

# DYNAMIC PROGRAMMING

Slides from Prof. Daniel Marx

# The SET COVER problem

**Input:** A set family  $\mathcal{F}$  over a universe U and an integer k

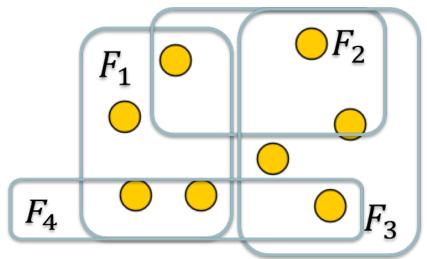
Parameter: |U|

**Question:** Is there a subfamily  $\mathcal{F}' \subseteq \mathcal{F}$  of at most k sets,

such that  $\bigcup_{F \in \mathcal{F}}, F = U$ ?

• The subfamily  $\mathcal{F}'$  covers the universe U

- SET COVER parameterized by the universe size is FPT
  - Algorithm with running time  $2^{|U|} \cdot (|U| + |\mathcal{F}|)^c$
  - Based on dynamic programming



# Dynamic programming for SET COVER

- Let  $\mathcal{F} = \{F_1, F_2, \dots, F_m\}$
- We define a DP table for  $X \subseteq U$  and  $j \in \{0,1,...,m\}$   $T[X,j] = \min \text{ nr. of sets from } F_1,...,F_j \text{ needed to cover } X$   $\text{Or } +\infty \text{ if impossible}$
- The value T[U, m] gives the minimum size of a set cover
  - To solve the problem, compute T using base cases and a recurrence

# Filling the dynamic programming table

• T[X,j] = min nr. of sets from  $F_1, ..., F_j$  needed to cover X

Base case: 
$$j = 0$$
  
 $T[X,j] = 0$  if  $X = \emptyset$ , otherwise it is  $+\infty$ 

Recursive step: 
$$j > 0$$
  

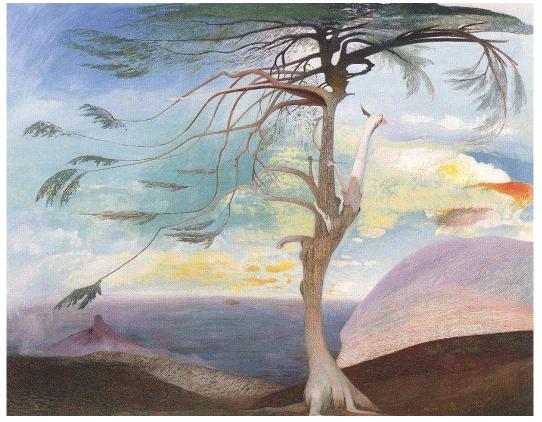
$$T[X,j] = \min(T[X,j-1], 1 + T[X \setminus F_j, j-1])$$

- Skip set  $F_j$ , or pay for  $F_j$  and afterwards cover  $X \setminus F_j$
- Each entry can be computed in polynomial time  $-(|\mathcal{F}|+1)\cdot 2^{|U|}$  entries in total

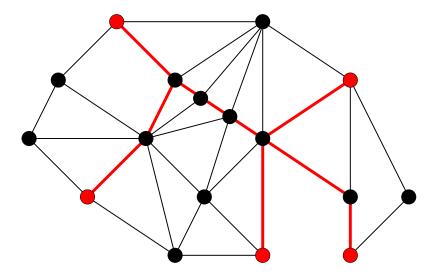
# More on dynamic programming

- Dynamic programming is a memory-intensive algorithmic paradigm that yields FPT algorithms in various situations
  - Here: dynamic programming over **subsets** of U
  - Later: dynamic programming over tree decompositions
- Research challenge:
  - Determine whether the  $2^{|U|}$  factor can be improved to  $(2-\epsilon)^{|U|}$  for some  $\epsilon>0$





**Task:** Given a graph G with weighted edges and a set S of k vertices, find a tree T of minimum weight that contains S.



Known to be NP-hard. For fixed k, we can solve it in polynomial time: we can guess the Steiner points and the way they are connected.

**Theorem:** Steiner Tree is FPT parameterized by k = |S|.

Solution by dynamic programming. For  $v \in V(G)$  and  $X \subseteq S$ ,

c(v, X) := minimum cost of a Steiner tree of X that contains v

d(u, v) :=distance of u and v

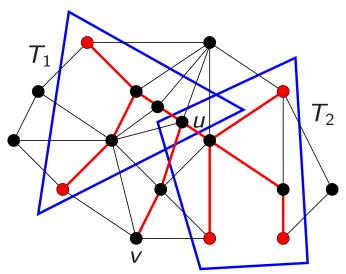
#### **Recurrence relation:**

$$c(v,X) = \min_{\substack{u \in V(G) \\ \emptyset \subset X' \subset X}} c(u,X' \setminus u) + c(u,(X \setminus X') \setminus u) + d(u,v)$$

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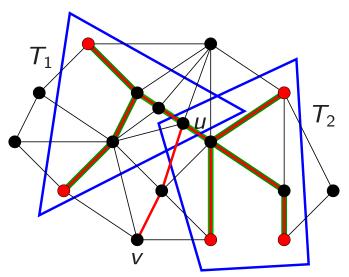
 $\leq$ : A tree  $T_1$  realizing  $c(u, X' \setminus u)$ , a tree  $T_2$  realizing  $c(u, (X \setminus X') \setminus u)$ , and the path uv gives a (superset of a) Steiner tree of X containing v.



#### **Recurrence relation:**

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Suppose T realizes c(v,X), let T' be the minimum subtree containing X. Let u be a vertex of T' closest to v. If |X| > 1, then there is a component C of  $T \setminus u$  that contains a subset  $\emptyset \subset X' \subset X$  of terminals. Thus T is the disjoint union of a tree containing  $X' \setminus u$  and u, a tree containing  $(X \setminus X') \setminus u$  and u, and the path uv.



#### **Recurrence relation:**

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#### **Running time:**

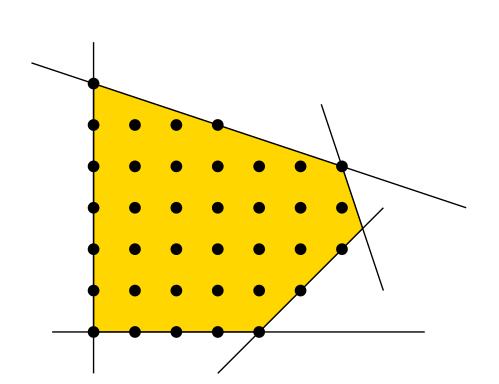
 $2^k|V(G)|$  variables c(v,X), determine them in increasing order of |X|. Variable c(v,X) can be determined by considering  $2^{|X|}$  cases. Total number of cases to consider:

$$\sum_{X \subseteq T} 2^{|X|} = \sum_{i=1}^k \binom{k}{i} 2^i \le (1+2)^k = 3^k.$$

Running time is  $O^*(3^k)$ .

**Note:** Running time can be reduced to  $O^*(2^k)$  with clever techniques.

# Integer Linear Programming



## Integer Linear Programming

**Linear Programming (LP):** important tool in (continuous) combinatorial optimization. Sometimes very useful for discrete problems as well.

$$\max c_1 x_1 + c_2 x_2 + c_3 x_3$$
 s.t.  $x_1 + 5x_2 - x_3 \leq 8$   $2x_1 - x_3 \leq 0$   $3x_2 + 10x_3 \leq 10$   $x_1, x_2, x_3 \in \mathbb{R}$ 

Fact: It can be decided if there is a solution (feasibility) and an optimum solution can be found in polynomial time.

## Integer Linear Programming

**Integer Linear Programming (ILP):** Same as LP, but we require that every  $x_i$  is integer.

Very powerful, able to model many NP-hard problems. (Of course, no polynomial-time algorithm is known.)

**Theorem:** ILP with p variables can be solved in time  $p^{O(p)} \cdot n^{O(1)}$ .

**Task:** Given strings  $s_1, ..., s_k$  of length L over alphabet  $\Sigma$ , and an integer d, find a string s (of length L) such that  $d(s, s_i) \le d$  for every  $1 \le i \le k$ .

**Note:**  $d(s, s_i)$  is the Hamming distance.

**Theorem:** CLOSEST STRING parameterized by k is FPT.

**Theorem:** CLOSEST STRING parameterized by *d* is FPT.

**Theorem:** CLOSEST STRING parameterized by *L* is FPT.

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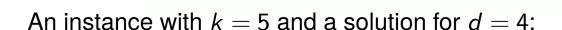
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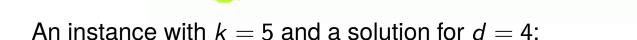
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- s<sub>1</sub> CBDCCACBB
- s<sub>2</sub> ABDBCABDB
- s<sub>3</sub> CDDBACCBD
- s<sub>4</sub> DDABACCBD
- s<sub>5</sub> ACDBDDCBC

**ADDBCACBD** 

Each column can be described by a partition  $\mathcal{P}$  of [k].



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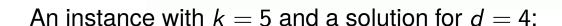


An instance with k = 5 and a solution for d = 4:

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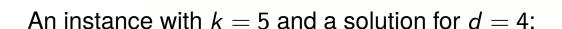
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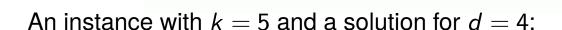
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The instance can be described by an integer  $c_{\mathcal{P}}$  for each partition  $\mathcal{P}$ : the number of columns with this type.

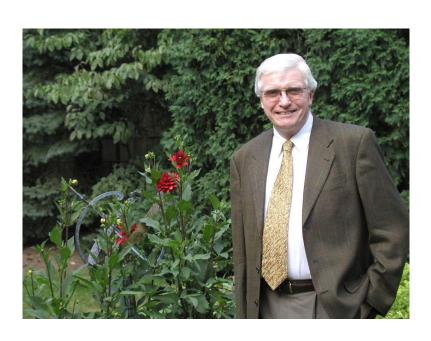
**Describing a solution:** If C is a class of  $\mathcal{P}$ , let  $x_{\mathcal{P},C}$  be the number of type  $\mathcal{P}$  columns where the solution agrees with class C.

There is a solution iff the following ILP has a feasible solution:

$$\sum_{C \in \mathcal{P}} x_{\mathcal{P},C} \leq c_{\mathcal{P}}$$
  $\forall partition \mathcal{P}$   $\sum_{i \notin C,C \in \mathcal{P}} x_{\mathcal{P},C} \leq d$   $\forall 1 \leq i \leq k$   $x_{\mathcal{P},C} \geq 0$   $\forall \mathcal{P},C$ 

Number of variables is  $\leq B(k) \cdot k$ , where B(k) is the no. of partitions of [k]  $\Rightarrow$  The ILP algorithm solves the problem in time  $f(k) \cdot n^{O(1)}$ .

# **Graph Minors**







Paul Seymour

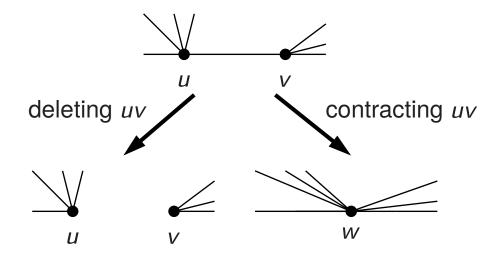
## **Graph Minors**



- Some consequences of the Graph Minors Theorem give a quick way of showing that certain problems are FPT.
- 6 However, the function f(k) in the resulting FPT algorithms can be HUGE, completely impractical.
- History: motivation for FPT.
- Parts and ingredients of the theory are useful for algorithm design.
- New algorithmic results are still being developed.

## **Graph Minors**

**Definition:** Graph H is a **minor** G ( $H \le G$ ) if H can be obtained from G by deleting edges, deleting vertices, and contracting edges.

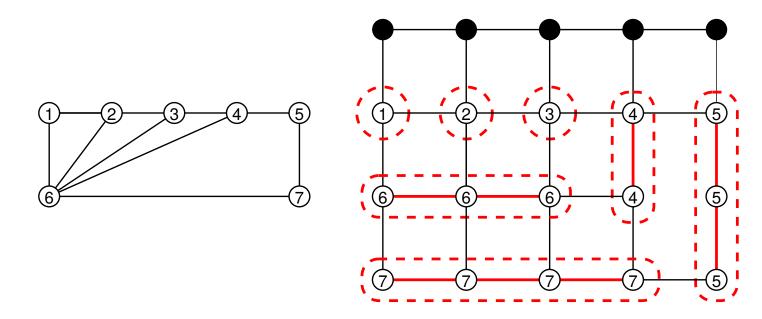


**Example:** A triangle is a minor of a graph G if and only if G has a cycle (i.e., it is not a forest).

## Graph minors

**Equivalent definition:** Graph H is a **minor** of G if there is a mapping  $\phi$  that maps each vertex of H to a connected subset of G such that

- $\phi(u)$  and  $\phi(v)$  are disjoint if  $u \neq v$ , and
- if  $uv \in E(G)$ , then there is an edge between  $\phi(u)$  and  $\phi(v)$ .



## Minor closed properties

**Definition:** A set  $\mathcal{G}$  of graphs is **minor closed** if whenever  $G \in \mathcal{G}$  and  $H \leq G$ , then  $H \in \mathcal{G}$  as well.

#### **Examples of minor closed properties:**

planar graphs acyclic graphs (forests) graphs having no cycle longer than k empty graphs

#### **Examples of not minor closed properties:**

complete graphs regular graphs bipartite graphs

### Forbidden minors

Let  $\mathcal{G}$  be a minor closed set and let  $\mathcal{F}$  be the set of "minimal bad graphs":  $H \in \mathcal{F}$  if  $H \notin \mathcal{G}$ , but every proper minor of H is in  $\mathcal{G}$ .

#### **Characterization by forbidden minors:**

$$G \in \mathcal{G} \iff \forall H \in \mathcal{F}, H \not\leq G$$

The set  $\mathcal{F}$  is the **obstruction set** of property  $\mathcal{G}$ .

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**Theorem:** [Wagner] A graph is planar if and only if it does not have a  $K_5$  or  $K_{3,3}$  minor.

In other words: the obstruction set of planarity is  $\mathcal{F} = \{K_5, K_{3,3}\}$ .

Does every minor closed property have such a finite characterization?

## Graph Minors Theorem

**Theorem:** [Robertson and Seymour] Every minor closed property  $\mathcal{G}$  has a finite obstruction set.

**Note:** The proof is contained in the paper series "Graph Minors I–XX".

Note: The size of the obstruction set can be astronomical even for simple

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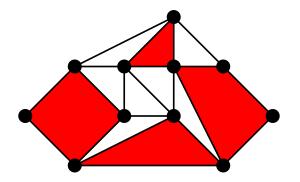
properties.

**Theorem:** [Robertson and Seymour] For every fixed graph H, there is an  $O(n^3)$  time algorithm for testing whether H is a minor of the given graph G.

**Corollary:** For every minor closed property  $\mathcal{G}$ , there is an  $O(n^3)$  time algorithm for testing whether a given graph G is in  $\mathcal{G}$ .

## **Applications**

PLANAR FACE COVER: Given a graph G and an integer k, find an embedding of planar graph G such that there are k faces that cover all the vertices.



#### One line argument:

For every fixed k, the class  $\mathcal{G}_k$  of graphs of yes-instances is minor closed.

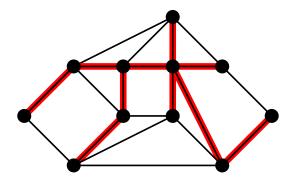


For every fixed k, there is a  $O(n^3)$  time algorithm for Planar Face Cover.

Note: non-uniform FPT.

## **Applications**

k-LEAF SPANNING TREE: Given a graph G and an integer k, find a spanning tree with **at least** k leaves.



Technical modification: Is there such a spanning tree for at least one component of *G*?

#### One line argument:

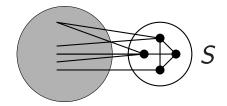
For every fixed k, the class  $\mathcal{G}_k$  of no-instances is minor closed.



For every fixed k, k-LEAF SPANNING TREE can be solved in time  $O(n^3)$ .

### 9 + k vertices

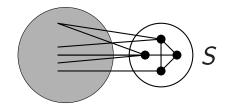
Let  $\mathcal{G}$  be a graph property, and let  $\mathcal{G} + kv$  contain graph G if there is a set  $S \subseteq V(G)$  of k vertices such that  $G \setminus S \in \mathcal{G}$ .



**Lemma:** If  $\mathcal{G}$  is minor closed, then  $\mathcal{G} + kv$  is minor closed for every fixed k.  $\Rightarrow$  It is (nonuniform) FPT to decide if G can be transformed into a member of  $\mathcal{G}$  by deleting k vertices.

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- If  $G = \text{forests} \Rightarrow G + kv = \text{graphs that can be made acyclic by the deletion of } k$ vertices  $\Rightarrow \text{FEEDBACK VERTEX SET is FPT.}$
- If  $\mathcal{G} = \text{planar graphs} \Rightarrow \mathcal{G} + kv = \text{graphs that can be made planar by the deletion of } k \text{ vertices } (k\text{-apex graphs}) \Rightarrow k\text{-Apex Graph is FPT.}$
- If  $\mathcal{G} = \text{empty graphs} \Rightarrow \mathcal{G} + kv = \text{graphs with vertex cover number at most } k \Rightarrow Vertex Cover is Fixed Parameter Algorithms p.65/98$