

# CSL7910: Special Topics in Computer Science 1

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## Problem Statement

In the Cluster Vertex Deletion problem, we are given an undirected graph  $G$  and an integer  $k$ , and the task is to find a set  $X$  of at most  $k$  vertices of  $G$  such that  $G - X$  is a cluster graph (a disjoint union of cliques). Using iterative compression, obtain an algorithm for Cluster Vertex Deletion running in time  $2^k n^{O(1)}$ .

**Input:** Undirected Graph  $G(V,E)$

**Parameter:**  $k$ , the number of vertex deletions

**Output:**  $\{X \mid X \subseteq V, |X| \leq k, G-X \text{ is a cluster graph}\}$

# Cluster Graph

A **cluster graph** is a graph formed from the disjoint union of complete graphs.

In other words, a graph  $G$  is a cluster graph if every connected component of  $G$  is a clique [2.5 problem]

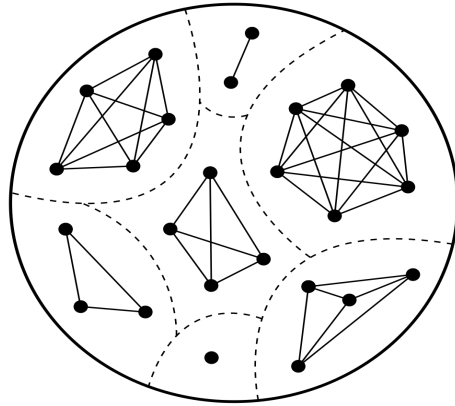


Image: Wikipedia

# Is solving Cluster Graph an NP Complete problem?

Polynomial Time algorithm to find if the given graph is a cluster graph.

Find the components of the input graph. A check if there is an edge between every pair of vertices in the component.

# CVD is NP Complete

In the Cluster Vertex Deletion problem, we are given a graph  $G$  and an integer  $k$ , and the task is to find a set  $X$  of at most  $k$  vertices of  $G$  such that  $G - X$  is a cluster graph

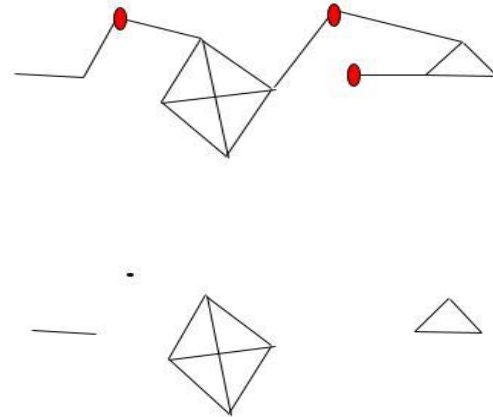
Brute Force Approach:

For  $i = 0, 1, 2, \dots, K$  remove  $i$  vertices

and check if it is a cluster graph

Time complexity:  $O(n^k n^2)$

[! Not of the form  $f(k) \cdot n^c$ ]

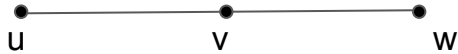


# Designing an Fixed Parameter Algorithm(FPT)

## Lemma 1:

A graph  $G$  is a cluster graph if and only if it does not have an induced path on three vertices

(sequence of three vertices  $u, v, w$  such that  $uv$  and  $vw$  are edges and  $uw \notin E(G)$ ).



## Proof:

If  $P_3$  is an induced subgraph of the given cluster graph, but  $uw$  is not an edge which contradicts that induced subgraph of a complete graph is a clique.  $P_3$  can't be in a clique.

If the cluster graph does not contain a substructure  $P_3$ , then every connected component is a clique



# Branching

If  $k=0$ , check if  $G$  is a cluster graph

If Yes:

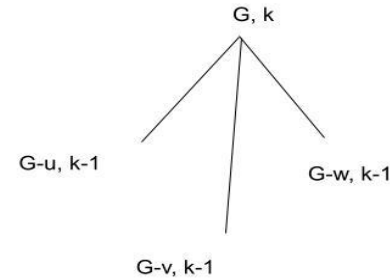
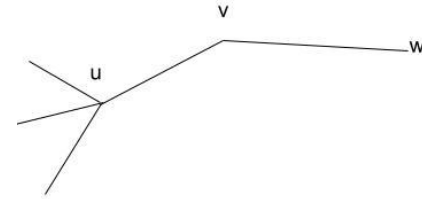
Return Yes

Else:

Find the substructure  $P_3$  in  $G$

Branch on each of the 3 vertices

Time complexity:  $O(3^k \cdot n^2)$



# Kernelization

Reduce the CVD problem into 3-Hitting Set Problem

Every induced path  $P_3$  is 3-element set. The problem becomes finding a hitting set that intersect at most  $k$  of these 3-element set

Kernelization results for 3-Hitting Set [2] is an  $O(k^2)$  - vertex problem kernel for unweighted Cluster Vertex Deletion, kernel can be found in  $O(n^3)$  time.

Time complexity:  $O(2.08^k + n^3)$



# Iterative Compression

Idea: Given a problem instance and corresponding solution of size  $k+1$ , either compress it to a solution of at most size  $k$  or prove that given solution is of minimum size and that there is no solution of size at most  $k$

Does it work for Cluster Vertex Deletion?

The induced subgraph of a cluster graph is also a cluster graph.

If we can't find a CVD of size at most  $k$ , in a subgraph induced by  $k+1$  vertices. Then, we can't find a CVD in the entire graph.

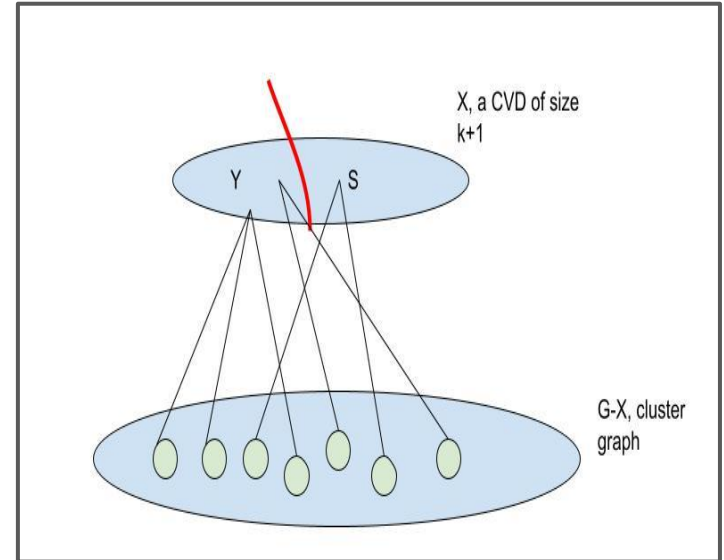
# Iterative Compression (contd.)

Given a ' $k+1$ ' sized solution, guess a subset  $Y$  of size ' $l$ '.

If  $G[S]$  induces a cluster graph then

add  $Y$  to the Solution of CVD and Delete  $Y$

Find a CVD of size atmost  $k+1-l$  which is DISJOINT from  $S$

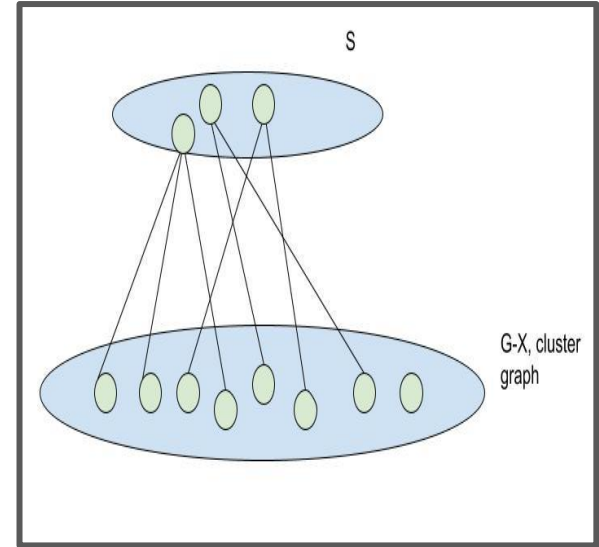


# DISJOINT PROBLEM

**Input:** An undirected graph  $G = (V, E)$ , and

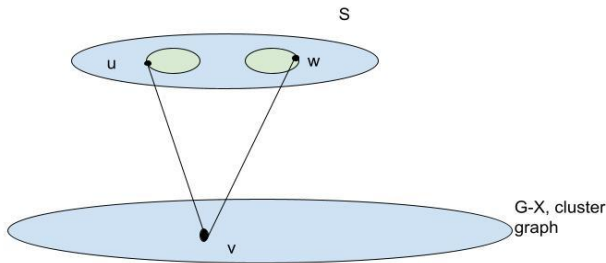
a vertex set  $S \subseteq V$  such that  $G[S]$  and  $G \setminus S$  are cluster graphs.

**Task:** Find a vertex set  $C \subseteq V \setminus S$  such that  $G \setminus C$  is a cluster graph



# Reduction Rule 1

What happens if  $\exists v \in V \setminus S$  that has neighbours in two cliques in  $S$ ?

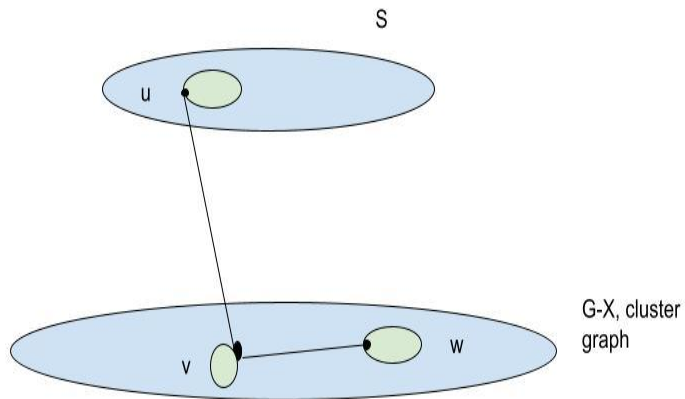


**RR1:** Delete all vertices in  $V \setminus S$  that are adjacent to more than one cluster in  $G[S]$ .

**POC:** If a vertex  $v \in V \setminus S$  is adjacent to vertices  $u$  and  $w$  in different clusters in  $S$ , then  $uvw$  induces a  $P_3$  that can only be removed by deleting  $v$ .  $uw$  can't be included

# Reduction Rule 2

What happens if  $\exists v \in V \setminus S$  that has some neighbours in  $S$  but not all vertices of the clique in  $S$



**RR2:** Delete all vertices in  $V \setminus S$  that are adjacent to some, but not all vertices of a cluster in  $X$ .

**POC:** If a vertex  $v \in R$  is adjacent to a vertex  $u$ , but not to a vertex  $w$  in a cluster in  $X$ , then  $uwv$  induces a  $P_3$ , which can only be removed by deleting  $v$

# Reduction Rule 3

**RR3:** Remove connected components that are complete graphs

**POC:** No optimal solution deletes vertices in such components

# What are we left with?

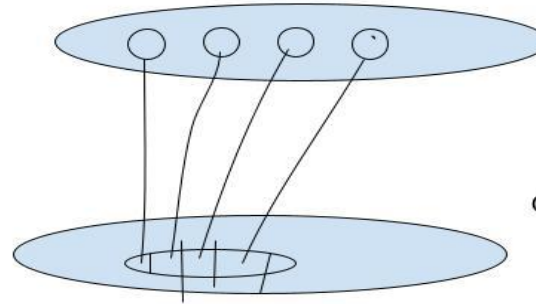
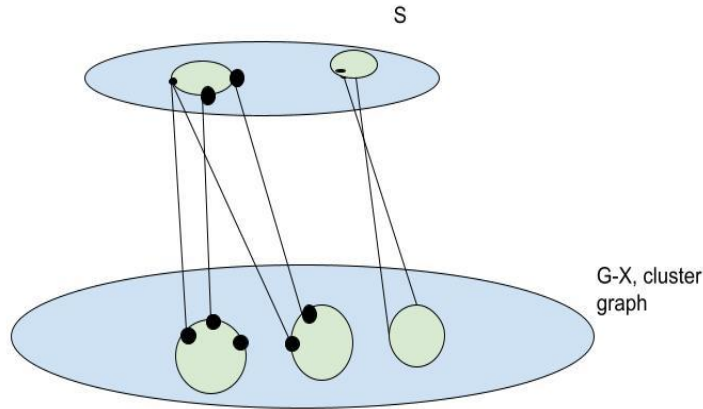
1. Removed all the vertices in cliques of  $G \setminus S$  that had neighbours in different cliques
2. Removed all the vertices in cliques had few neighbours in the clique
3. Removed components that are complete graphs

Any vertex in  $V \setminus S$  will be adjacent to exactly one clique entirely (all vertices) or none of the cliques.

# Reduction Rule 4

Take a clique in  $G-X$ , partition its vertices into sets of vertices that are neighbours of  $c_1, c_2, \dots, c_l$  or none.

We can select only one of the set partitions of  $B$  otherwise it will create a P3

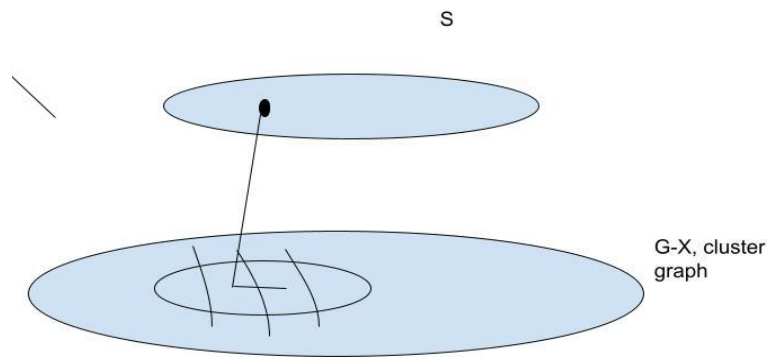




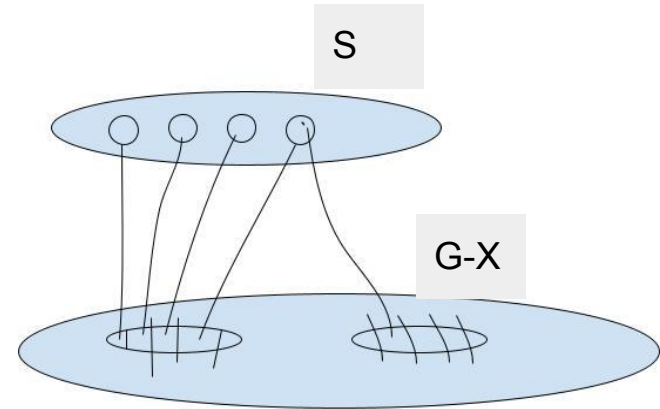
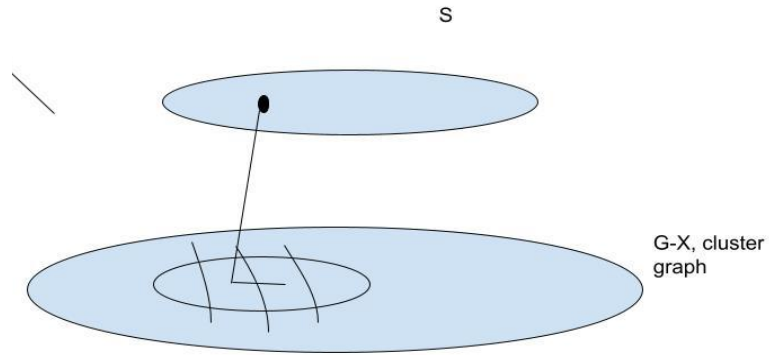
## Lemma 2.

In CVD compression solution, for each cluster in  $G[V \setminus X]$ , the vertices of at most one class are present.

**Proof:** If  $v \in R$  is adjacent to some  $w \in S$ , and  $u$  is a vertex from the same cluster as  $v$ , but from a different class, then  $uvw$  is a  $P_3$ ; therefore, we cannot keep vertices from two different classes within a cluster.

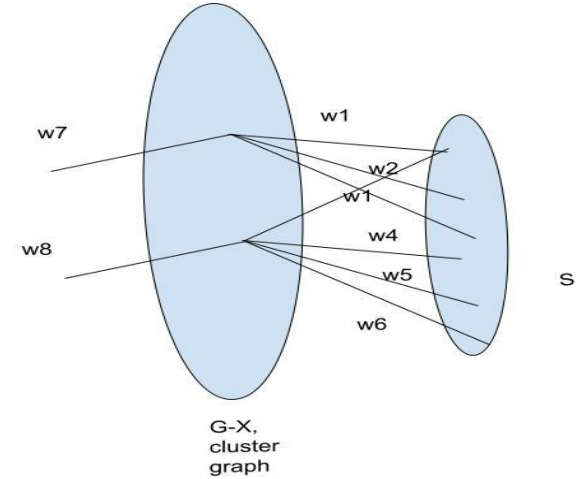


# Maximum Matching in Bipartite Graph



## Construction of the Bipartite Graph H

1. Add a vertex for every cluster in  $G-X$
  2. Add a vertex for every cluster in  $G[S]$
  3. For every class that is adjacent to a cluster in  $G[S]$  add an edge
  4. If there is a class in  $G-X$  that is not adjacent to any cluster in  $G[S]$
- add an extra vertex and connected it the cluster/vertex it is part of



# Lemma 3

A maximum bipartite matching in  $H$  results in a Maximum Induced Cluster Graph

**Proof:**

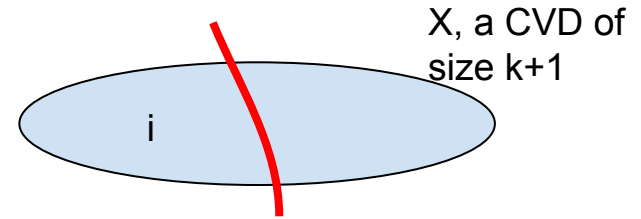
Each edge in a matching corresponds to a class in a cluster of  $G[V \setminus X]$ . The CVD compression solution is to delete all vertices in  $G \setminus X$  but those of the selected classes. The matching cannot select two classes within the same cluster, since the corresponding edges have an endpoint in common;

Similarly, it cannot select two classes that share a connection to the same cluster in  $G[S]$ . Therefore, a matching yields a feasible solution. By Lemma 3, an CVD compression solution corresponds to an assignment of each cluster to one of its classes or to nothing, and therefore, it corresponds to a matching. Finally, the weight of a matching corresponds to the number vertices not deleted from  $V \setminus X$ , and therefore a maximum matching corresponds a maximum induced cluster graph

# Analysis

The problem of finding matching in a bipartite graph can be found in polynomial time.

$$\sum_{i=0}^{k+1} \binom{k+1}{i} n^i \leq 2^{k+1} n^{k+1}$$



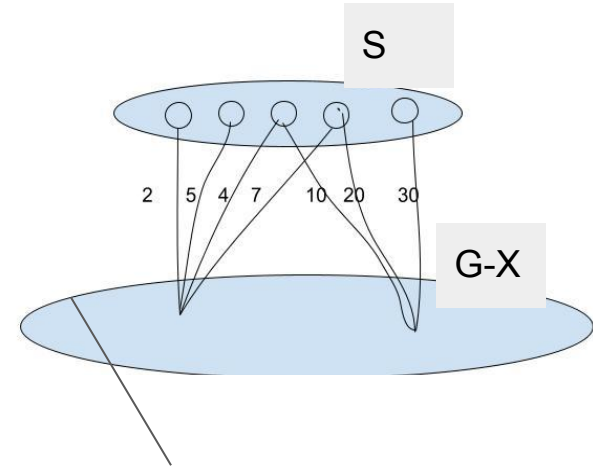
# Application of CVD

Remove spurious relation and the remaining is a cluster graph(DNA)

Largest clique in a graph

# Variants of CVD in FPT

1. Weighted cluster vertex deletion
2. d-cluster vertex deletion problem where the maximum number of cliques to be generated is already specified
3. Weighted d-cluster vertex deletion



# References

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