

# COMPSCI 720: Parameterized and approximation algorithms

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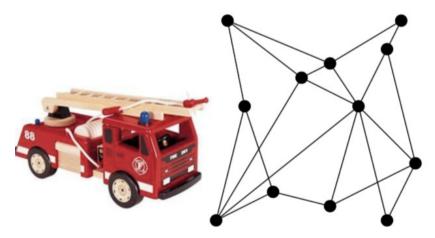
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Both books available online from UoA library website.

Fire station problem (Dominating Set)

Every city needs a fire station – or at least one in the neighboring city.

How many fire stations are needed?

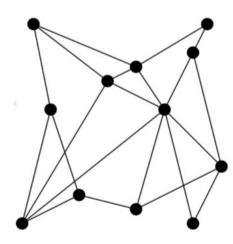


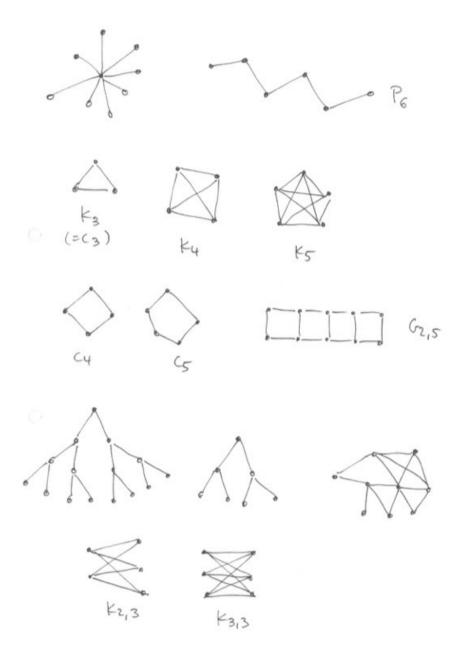
Fire station problem (Dominating Set)

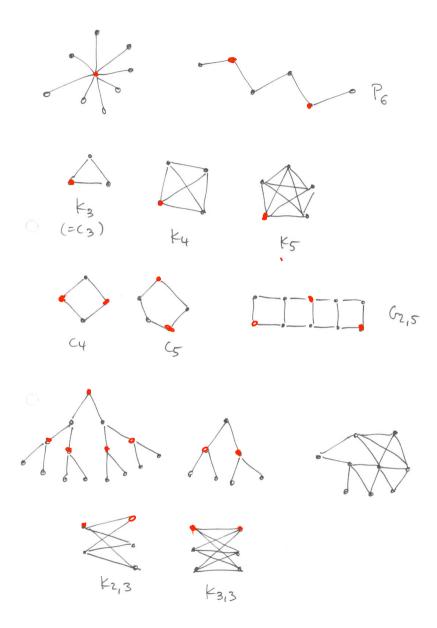
Model the problem as a graph:

Each city is a vertex and two vertices are connected if the two corresponding cities are neighbors of each other.

Example 1.







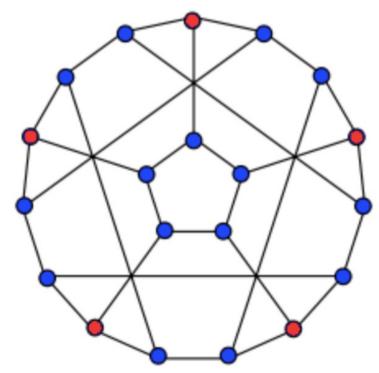
## Dominating Set (DS)

Instance. A graph G=(V, E) and a non-negative integer k.

Question. Is there a subset V' of V such that  $|V'| \le k$  and each vertex in  $V \setminus V'$  is adjacent to at least one vertex in V'?

Let V' contain the five red vertices. Then each vertex not in V' is adjacent to a red vertex.

Can we do better?

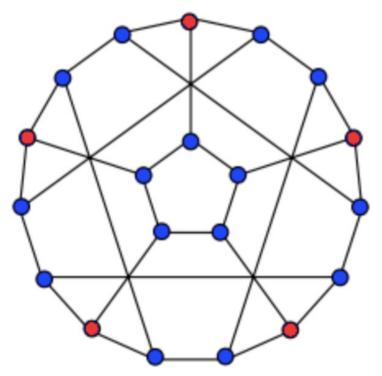


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Let V' contain the five red vertices. Then each vertex not in V' is adjacent to a red vertex. Can we do better? No, V' is a minimum-size dominating set



DS: easy or hard?

Easy – Then there is an algorithm that solves DS in **polynomial** running time. e.g.  $O(n^c)$ , where n is the number of vertices (cities) and c is a constant No such algorithm known for DS!

Hard – Then all known algorithms to solve DS have **exponential** running time. (Brute force: Check all  $2^n$  subsets of the vertices. No significantly better algorithm known!)

Indeed, DS is NP-complete!

## Running times

	Polynomial (e.g. n²)		Exponential (e.g. 2 <sup>n</sup> )	
	$n^2$	time	2 <sup>n</sup>	time
n=2	4	0.04 μs	4	0.04 μs
n=10	100	1 μs	1024	10 μs
n=50	2500	25 μs	10 <sup>15</sup>	116 days
n=100	10000	0.1 ms	10 <sup>30</sup>	3 (10 <sup>14</sup> ) years

## Dealing with NP-completeness

- Never a reason to give up!
- Parameterized algorithms
- Heuristics
- Approximation algorithms (later in course)

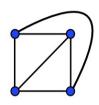


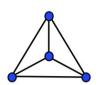
"I can't find an efficient algorithm, because no such algorithm is possible!"

Here: Focus on the design of parameterized algorithms (not so much the complexity theoretic background).

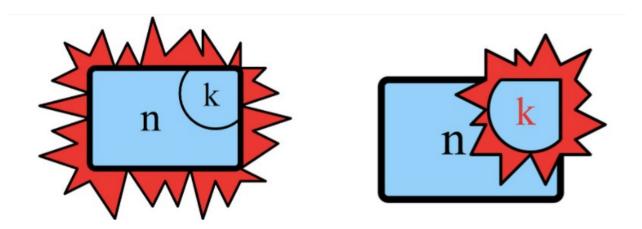
## Dominating Set on planar graphs (DS-p)







- NP-complete (no algorithm know that solves the problem in polynomial time).
- BUT: if there exists a *small* dominating set (if only a few fire stations are needed) then the problem can be solved *reasonably quickly*.
- Why? Running time is exponential but the exponential explosion is due to a particular parameter, here *k* (i.e. number of fire stations needed).
- Recognizing if a graph is planar can be done in polynomial time (Kuratowski's Theorem).

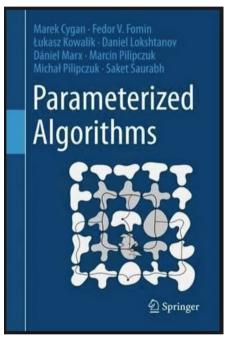


- If the parameter (k) is small, the running time is (still) exponential but often not too bad in practice.
- Name of the game in parameterized algorithmics: find parameters such that exponential explosion is restricted to these parameters (multidimensional approach, e.g. for DS-p: size of graph and size of solution). Traditional approach is one-dimensional, i.e. running time is measured with respect to input size only.
- Possible parameters: the size of a solution itself (e.g. for DS-p it is the minimum size of a dominating set), tree-width of a graph, diameter of a graph, maximum vertex-degree ....

#### Definitions

- A parameterized problem is a language  $L \subseteq \Sigma^* \times \mathbb{N}$ , where  $\Sigma$  is a fixed and finite alphabet. E.g.  $(x,k) \in \Sigma^* \times \mathbb{N}$ , where k is called the parameter.  $((x,k) \in L$  if and only if (x,k) is a yes-instance)
- A parameterized problem  $L \subseteq \Sigma^* \times \mathbb{N}$  is called fixed-parameter tractable (FPT) for parameter k if there exists an algorithm that solves the problem in time  $O(f(k) |x|^c)$ , i.e. one can decide in time  $f(k) |x|^c$  whether or not  $(x,k) \in L$ . [ Alternative definition uses  $f(k) + |x|^c$  ]

c constant; f(k) computable function only depends on k (e.g.  $2^k$ ,  $3^k$ ,  $2^{2^k}$ )



Chapter 2.

(1) Kernelization

Idea. pre-processing/reducing/shrinking the input data to the stuff that is difficult to solve

The next three slides are from a presentation by Professor Rod Downey (Victoria University of Wellington).

http://homepages.mcs.vuw.ac.nz/~downey/wa\_2011.pdf

https://royalsociety.org.nz/what-we-do/medals-and-awards/medals-and-awards-news/2018-rutherford-medal-solving-cant-compute-and-is-that-random-sequence-really-random/

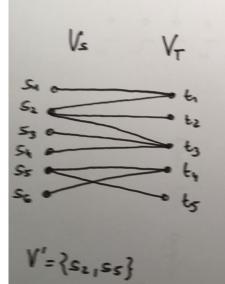


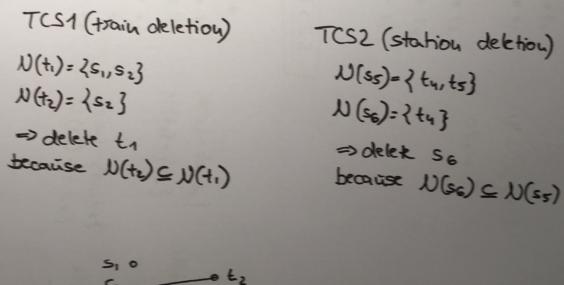
### KARSTEN WEIHE'S TRAIN PROBLEM

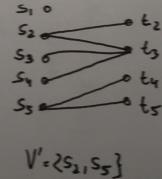
► TRAIN COVERING BY STATIONS Instance: A bipartite graph  $G = (V_S \cup V_T, E)$ , where the set of vertices  $V_S$  represents railway stations and the set of vertices  $V_T$  represents trains. E contains an edge  $(s,t), s \in V_s, t \in V_T$ , iff the train t stops at the station s. Problem: Find a minimum set  $V' \subseteq V_S$  such that V' covers  $V_T$ , that is, for every vertex  $t \in V_T$ , there is some  $s \in V'$ such that  $(s,t) \in E$ .

## WEIHE'S SOLUTION

- ▶ REDUCTION RULE TCS1: Let N(t) denote the neighbours of t in  $V_S$ . If  $N(t) \subseteq N(t')$  then remove t' and all adjacent edges of t' from G. If there is a station that covers t, then this station also covers t'.
- ▶ REDUCTION RULE TCS2: Let N(s) denote the neighbours of s in  $V_T$ . If  $N(s) \subseteq N(s')$  then remove s and all adjacent edges of s from G. If there is a train covered by s, then this train is also covered by s'.







- European train schedule, gave a graph consisting of around 1.6 · 10<sup>5</sup> vertices and 1.6 · 10<sup>6</sup> edges.
- Solved in minutes.
- This has also been applied in practice as a subroutine in practical heuristical algorithms.

#### Idea of kernelization

- in polynomial time, pre-process/reduce/shrink the input data to the stuff that is difficult to solve
- more precisely, reduce the parameterized problem to a kernel whose size depends solely on the parameter
- pre-processing is often used to design heuristics. But we want to do it in a controlled way, so that we get a provable performance guarantee
- want to design reductions such that, at each step, enough information is preserved to obtain an optimal solution

## Vertex Cover (VC)

Instance. A graph G=(V, E) and a non-negative integer k.

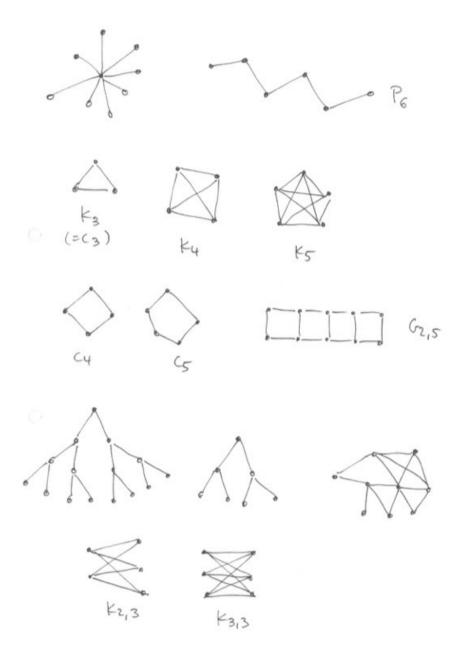
Question. Is there a subset V' of V such that  $|V'| \le k$  and for each edge  $\{u,v\}$  in E, at least one of u and v is in V'?

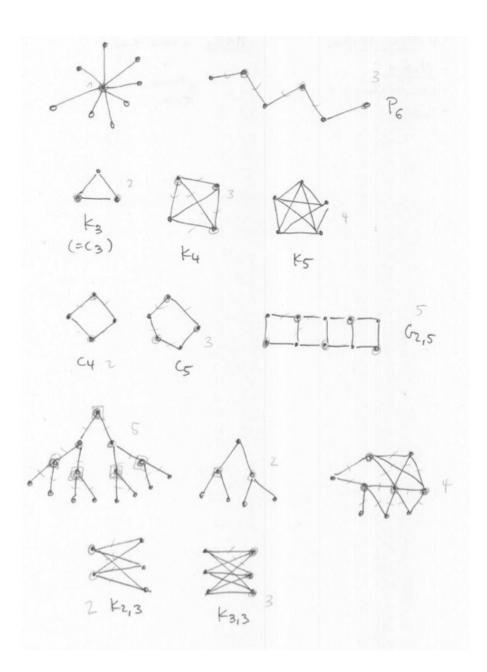
#### Subtle difference between VC and DS.

VC: Vertices cover edges.

DS: Vertices cover vertices.

#### Example 2.

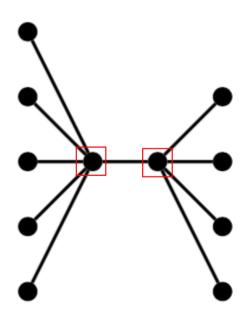




#### Observation.

A vertex cover *VC* of a graph *G* has the following property:

For every vertex v either v or all of its neighbours are contained in VC.



## Reductions for VC: Consider an instance (G=(V,E), k) of VC

 VC1: If (G, k) contains an isolated vertex v, delete v from G; new instance is (G-v, k)

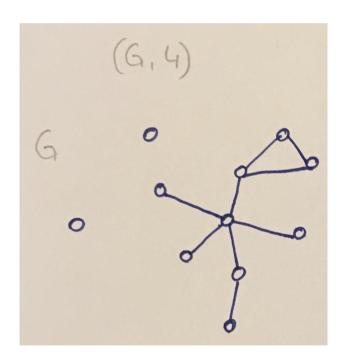
[safe because *v* does not cover any edge]

 VC2: If G has a vertex v of degree at least k+1, delete v from G and decrement k by one; new instance is (G-v, k-1)

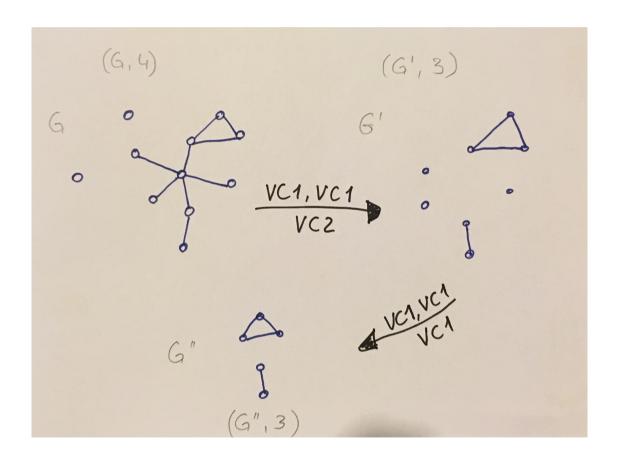
[safe because v must be in VC, otherwise all its at least k+1 neighbors are in VC; a contradiction (VC cannot have more than k vertices)]

Repeat VC1 and VC2 until no more reduction is possible.

Example 3. Apply VC1 and VC2 to the instance (G, 4) until no more reduction is possible.



Example 3. Apply VC1 and VC2 to the instance (G, 4) until no more reduction is possible.



Lemma 1. If a graph has maximum degree d, then a set of k vertices can cover at most kd edges.

(Why?) Each vertex covers at most d edges.

Lemma 2. If (G, k) is a yes-instance of VC and neither VC1 nor VC2 can be applied to (G, k), then  $|V(G)| \le k^2 + k$  and  $|E(G)| \le k^2$ .

#### Proof.

- 1. As G is fully reduced under VC2, every vertex of G has at most degree k.
- 2. As G is a yes-instance, there exists a set of at most k vertices that covers **all** edges of G.
- 3. Hence, by Lemma 1, we have  $|E(G)| \le k^2$ .
- 4. As G is fully reduced under VC1, there are no isolated vertices in G.
- 5. As G is a yes-instance, there is a set S of at most k vertices that are (collectively) incident with all edges, i.e.  $|E(G)| \le k^2$ . Each edge is incident with a vertex in S and at most one vertex not in S.
- 6. Hence  $|V(G)| \le k + k^2$ .

Think about the contrapositive!

Theorem 3. Vertex Cover has a kernel with  $O(k^2)$  vertices and  $O(k^2)$  edges. **Proof.** Follows immediately from Lemma 2.

Input. A graph G=(V,E) and  $k \in \mathbb{N}$ 

Output. Vertex cover C of G of size at most k if it exists.

Set *C*={}.

Set H to be the set of vertices of G whose degree is > k.

If |H| > k,

**then**: return "(G,k) is a no-instance"

else:  $C := C \cup H$ 

j:=k-|H|

Delete all vertices in H from G and all isolated vertices.

If resulting graph G' has more than  $j^2$  edges or more than  $j^2+j$  vertices,

**then:** return "(G,k) is a no-instance"

**else:** compute a VC for G' of size at most j (later)

#### So what?

- Start with a graph *G*, apply VC1 and VC2 repeatedly until no further reduction is possible. Let *G'* be the resulting graph. (*How long does this take?*)
- The size of *G'* only depends on *k*.
- VC1 and VC2 in combination with a bounded-search tree algorithm (later) gives an algorithm that decides in time  $O(n^2+2^kk^2)$  if (G,k) is a yes-instance of VC, where n=|V(G)|.
- Hence VC for a graph G is fixed-parameter tractable, when parameterized by the size of a smallest vertex cover for G.
- Running time  $O(n^2+2^kk^2)$  can be further improved to  $O(1.28^k+nk)$
- Independent of the size of G (e.g. n=1000000), we can now solve VC on a reduced graph with at most  $k^2$  edges.

#### Kernelization – formal definition

Let L be a parameterized problem consisting of input pairs (x, k). A kernelization algorithm A for L replaces an instance (x, k), with a reduced instance (x', k'), called kernel, such that

- A runs in time polynomial in |x|
- $k' \leq k$
- $(x,k) \in L$  if and only if  $(x',k') \in L$
- $|x'| \le g(k)$ , for some computable function g that only depends on k

Revisit formal definition: Give VC1 and VC2 a kernelization algorithm?

Let L be a parameterized problem consisting of input pairs (x, k). A kernelization algorithm A for L replaces an instance (x, k), with a reduced instance (x', k'), called kernel, such that

A runs in time polynomial in |x|

Reductions VC1 and VC2 can be applied in time  $O(n^2)$ 

• *k*′ ≤ *k* 

VC1 and VC2 reduce parameter k or leave it unchanged

•  $(x,k) \in L$  if and only if  $(x',k') \in L$ 

VC1 and VC2 are safe.

•  $|x'| \le g(k)$ , for some computable function g that only depends on k By Lemma 2., we have  $|x'| \le k^2 + k$  vertices.

Theorem. A problem Q is fixed-parameter tractable if and only if it admits a kernelization.

#### *Proof.* [We only prove one implication!]

Let I=(x,k) be an instance of Q, and let (x',k') be the instance of Q resulting from kernelizing I.

Suppose Q admits a kernelization. Then the reduction can be performed in time  $O(|x|^c)$  for some constant c.

Since the size of x' only depends on k and is independent of the input size, we can use brute-force to decide whether or not (x',k') is a yes-instance in time

$$O(|x|^c + f(g(k))).$$

Hence, Q is fixed-parameter tractable.

# 3-Hitting Set (3-HS)

Instance. A finite set S and a collection  $C=\{C_1, C_2, ..., C_m\}$  of subsets of S such that each element in C has size at most three, and a non-negative integer k.

Question. Is there a subset H of S such that  $|H| \le k$  and H contains at least one element from each subset in C?

VC is equivalent to 2-Hitting Set!

3-Hitting Set (3-HS)

### Example 4.

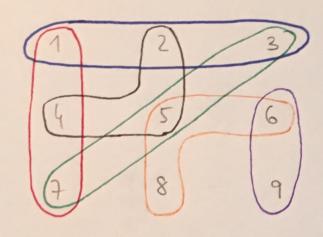
$$S=\{s_1,s_2,...,s_9\}, k=3$$

$$C = \{\{s_1, s_2, s_3\}, \{s_1, s_4, s_7\}, \{s_2, s_4, s_5\}, \{s_3, s_5, s_7\}, \{s_5, s_6, s_8\}, \{s_6, s_9\}\}$$

Does C have a hitting set H of size at most 3?

5- 25,,52, ...,59}

(={\ S\_1, S\_2, S\_3}, \ S\_1, S\_4, S\_5, \ S\_2, S\_4, S\_5\, \ S\_3, S\_5, S\_7\, \ S\_5, S\_6, S\_8\, \ S\_6, S\_9\}



Hn= 253, S4, S6}

Hz= (Sz, Sc, Sa)

H3=25,55,56}

Others ?

#### Reductions for 3-HS

- HS1. If C contains a singleton  $\{s_i\}$ , then add  $s_i$  to H. Delete  $\{s_i\}$  from C. Decrement parameter k by 1.
  - s; must be an element of every hitting set
- HS2. For each pair  $C_i, C_j \in C$  with  $i \neq j$ , if  $C_i \subseteq C_j$  delete  $C_j$  from C.
  - a hitting set that contains an element in  $C_i$  also contains an element in  $C_j$

#### Reductions for 3-HS

• HS3. For each pair  $s_i, s_j \in S$  with  $i \neq j$ , if there are more than k 3-element subsets in C that all contain  $s_i$  and  $s_j$ , then delete all elements from C that contain  $s_i$  and  $s_i$ . Add  $\{s_i, s_i\}$  to C.

If  $s_i$  and  $s_j$  not in H, then there are at least k+1 other elements in H (Why?); hitting set becomes too big.

E.g. 
$$k=3$$
; if  $\{\{1,2,3\},\{1,2,4\},\{1,2,5\},\{1,2,6\}\}\subseteq C$ 

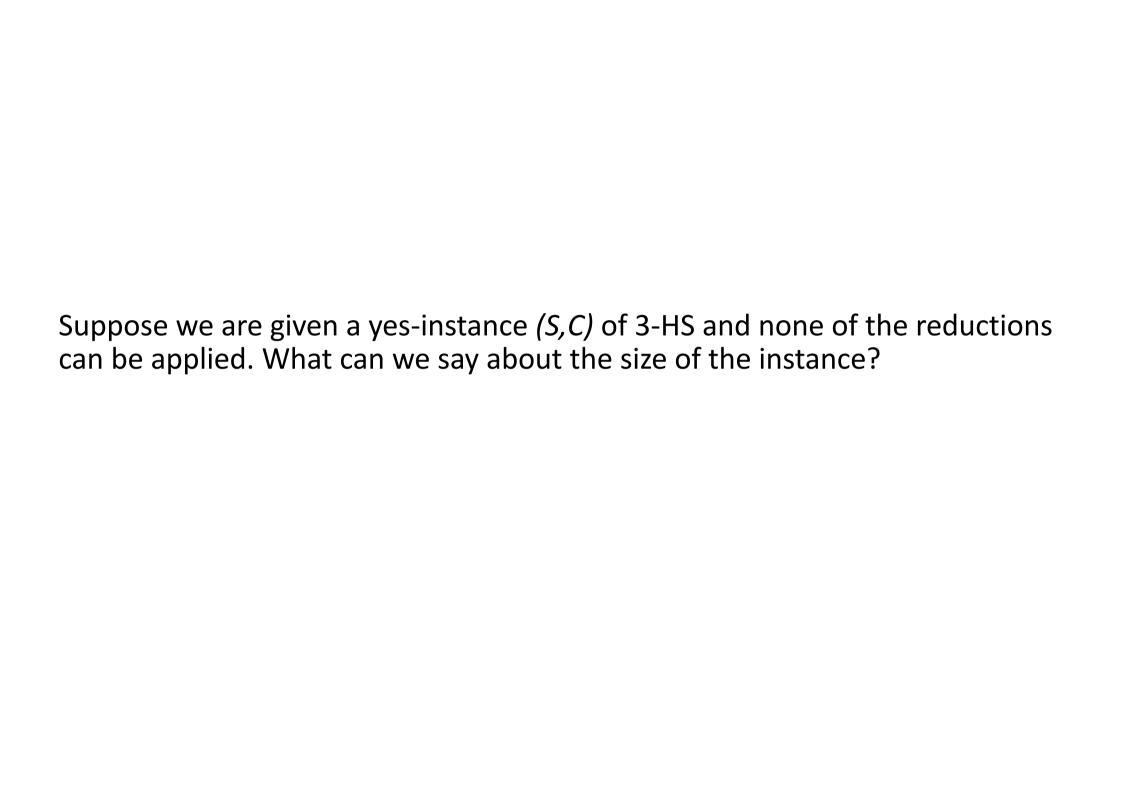
Once applied exhaustively, there are at most k subsets of the form  $\{s_i, s_j, *\}$  in C (otherwise, we can apply HS3 again).

#### Reductions for 3-HS

HS4. For each s<sub>i</sub> ∈ S<sub>i</sub>, if there are
 more than k 2-element subsets in C or
 more than k<sup>2</sup> 3-element subsets in C

that contain  $s_i$ , then add  $s_i$  to H. Delete all elements from C that contain  $s_i$ . Decrement k by 1.

If there are more than k subsets of the form  $\{s_i, *\}$  or more than  $k^2$  subsets of the form  $\{s_i, *, *\}$ , add  $s_i$  to H; otherwise H becomes too big.



Claim 1. For each pair  $s_i, s_j \in S$  with  $i \neq j$ , there are at most k 3-element subsets in C that have the form  $\{s_i, s_i, *\}$ .

#### Proof.

Suppose that there are more than k such sets in C. Let C' be the subset of C that contains all triples of the form  $\{s_i, s_j, *\}$ . Since C is a set (no element appears multiple times), the third coordinates of all triples in C' are pairwise distinct. Hence, as |C'| > k, a hitting set with at most k elements contains  $s_i$  or  $s_i$ . But then we can replace C' in C with a single element  $\{s_i, s_i\}$ .

Claim 1 motivates HS3.

Claim 2. For each element  $s_i \in S$ , there are at most k 2-element subsets in C that have the form  $\{s_i, *\}$ .

Proof. [Analogous to the proof of Claim 1]

Suppose that there are more than k such sets in C. Let C' be the subset of C that contains all subsets in C of the form  $\{s_i, *\}$ . Since C is a set (no element appears multiple times), the second coordinates of all elements in C' are pairwise distinct. Hence, as |C'| > k, a hitting set with at most k elements contains  $s_i$  and we can delete C' from C, decrement k by one and add  $s_i$  to H.

Claim 3. For each element  $s_i \in S$ , there are at most  $k^2$  3-element subsets in C that have the form  $\{s_i, *, *\}$ .

Proof. Suppose that there are more than  $k^2$  such sets in C. Let C' be the subset of C that contains all triples of the form  $\{s_i, *, *\}$ . By Claim 1, we may assume that  $s_i$  occurs together with  $s_i$  in a triple  $\{s_i, s_i, *\}$  in C' at most k times.

.

$$\{s_{i}, s_{j_{1}}, *\} \leq k \text{ times}$$
  
 $\{s_{i}, s_{j_{2}}, *\} \leq k \text{ times}$   
 $\{s_{i}, s_{j_{3}}, *\} \leq k \text{ times}$   
 $\{s_{i}, s_{j_{k}}, *\} \leq k \text{ times}$ 

Claim 3. For each element  $s_i \in S$ , there are at most  $k^2$  3-element subsets in C that have the form  $\{s_i, *, *\}$ .

Proof. Suppose that there are more than  $k^2$  such sets in C. Let C' be the subset of C that contains all triples of the form  $\{s_i, *, *\}$ . By Claim 1, we may assume that  $s_i$  occurs together with  $s_j$  in a triple  $\{s_i, s_j, *\}$  in C' at most k times. Hence, if  $|C'| > k^2$ , then the subsets in C' cannot be covered by a hitting set of size at most k that does not contain  $s_i$ . Hence  $s_i$  is in a hitting set for C and C' can be deleted from C.

Claims 2 and 3 motivate HS4.

Theorem 4. 3-HS has a kernel with  $|C| = O(k^3)$  that can be found in polynomial time.

Proof. (based on HS4)

Let  $s_i \in S$ .

Claim 2 gives us that there are at most k elements in C that have the form  $\{s_i, *\}$ .

Claim 3 gives us that there are at most  $k^2$  elements in C that have the form  $\{s_i, *, *\}$ .

Taken together, there are at most  $k + k^2$  subsets in C that contain  $s_i$ .

Provided C has a hitting set of size at most k, there are at most  $k(k + k^2) = k^2 + k^3$  elements in C. Hence  $|C| = O(k^3)$ . (If there are more elements we have a noinstance.)

For a poly-time algorithm, count how many times  $s_i$  appears in an element in C.

Theorem 4. 3-HS has a kernel with  $|C| = O(k^3)$  that can be found in polynomial time.

#### What does Theorem 4 say?

If an instance (S,C) of 3-HS has a hitting set of size at most k, then C can be reduced to a set of size  $O(k^3)$  by applying the reductions HS1-HS4.

#### Think about the contrapositive!

If C has more than  $O(k^3)$  elements after applying the reductions HS1-HS4, then (S,C) is a no-instance. The algorithm can stop and return `no' without performing any exhaustive search.

## Generalization to d-Hitting Set

Instance. A finite set S and a collection  $C = \{C_1, C_2, ..., C_m\}$  of subsets of S such that each element in C has size at most d, and a non-negative integer k.

Question. Is there a subset H of S such that  $|H| \le k$  and H contains at least one element from each subset in C?

Theorem 4. For a **fixed** d>3, d-HS has a kernel with  $|C| = O(k^d)$ .