

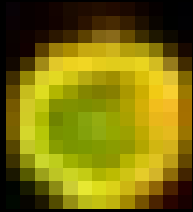
Practical Fixed-Parameter Algorithms for Graph-Modeled Data Clustering

Sebastian Wernicke*

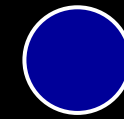
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Ernst-Abbe-Platz 2, D-07743 Jena, Fed. Rep. of Germany
wernicke@minet.uni-jena.de

Prologue

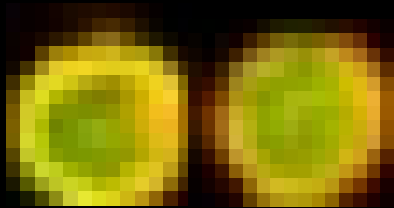
Graph-Modeled Clustering



Data Point



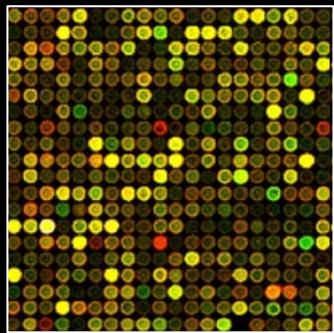
Vertex / Node



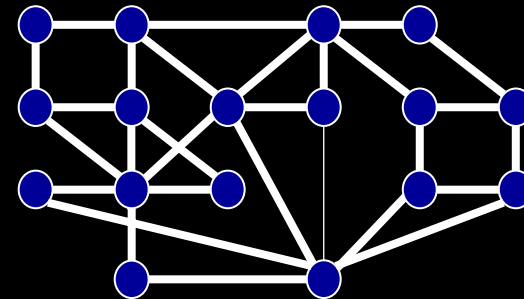
Correlation



Edge / Link

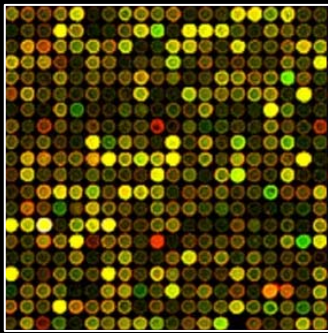


Data

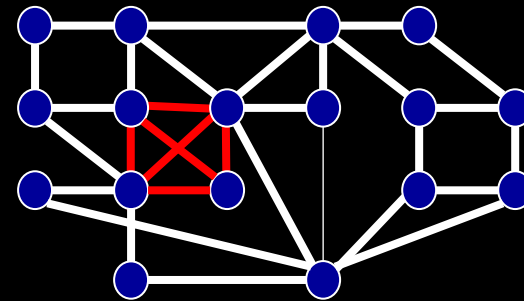


Graph / Network

Graph-Based Clustering



Cluster(s)
in Data



Dense / Complete
Subgraph(s)

Potential
Tasks to
Face

Find a large complete subgraph (clique).
Partition the graph into disjoint cliques.
Cover the graph with cliques.

Toolbox for Hard Problems

Brute-Force

Sidestepping

Heuristics

Approximation
Algorithms

(Integer) Linear
Programming



Toolbox for Hard Problems

Brute-Force

Sidestepping

Heuristics



Approximation
Algorithms

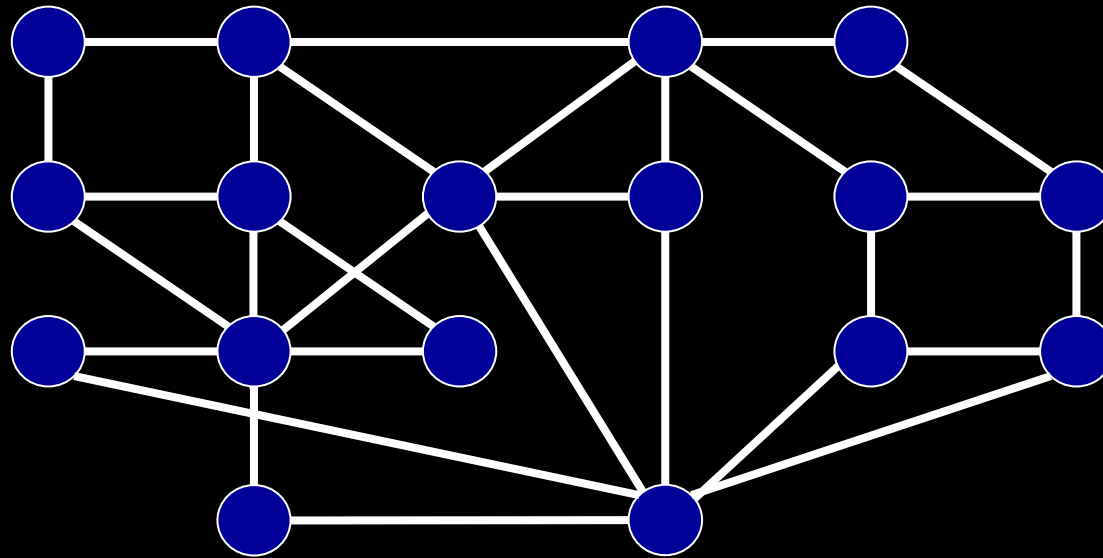
(Integer) Linear
Programming

Fixed-
Parameter
Algorithms

- data reduction
- search trees
- and more!

Part I
Fixed-Parameter Tractability Primer

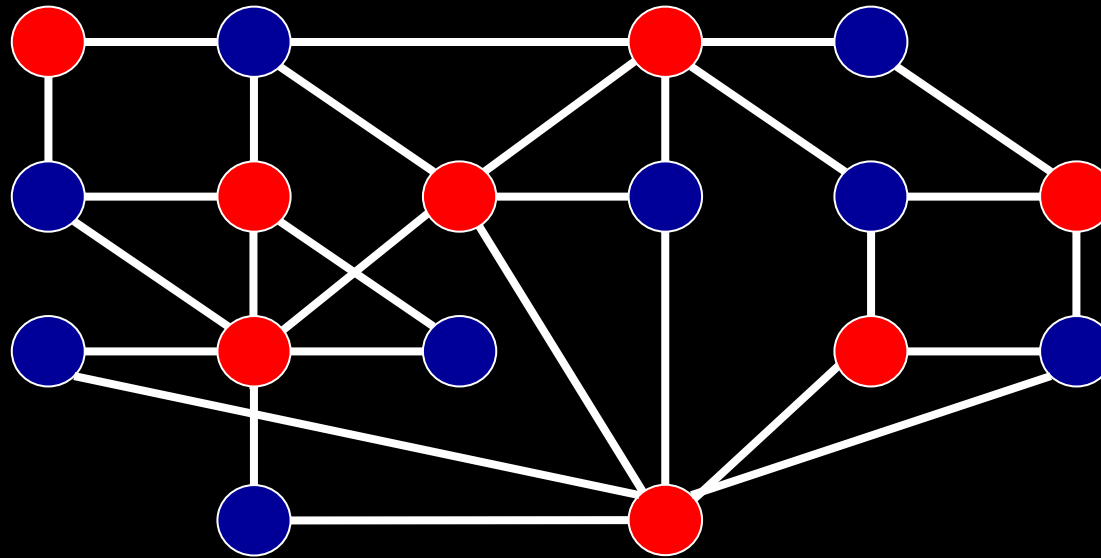
Drosophila of Fixed-Parameter Algorithmics: Vertex Cover



Given a graph G , find a set C of at most k vertices such that each edge has at least one endpoint in C .



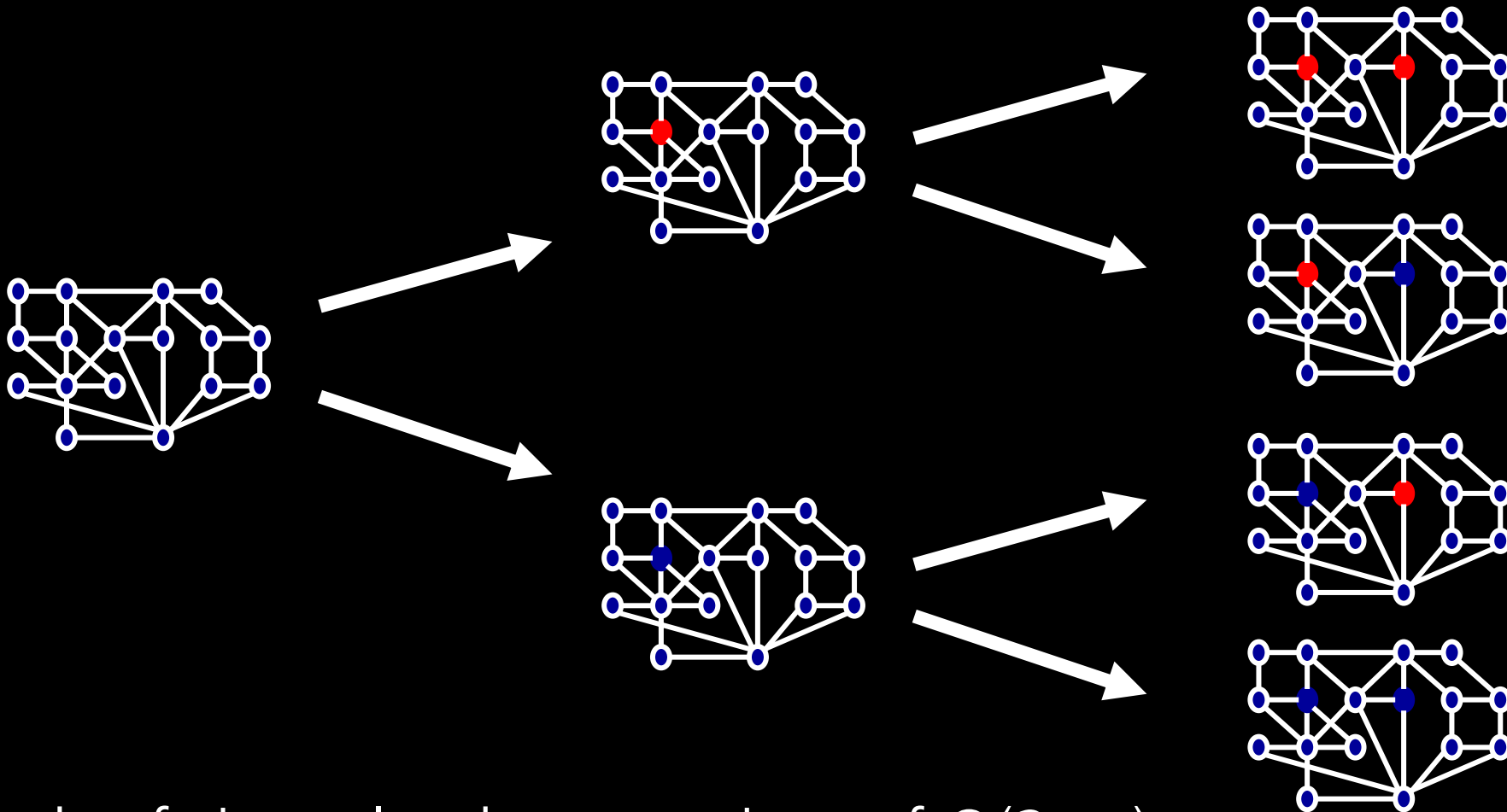
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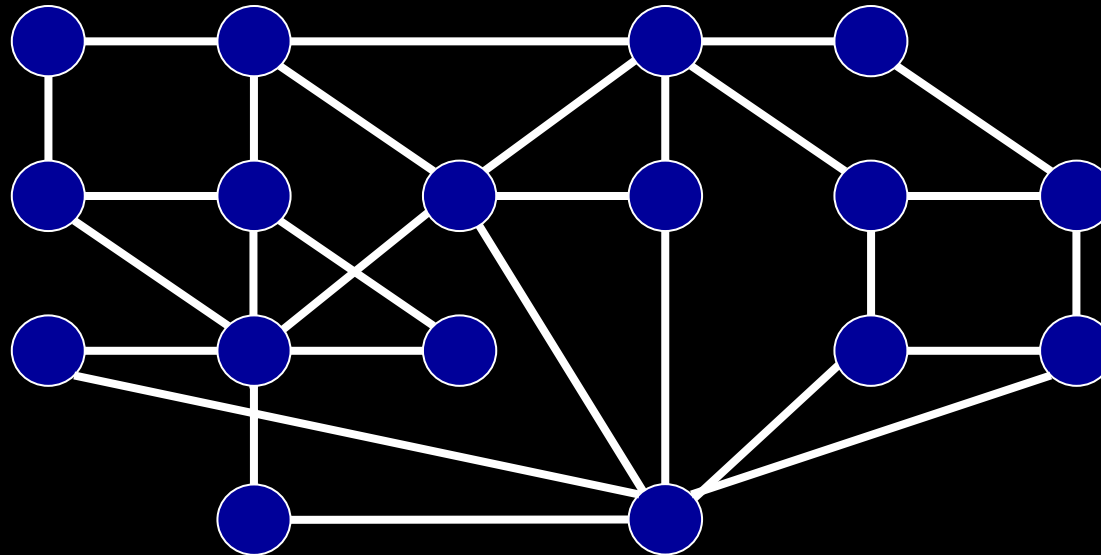


Vertex Cover – Naïve Brute-Force Approach



Graph of size n leads to runtime of $O(2^n m)$.

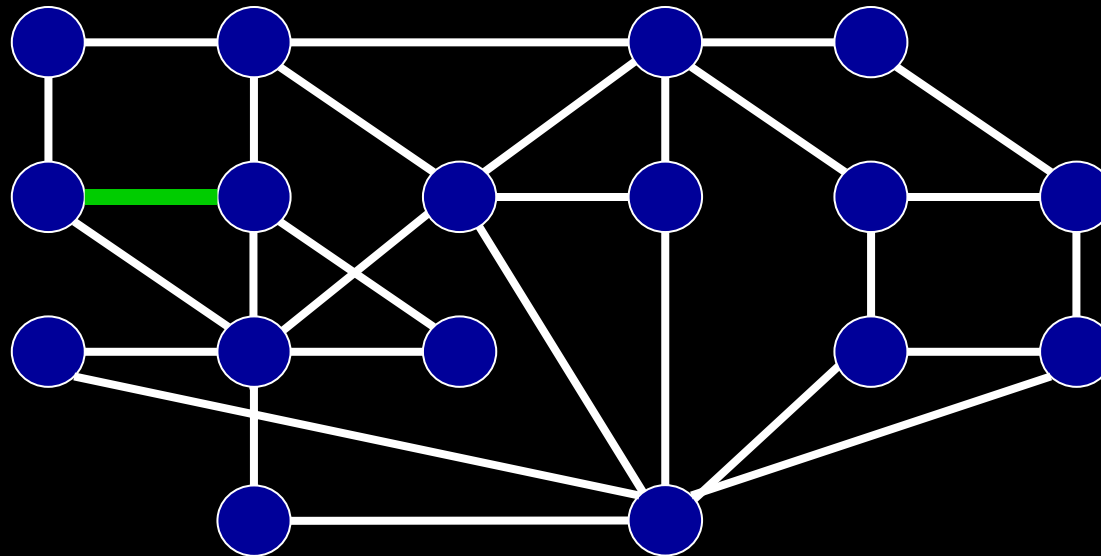
Vertex Cover – Better Approach



Given graph G , find a set C of at most k vertices such that each edge has at least one endpoint in C .

Main idea: For each edge, one endpoint *must* be in C .

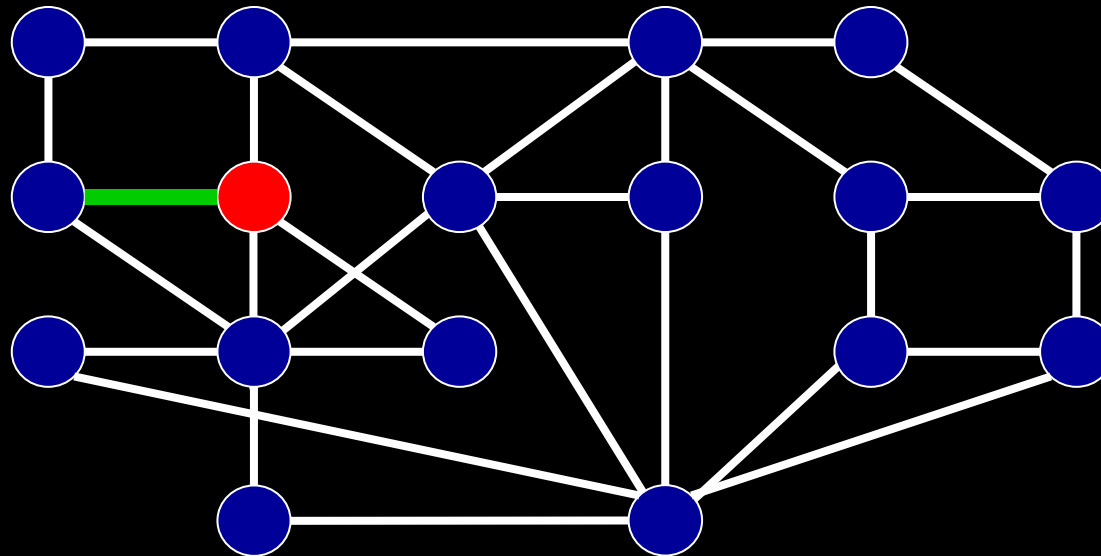
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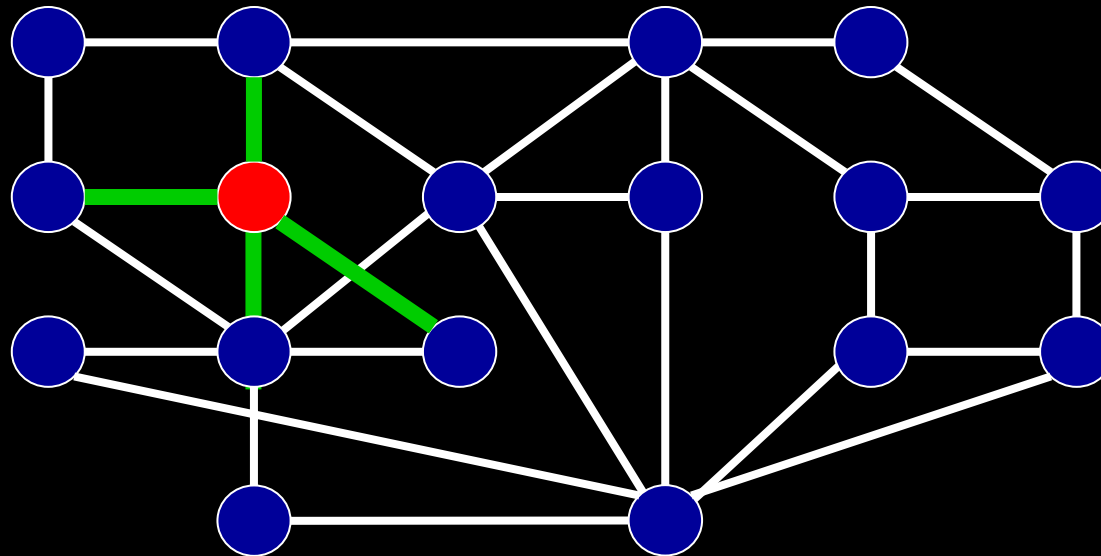
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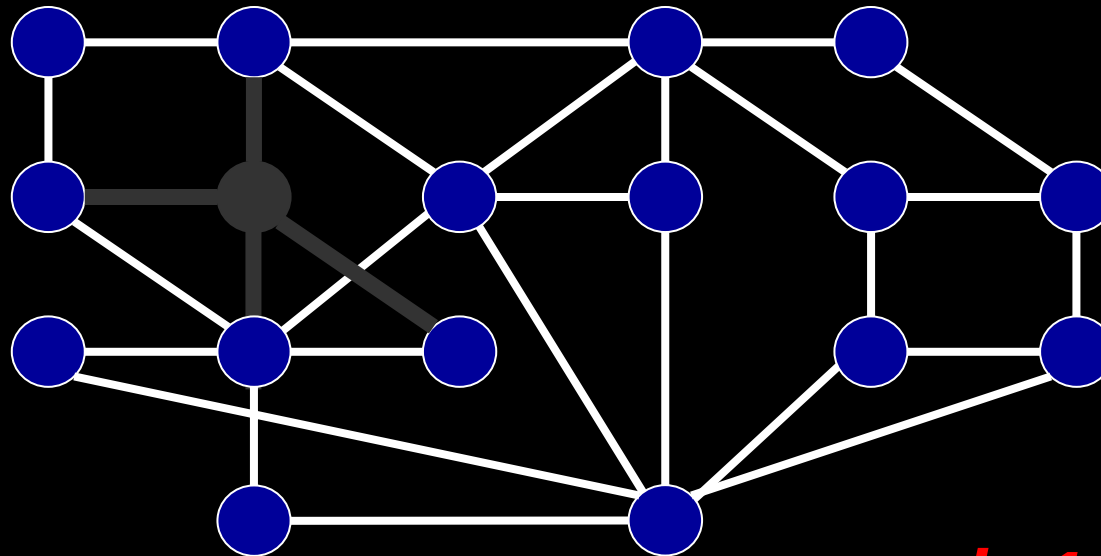
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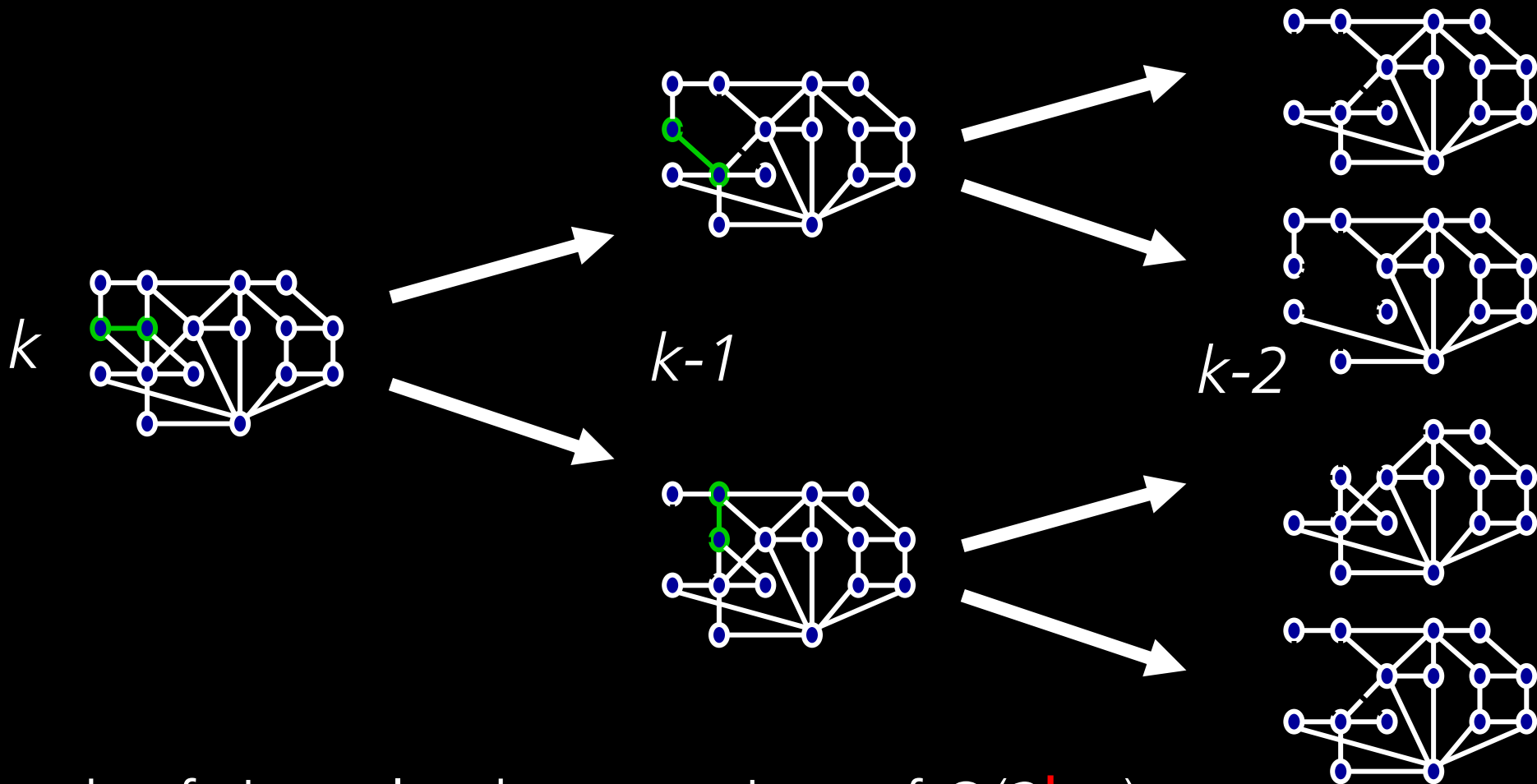


~~k~~
 $k-1$

Given graph G , find a set C of at most ~~k~~ vertices such that each edge has at least one endpoint in C .

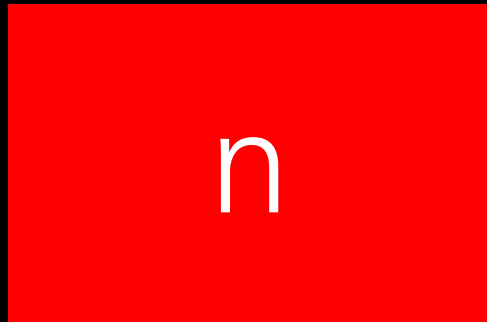
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Vertex Cover – Fixed-Parameter Approach

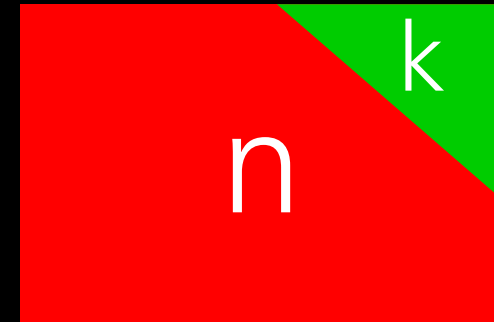


Graph of size n leads to runtime of $O(2^k m)$.

Fixed-Parameter Tractability



Classical complexity theory:
One dimensional. Hard
problems take *exponential
time in **instance size*** to
solve.

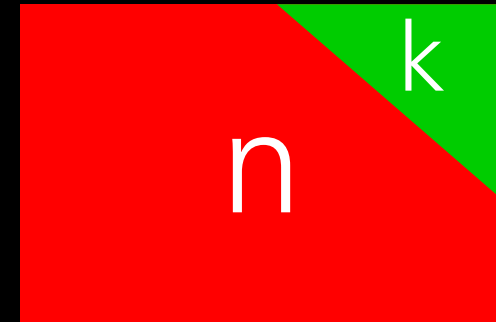


The fixed-parameter
approach: Two-
dimensional. Hard problem
takes *exponential time in
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Fixed-Parameter Tractability



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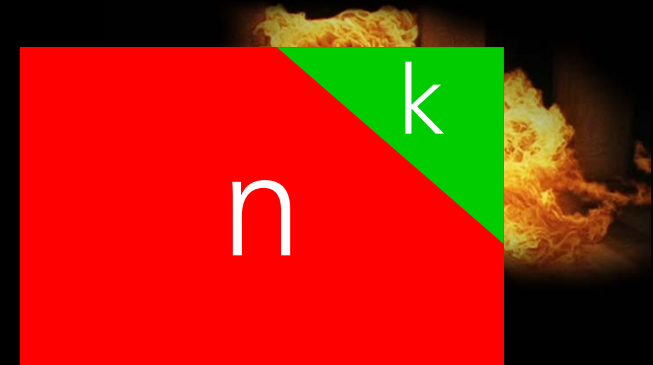


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Fixed-Parameter Tractability

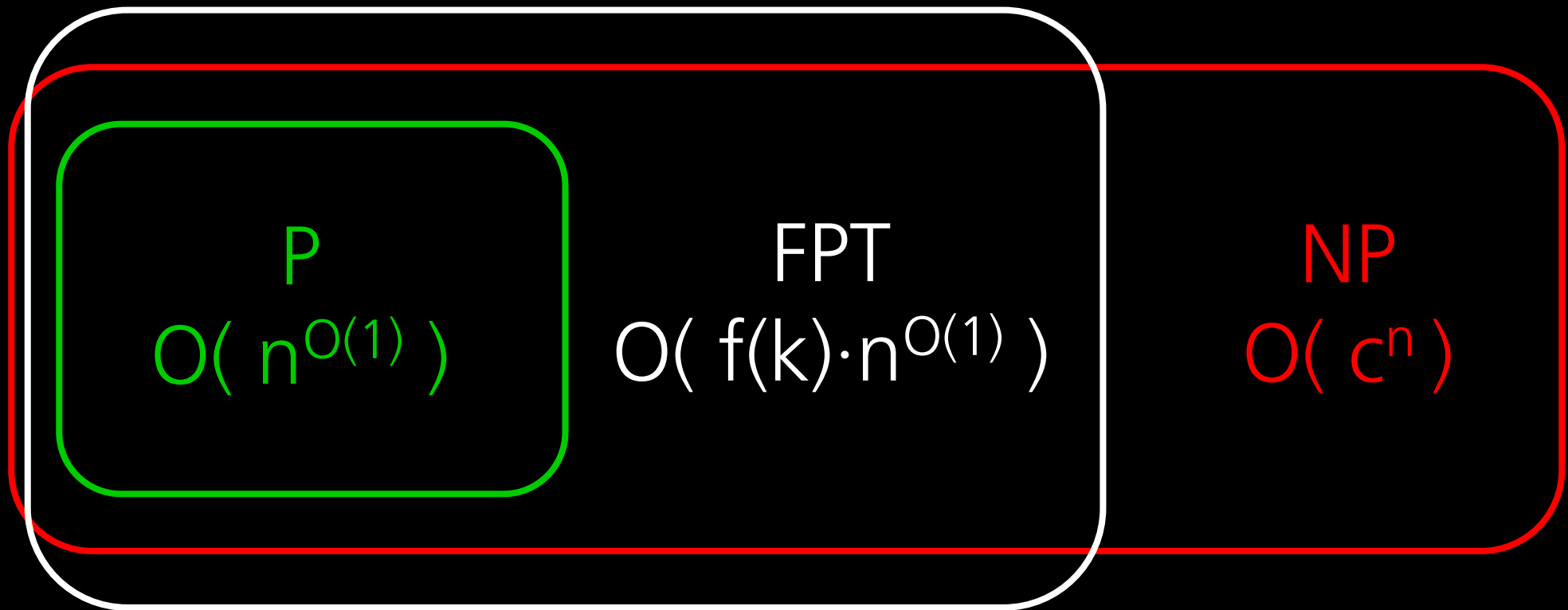


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The fixed-parameter
approach: Two-
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Fixed-Parameter Tractability



Key Facts About Fixed-Parameter Tractability

Data Reduction (Kernels)

If a problem is fixed-parameter tractable, we can polynomial-time reduce an instance of size n to an instance of size $f(k)$ (called *kernel*). Example: A VERTEX COVER instance can be reduced to an instance of size $2k$, where k is the size of a minimum cover. [Chen et al., J. Alg., 2001]

Search Tree

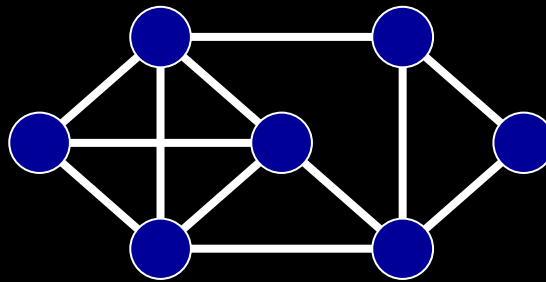
Search trees are depth-bounded by the parameter k .

Intractability

Not all problems are fixed-parameter tractable.

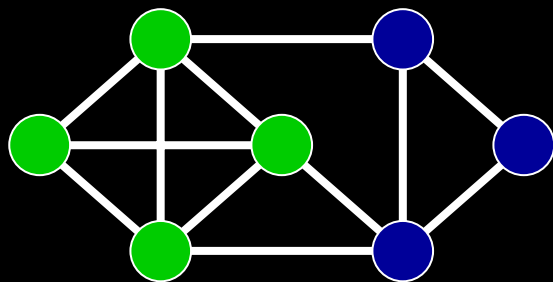
Part II
Case Studies

Case Studies



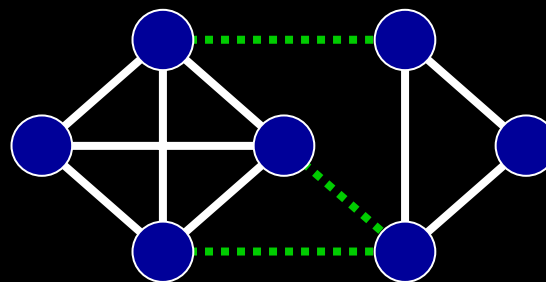
Input

CLIQUE



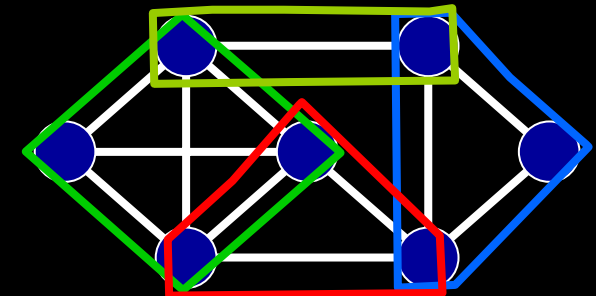
Task: Find the largest clique in the graph.

CLUSTER EDITING



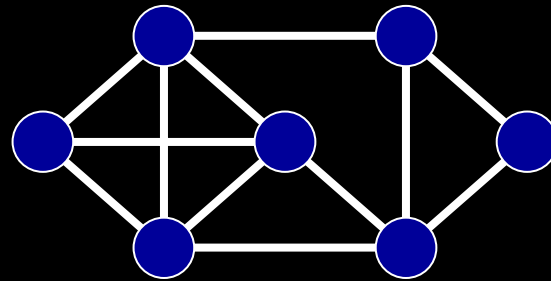
Task: Add / remove few edges so that cliques remain

CLIQUE COVER

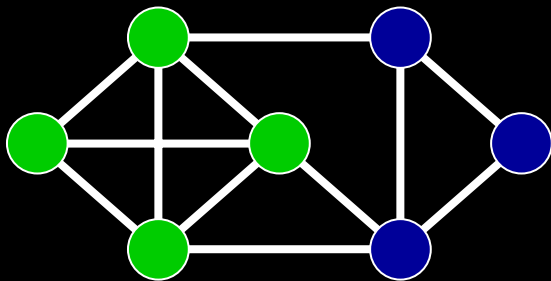


Task: Find few cliques to cover all edges.

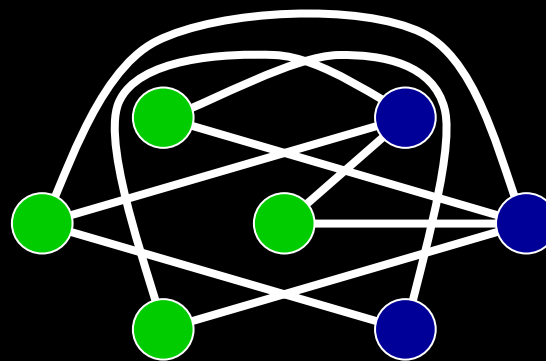
CLIQUE – Main Idea



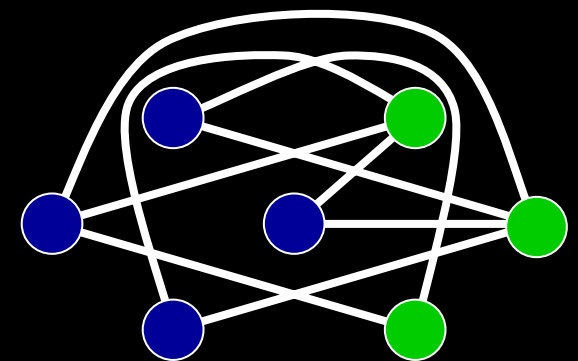
Input: An n -vertex graph



Graph has clique of size $(n-k)$.



Complement graph has independent set of size $(n-k)$.



Complement graph has vertex cover of size k .

CLIQUE – Results

Vertex Cover

Can be solved optimally with a search-tree of size $O(1.28^k)$.

Workhorse in practice is data reduction.

[Abu-Khzam et al., Proc. ALENEX 2004]

Experimental

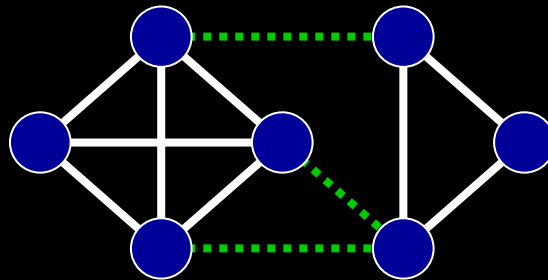
Graphs with 10^5 vertices and with $k \leq 300$ can be solved in practice

[Cheetham et al., J. Com. Sys. Sci, 2003]

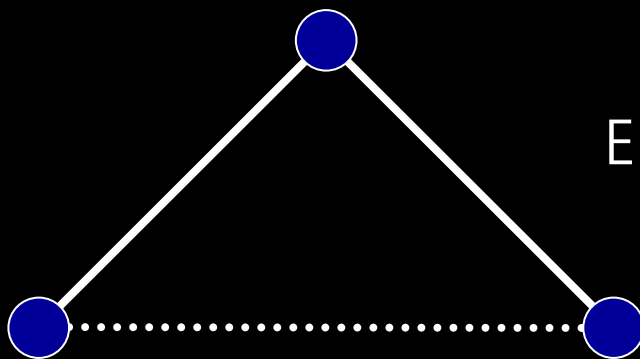
Intractability

Parameterized by the clique size is not fixed-parameter tractable. Graph must contain a large clique.

CLUSTER EDITING – Main Idea



Task: Add / remove few edges so that cliques remain



Basic search tree strategy:
Either delete one of the edges
or add the missing one

CLUSTER EDITING – Results

Data
Reduction

With parameter k , an instance can be reduced to a kernel of size k^3 in polynomial time.

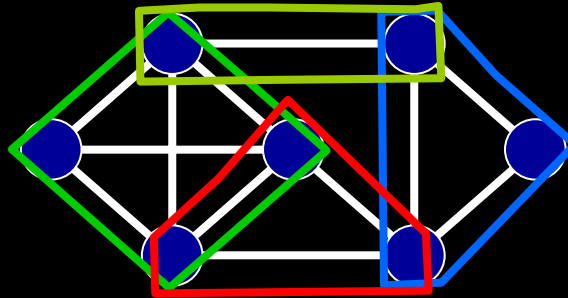
Worst-Case
Running Time

For CLUSTER EDITING, $O(1.92^k + |V|^3)$.
[Gramm et al., Algorithmica, 2004]

When only edge deletions are allowed (CLUSTER DELETION), this reduces to $O(1.77^k + |V|^3)$.

[Gramm et al., Theory of Computing Systems, 2005]

CLIQUE COVER – Results



Task: Find few cliques
to cover all edges.

Workhorse

Interleaving

Data reduction rules prove quite effective although they only guarantee a kernel of size $O(2^k)$.

Combining search trees and data reduction can solve instances with clique cover sizes of about 150.

[Gramm et al., Proc. ALENEX 2006]

Part III
Epilogue

Conclusion

Fixed-Parameter Tractability belongs into the toolbox of algorithm designers - also in the area of graph-modeled data clustering.

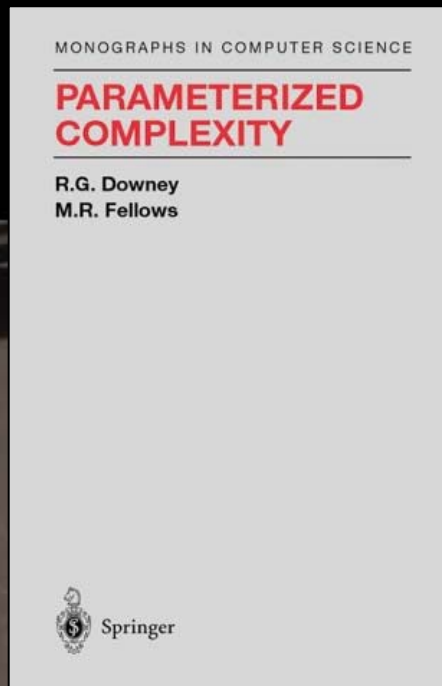
Reduction rules and data reduction may be very effective despite of seemingly impractical worst-case bounds.

More experimental work and algorithm engineering needs to be done in cooperation with theorists. Be invited!

Further Reading



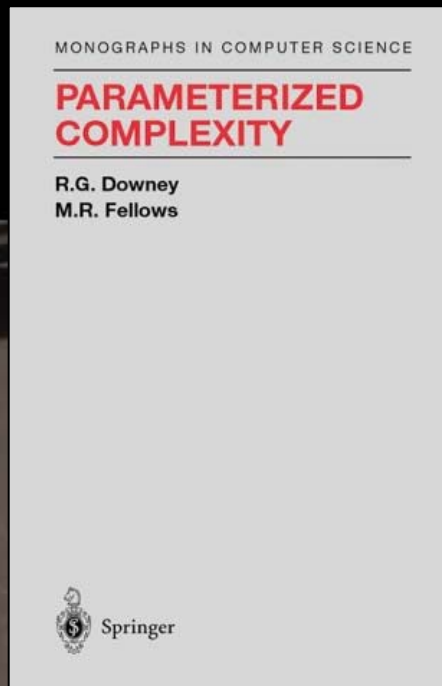
Further Reading



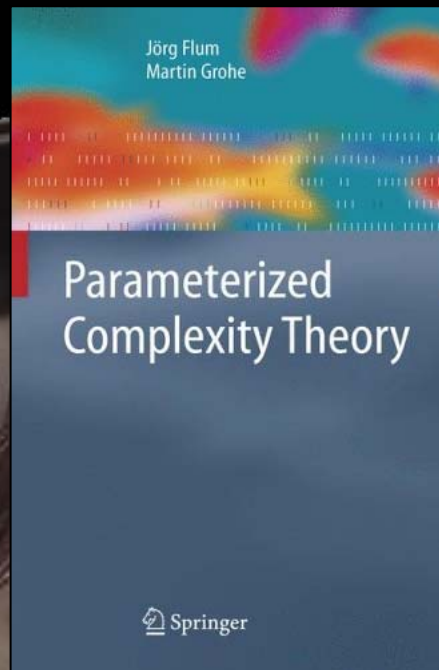
DOWNEY / FELLOWS:
PARAMETERIZED
COMPLEXITY (1999)
FIRST BOOK ON FPT



Further Reading

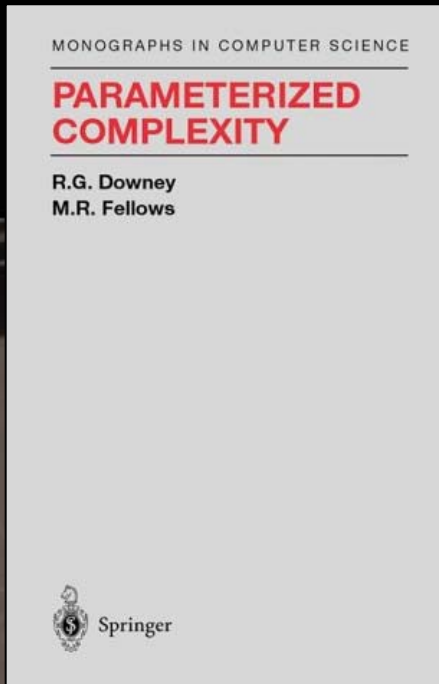


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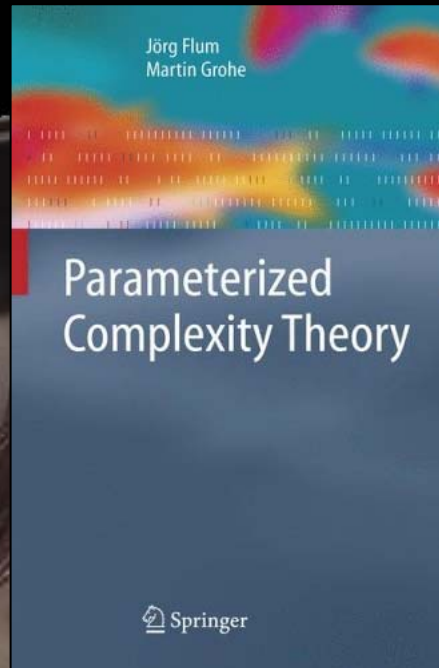


FLUM / GROHE:
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THEORY (2006)
COVERS THEORY

Further Reading



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PARAM. COMPLEXITY
THEORY (2006)
COVERS THEORY



NIEDERMEIER:
INVITATION TO FIXED-
PARAM. ALGOR. (2006)
COVERS ALGORITHMICS

Group Members Working on Fixed-Parameter Algorithms



Rolf Niedermeier
Chair



Michael
Dom



Jiong
Guo



Falk
Hüffner



Hannes
Moser



Sebastian
Wernicke



Jochen
Alber



Jens
Gramm

<http://theinf1.informatik.uni-jena.de/>

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