

# Exponential Time Hypothesis part 1

Marek Cygan

Institute of Informatics  
University of Warsaw

19th August 2014, Będlewo

# Introduction

- We already know that  $W[1]$ -hard problems are different than the ones which are FPT.
- However, within the FPT class we have different dependency on the parameter in best known algorithms:  
 $2^{\mathcal{O}(\sqrt{k})}, 2^{\mathcal{O}(k)}, 2^{\mathcal{O}(k \log k)}, 2^{2^{\mathcal{O}(k)}}$ .
- Similarly for  $W[1]$ -hard problems:  
 $f(k)n^{\mathcal{O}(k)}, f(k)n^{\mathcal{O}(\sqrt{k})}, f(k)n^{\mathcal{O}(\log \log k)}$ .

# Introduction

- It seems that there are some inner hierarchies of subclasses, corresponding to different classes of running times.
- To separate these classes, we need a methodology for proving lower bounds.
- Ideally, if  $2^{f(k)} n^{\mathcal{O}(1)}$  is known, we want  $2^{o(f(k))} n^{\mathcal{O}(1)}$  lower bound.
- Dream, for  $2^{f(k)} n^{\mathcal{O}(1)}$  algorithm,  $2^{(1-\epsilon)f(k)} n^{\mathcal{O}(1)}$  lower bound.
- $P \neq NP$  or even  $FPT \neq W[1]$  is (probably) not strong enough to serve as a source of hardness.

# Introduction

- It seems that there are some inner hierarchies of subclasses, corresponding to different classes of running times.
- To separate these classes, we need a methodology for proving lower bounds.
- Ideally, if  $2^{f(k)} n^{\mathcal{O}(1)}$  is known, we want  $2^{o(f(k))} n^{\mathcal{O}(1)}$  lower bound.
- Dream, for  $2^{f(k)} n^{\mathcal{O}(1)}$  algorithm,  $2^{(1-\epsilon)f(k)} n^{\mathcal{O}(1)}$  lower bound.
- $P \neq NP$  or even  $FPT \neq W[1]$  is (probably) not strong enough to serve as a source of hardness.

If we are pessimistic enough, sometimes dreams come true.

# What this lecture is about?

## Exponential Time Hypothesis (ETH)

Informally: 3-CNF SAT cannot be solved in time subexponential in the number of variables.

# What this lecture is about?

## Exponential Time Hypothesis (ETH)

Informally: 3-CNF SAT cannot be solved in time subexponential in the number of variables.

### Outline:

- 1 Formal definition of ETH.
- 2 Sparsification lemma.
- 3 ETH and classic NP-hardness reductions:  $2^{o(n)}$ ,  $2^{o(k)} n^{O(1)}$ .
- 4 ETH and planar problems:  $2^{o(\sqrt{n})}$ ,  $2^{o(\sqrt{k})} n^{O(1)}$ .
- 5 ETH and  $W[1]$ -hard problems:  $f(k) n^{o(k)}$ .
- 6 Glimpse on less common lower bounds.

# CNF SAT

A formula on variables  $x_1, \dots, x_n$  in *conjunctive normal form* (CNF):

$$\varphi = C_1 \wedge C_2 \wedge \dots \wedge C_m,$$

where  $C_i = l_1^i \vee l_2^i \vee \dots \vee l_r^i$

and each  $l_j^i$  is a literal, i.e., variable or its negation.

- Example:  $(x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee x_4) \wedge (\neg x_2)$
- If each clause has at most  $q$  literals, we have  $q$ -CNF formula.
- CNF-SAT: given a CNF formula decide if it is satisfiable.
- $q$ -SAT: same for  $q$ -CNF.

# CNF SAT

- 3-SAT is NP-complete, hence no poly time algorithm,
- exhaustive search gives  $\mathcal{O}^*(2^n)$  time algorithm for CNF-SAT,  
 $\mathcal{O}^*(\ )$  notation suppresses polynomial factors
- $q$ -SAT is solvable in  $\mathcal{O}^*(2^{\gamma_q n})$  where  $\gamma_q = 1 - \Theta(1/q)$ ,
- current best algorithm for 3-SAT:  $\mathcal{O}(2^{0.386n})$ .

Current status exposes two barriers hard to break:

- 1  $2^{o(n)}$  algorithm for 3-SAT,
- 2  $\mathcal{O}^*((2 - \epsilon)^n)$  algorithm for general CNF-SAT for some  $\epsilon > 0$ .



# ETH and SETH

## Definition

For  $q \geq 3$  let  $\delta_q$  be infimum of the set  
 $\{c : \text{there is } \mathcal{O}^*(2^{cn}) \text{ algorithm for } q\text{-SAT}\}$

# ETH and SETH

## Definition

For  $q \geq 3$  let  $\delta_q$  be infimum of the set  
 $\{c : \text{there is } \mathcal{O}^*(2^{cn}) \text{ algorithm for } q\text{-SAT}\}$

## Conjectures

- Exponential Time Hypothesis (ETH):  $\delta_3 > 0$ .
- Strong Exponential Time Hypothesis (SETH):  $\lim_{q \rightarrow \infty} \delta_q = 1$ .

# ETH and SETH

## Definition

For  $q \geq 3$  let  $\delta_q$  be infimum of the set  
 $\{c : \text{there is } \mathcal{O}^*(2^{cn}) \text{ algorithm for } q\text{-SAT}\}$

## Conjectures

- Exponential Time Hypothesis (ETH):  $\delta_3 > 0$ .
- Strong Exponential Time Hypothesis (SETH):  $\lim_{q \rightarrow \infty} \delta_q = 1$ .

## Why not the following?

- ETH: no  $2^{o(n)}$  algorithm for 3-SAT,
- SETH: no  $\mathcal{O}^*((2 - \epsilon)^n)$  algorithm for CNF-SAT for any  $\epsilon > 0$ .

# Randomized ETH

- The default version of ETH/SETH refutes existence of deterministic algorithms.

# Randomized ETH

- The default version of ETH/SETH refutes existence of deterministic algorithms.
- One could define  $\delta_q^* \leq \delta_q$  as infimum over randomized two-sided error algorithms.

## Two-sided error algorithms

A randomized algorithm with two-sided error is an algorithm which for any YES-instance and any NO-instance returns an incorrect answer with prob. at most constant  $p < 1/2$ .

# Randomized ETH

- The default version of ETH/SETH refutes existence of deterministic algorithms.
- One could define  $\delta_q^* \leq \delta_q$  as infimum over randomized two-sided error algorithms.

## Randomized versions

- ETH:  $\delta_3^* > 0$ .
- SETH:  $\lim_{q \rightarrow \infty} \delta_q^* = 1$ .

# Randomized ETH

## Randomized versions

- ETH:  $\delta_3^* > 0$ .
- SETH:  $\lim_{q \rightarrow \infty} \delta_q^* = 1$ .
- Reasonable from the point of view of research on satisfiability.
- But inconvenient from pragmatic point of view: dealing with error probability in reductions.
- Usually for a chain of reductions excluding some deterministic algorithm, the same chain works for randomized algorithms.
- During this school we assume deterministic versions (discussion on randomized versions in the book).

# Plausibility

- ETH is generally considered as plausible.
- Many believe that SETH can be refuted anytime.
- Still, SETH based lower bounds make sense: existence of better algorithm would constitute a major breakthrough in the complexity of satisfiability.
- In other words, if you do not believe in a lower bound  $X$  which is SETH based, then refute SETH first before trying to disprove  $X$ .



## Take-home message

$\text{SETH} \implies \text{ETH} \implies \text{W}[1] \neq \text{FPT} \implies \text{P} \neq \text{NP}$

$\text{FPT} \neq \text{W}[1] \implies \text{no } \mathcal{O}^*(f(k))$

$\text{ETH} \implies \text{no } \mathcal{O}^*(2^{o(f(k))})$

$\text{SETH} \implies \text{no } \mathcal{O}^*(2^{(1-\epsilon)f(k)}).$

# Sparsification Lemma

Consider a reduction

Given a 3-SAT formula  $\varphi$  with  $n$  vars and  $m$  clauses, produces equivalent instance of problem  $X$  of size  $\mathcal{O}(n + m)$ .

# Sparsification Lemma

## Consider a reduction

Given a 3-SAT formula  $\varphi$  with  $n$  vars and  $m$  clauses, produces equivalent instance of problem  $X$  of size  $\mathcal{O}(n + m)$ .

## Lemma

Under ETH there is no  $\mathcal{O}^*(2^{o(|I|^{1/3})})$  time algorithm for  $X$ .

# Sparsification Lemma

## Consider a reduction

Given a 3-SAT formula  $\varphi$  with  $n$  vars and  $m$  clauses, produces equivalent instance of problem  $X$  of size  $\mathcal{O}(n + m)$ .

## Lemma

Under ETH there is no  $\mathcal{O}^*(2^{o(|I|^{1/3})})$  time algorithm for  $X$ .

Proof by contradiction:

- Assume there is such an algorithm  $\mathcal{A}$ .
- For a given 3-SAT formula  $\varphi$ : (i) apply the reduction, (ii) use  $\mathcal{A}$ .
- Pipelined algorithm works in  $2^{o(|I|^{1/3})} = 2^{o((n+m)^{1/3})} = 2^{o(n)}$ .

# Sparsification Lemma

## Consider a reduction

Given a 3-SAT formula  $\varphi$  with  $n$  vars and  $m$  clauses, produces equivalent instance of problem  $X$  of size  $\mathcal{O}(n + m)$ .

## Lemma

Under ETH there is no  $\mathcal{O}^*(2^{o(|I|^{1/3})})$  time algorithm for  $X$ .

Proof by contradiction:

- Assume there is such an algorithm  $\mathcal{A}$ .
- For a given 3-SAT formula  $\varphi$ : (i) apply the reduction, (ii) use  $\mathcal{A}$ .
- Pipelined algorithm works in  $2^{o(|I|^{1/3})} = 2^{o((n+m)^{1/3})} = 2^{o(n)}$ .

If we knew that  $m = \mathcal{O}(n)$ , we would get a much stronger lower bound of  $\mathcal{O}^*(2^{o(|I|)})$ .

# Sparsification Lemma

## Sparsification Lemma informally

There is subexponential time algorithm which reduces the number of clauses of a  $q$ -SAT formula to  $O(n)$ , for any constant  $q$ .

# Sparsification Lemma

## Sparsification Lemma informally

There is subexponential time algorithm which reduces the number of clauses of a  $q$ -SAT formula to  $O(n)$ , for any constant  $q$ .

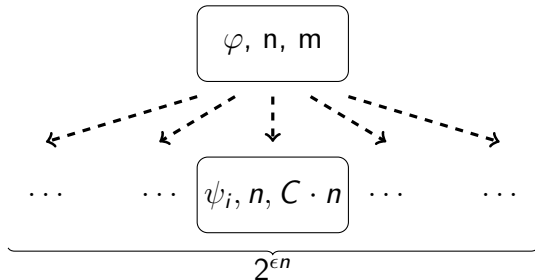
## Sparsification Lemma

For any  $q \geq 3$  and  $\epsilon > 0$  there is a constant  $C = C(q, \epsilon)$  such that any  $q$ -CNF formula  $\varphi$  can be expressed as  $\varphi = \bigvee_{i=1}^t \psi_i$ , where  $t \leq 2^{\epsilon n}$  and each  $\psi_i$  is a  $q$ -CNF formula with the same set of variables and at most  $C \cdot n$  clauses. Such disjunction can be computed in time  $O^*(2^{\epsilon n})$ .

# Sparsification Lemma

## Sparsification Lemma

For any  $q \geq 3$  and  $\epsilon > 0$  there is a constant  $C = C(q, \epsilon)$  such that any  $q$ -CNF formula  $\varphi$  can be expressed as  $\varphi = \bigvee_{i=1}^t \psi_i$ , where  $t \leq 2^{\epsilon n}$  and each  $\psi_i$  is a  $q$ -CNF formula with the same set of variables and at most  $C \cdot n$  clauses. Such disjunction can be computed in time  $\mathcal{O}^*(2^{\epsilon n})$ .





# Sparsification Lemma

## Theorem

Unless ETH fails, there is a constant  $c > 0$ , such that no algorithm solves 3-SAT in time  $\mathcal{O}^*(2^{c(n+m)})$ .

# Sparsification Lemma

## Theorem

Unless ETH fails, there is a constant  $c > 0$ , such that no algorithm solves 3-SAT in time  $\mathcal{O}^*(2^{c(n+m)})$ .

Proof by contradiction:

- Assume for every  $c > 0$  there is an algorithm  $\mathcal{A}_c$  solving 3-SAT in  $\mathcal{O}^*(2^{c(n+m)})$ .
- Consider any  $d > 0$ . We want to solve 3-SAT in time  $\mathcal{O}^*(2^{dn})$ .
- Use Sparsification Lemma for  $\epsilon = d/2$ . Denote  $C = C(3, \epsilon)$ .
- Solve each  $\phi_i$  by  $\mathcal{A}_{c'}$ , where  $c' = \frac{d}{2(C+1)}$ .
- The total running time is  $\mathcal{O}^*(2^\epsilon) + \mathcal{O}^*(2^\epsilon \cdot 2^{\frac{d}{2(C+1)} \cdot (C+1)n}) = \mathcal{O}^*(2^{dn})$ .

# Sparsification Lemma

## Theorem

Unless ETH fails, there is a constant  $c > 0$ , such that no algorithm solves 3-SAT in time  $\mathcal{O}^*(2^{c(n+m)})$ .

## Corollary

Unless ETH fails, there is no  $2^{o(n+m)}$  time algorithm for 3-SAT.

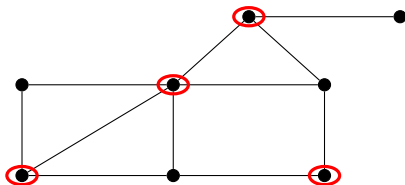
# Combining ETH with classic NP-hardness reductions

# Vertex Cover

## Vertex Cover

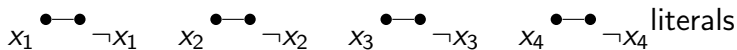
**Input:** undirected  $G$ , integer  $k$

**Question:** is there a subset  $X \subseteq V(G)$  of size at most  $k$ , such that for each  $uv \in E(G)$  we have  $\{u, v\} \cap X \neq \emptyset$ .



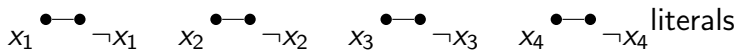
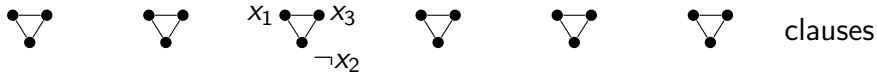
# Vertex Cover

Let us inspect the standard reduction from 3-SAT to Vertex Cover.



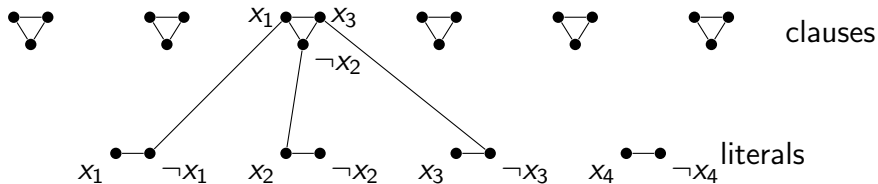
# Vertex Cover

Let us inspect the standard reduction from 3-SAT to Vertex Cover.



# Vertex Cover

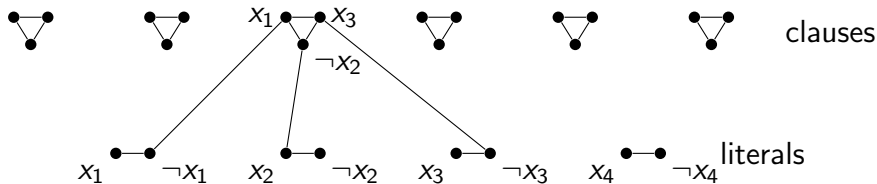
Let us inspect the standard reduction from 3-SAT to Vertex Cover.





# Vertex Cover

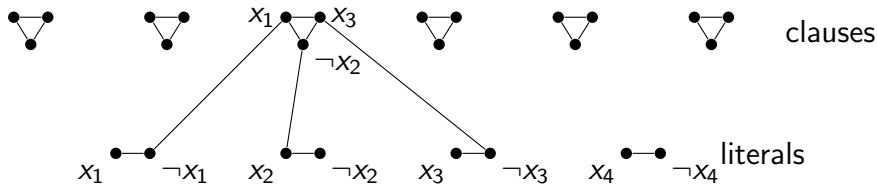
Let us inspect the standard reduction from 3-SAT to Vertex Cover.



The formula is satisfiable iff the created graph has vertex cover of size  $n + 2m$ .

# Vertex Cover

Let us inspect the standard reduction from 3-SAT to Vertex Cover.



The formula is satisfiable iff the created graph has vertex cover of size  $n + 2m$ .

Created instance has  $O(n + m)$  vertices and edges.

## Lemma

Unless ETH fails, there is no  $2^{o(n+m)}$  time algorithm for 3-SAT.

## Reduction

Given a 3-SAT formula  $\varphi$  with  $n$  variables and  $m$  clauses one can in polynomial time create a graph with  $O(n + m)$  vertices, which has a vertex cover of size  $2n + m$  iff  $\varphi$  is satisfiable.

## Lemma

Unless ETH fails, there is no  $2^{o(n+m)}$  time algorithm for 3-SAT.

## Reduction

Given a 3-SAT formula  $\varphi$  with  $n$  variables and  $m$  clauses one can in polynomial time create a graph with  $O(n + m)$  vertices, which has a vertex cover of size  $2n + m$  iff  $\varphi$  is satisfiable.

## Corollary

Unless ETH fails, there is no  $2^{o(|V(G)|+|E(G)|)}$  time algorithm VC.

## Lemma

Unless ETH fails, there is no  $2^{o(n+m)}$  time algorithm for 3-SAT.

## Reduction

Given a 3-SAT formula  $\varphi$  with  $n$  variables and  $m$  clauses one can in polynomial time create a graph with  $O(n + m)$  vertices, which has a vertex cover of size  $2n + m$  iff  $\varphi$  is satisfiable.

## Corollary

Unless ETH fails, there is no  $2^{o(|V(G)|+|E(G)|)}$  time algorithm VC.

## Corollary 2

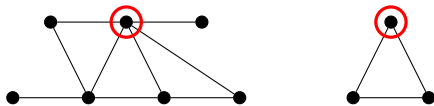
Unless ETH fails, there is no  $\mathcal{O}^*(2^{o(k)})$  time FPT algorithm for VC.

# FVS

## Feedback Vertex Set (FVS)

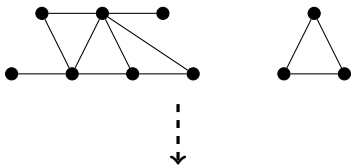
**Input:** undirected  $G$ , integer  $k$

**Question:** is there a subset  $X \subseteq V(G)$  of size at most  $k$ , such that  $G \setminus X$  is a forest

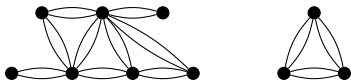


# FVS

VC instance

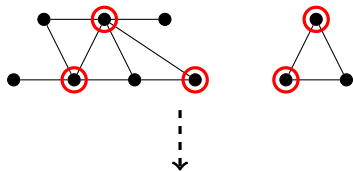


FVS instance

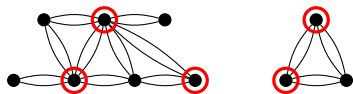


# FVS

VC instance

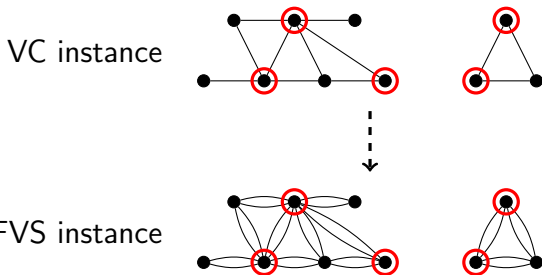


FVS instance





# FVS



## Lemma

Unless ETH fails, there is neither  $2^{o(n+m)}$  nor  $\mathcal{O}^*(2^{o(k)})$  time algorithm for FVS.

## Theorem

Unless ETH fails, there is no  $2^{o(n+m)}$  time algorithm for:

- Vertex Cover,
- Feedback Vertex Set,
- Odd Cycle Transversal (exercise),
- Hamiltonian Cycle,
- 3-Coloring (exercise).

# Planar lower bounds

## Planar Vertex Cover

**Input:** undirected **planar**  $G$ , integer  $k$

**Question:** is there a VC of size at most  $k$ .

# Planar lower bounds

## Planar Vertex Cover

**Input:** undirected **planar**  $G$ , integer  $k$

**Question:** is there a VC of size at most  $k$ .

Planar graphs have treewidth  $\mathcal{O}(\sqrt{n})$  (algorithmic proof, Wed).

# Planar lower bounds

## Planar Vertex Cover

**Input:** undirected **planar**  $G$ , integer  $k$

**Question:** is there a VC of size at most  $k$ .

Planar graphs have treewidth  $\mathcal{O}(\sqrt{n})$  (algorithmic proof, Wed).

VC can be solved in time  $\mathcal{O}^*(2^{\text{tw}})$ ,  
hence Planar VC can be solved in time  $2^{\mathcal{O}(\sqrt{n})}$ .

# Planar lower bounds

## Planar Vertex Cover

**Input:** undirected **planar**  $G$ , integer  $k$

**Question:** is there a VC of size at most  $k$ .

Planar graphs have treewidth  $\mathcal{O}(\sqrt{n})$  (algorithmic proof, Wed).

VC can be solved in time  $\mathcal{O}^*(2^{\text{tw}})$ ,  
hence Planar VC can be solved in time  $2^{\mathcal{O}(\sqrt{n})}$ .

Bidimensionality (Wed): planar VC can be solved in time  $\mathcal{O}^*(2^{\sqrt{k}})$ .

# Planar lower bounds

How can we prove  $2^{\Theta(\sqrt{n})}$  lower bound for Planar VC?

# Planar lower bounds

How can we prove  $2^{\Theta(\sqrt{n})}$  lower bound for Planar VC?

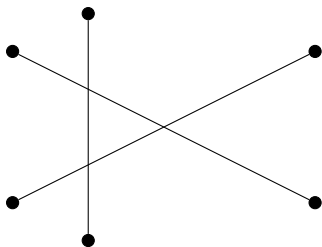
Idea: transform a Vertex Cover instance into a planar one, increasing the size to at most  $\mathcal{O}((n + m)^2)$ .



# Planar lower bounds

How can we prove  $2^{\Omega(\sqrt{n})}$  lower bound for Planar VC?

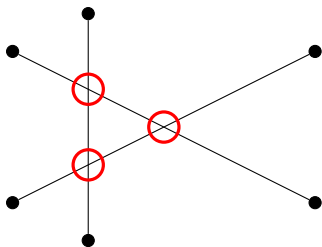
Idea: transform a Vertex Cover instance into a planar one, increasing the size to at most  $\mathcal{O}((n + m)^2)$ .



# Planar lower bounds

How can we prove  $2^{\Theta(\sqrt{n})}$  lower bound for Planar VC?

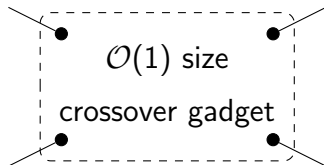
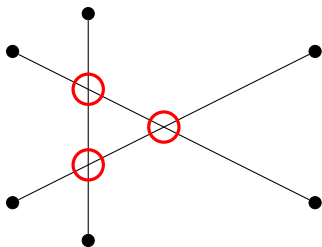
Idea: transform a Vertex Cover instance into a planar one, increasing the size to at most  $\mathcal{O}((n + m)^2)$ .



# Planar lower bounds

How can we prove  $2^{\Omega(\sqrt{n})}$  lower bound for Planar VC?

Idea: transform a Vertex Cover instance into a planar one, increasing the size to at most  $\mathcal{O}((n + m)^2)$ .



# ETH lower bounds methodology

## Observation

In order to exclude an algorithm with running time  $\mathcal{O}^*(2^{o(f(k))})$ , we need to provide a reduction from 3-SAT to  $B$  that outputs instances with parameter  $k$  bounded by  $g(n + m)$ , where  $g$  is the inverse of  $f$ .

# ETH lower bounds methodology

## Observation

In order to exclude an algorithm with running time  $\mathcal{O}^*(2^{o(f(k))})$ , we need to provide a reduction from 3-SAT to  $B$  that outputs instances with parameter  $k$  bounded by  $g(n + m)$ , where  $g$  is the inverse of  $f$ .

- for  $\mathcal{O}^*(2^{o(k)})$  we need  $k = \mathcal{O}(n + m)$ ,
- for  $\mathcal{O}^*(2^{o(\sqrt{k})})$  we need  $k = \mathcal{O}((n + m)^2)$ ,
- for  $\mathcal{O}^*(2^{o(k^2)})$  we need  $k = \mathcal{O}(\sqrt{n + m})$ ,
- for  $\mathcal{O}^*(2^{2^{o(k)}})$  we need  $k = \mathcal{O}(\log(n + m))$ .

# ETH for $W[1]$ -hard problems

Theorem (implies  $W[1] \neq FPT$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

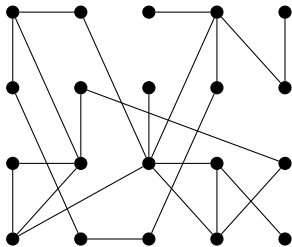
## Theorem (implies $W[1] \neq \text{FPT}$ )

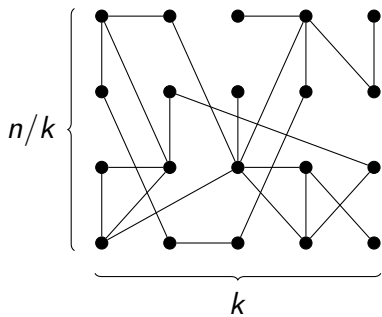
Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

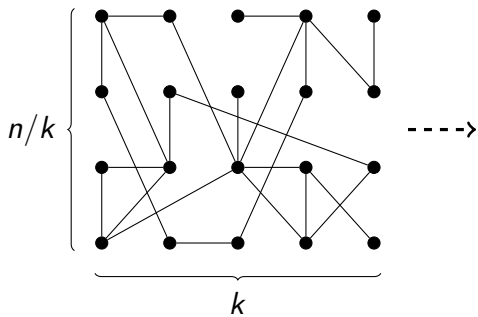
Proof:

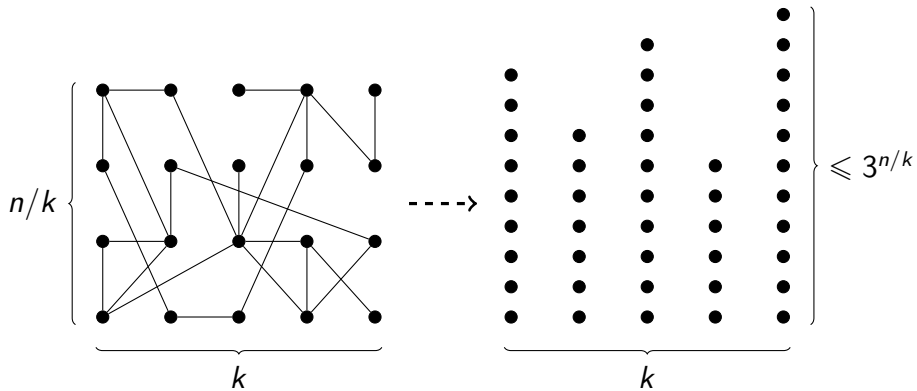
- A reduction from 3-Coloring (for which there is no  $2^{\mathcal{O}(n+m)}$  time algorithm under ETH).
- In 3-Coloring we have  $n$  variables with 3 possible choices each.
- In Clique we have  $k$  variables with  $N$  possible choices each.
- We want  $3^n \approx N^k$ , so  $N \approx 3^{n/k}$ .
- But how to choose  $k$ ? (decide later, but it has to be  $\omega(1)$ )

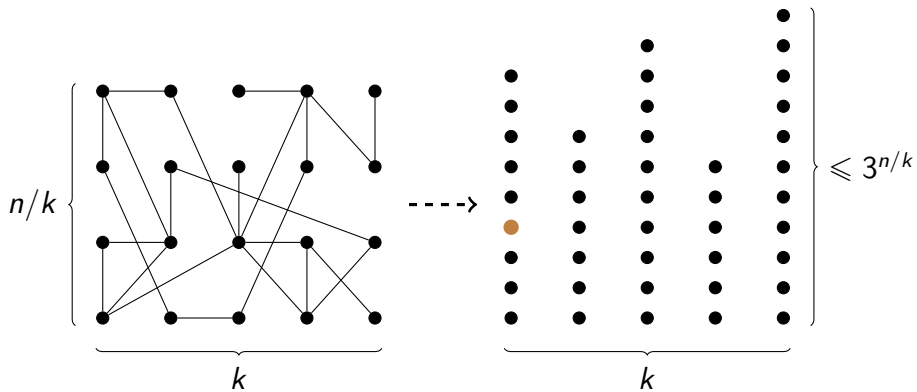


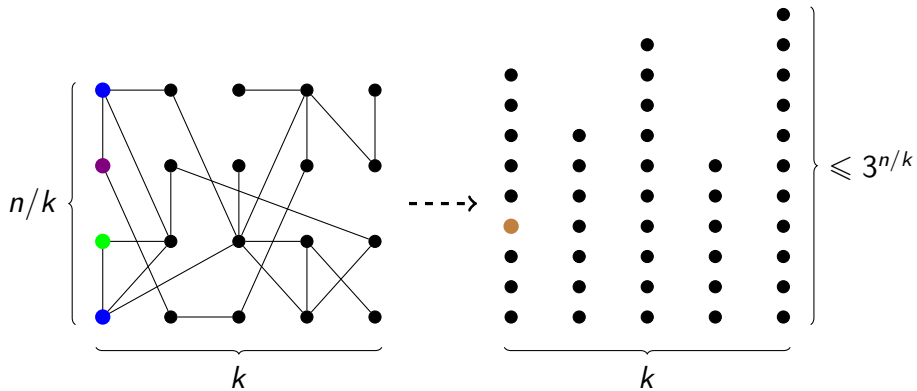


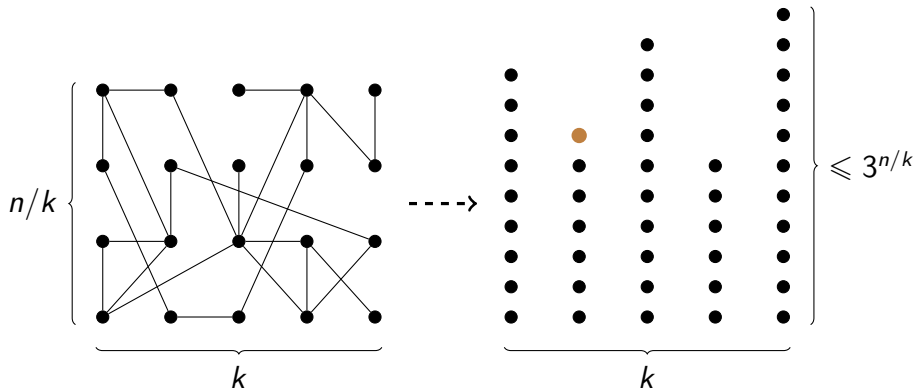


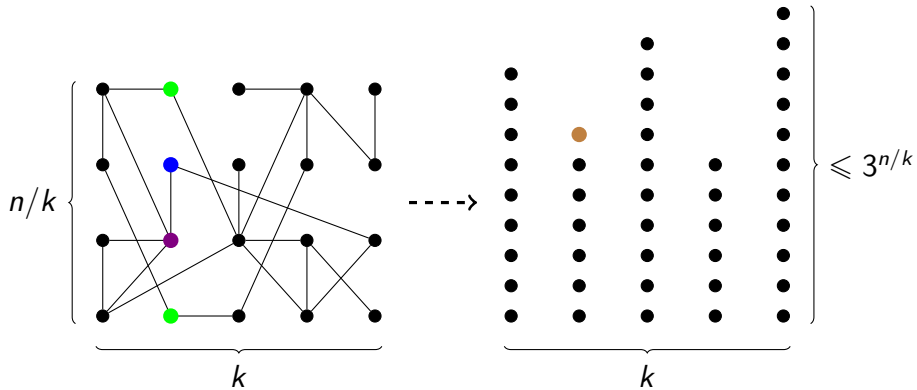




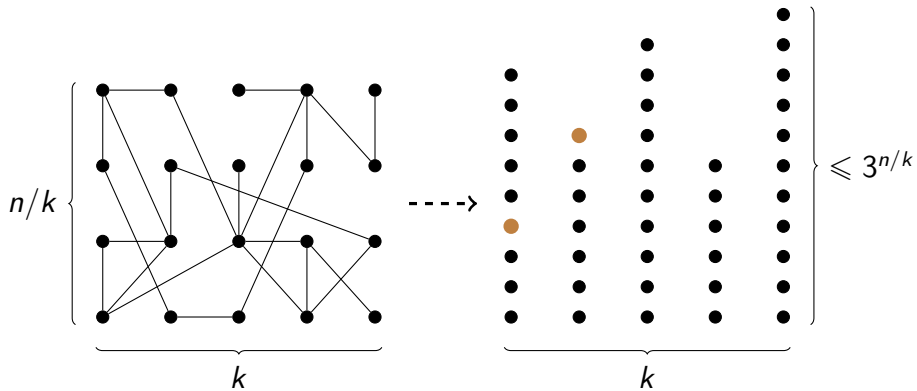


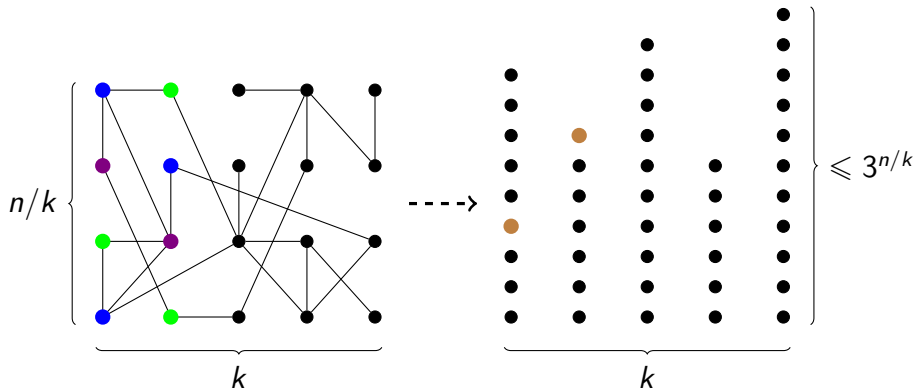


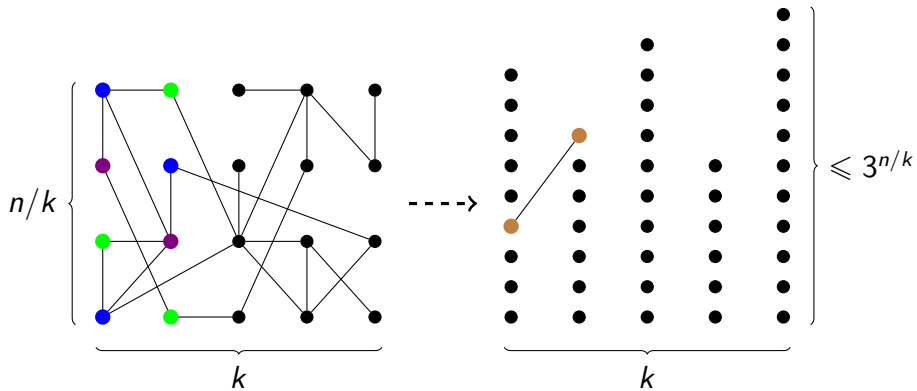


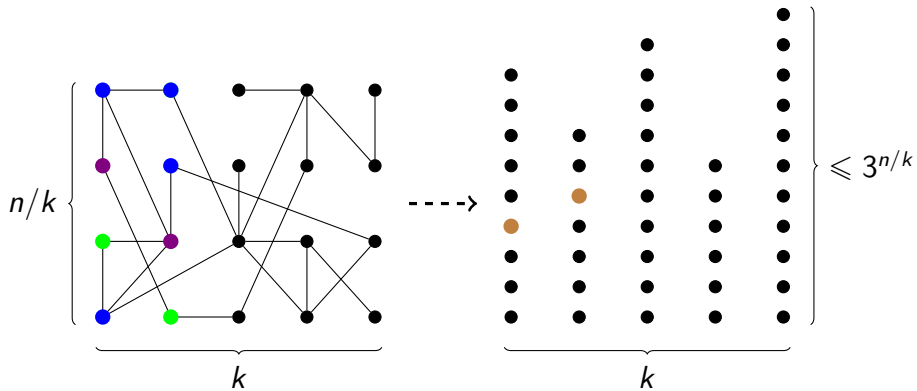


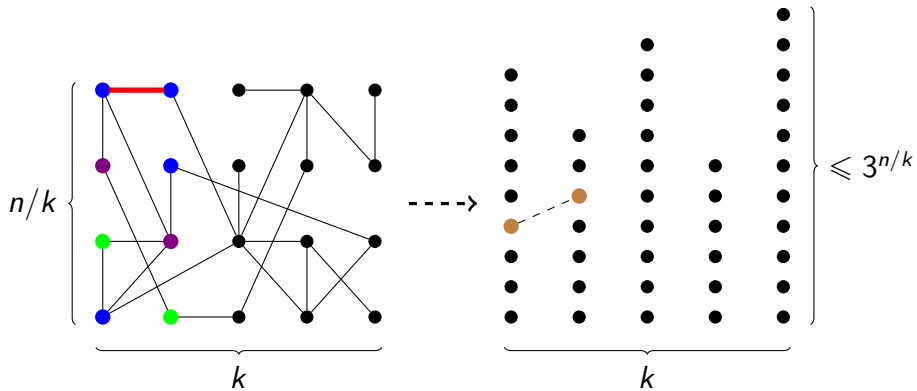


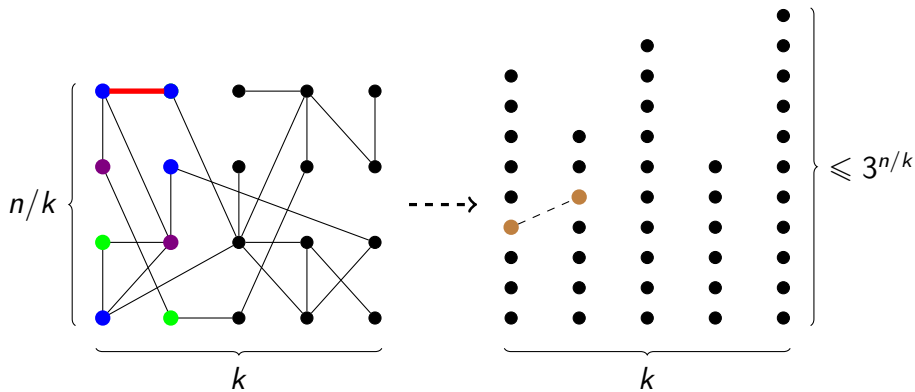












Left graph admits 3-coloring iff right graph contains  $k$ -clique.

Theorem (implies  $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

Proof continued:

- Created graph has  $N = k \cdot 3^{n/k}$  vertices.

Theorem (implies  $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

Proof continued:

- Created graph has  $N = k \cdot 3^{n/k}$  vertices.
- Lets try  $k = \log n$ .



## Theorem (implies $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

Proof continued:

- Created graph has  $N = k \cdot 3^{n/k}$  vertices.
- Lets try  $k = \log n$ .
- $2^k \cdot N^{o(k)} = n \cdot (\log n)^{\log n} \cdot 3^{n \cdot o(\log n) / \log n} = \mathcal{O}^*(2^{o(n)})$ .

## Theorem (implies $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

Proof continued:

- Created graph has  $N = k \cdot 3^{n/k}$  vertices.
- Lets try  $k = \log n$ .
- $2^k \cdot N^{o(k)} = n \cdot (\log n)^{\log n} \cdot 3^{n \cdot o(\log n) / \log n} = \mathcal{O}^*(2^{o(n)})$ .
- Similarly  $k = \log \log n$  implies no  $2^{2^k} N^{o(k)}$  time algorithm.

## Theorem (implies $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

Proof continued:

- Created graph has  $N = k \cdot 3^{n/k}$  vertices.
- Lets try  $k = \log n$ .
- $2^k \cdot N^{o(k)} = n \cdot (\log n)^{\log n} \cdot 3^{n \cdot o(\log n) / \log n} = \mathcal{O}^*(2^{o(n)})$ .
- Similarly  $k = \log \log n$  implies no  $2^{2^k} N^{o(k)}$  time algorithm.
- To exclude all computable  $f(k)$ , one needs roughly  $f^{-1}(n)$  (technical difficulties omitted).

## Theorem (implies $W[1] \neq FPT$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

- A reduction from Clique to  $X$  which outputs parameter  $g(k)$  implies lower bound of  $f(k)n^{o(g^{-1}(k))}$ .

## Theorem (implies $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

- A reduction from Clique to  $X$  which outputs parameter  $g(k)$  implies lower bound of  $f(k)n^{o(g^{-1}(k))}$ .
- Reductions from Clique to Multicolored Clique, Independent Set, and Dominating Set from lecture 4 result in instances with parameters  $\mathcal{O}(k)$ .

## Theorem (implies $W[1] \neq \text{FPT}$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

- A reduction from Clique to  $X$  which outputs parameter  $g(k)$  implies lower bound of  $f(k)n^{o(g^{-1}(k))}$ .
- Reductions from Clique to Multicolored Clique, Independent Set, and Dominating Set from lecture 4 result in instances with parameters  $\mathcal{O}(k)$ .
- Consequently (under ETH) there is no  $f(k)n^{o(k)}$  algorithm for those problems.

## Theorem (implies $W[1] \neq FPT$ )

Unless ETH fails, there is no  $f(k)n^{o(k)}$  time algorithm for Clique.

- A reduction from Clique to  $X$  which outputs parameter  $g(k)$  implies lower bound of  $f(k)n^{o(g^{-1}(k))}$ .
- Reductions from Clique to Multicolored Clique, Independent Set, and Dominating Set from lecture 4 result in instances with parameters  $\mathcal{O}(k)$ .
- Consequently (under ETH) there is no  $f(k)n^{o(k)}$  algorithm for those problems.
- For Odd Set the reduction does not give so strong lower bound (Ex: check what is implied by the reduction from lecture 4).

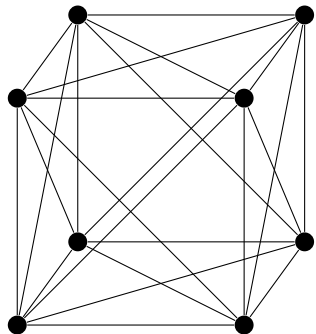
ETH used for somewhat  
exotic lower bounds



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

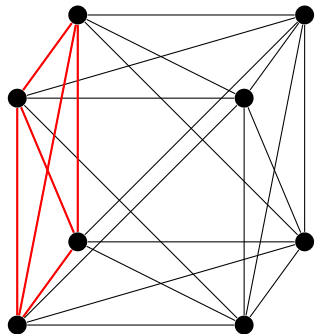
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

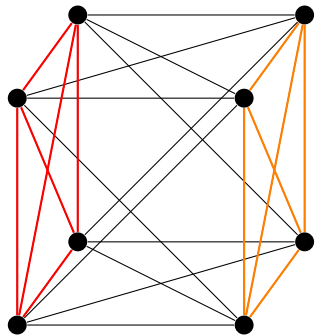
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

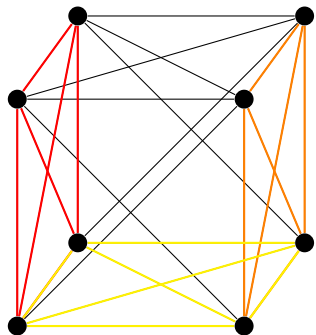
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

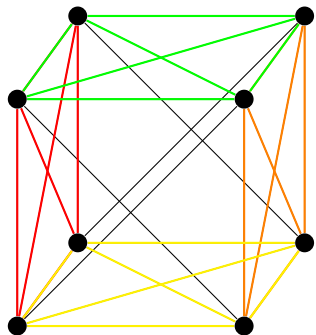
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

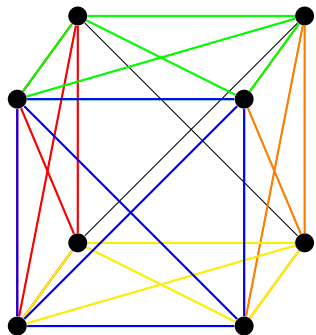
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

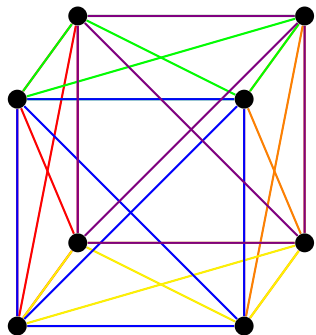
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

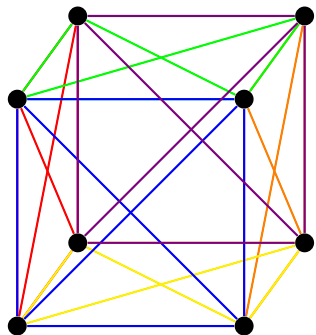
**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



## Edge Clique Cover (ECC)

**Input:** Graph  $G$ , integer  $k$

**Question:** Does there exist a collection of  $k$  cliques  $C_1, C_2, \dots, C_k$  in  $G$ , such that  $E(G) = \bigcup_{i=1}^k E(C_i)$ ?



6 cliques, but not optimal

puzzle: find a solution with 5 cliques



# ECC

Book, section 2.3.3

Edge Clique Cover admits a kernel with at most  $2^k$  vertices.

# ECC

Book, section 2.3.3

Edge Clique Cover admits a kernel with at most  $2^k$  vertices.

Exercise

By casting the problem as Set Cover, we obtain  $\mathcal{O}^*(2^{2^{\mathcal{O}(k)}})$  algorithm.

# ECC

Book, section 2.3.3

Edge Clique Cover admits a kernel with at most  $2^k$  vertices.

Exercise

By casting the problem as Set Cover, we obtain  $\mathcal{O}^*(2^{2^{\mathcal{O}(k)}})$  algorithm.

Theorem

Unless ETH fails, there is no  $\mathcal{O}^*(2^{2^{\mathcal{O}(k)}})$  time algorithm for ECC.  
Proof by reducing formula with  $n$  vars,  $m$  clauses to poly size ECC instance with  $k = \mathcal{O}(\log(n + m))$ .

# Closest Substring

## Closest Substring

**Input:** strings  $s_1, \dots, s_t \in \Sigma^*$ , integers  $L, d$ .

**Question:** Is there  $s \in \Sigma^L$ , such that for every  $1 \leq i \leq t$  there is a substring  $s'_i$  of  $s_i$  of length  $L$  with  $d_H(s, s'_i) \leq d$ .

# Closest Substring

## Closest Substring - important parts

**Input:**  $t$  strings of total size  $n$

**Question:** fairly natural question about Hamming distance

# Closest Substring

## Closest Substring - important parts

**Input:**  $t$  strings of total size  $n$

**Question:** fairly natural question about Hamming distance

## Lemma

Closest Substring over constant size alphabet can be solved in time  $f(d, t)n^{\mathcal{O}(\log \log t)}$ . Mover under ETH there is no  $f(d, t)n^{\mathcal{O}(\log \log t)}$  time algorithm for any  $f$ .

# Summary

## Take-home message 1

$\text{SETH} \implies \text{ETH} \implies \text{W}[1] \neq \text{FPT} \implies \text{P} \neq \text{NP}$

$\text{FPT} \neq \text{W}[1] \implies \text{no } \mathcal{O}^*(f(k))$

$\text{ETH} \implies \text{no } \mathcal{O}^*(2^{o(f(k))})$

$\text{SETH} \implies \text{no } \mathcal{O}^*(2^{(1-\epsilon)f(k)})$ .

## Take-home message 2

When showing lower bounds, we have to come up with a reduction and track what happens with the parameter in the new instance.

# Summary

- We have learned that ETH is a tool that allows proving exciting lower bounds.
- More on this subject on Thursday (lectures by Michał and Daniel L.).
- Exercises: 14.1, 14.2.(d) (+OCT), show  $\mathcal{O}^*(2^{2^{O(k)}})$  algorithm for ECC (after kernelization), check what lower bound (under ETH) is implied for Odd Set by reduction from lecture 4.

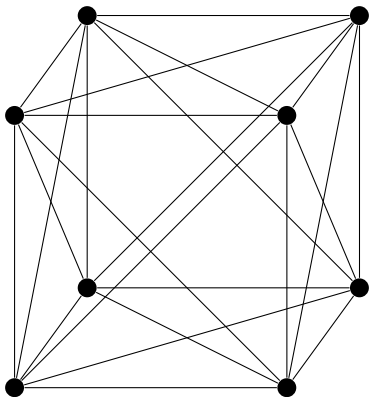


# Summary

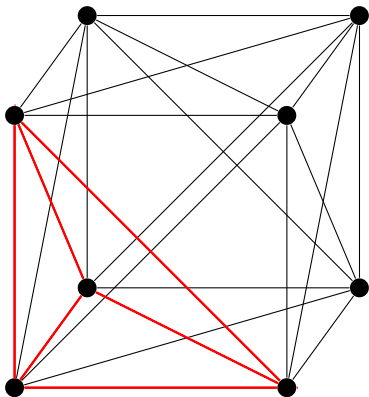
- We have learned that ETH is a tool that allows proving exciting lower bounds.
- More on this subject on Thursday (lectures by Michał and Daniel L.).
- Exercises: 14.1, 14.2.(d) (+OCT), show  $\mathcal{O}^*(2^{2^{O(k)}})$  algorithm for ECC (after kernelization), check what lower bound (under ETH) is implied for Odd Set by reduction from lecture 4.

Thank you for your attention

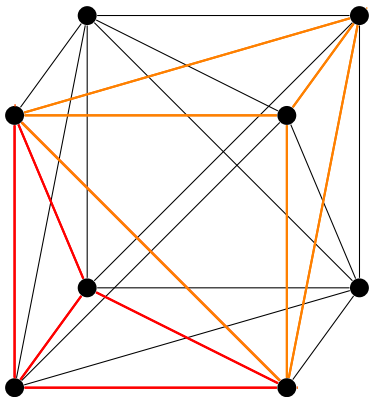
# Hidden slide



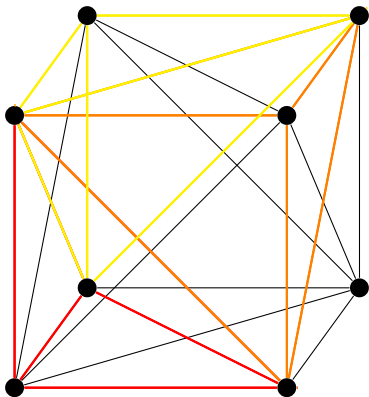
# Hidden slide



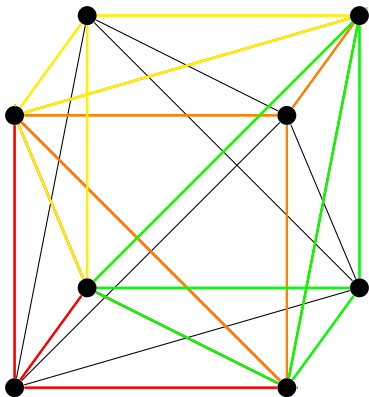
# Hidden slide



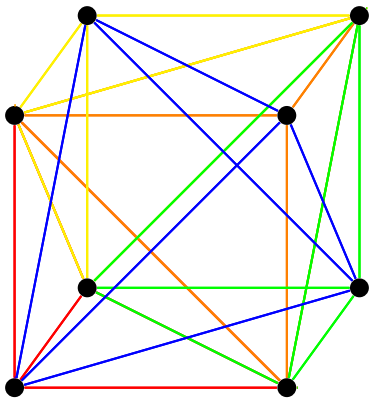
# Hidden slide



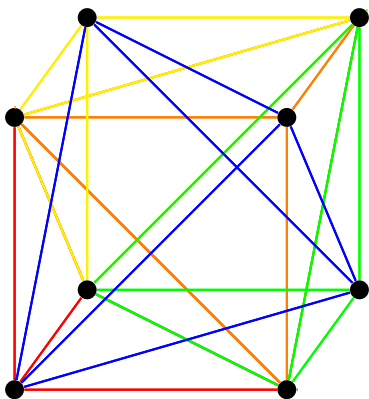
# Hidden slide



# Hidden slide



# Hidden slide



Magical solution with 5 cliques