# 11. Kernel Lower Bounds 

## COMP6741: Parameterized and Exact Computation

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## Outline

(1) Reminder
(2) Further Examples of Kernels

- Kernel for Hamiltonian Cycle
- Kernel for Edge Clique Cover
(3) Frequently Arising Issues
(4) Kernel Lower Bounds
- Compositions
- Polynomial Parameter Transformations
(5) Further Reading


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## Kernelization

## Definition 1

A kernelization (kernel) for a parameterized problem $\Pi$ is a polynomial time algorithm, which, for any instance $I$ of $\Pi$ with parameter $k$, produces an equivalent instance $I^{\prime}$ of $\Pi$ with parameter $k^{\prime}$ such that $\left|I^{\prime}\right| \leq f(k)$ and $k^{\prime} \leq f(k)$ 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 $\Pi$ is fixed-parameter tractable (FPT) if there is an algorithm solving $\Pi$ in time $f(k) \cdot \operatorname{poly}(n)$, where $n$ is the instance size, $k$ is the parameter, poly is a polynomial function, and $f$ is a computable function.

## Theorem 3

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

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## Hamiltonian Cycle I

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

$$
\begin{array}{ll}
\text { vc-HAmiltonian CyCle } \\
\text { Input: } & \text { A graph } G=(V, E) . \\
\text { Parameter: } & k=v c(G) \text {, the size of a smallest vertex cover of } G . \\
\text { Question: } & \text { Does } G \text { have a Hamiltonian cycle? }
\end{array}
$$

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

## Hamiltonian Cycle II

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

## Hamiltonian Cycle III

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


## (Too-large)

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

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## 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

> | 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 4

A clique $S$ in a graph $G$ is a maximal clique if there is no other clique $S^{\prime}$ in $G$ with $S \subset S^{\prime}$.

## Lemma 5

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 $\mathcal{C}^{\prime}$ of size at most $k$ such that each $S \in \mathcal{C}^{\prime}$ is a maximal clique.

## Proof sketch.

$(\Rightarrow)$ : Replace each clique $S \in \mathcal{C}$ by a maximal clique $S^{\prime}$ with $S \subseteq S^{\prime}$.
$(\Leftarrow)$ : Trivial, since $\mathcal{C}^{\prime}$ 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?

## Simplification rules for Edge Clique Cover II

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 6

(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.

## Simplification rules for Edge Clique Cover II

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## (Isolated-Edge)

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

## Simplification rules for Edge Clique Cover III

## (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 7

(Twins) is sound.

## Simplification rules for Edge Clique Cover III

## (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 7

(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 $\{S \backslash\{v\}: S \in \mathcal{C}\}$ is an edge clique cover of $G-v$ of size at most $k$.
$(\Leftarrow)$ : Let $\mathcal{C}^{\prime}$ 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 $\mathcal{C}_{\neg u}=\mathcal{C} \backslash \mathcal{C}_{u}$. Note that each set in
$\mathcal{C}_{u}^{\prime}=\left\{S \cup\{v\}: S \in \mathcal{C}_{u}\right\}$ 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, $\mathcal{C}_{u}^{\prime} \cup \mathcal{C}_{\neg u}$ is an edge clique cover of $G$ of size at most $k$.

## Simplification rules for Edge Clique Cover IV

## (Size-V)

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

## Lemma 8

(Size-V) is sound.

## Simplification rules for Edge Clique Cover IV

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

## Lemma 8

(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.

## Kernel for Edge Clique Cover

## Theorem 9

Edge Clique Cover has a kernel with $O\left(2^{k}\right)$ vertices and $O\left(4^{k}\right)$ edges.

## Corollary 10

Edge Clique Cover is FPT.

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## Frequent issues designing simplification rules

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 11

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 5, we may assume the clique covering the edge $\{u, v\}$ is a maximal clique. But, $S$ is the unique maximal clique covering $\{u, v\}$.

## Frequent issues designing simplification rules

## (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 $S \backslash\{u, v\}$ are covered by $S$ but might still be needed in other cliques.

## Frequent issues designing simplification rules

We could design a kernelization for a more general problem.
Generalized Edge Clique Cover
Input: $\quad$ 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 $G \leftarrow(V, E \backslash\{u, v\}), R \leftarrow R \backslash\{\{x, y\}: x, y \in S\}$, and $k \leftarrow k-1$.

## Frequent issues designing simplification rules

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 11
(Optimistic-Degree- $(\geq k)$ ) is not sound for Vertex Cover.

## Frequent issues designing simplification rules

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

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

## Lemma 11

(Optimistic-Degree- $(\geq k)$ ) is not sound for Vertex Cover.

## Proof.

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


Since $M=\left\{\left\{a_{i}, b_{i}\right\}: 1 \leq i \leq 3\right\}$ 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, $\left\{a_{i}: 1 \leq i \leq 3\right\}$ is a vertex cover of size 3.
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## Frequent issues designing simplification rules

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

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## 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)\mathrm{ , 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.

## Intuition by example

- 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 $k$ :

$$
\left(G_{1}, k\right),\left(G_{2}, k\right), \ldots,\left(G_{q}, k\right)
$$

- Let $G=G_{1} \oplus G_{2} \oplus \cdots \oplus G_{q}$ be the disjoint union of all these graphs.
- Note that $(G, k)$ is a Yes-instance if and only if at least one of $\left(G_{i}, k\right), 1 \leq i \leq q$, 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.


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## Distillation

## Definition 11

Let $\Pi_{1}, \Pi_{2}$ be two problems. An OR-distillation (resp., AND-distillation) from $\Pi_{1}$ into $\Pi_{2}$ is a polynomial time algorithm $D$ whose input is a sequence $I_{1}, \ldots, I_{q}$ of instances for $\Pi_{1}$ and whose output is an instance $I^{\prime}$ for $\Pi_{2}$ such that

- $\left|I^{\prime}\right| \leq \operatorname{poly}\left(\max _{1 \leq i \leq q}\left|I_{i}\right|\right)$, and
- $I^{\prime}$ is a Yes-instance for $\Pi_{2}$ if and only if for at least one (resp., for each) $i \in\{1, \ldots, q\}$ we have that $I_{i}$ is a Yes-instance for $\Pi_{1}$.


## NP-complete problems don't have distillations

## Theorem 12 ([Fortnow, Santhanam, 2008])

If any NP-complete problem has an OR-distillation, then coNP $\subseteq$ NP/poly. ${ }^{1}$
Note: coNP $\subseteq$ NP/poly is not believed to be true and it would imply that the polynomial hierarchy collapses to the third level: $\mathrm{PH} \subseteq \Sigma_{3}^{p}$.

## Theorem 13 ([Drucker, 2012])

If any NP-complete problem has an AND-distillation, then coNP $\subseteq$ NP/poly.

[^0]
## Composition algorithms

## Definition 14

Let $\Pi$ be a parameterized problem. An OR-composition (resp., AND-composition) of $\Pi$ is a polynomial time algorithm $A$ that receives as input a finite sequence $I_{1}, \ldots, I_{q}$ of $\Pi$ with parameters $k_{1}=\cdots=k_{q}=k$ and outputs an instance $I^{\prime}$ for $\Pi$ with parameter $k^{\prime}$ such that

- $k^{\prime} \leq \operatorname{poly}(k)$, and
- $I^{\prime}$ is a Yes-instance for $\Pi$ if and only if for at least one (resp., for each) $i \in\{1, \ldots, q\}, I_{i}$ is a Yes-instance for $\Pi$.


## Tool for showing kernel lower bounds

## Theorem 15 (Composition Theorem)

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

## Tool for showing kernel lower bounds

## Theorem 15 (Composition Theorem)

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

## Proof sketch.

Suppose $\Pi$ has an OR/AND-composition and a polynomial kernel. Then, one can obtain an OR/AND-distillation from $\Pi$ into $\operatorname{OR}(\Pi) / A N D(\Pi)$.

## Tool for showing kernel lower bounds

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## Proof sketch.

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

$$
\begin{array}{rccrl}
I_{1} & I_{2} & \ldots & I_{q} & q \text { instances of size } \leq n=\max _{1 \leq i \leq q}\left|I_{i}\right| \\
\left\{I_{i}: k_{i}=0\right\} \ldots\left\{I_{i}: k_{i}=n\right\} & \text { group by parameter } \\
I_{0}^{\prime} & I_{1}^{\prime} & \ldots & I_{n}^{\prime} & \text { After OR-composition: } n+1 \text { instances with } k_{i}^{\prime} \leq \operatorname{poly}(n) \\
I_{0}^{\prime \prime} & I_{1}^{\prime \prime} & \ldots & I_{n}^{\prime \prime} & \text { After kernelization: } n+1 \text { instances of size } \operatorname{poly}(n) \text { each } \\
& & & & \text { This is an instance of } \operatorname{OR}(\Pi) \text { of size } \operatorname{poly}(n) .
\end{array}
$$

## Long Path has no polynomial kernel I

## Theorem 16

Long Path has no polynomial kernel unless NP $\subseteq$ 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: $\left(G_{1}, k\right), \ldots,\left(G_{q}, k\right)$, and it produces the instance $\left(G_{1} \oplus \cdots \oplus G_{q}, k\right)$, which is a Yes-instance if and only if at least one of $\left(G_{1}, k\right), \ldots,\left(G_{q}, k\right)$ is a Yes-instance.

## var-SAT has no poly kernel I

## var-SAT

Input:
Parameter:
Question: Is there an assignment to $\operatorname{var}(F)$ satisfying all clauses of $F$ ?

## Example:

$$
\left(x_{1} \vee x_{2}\right) \wedge\left(\neg x_{2} \vee x_{3} \vee \neg x_{4}\right) \wedge\left(x_{1} \vee x_{4}\right) \wedge\left(\neg x_{1} \vee \neg x_{3} \vee \neg x_{4}\right)
$$

or

$$
\left\{\left\{x_{1}, x_{2}\right\},\left\{\neg x_{2}, x_{3}, \neg x_{4}\right\},\left\{x_{1}, x_{4}\right\},\left\{\neg x_{1}, \neg x_{3}, \neg x_{4}\right\}\right\}
$$

## var-SAT has no poly kernel II

## Theorem 17

var-SAT has no polynomial kernel unless NP $\subseteq$ coNP/poly.

## Proof.

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

- Let $F_{1}, \ldots, F_{q}$ be CNF formulas, $\left|F_{i}\right| \leq m,\left|\operatorname{var}\left(F_{i}\right)\right|=n$.
- We can decide whether one of the formulas is satisfiable in time poly $\left(m t 2^{n}\right)$. Hence, if $q>2^{n}$, the check is polynomial. If some formula is satisfiable, we output this formula, otherwise we output $F_{1}$.


## var-SAT has no poly kernel III

## Proof (continued).

- It remains the case $q \leq 2^{n}$. We assume $\operatorname{var}\left(F_{1}\right)=\cdots=\operatorname{var}\left(F_{q}\right)$, otherwise we change the names of variables.
- Let $s=\left\lceil\log _{2} q\right\rceil$. Since $q \leq 2^{n}$, we have that $s \leq n$.
- We take a set $Y=\left\{y_{1}, \ldots, y_{s}\right\}$ of new variables. Let $C_{1}, \ldots, C_{2^{s}}$ 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 $F_{i}^{\prime}=\left\{C \cup C_{i}: C \in F_{i}\right\}$.
- We define $F=\bigcup_{i=1}^{q} F_{i}^{\prime} \cup\left\{C_{i}: q+1 \leq i \leq 2^{s}\right\}$.
- 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.


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## Another tool for showing kernel lower bounds I

## Definition 18

Let $\Pi_{1}, \Pi_{2}$ be parameterized problems. A polynomial parameter transformation from $\Pi_{1}$ to $\Pi_{2}$ is a polynomial time algorithm, which, for any instance $I_{1}$ of $\Pi_{1}$ with parameter $k_{1}$, produces an equivalent instance $I_{2}$ of $\Pi_{2}$ with parameter $k_{2}$ such that $k_{2} \leq \operatorname{poly}\left(k_{1}\right)$.

## Another tool for showing kernel lower bounds II

## Theorem 19

Let $\Pi_{1}, \Pi_{2}$ be parameterized problems such that $\Pi_{1}$ is NP-complete, $\Pi_{2}$ is in NP, and there is a polynomial parameter transformation from $\Pi_{1}$ to $\Pi_{2}$. If $\Pi_{2}$ has a polynomial kernel, then $\Pi_{1}$ has a polynomial kernel.

Remark: If we know that an NP-complete parameterized problem $\Pi_{1}$ has no polynomial kernel (unless NP $\subseteq$ coNP/poly), we can use the theorem to show that some other NP-complete parameterized problem $\Pi_{2}$ has no polynomial kernel (unless NP $\subseteq$ coNP/poly) by giving a polynomial parameter transformation from $\Pi_{1}$ to $\Pi_{2}$.

## Another tool for showing kernel lower bounds III

## Proof.

- We show that under the assumptions of the theorem $\Pi_{1}$ has a polynomial kernel.
- Let $I_{1}$ be an instance of $\Pi_{1}$ with parameter $k_{1}$.
- We obtain in polynomial time an equivalent instance $I_{2}$ of $\Pi_{2}$ with parameter $k_{2} \leq \operatorname{poly}\left(k_{1}\right)$.
- We apply $\Pi_{2}$ 's kernelization and obtain $I_{2}^{\prime}$ of size $\leq \operatorname{poly}\left(k_{1}\right)$.
- Since $\Pi_{2}$ is in NP and $\Pi_{1}$ is NP-complete, there exists a polynomial time reduction that maps $I_{2}^{\prime}$ to an equivalent instance $I_{1}^{\prime}$ of $\Pi_{1}$.
- The size of $I_{1}^{\prime}$ is polynomial in $k_{1}$.


## 2CNF-Backdoor Evaluation I

## Definition 20

## 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 21

A 2CNF-backdoor of a CNF formula $F$ is a set of variables $B \subseteq \operatorname{var}(F)$ 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 $\alpha$, and removing the literals set to 0 from all remaining clauses.

## 2CNF-Backdoor Evaluation II

2CNF-Backdoor Evaluation
Input: $\quad$ A CNF formula $F$ and a 2CNF-backdoor $B$ of $F$
Parameter: $\quad k=|B|$
Question: Is $F$ satisfiable?
Note: the problem is FPT by trying all assignments to $B$ and evaluating the resulting formulas.

## 2CNF-Backdoor Evaluation III

```
Theorem 22
2CNF-Backdoor Evaluation has no polynomial kernel unless
NP\subseteq coNP/poly.
```


## Proof.

We give a polynomial parameter transformation from var-SAT to 2CNF-Backdoor Evaluation.
Let $F$ be an instance for var-SAT.
Then, $(F, B=\operatorname{var}(F))$ is an equivalent instance for 2CNF-BACKDOor
Evaluation with $|B| \leq|\operatorname{var}(F)|$.

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## Further Reading

- Chapter 15, Lower bounds for kernelization in Marek Cygan, Fedor V. Fomin, Łukasz Kowalik, Daniel Lokshtanov, Dániel Marx, Marcin Pilipczuk, MichałPilipczuk, 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).


[^0]:    ${ }^{1} \mathrm{NP} /$ 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 poly $(n)$ such that, for every input $I$ of length $n$, the machine $M$ correctly decides the problem with input $I$, given $I$ and $A$.

