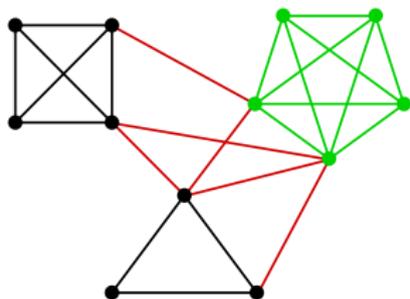


Solving CLIQUE:

- ▶ Find the  $k$  added edges:  $O(1.53^k + n^3)$  time [GRAMM et al., Algorithmica 2004].

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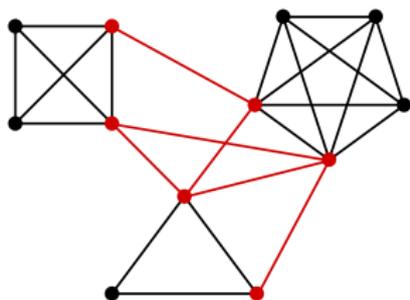


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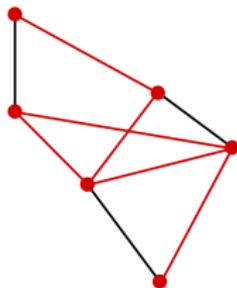


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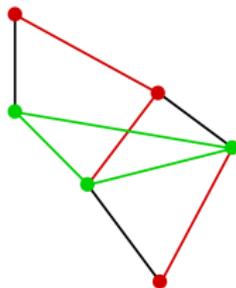


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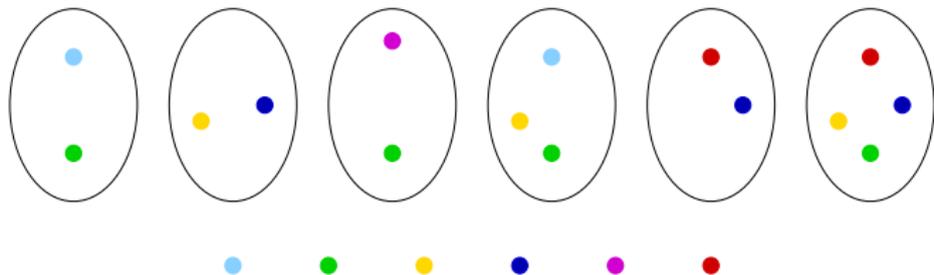
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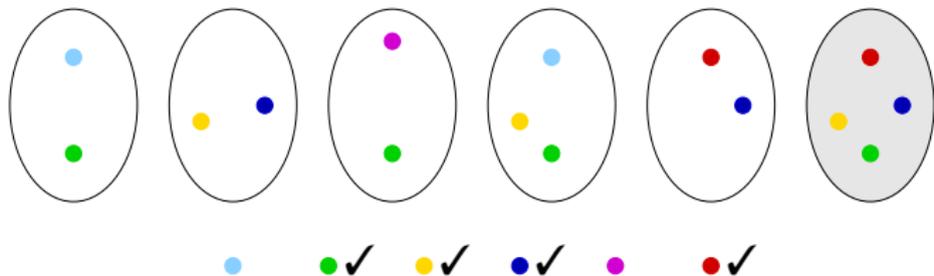
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CLIQUE for a cluster graph with  $k$  edges added can be solved in  $O(1.53^k + n^3)$  time.

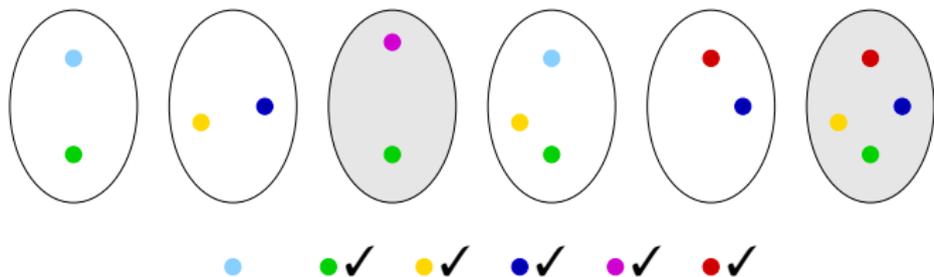
# SET COVER



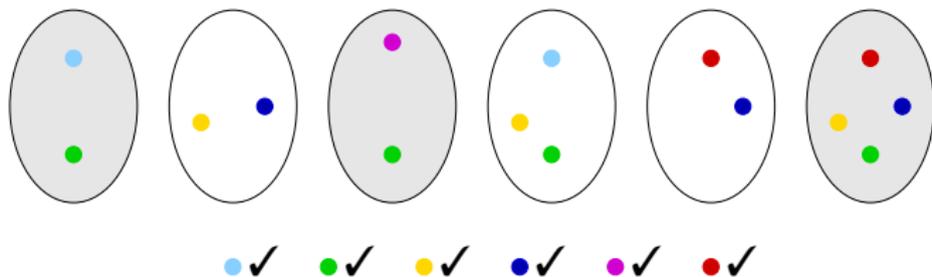
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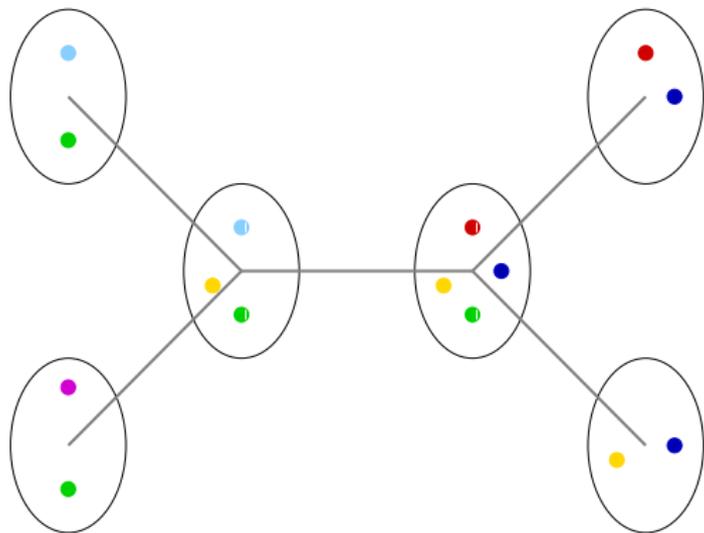
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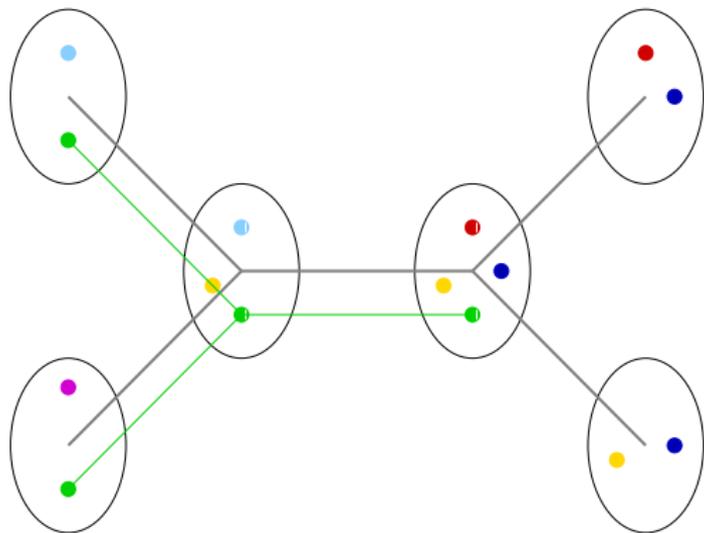
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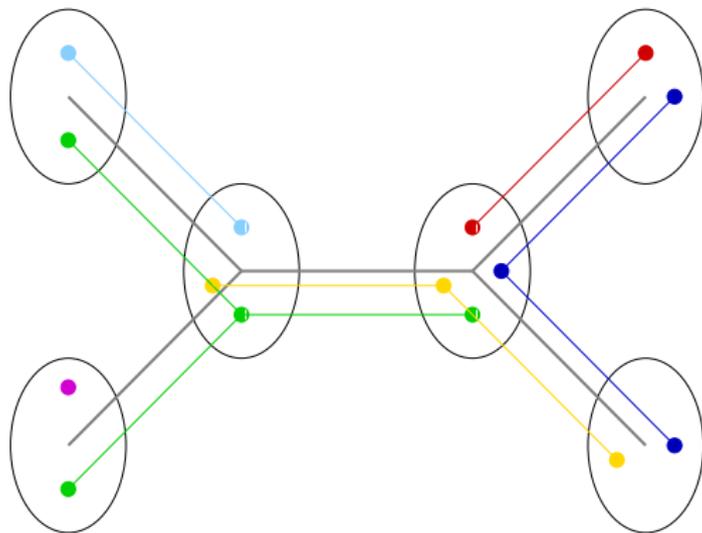
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# Parameterizing TREE-LIKE WEIGHTED SET COVER

[GUO&NIEDERMEIER, Manuscript, June 2004]

- ▶ TREE-LIKE WEIGHTED SET COVER is NP-complete, even with bounded number of occurrences per element.
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**Triviality:** Subset trees that are paths.

**Distance from Triviality:** Number of leaves of the subset tree.

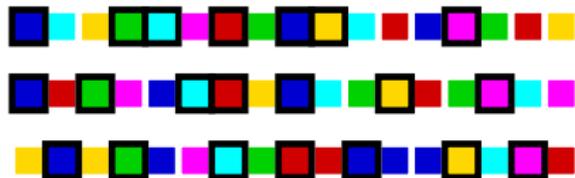
## Theorem

TREE-LIKE WEIGHTED SET COVER *with occurrence bounded by  $d$  can be solved in  $O(2^{dk^2} \cdot m^2 n)$  time, where  $k$  denotes the number of the leaves of the subset tree.*

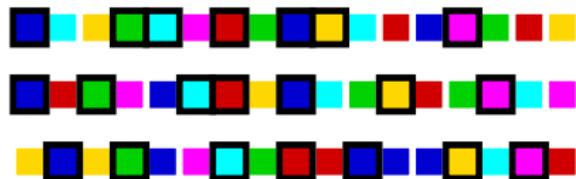
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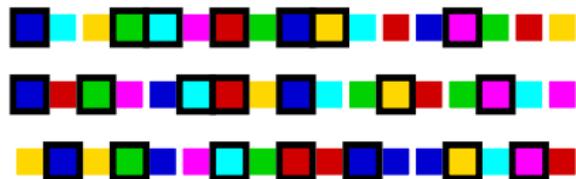


# LONGEST COMMON SUBSEQUENCE



- ▶ LONGEST COMMON SUBSEQUENCE is NP-complete and  $W[1]$ -hard for parameter “number of strings”

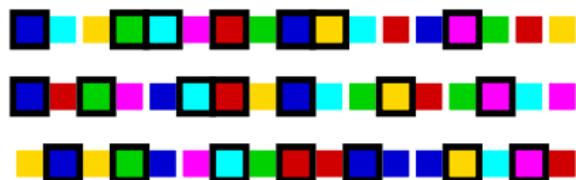
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**Triviality:** Strings are permutations.

**Distance from Triviality:** Maximum occurrence number.

## Theorem

LONGEST COMMON SUBSEQUENCE of  $k$  strings can be solved in  $O(2^{2k \log s} \cdot k \cdot n^2)$  time, where  $s$  denotes the maximum occurrence number of a letter in an input string.

# Summary

Distance from triviality—a natural way of parameterizing a hard problem  $X$ :

1. Determine efficiently solvable special cases of  $X$ —the triviality.
2. Identify useful distance measures from the triviality—the (structural) parameter.
  - ▶ Mostly structural results: How can we extend the range of tractability?
  - ▶ Might also lead to efficient practical implementations if the parameter is small.