

11. Kernel Lower Bounds

COMP6741: Parameterized and Exact Computation

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1 Reminder

Kernelization

Definition 1. A *kernelization* (*kernel*) for a parameterized problem Π is a **polynomial time** algorithm, which, for any instance I of Π with parameter k , produces an **equivalent** instance I' of Π with parameter k' such that $|I'| \leq f(k)$ and $k' \leq f(k)$ for a computable function f . We refer to the function f as the *size* of the kernel.

Fixed-parameter tractability

Definition 2. A parameterized problem Π is *fixed-parameter tractable* (FPT) if there is an algorithm solving Π in time $f(k) \cdot \text{poly}(n)$, where n is the instance size, k is the parameter, poly is a polynomial function, and f is a computable function.

Theorem 3. Let Π be a decidable parameterized problem. Π has a kernelization $\Leftrightarrow \Pi$ is FPT.

2 Further Examples of Kernels

2.1 Kernel for Hamiltonian Cycle

A *Hamiltonian cycle* of G is a subgraph of G that is a cycle on $|V(G)|$ vertices.

vc-HAMILTONIAN CYCLE

Input: A graph $G = (V, E)$.
Parameter: $k = vc(G)$, the size of a smallest vertex cover of G .
Question: Does G have a Hamiltonian cycle?

Thought experiment: Imagine a very large instance where the parameter is tiny. How can you simplify such an instance?

Issue: We do not actually know a vertex cover of size k .

- Obtain a vertex cover of size $\leq 2k$ by applying VERTEX COVER-kernelizations to $(G, 0), (G, 1), \dots$ until the first instance where no trivial NO-instance is returned.
- If C is a vertex cover of size $\leq 2k$, then $I = V \setminus C$ is an independent set of size $\geq |V| - 2k$.
- No two consecutive vertices in the Hamiltonian Cycle can be in I .
- A kernel with $\leq 4k$ vertices can now be obtained with the following simplification rule.

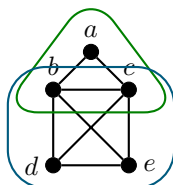
(Too-large)

Compute a vertex cover C of size $\leq 2k$ in polynomial time. If $2|C| < |V|$, then return No

2.2 Kernel for Edge Clique Cover

Definition 4. An *edge clique cover* of a graph $G = (V, E)$ is a set of cliques in G covering all its edges. In other words, if $\mathcal{C} \subseteq 2^V$ is an edge clique cover then each $S \in \mathcal{C}$ is a clique in G and for each $\{u, v\} \in E$ there exists an $S \in \mathcal{C}$ such that $u, v \in S$.

Example: $\{\{a, b, c\}, \{b, c, d, e\}\}$ is an edge clique cover for this graph.



EDGE CLIQUE COVER

Input: A graph $G = (V, E)$ and an integer k
 Parameter: k
 Question: Does G have an edge clique cover of size at most k ?

The *size* of an edge clique cover \mathcal{C} is the number of cliques contained in \mathcal{C} and is denoted $|\mathcal{C}|$.

Helpful properties

Definition 5. A clique S in a graph G is a *maximal* clique if there is no other clique S' in G with $S \subset S'$.

Lemma 6. A graph G has an edge clique cover \mathcal{C} of size at most k if and only if G has an edge clique cover \mathcal{C}' of size at most k such that each $S \in \mathcal{C}'$ is a maximal clique.

Proof sketch. (\Rightarrow): Replace each clique $S \in \mathcal{C}$ by a maximal clique S' with $S \subseteq S'$.

(\Leftarrow): Trivial, since \mathcal{C}' is an edge clique cover of size at most k . □

Simplification rules for Edge Clique Cover

Thought experiment: Imagine a very large instance where the parameter is tiny. How can you simplify such an instance?

The instance could have many degree-0 vertices.

(Isolated)

If there exists a vertex $v \in V$ with $d_G(v) = 0$, then set $G \leftarrow G - v$.

Lemma 7. (*Isolated*) is sound.

Proof sketch. Since no edge is incident to v , a smallest edge clique cover for $G - v$ is a smallest edge clique cover for G , and vice-versa. □

(Isolated-Edge)

If $\exists uv \in E$ such that $d_G(u) = d_G(v) = 1$, then set $G \leftarrow G - \{u, v\}$ and $k \leftarrow k - 1$.

(Twins)

If $\exists u, v \in V, u \neq v$, such that $N_G[u] = N_G[v]$, then set $G \leftarrow G - v$.

Lemma 8. *(Twins) is sound.*

Proof. We need to show that G has an edge clique cover of size at most k if and only if $G - v$ has an edge clique cover of size at most k .

(\Rightarrow): If \mathcal{C} is an edge clique cover of G of size at most k , then $\{S \setminus \{v\} : S \in \mathcal{C}\}$ is an edge clique cover of $G - v$ of size at most k .

(\Leftarrow): Let \mathcal{C}' be an edge clique cover of $G - v$ of size at most k . Partition \mathcal{C}' into $\mathcal{C}_u = \{S \in \mathcal{C}' : u \in S\}$ and $\mathcal{C}_{-u} = \mathcal{C}' \setminus \mathcal{C}_u$. Note that each set in $\mathcal{C}'_u = \{S \cup \{v\} : S \in \mathcal{C}_u\}$ is a clique since $N_G[u] = N_G[v]$ and that each edge incident to v is contained in at least one of these cliques. Now, $\mathcal{C}'_u \cup \mathcal{C}_{-u}$ is an edge clique cover of G of size at most k . \square

(Size-V)

If the previous simplification rules do not apply and $|V| > 2^k$, then return No.

Lemma 9. *(Size-V) is sound.*

Proof. For the sake of contradiction, assume neither (Isolated) nor (Twins) are applicable, $|V| > 2^k$, and G has an edge clique cover \mathcal{C} of size at most k . Since $2^{\mathcal{C}}$ (the set of all subsets of \mathcal{C}) has size at most 2^k , and every vertex belongs to at least one clique in \mathcal{C} by (Isolated), we have that there exists two vertices $u, v \in V$ such that $\{S \in \mathcal{C} : u \in S\} = \{S \in \mathcal{C} : v \in S\}$. But then, $N_G[u] = \bigcup_{S \in \mathcal{C}: u \in S} S = \bigcup_{S \in \mathcal{C}: v \in S} S = N_G[v]$, contradicting that (Twin) is not applicable. \square

Kernel for Edge Clique Cover

Theorem 10. EDGE CLIQUE COVER has a kernel with $O(2^k)$ vertices and $O(4^k)$ edges.

Corollary 11. EDGE CLIQUE COVER is FPT.

3 Frequently Arising Issues

Issue 1: A kernelization needs to produce an instance of the same problem.

How could we turn the following lemma into a simplification rule?

Lemma 12. *If there is an edge $\{u, v\} \in E$ such that $S = N_G[u] \cap N_G[v]$ is a clique, then there is a smallest edge clique cover \mathcal{C} with $S \in \mathcal{C}$.*

Proof. By Lemma 6, we may assume the clique covering the edge $\{u, v\}$ is a maximal clique. But, S is the unique maximal clique covering $\{u, v\}$. \square

(Neighborhood-Clique)

If there exists $\{u, v\} \in E$ such that $S = N_G[u] \cap N_G[v]$ is a clique, then ...???

Edges with both endpoints in $S \setminus \{u, v\}$ are covered by S but might still be needed in other cliques.

We could design a kernelization for a more general problem.

GENERALIZED EDGE CLIQUE COVER

Input: A graph $G = (V, E)$, a set of edges $R \subseteq E$, and an integer k

Parameter: k

Question: Is there a set \mathcal{C} of at most k cliques in G such that each $e \in R$ is contained in at least one of these cliques?

(Neighborhood-Clique)

If there exists $\{u, v\} \in R$ such that $S = N_G[u] \cap N_G[v]$ is a clique, then set $G \leftarrow (V, E \setminus \{u, v\})$, $R \leftarrow R \setminus \{\{x, y\} : x, y \in S\}$, and $k \leftarrow k - 1$.

Issue 2: A proposed simplification rule might not be sound.

Consider the following simplification rule for VERTEX COVER.

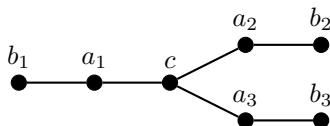
(Optimistic-Degree- $(\geq k)$)

If $\exists v \in V$ such that $d_G(v) \geq k$, then set $G \leftarrow G - v$ and $k \leftarrow k - 1$.

To show that a simplification rule is not sound, we exhibit a counter-example.

Lemma 13. *(Optimistic-Degree- $(\geq k)$) is not sound for VERTEX COVER.*

Proof. Consider the instance consisting of the following graph and $k = 3$.



Since $M = \{\{a_i, b_i\} : 1 \leq i \leq 3\}$ is a matching, a vertex cover contains at least one endpoint of each edge in M . The rule would add c to the vertex cover, leading to a vertex cover of size at least 4. However, $\{a_i : 1 \leq i \leq 3\}$ is a vertex cover of size 3. \square

Issue 3: A problem might be FPT, but only an exponential kernel might be known / possible to achieve.

4 Kernel Lower Bounds

Polynomial vs. exponential kernels

- For some FPT problems, only exponential kernels are known.
- Could it be that all FPT problems have polynomial kernels?
- We will see that polynomial kernels for some fixed-parameter tractable parameterized problems would contradict complexity-theoretic assumptions.

Intuition by example

LONG PATH

Input: A graph $G = (V, E)$, and an integer $k \leq |V|$.

Parameter: k

Question: Does G have a path of length at least k (as a subgraph)?

LONG PATH is NP-complete but FPT.

- Assume LONG PATH has a k^c kernel, where $c = O(1)$.
- Set $q = k^c + 1$ and consider q instances with the same parameter k :

$$(G_1, k), (G_2, k), \dots, (G_q, k).$$

- Let $G = G_1 \oplus G_2 \oplus \dots \oplus G_q$ be the disjoint union of all these graphs.
- Note that (G, k) is a YES-instance if and only if at least one of $(G_i, k), 1 \leq i \leq q$, is a YES-instance.
- Kernelizing (G, k) gives an instance of size k^c , i.e., on average less than one bit per original instance.
- “The kernelization must have solved at least one of the original NP-hard instances in polynomial time”.
- Note that this is not a rigorous argument, and we will make this more formal now.

4.1 Compositions

Distillation

Definition 14. Let Π_1, Π_2 be two problems. An *OR-distillation* (resp., *AND-distillation*) from Π_1 into Π_2 is a polynomial time algorithm D whose input is a sequence I_1, \dots, I_q of instances for Π_1 and whose output is an instance I' for Π_2 such that

- $|I'| \leq \text{poly}(\max_{1 \leq i \leq q} |I_i|)$, and
- I' is a YES-instance for Π_2 if and only if for at least one (resp., for each) $i \in \{1, \dots, q\}$ we have that I_i is a YES-instance for Π_1 .

NP-complete problems don't have distillations

Theorem 15 ([Fortnow, Santhanam, 2008]). *If any NP-complete problem has an OR-distillation, then $\text{coNP} \subseteq \text{NP/poly}$.*¹

Note: $\text{coNP} \subseteq \text{NP/poly}$ is not believed to be true and it would imply that the polynomial hierarchy collapses to the third level: $\text{PH} \subseteq \Sigma_3^p$.

Theorem 16 ([Drucker, 2012]). *If any NP-complete problem has an AND-distillation, then $\text{coNP} \subseteq \text{NP/poly}$.*

Composition algorithms

Definition 17. Let Π be a parameterized problem. An *OR-composition* (resp., *AND-composition*) of Π is a polynomial time algorithm A that receives as input a finite sequence I_1, \dots, I_q of Π with parameters $k_1 = \dots = k_q = k$ and outputs an instance I' for Π with parameter k' such that

- $k' \leq \text{poly}(k)$, and
- I' is a YES-instance for Π if and only if for at least one (resp., for each) $i \in \{1, \dots, q\}$, I_i is a YES-instance for Π .

Tool for showing kernel lower bounds

Theorem 18 (Composition Theorem). *Let Π be an NP-complete parameterized problem such that for each instance I of Π with parameter k , the value of the parameter k can be computed in polynomial time and $k \leq |I|$. If Π has an OR-composition or an AND-composition, then Π has no polynomial kernel, unless $\text{coNP} \subseteq \text{NP/poly}$.*

Proof sketch. Suppose Π has an OR/AND-composition and a polynomial kernel. Then, one can obtain an OR/AND-distillation from Π into $\text{OR}(\Pi)/\text{AND}(\Pi)$.

$$\begin{array}{llll}
 I_1 & I_2 & \dots & I_q \quad q \text{ instances of size } \leq n = \max_{1 \leq i \leq q} |I_i| \\
 \{I_i : k_i = 0\} \dots \{I_i : k_i = n\} & & & \text{group by parameter} \\
 I'_0 & I'_1 & \dots & I'_n \quad \text{After OR-composition: } n+1 \text{ instances with } k'_i \leq \text{poly}(n) \\
 I''_0 & I''_1 & \dots & I''_n \quad \text{After kernelization: } n+1 \text{ instances of size } \text{poly}(n) \text{ each} \\
 & & & \text{This is an instance of } \text{OR}(\Pi) \text{ of size } \text{poly}(n).
 \end{array}$$

□

Long Path has no polynomial kernel

Theorem 19. *LONG PATH has no polynomial kernel unless $\text{NP} \subseteq \text{coNP/poly}$.*

Proof. Clearly, k can be computed in polynomial time and $k \leq |V|$. We give an OR-composition for LONG PATH, which will prove the theorem by the previous lemma. It receives as input a sequence of instances for LONG PATH: $(G_1, k), \dots, (G_q, k)$, and it produces the instance $(G_1 \oplus \dots \oplus G_q, k)$, which is a YES-instance if and only if at least one of $(G_1, k), \dots, (G_q, k)$ is a YES-instance. □

¹NP/poly is the class of all decision problems for which there exists a polynomial-time nondeterministic Turing Machine M with the following property: for every $n \geq 0$, there is an advice string A of length $\text{poly}(n)$ such that, for every input I of length n , the machine M correctly decides the problem with input I , given I and A .

var-SAT has no poly kernel

var-SAT

Input: A propositional formula F in conjunctive normal form (CNF)
 Parameter: $n = |\text{var}(F)|$, the number of variables in F
 Question: Is there an assignment to $\text{var}(F)$ satisfying all clauses of F ?

Example:

$$(x_1 \vee x_2) \wedge (\neg x_2 \vee x_3 \vee \neg x_4) \wedge (x_1 \vee x_4) \wedge (\neg x_1 \vee \neg x_3 \vee \neg x_4)$$

or

$$\{\{x_1, x_2\}, \{\neg x_2, x_3, \neg x_4\}, \{x_1, x_4\}, \{\neg x_1, \neg x_3, \neg x_4\}\}$$

Theorem 20. *var-SAT has no polynomial kernel unless $\text{NP} \subseteq \text{coNP/poly}$.*

Proof. Clearly, $\text{var}(F)$ can be computed in polynomial time and $n = |\text{var}(F)| \leq |F|$. We give an OR-composition for var-SAT, which will prove the theorem by the previous lemma.

- Let F_1, \dots, F_q be CNF formulas, $|F_i| \leq m$, $|\text{var}(F_i)| = n$.
- We can decide whether one of the formulas is satisfiable in time $\text{poly}(mt2^n)$. Hence, if $q > 2^n$, the check is polynomial. If some formula is satisfiable, we output this formula, otherwise we output F_1 .
- It remains the case $q \leq 2^n$. We assume $\text{var}(F_1) = \dots = \text{var}(F_q)$, otherwise we change the names of variables.
- Let $s = \lceil \log_2 q \rceil$. Since $q \leq 2^n$, we have that $s \leq n$.
- We take a set $Y = \{y_1, \dots, y_s\}$ of new variables. Let C_1, \dots, C_{2^s} be the sequence of all 2^s possible clauses containing exactly s literals over the variables in Y .
- For $1 \leq i \leq q$ we let $F'_i = \{C \cup C_i : C \in F_i\}$.
- We define $F = \bigcup_{i=1}^q F'_i \cup \{C_i : q+1 \leq i \leq 2^s\}$.
- Claim: F is satisfiable if and only if F_i is satisfiable for some $1 \leq i \leq q$.
- Hence we have an OR-composition. □

4.2 Polynomial Parameter Transformations

Another tool for showing kernel lower bounds

Definition 21. Let Π_1, Π_2 be parameterized problems. A *polynomial parameter transformation* from Π_1 to Π_2 is a polynomial time algorithm, which, for any instance I_1 of Π_1 with parameter k_1 , produces an **equivalent** instance I_2 of Π_2 with parameter k_2 such that $k_2 \leq \text{poly}(k_1)$.

Theorem 22. *Let Π_1, Π_2 be parameterized problems such that Π_1 is NP-complete, Π_2 is in NP, and there is a polynomial parameter transformation from Π_1 to Π_2 . If Π_2 has a polynomial kernel, then Π_1 has a polynomial kernel.*

Remark: If we know that an NP-complete parameterized problem Π_1 has no polynomial kernel (unless $\text{NP} \subseteq \text{coNP/poly}$), we can use the theorem to show that some other NP-complete parameterized problem Π_2 has no polynomial kernel (unless $\text{NP} \subseteq \text{coNP/poly}$) by giving a polynomial parameter transformation from Π_1 to Π_2 .

Proof. • We show that under the assumptions of the theorem Π_1 has a polynomial kernel.

- Let I_1 be an instance of Π_1 with parameter k_1 .
- We obtain in polynomial time an equivalent instance I_2 of Π_2 with parameter $k_2 \leq \text{poly}(k_1)$.
- We apply Π_2 's kernelization and obtain I'_2 of size $\leq \text{poly}(k_1)$.
- Since Π_2 is in NP and Π_1 is NP-complete, there exists a polynomial time reduction that maps I'_2 to an equivalent instance I'_1 of Π_1 .
- The size of I'_1 is polynomial in k_1 . □

2CNF-Backdoor Evaluation

Definition 23. A CNF formula F is a 2CNF formula if each clause of F has at most 2 literals.

Note: SAT is polynomial time solvable when the input is restricted to be a 2CNF formula.

Definition 24. A 2CNF-backdoor of a CNF formula F is a set of variables $B \subseteq \text{var}(F)$ such that for each assignment $\alpha : B \rightarrow \{0, 1\}$, the formula $F[\alpha]$ is a 2CNF formula. Here, $F[\alpha]$ is obtained by removing all clauses containing a literal set to 1 by α , and removing the literals set to 0 from all remaining clauses.

2CNF-BACKDOOR EVALUATION
Input: A CNF formula F and a 2CNF-backdoor B of F
Parameter: $k = B $
Question: Is F satisfiable?

Note: the problem is FPT by trying all assignments to B and evaluating the resulting formulas.

Theorem 25. 2CNF-BACKDOOR EVALUATION has no polynomial kernel unless $\text{NP} \subseteq \text{coNP}/\text{poly}$.

Proof. We give a polynomial parameter transformation from var-SAT to 2CNF-BACKDOOR EVALUATION. Let F be an instance for var-SAT . Then, $(F, B = \text{var}(F))$ is an equivalent instance for 2CNF-BACKDOOR EVALUATION with $|B| \leq |\text{var}(F)|$. \square

5 Further Reading

- Chapter 15, *Lower bounds for kernelization* in Marek Cygan, Fedor V. Fomin, Lukasz Kowalik, Daniel Lokshantov, Dániel Marx, Marcin Pilipczuk, Michał Pilipczuk, and Saket Saurabh. Parameterized Algorithms. Springer, 2015.
- Chapter 30 (30.1–30.4), *Kernelization Lower Bounds* in Rodney G. Downey and Michael R. Fellows. Fundamentals of Parameterized Complexity. Springer, 2013.
- Neeldhara Misra, Venkatesh Raman, and Saket Saurabh. *Lower bounds on kernelization*. Discrete Optimization 8(1): 110-128 (2011).