# Fixed Parameter Algorithms and Kernelization 

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## Classical complexity

A brief review:
6 We usually aim for polynomial-time algorithms: the running time is $O\left(n^{c}\right)$, where $n$ is the input size.

6 Classical polynomial-time algorithms: shortest path, mathching, minimum spanning tree, 2SAT, convext hull, planar drawing, linear programming, etc.

6 It is unlikely that polynomial-time algorithms exist for NP-hard problems.
(6) Unfortunately, many problems of interest are NP-hard: Hamiltonian cycle, 3-coloring, 3SAT, etc.
(6) We expect that these problems can be solved only in exponential time (i.e., $c^{n}$ ).

Can we say anything nontrivial about NP-hard problems?

## Parameterized complexity

Main idea: Instead of expressing the running time as a function $T(n)$ of $n$, we express it as a function $T(n, k)$ of the input size $n$ and some parameter $k$ of the input.

In other words: we do not want to be efficient on all inputs of size $n$, only for those where $k$ is small.

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In other words: we do not want to be efficient on all inputs of size $n$, only for those where $k$ is small.

What can be the parameter $k$ ?
6 The size $k$ of the solution we are looking for.
(6) The maximum degree of the input graph.

6 The diameter of the input graph.
(6) The length of clauses in the input Boolean formula.
© ...

## Parameterized complexity

| Problem: | Minimum Vertex Cover | MAXImUm Independent SET |
| :--- | :--- | :--- |
| Input: | Graph $G$, integer $k$ | Graph $G$, integer $k$ |
| Question: | Is it possible to cover | Is it possible to find |
| the edges with $k$ vertices? | $k$ independent vertices? |  |
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## Parameterized complexity



## Bounded search tree method

Algorithm for Minimum Vertex Cover:


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## Bounded search tree method

Algorithm for Minimum Vertex Cover:


Height of the search tree is $\leqslant k \Rightarrow$ number of leaves is $\leqslant 2^{k} \Rightarrow$ complete search requires $2^{k}$. poly steps.

## Fixed-parameter tractability

Definition: A parameterization of a decision problem is a function that assigns an integer parameter $k$ to each input instance $x$.

The parameter can be
© explicit in the input (for example, if the parameter is the integer $k$ appearing in the input ( $G, k$ ) of Vertex Cover), or
(6) implicit in the input (for example, if the parameter is the diameter $d$ of the input graph $G$ ).

## Main definition:

A parameterized problem is fixed-parameter tractable (FPT) if there is an $f(k) n^{c}$ time algorithm for some constant $c$.

## Fixed-parameter tractability

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Example: Minimum Vertex Cover parameterized by the required size $k$ is FPT: we have seen that it can be solved in time $O\left(2^{k}+n^{2}\right)$.

Better algorithms are known: e.g, $O\left(1.2832^{k} k+k|V|\right)$.
Main goal of parameterized complexity: to find FPT problems.

## FPT problems

Examples of NP-hard problems that are FPT:
6 Finding a vertex cover of size $k$.
6 Finding a path of length $k$.
6 Finding $k$ disjoint triangles.
(6) Drawing the graph in the plane with $k$ edge crossings.
© Finding disjoint paths that connect $k$ pairs of points.

## FPT algorithmic techniques

6 Significant advances in the past 20 years or so (especially in recent years).
6 Powerful toolbox for designing FPT algorithms:


## Books



Downey-Fellows: Parameterized Complexity, Springer, 1999


Flum-Grohe: Parameterized Complexity Theory, Springer, 2006


Niedermeier: Invitation to Fixed-Parameter Algorithms, Oxford University Press, 2006.


## Kernelization



## Kernelization

Definition: Kernelization is a polynomial-time transformation that maps an instance ( $I, k$ ) to an instance ( $I^{\prime}, k^{\prime}$ ) such that
(6) $(I, k)$ is a yes-instance if and only if $\left(I^{\prime}, k^{\prime}\right)$ is a yes-instance,
(6) $k^{\prime} \leqslant k$, and
© $\left|I^{\prime}\right| \leqslant f(k)$ for some function $f(k)$.

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6 $\left|I^{\prime}\right| \leqslant f(k)$ for some function $f(k)$.
Simple fact: If a problem has a kernelization algorithm, then it is FPT.
Proof: Solve the instance ( $I^{\prime}, k^{\prime}$ ) by brute force.

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(6) $k^{\prime} \leqslant k$, and

6 $\left|I^{\prime}\right| \leqslant f(k)$ for some function $f(k)$.
Simple fact: If a problem has a kernelization algorithm, then it is FPT.
Proof: Solve the instance ( $I^{\prime}, k^{\prime}$ ) by brute force.
Converse: Every FPT problem has a kernelization algorithm.
Proof: Suppose there is an $f(k) n^{c}$ algorithm for the problem.
(6) If $f(k) \leqslant n$, then solve the instance in time $f(k) n^{c} \leqslant n^{c+1}$, and output a trivial yes- or no-instance.
6. If $n<f(k)$, then we are done: a kernel of size $f(k)$ is obtained.

## Kernelization for Vertex Cover

General strategy: We devise a list of reduction rules, and show that if none of the rules can be applied and the size of the instance is still larger than $f(k)$, then the answer is trivial.

Reduction rules for Vertex Cover instance ( $G, k$ ):
Rule 1: If $v$ is an isolated vertex $\Rightarrow(G \backslash v, k)$
Rule 2: If $d(v)>k \Rightarrow(G \backslash v, k-1)$

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Rule 1: If $v$ is an isolated vertex $\Rightarrow(G \backslash v, k)$
Rule 2: If $d(v)>k \Rightarrow(G \backslash v, k-1)$
If neither Rule 1 nor Rule 2 can be applied:
(6) If $|V(G)|>k(k+1) \Rightarrow$ There is no solution (every vertex should be the neighbor of at least one vertex of the cover).
© Otherwise, $|V(G)| \leqslant k(k+1)$ and we have a $k(k+1)$ vertex kernel.

## Kernelization for Vertex Cover

Let us add a third rule:
Rule 1: If $v$ is an isolated vertex $\Rightarrow(G \backslash v, k)$
Rule 2: If $d(v)>k \Rightarrow(G \backslash v, k-1)$
Rule 3: If $d(v)=1$, then we can assume that its neighbor $u$ is in the solution $\Rightarrow(G \backslash(u \cup v), k-1)$.

If none of the rules can be applied, then every vertex has degree at least 2.
$\Rightarrow|V(G)| \leqslant|E(G)|$
6. If $|E(G)|>k^{2} \Rightarrow$ There is no solution (each vertex of the solution can cover at most $k$ edges).
(6) Otherwise, $|V(G)| \leqslant|E(G)| \leqslant k^{2}$ and we have a $k^{2}$ vertex kernel.

## Kernelization for Vertex Cover

Let us add a fourth rule:
Rule 4a: If $v$ has degree 2, and its neighbors $u_{1}$ and $u_{2}$ are adjacent, then we can assume that $u_{1}, u_{2}$ are in the solution $\Rightarrow\left(G \backslash\left\{u_{1}, u_{2}, v\right\}, k-2\right)$.


## Kernelization for Vertex Cover

Let us add a fourth rule:
Rule 4b: If $v$ has degree 2, then $G^{\prime}$ is obtained by identifying the two neighbors of $v$ and deleting $v \Rightarrow\left(G^{\prime}, k-1\right)$.


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Rule 4b: If $v$ has degree 2, then $G^{\prime}$ is obtained by identifying the two neighbors of $v$ and deleting $v \Rightarrow\left(G^{\prime}, k-1\right)$.

Correctness:


Let $S^{\prime}$ be a vertex cover of size $k-1$ for $G^{\prime}$.
If $u \in S \Rightarrow\left(S^{\prime} \backslash u\right) \cup\left\{u_{1}, u_{2}\right\}$ is a vertex cover of size $k$ for $G$.
If $u \notin S \Rightarrow S^{\prime} \cup v$ is a vertex cover of size $k$ for $G$.

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If $u_{1}, u_{2} \in S \Rightarrow\left(S \backslash\left\{u_{1}, u_{2}, v\right\}\right) \cup u$ is a vertex cover of size $k-1$ for $G^{\prime}$.
If exactly one of $u_{1}$ and $u_{2}$ is in $S$, then $v \in S \Rightarrow\left(S \backslash\left\{u_{1}, u_{2}, v\right\}\right) \cup u$ is a vertex cover of size $k-1$ for $G^{\prime}$.
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Kernel size:


6 If $|E(G)|>k^{2} \Rightarrow$ There is no solution (each vertex of the solution can cover at most $k$ edges).
(6) Otherwise, $|V(G)| \leqslant 2|E(G)| / 3 \leqslant \frac{2}{3} k^{2}$ and we have a $\frac{2}{3} k^{2}$ vertex kernel.

## Covering Points with Lines

Task: Given a set $P$ of $n$ points in the plane and an integer $k$, find $k$ lines that cover all the points.


Note: We can assume that every line of the solution covers at least 2 points, thus there are at most $n^{2}$ candidate lines.

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## Reduction Rule:

If a candidate line covers a set $S$ of more than $k$ points $\Rightarrow(P \backslash S, k-1)$.
If this rule cannot be applied and there are still more than $k^{2}$ points, then there is no solution $\Rightarrow$ Kernel with at most $k^{2}$ points.

## Kernelization

6 Kernelization can be thought of as a polynomial-time preprocessing before attacking the problem with whatever method we have. "It does no harm" to try kernelization.

6 Some kernelizations use lots of simple reduction rules and require a complicated analysis to bound the kernel size...

6 ... while other kernelizations are based on surprising nice tricks (Next: Crown Reduction and the Sunflower Lemma).

6 Possibility to prove lower bounds.

## Crown Reduction

Definition: A crown decomposition is a partition $C \cup H \cup B$ of the vertices such that

6 $C$ is an independent set,
6 there is no edge between $C$ and $B$,
6 there is a matching between $C$ and $H$ that covers $H$.


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Crown rule for Vertex Cover:
The matching needs to be covered and we can assume that it is covered by $H$ (makes no sense to use vertices of $C$ )
$\Rightarrow(G \backslash(H \cup C), k-|H|)$.

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## Crown Reduction

Key lemma:
Lemma: Given a graph $G$ without isolated vertices and an integer $k$, in polynomial time we can either

6 find a matching of size $k+1$,
(6) find a crown decomposition,
© or conclude that the graph has at most $3 k$ vertices.

## Crown Reduction

Key lemma:
Lemma: Given a graph $G$ without isolated vertices and an integer $k$, in polynomial time we can either

6 find a matching of size $k+1, \Rightarrow$ No solution!
© find a crown decomposition, $\Rightarrow$ Reduce!
6 or conclude that the graph has at most $3 k$ vertices.
$\Rightarrow 3 \mathrm{k}$ vertex kernel!
This gives a 3k vertex kernel for Vertex Cover.

## Proof

Lemma: Given a graph $G$ without isolated vertices and an integer $k$, in polynomial time we can either
(6) find a matching of size $k+1$,
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6 or conclude that the graph has at most $3 k$ vertices.

For the proof, we need the classical Kőnig's Theorem.
$\tau(G)$ : size of the minimum vertex cover
$\nu(G)$ : size of the maximum matching (independent set of edges)
Theorem: [Kőnig, 1931] If $G$ is bipartite, then

$$
\tau(G)=\nu(G)
$$

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Proof: Find (greedily) a maximal matching; if its size is at least $k+1$, then we are done. The rest of the graph is an independent set $l$.


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Find a maximum matching/minimum vertex cover in the bipartite graph between $X$ and $I$.


## Proof

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Case 1: The minimum vertex cover contains at least one vertex of $X$
$\Rightarrow$ There is a crown decomposition.


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Lemma: Given a graph $G$ without isolated vertices and an integer $k$, in polynomial time we can either
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6 or conclude that the graph has at most $3 k$ vertices.

## Proof:

Case 1: The minimum vertex cover contains at least one vertex of $X$
$\Rightarrow$ There is a crown decomposition.
Case 2: The minimum vertex cover contains only
 vertices of $I \Rightarrow$ It contains every vertex of $I$
$\Rightarrow$ There are at most $2 k+k$ vertices.

## Dual of Vertex Coloring

Parameteric dual of $k$-Coloring. Also known as SAVING $k$ Colors.
Task: Given a graph $G$ and an integer $k$, find a vertex coloring with $|V(G)|-k$ colors.

Crown rule for Dual of Vertex Coloring:

## Dual of Vertex Coloring

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Task: Given a graph $G$ and an integer $k$, find a vertex coloring with $|V(G)|-k$ colors.

Crown rule for Dual of Vertex Coloring:
Suppose there is a crown decomposition for the complement graph $\bar{G}$.
6 $C$ is a clique in $G$ : each vertex needs a distinct color.

6 Because of the matching, it is possible to color $H$ using only these $|C|$ colors.

6 These colors cannot be used for $B$.
© $(G \backslash(H \cup C), k-|H|)$


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## Crown Reduction for Dual of Vertex Coloring

Use the key lemma for the complement $\bar{G}$ of $G$ :
Lemma: Given a graph $G$ without isolated vertices and an integer $k$, in polynomial time we can either
(6) find a matching of size $k+1, \Rightarrow$ YES: we can save $k$ colors!
(6) find a crown decomposition, $\Rightarrow$ Reduce!

6 or conclude that the graph has at most $3 k$ vertices.

$$
\Rightarrow 3 k \text { vertex kernel! }
$$

This gives a 3k vertex kernel for Dual of Vertex Coloring.

## Sunflower Lemma



## Sunflower lemma

Definition: Sets $S_{1}, S_{2}, \ldots, S_{k}$ form a sunflower if the sets $S_{i} \backslash\left(S_{1} \cap S_{2} \cap \cdots \cap S_{k}\right)$ are disjoint.


Lemma: [Erdős and Rado, 1960] If the size of a set system is greater than $(p-1)^{d} \cdot d!$ and it contains only sets of size at most $d$, then the system contains a sunflower with $p$ petals. Furthermore, in this case such a sunflower can be found in polynomial time.

## Sunflowers and d-Hitting Set

$d$-Hitting Set: Given a collection $\mathcal{S}$ of sets of size at most $d$ and an integer $k$, find a set $S$ of $k$ elements that intersects every set of $\mathcal{S}$.


Reduction Rule: If $k+1$ sets form a sunflower, then remove these sets from $\mathcal{S}$ and add the center $C$ to $\mathcal{S}(S$ does not hit one of the petals, thus it has to hit the center).

Note: if the center is empty (the sets are disjoint), then there is no solution.
If the rule cannot be applied, then there are at most $O\left(k^{d}\right)$ sets.

## Sunflowers and d-Hitting Set

$d$-Hitting Set: Given a collection $\mathcal{S}$ of sets of size at most $d$ and an integer $k$, find a set $S$ of $k$ elements that intersects every set of $\mathcal{S}$.


Reduction Rule (variant): Suppose more than $k+1$ sets form a sunflower.
(6) If the sets are disjoint $\Rightarrow$ No solution.

6 Otherwise, keep only $k+1$ of the sets.

If the rule cannot be applied, then there are at most $O\left(k^{d}\right)$ sets.

## Conclusions

6. Many nice techniques invented so far - and probably many more to come.
(6) A single technique might provide the key for several problems.

6 How to find new techniques? By attacking the open problems!
6 Theory is incomplete if there is no way to say sorry we cant! - recently theory has evolved to say problems do not have polynomial kernels!!!

