



Parameterized complexity of constraint satisfaction problems

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Constraint satisfaction problems

Let \mathcal{R} be a set Boolean of relations. An \mathcal{R} -formula is a conjunction of relations in \mathcal{R} :

$$R_1(x_1, x_4, x_5) \wedge R_2(x_2, x_1) \wedge R_1(x_3, x_3, x_3) \wedge R_3(x_5, x_1, x_4, x_1)$$

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- Given: an \mathcal{R} -formula φ
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- Given: an \mathcal{R} -formula φ
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$$\mathcal{R} = \{a \neq b\} \Rightarrow \mathcal{R}\text{-SAT} = \text{2-coloring of a graph}$$

$$\mathcal{R} = \{a \vee b, a \vee \bar{b}, \bar{a} \vee \bar{b}\} \Rightarrow \mathcal{R}\text{-SAT} = \text{2SAT}$$

$$\mathcal{R} = \{a \vee b \vee c, a \vee b \vee \bar{c}, a \vee \bar{b} \vee \bar{c}, \bar{a} \vee \bar{b} \vee \bar{c}\} \Rightarrow \mathcal{R}\text{-SAT} = \text{3SAT}$$

Question: \mathcal{R} -SAT is polynomial time solvable for which \mathcal{R} ?

It is **NP**-complete for which \mathcal{R} ?

Schaefer's Dichotomy Theorem (1978)

For every \mathcal{R} , the \mathcal{R} -SAT problem is polynomial time solvable if one of the following holds, and **NP**-complete otherwise:

- ⑥ Every relation is satisfied by the all 0 assignment
- ⑥ Every relation is satisfied by the all 1 assignment
- ⑥ Every relation can be expressed by a 2SAT formula
- ⑥ Every relation can be expressed by a Horn formula
- ⑥ Every relation can be expressed by an anti-Horn formula
- ⑥ Every relation is an affine subspace over $GF(2)$

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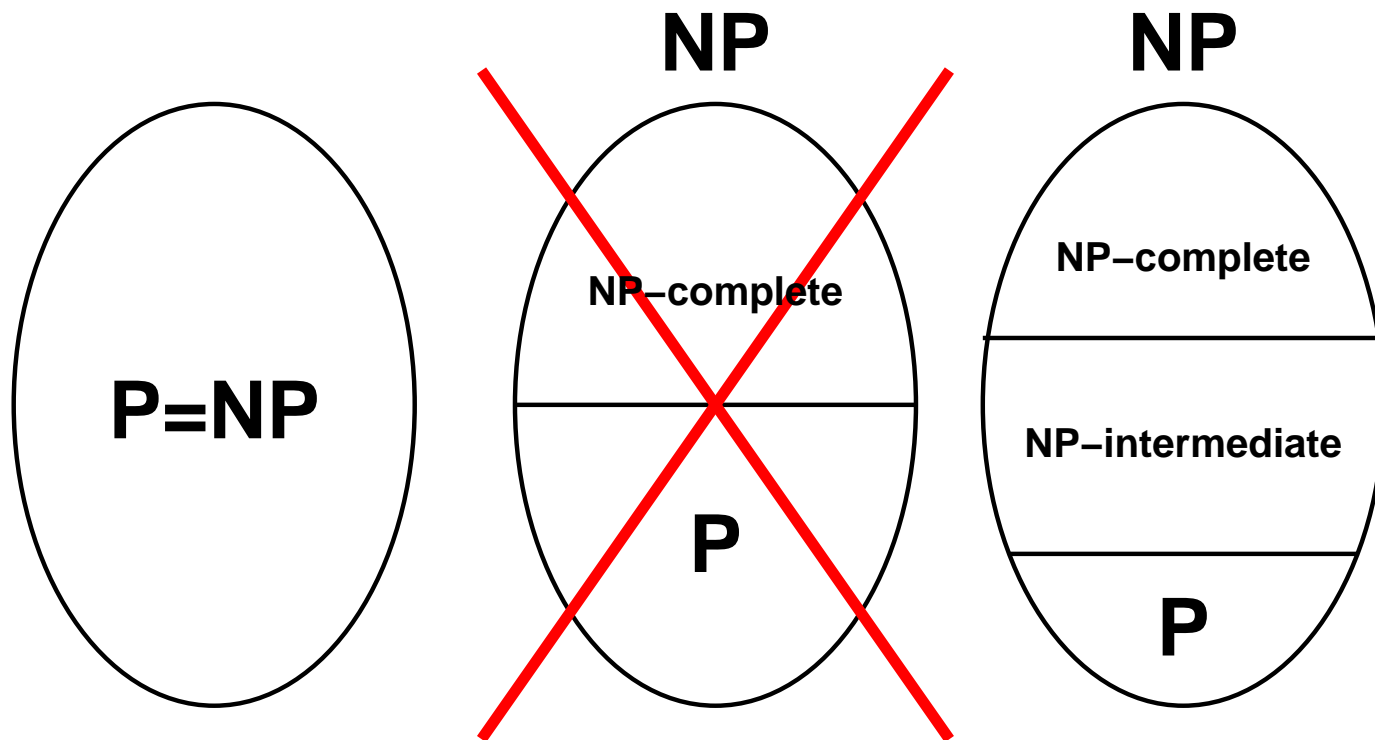
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Why is it surprising?

Ladner's Theorem (1975)

If $P \neq NP$, then there is a language $L \in NP \setminus P$ that is not NP-complete.



Other dichotomy results

- ⑥ Approximability of MAX-SAT, MIN-UNSAT [Khanna et al., 2001]
- ⑥ Approximability of MAX-ONES, MIN-ONES [Khanna et al., 2001]
- ⑥ Generalization to 3 valued variables [Bulatov, 2002]
- ⑥ Inverse satisfiability [Kavvadias and Sideri, 1999]
- ⑥ etc.

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Our contribution: parameterized analogue of Schaefer's dichotomy theorem.

Parameterized version

Parameterized \mathcal{R} -SAT

- ⌚ **Input:** a \mathcal{R} -formula φ , an integer k
- ⌚ **Parameter:** k
- ⌚ **Question:** Does φ has a satisfying assignment of weight exactly k ?

For which \mathcal{R} is there an $f(k) \cdot n^c$ algorithm for \mathcal{R} -SAT?

Main theorem: For every constraint family \mathcal{R} , the parameter \mathcal{R} -SAT problem is either fixed-parameter tractable or W[1]-complete.

(+ simple characterization of FPT cases)

Technical notes

- ⌚ Are constants allowed in the formula?

E.g., $R(x_1, 0, 1) \wedge R(1, x_2, x_3)$

- ⌚ Can a variable appear multiple times in a constraint?

E.g., $R(x_1, x_1, x_2) \wedge R(x_3, x_3, x_3)$

- ⌚ Constraints that are not satisfied by the all **0** assignment can be handled easily (bounded search tree).

Weak separability

Definition: \mathcal{R} is weakly separable if

1. the union of two disjoint satisfying assignments is also satisfying, and
2. if a satisfying assignment contains a smaller satisfying assignment, then their difference is also satisfying.

Example of 1:

$$R(\mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{0}, \mathbf{0}, \mathbf{0}, \mathbf{0}, \mathbf{0}) = 1$$

$$R(\mathbf{0}, \mathbf{0}, \mathbf{0}, \mathbf{0}, \mathbf{1}, \mathbf{1}, \mathbf{0}, \mathbf{0}, \mathbf{0}) = 1$$

⇓

$$R(\mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{0}, \mathbf{0}, \mathbf{0}) = 1$$

Example of 2:

$$R(\mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{0}, \mathbf{0}) = 1$$

$$R(\mathbf{0}, \mathbf{0}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{1}, \mathbf{0}, \mathbf{0}) = 1$$

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Main theorem: \mathcal{R} -SAT is FPT if and only if every constraint is weakly separable, and W[1]-complete otherwise.

Weak separability: examples

The constraint EVEN is weakly separable:

Property 1:

$$R(\overbrace{1, 1, 1, 1}^{\text{even}}, 0, 0, 0, 0, 0) = 1$$

$$R(0, 0, 0, 0, \underbrace{1, 1}_{\text{even}}, 0, 0, 0) = 1$$

⇓

$$R(\underbrace{1, 1, 1, 1, 1, 1}_{\text{even}}, 0, 0, 0) = 1$$

Property 2:

$$R(\overbrace{1, 1, 1, 1, 1, 1}^{\text{even}}, 0, 0) = 1$$

$$R(0, 0, \underbrace{1, 1, 1, 1}_{\text{even}}, 0, 0) = 1$$

⇓

$$R(\underbrace{1, 1}_{\text{even}}, 0, 0, 0, 0, 0, 0) = 1$$

More generally: every **affine** constraint is weakly separable.

Weak separability: examples (cont.)

The following constraint is trivially weakly separable:

$$R(0, 0, 0, 0, 0) = 1$$

$$R(1, 1, 1, 0, 0) = 1$$

$$R(0, 1, 1, 1, 0) = 1$$

$$R(0, 0, 1, 1, 1) = 1$$

$$R(x_1, x_2, x_3, x_4, x_5) = 0 \text{ otherwise.}$$

Reason: Property 1 and 2 vacuously hold, no disjoint sets, no subsets.

More generally: if the non-zero satisfying assignments are **intersecting** and form a **clutter**, then it is weakly separable.

Example: $R(x_1, \dots, x_n) = 1$ if and only if 0 or exactly t out of n variables are 1
($t > n/2$)

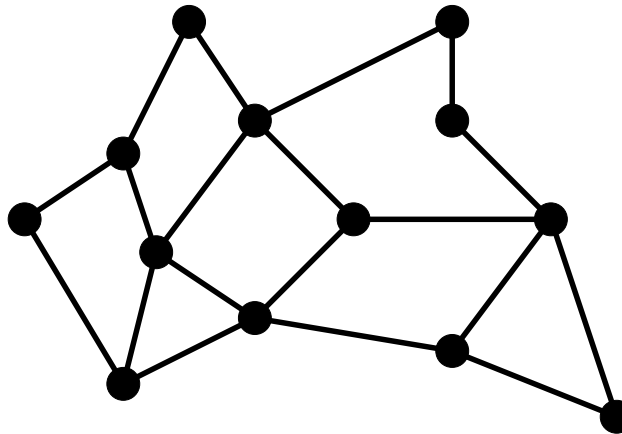
Parameterized vs. classical

The easy and hard cases are different in the classical and the parameterized version:

Constraint	Classical	Parameterized
$x \vee y$	in P	FPT (Vertex Cover)
$\bar{x} \vee \bar{y}$	in P	W[1]-complete (Max. Independent Set)
affine	in P	FPT
2-in-3	NP-complete	FPT

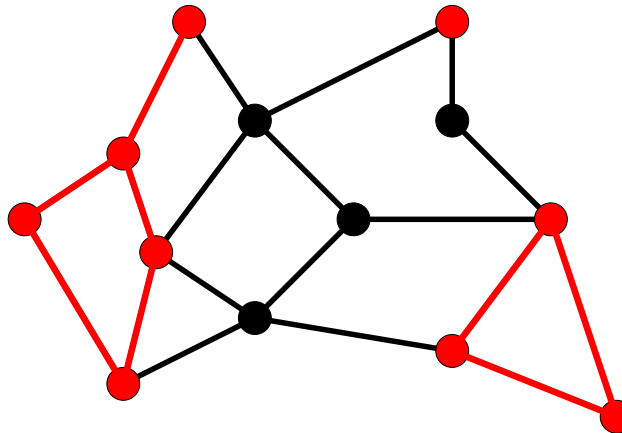
Bounded number of occurrences

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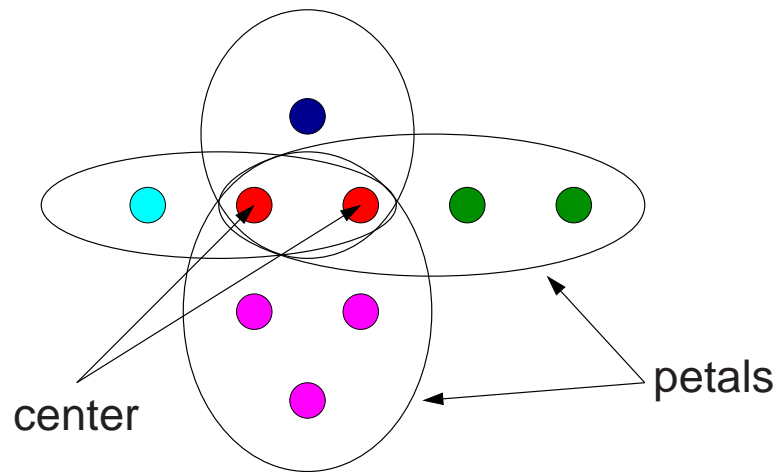
Every satisfying assignment is composed of **connected satisfying assignments**.

Lemma: There are at most $(rd)^{k^2} \cdot n$ connected satisfying assignment of size at most k . (r is the maximum arity, d is the maximum no. of occurrences)

Algorithm: Use color coding to put together the connected assignments to obtain a size k assignment.

The sunflower lemma

Definition: Sets S_1, S_2, \dots, S_k form a **sunflower** if the sets $S_i \setminus (S_1 \cap S_2 \cap \dots \cap S_k)$ are disjoint.



Lemma (Erdős and Rado, 1960): If the size of a set system is greater than $(p - 1)^\ell \cdot \ell!$ and it contains only sets of size at most ℓ , then the system contains a sunflower with k petals.

Sunflower of clauses

Definition: A **sunflower** is a set of k clauses such that for every i

- ⊗ either the same variable appears at position i in every clause,
- ⊗ or every clause “owns” its i th variable.

$$R(x_1, x_2, x_3, x_4, x_5, x_6)$$

$$R(x_1, x_2, x_3, x_7, x_8, x_9)$$

$$R(x_1, x_2, x_3, x_{10}, x_{11}, x_{12})$$

$$R(x_1, x_2, x_3, x_{13}, x_{14}, x_{15})$$

Lemma: If a variable occurs more than $c_{\mathcal{R}}(k)$ times in an \mathcal{R} -formula, then there is a sunflower of clauses with more than k petals in the formula.

Plucking the sunflower

For weakly separable constraints, the formula can be reduced if there is a sunflower with $k + 1$ petals. Example:

$$k + 1 \left\{ \begin{array}{l} \text{EVEN}(x_1, x_2, x_3, x_4, x_5, x_6) \\ \text{EVEN}(x_1, x_2, x_3, x_7, x_8, x_9) \\ \text{EVEN}(x_1, x_2, x_3, x_{10}, x_{11}, x_{12}) \\ \text{EVEN}(x_1, x_2, x_3, x_{13}, x_{14}, x_{15}) \end{array} \right.$$

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The algorithm

- ⑥ If there is a variable that occurs more than $c_{\mathcal{R}}(k)$ times:
 - △ Find a sunflower with $k + 1$ petals
 - △ Pluck the sunflower \Rightarrow shorter formula
- ⑥ If every variable occurs at most $c_{\mathcal{R}}(k)$ times:
 - △ Apply the bounded occurrence algorithm

Running time: $2^{k^{r+2} \cdot 2^{2^{O(r)}}} \cdot n \log n$, where r is the maximum arity in the constraint family \mathcal{R} .

Hardness results: case 1

Definition: R is weakly separable if

1. the union of two disjoint satisfying assignments is also satisfying, and
2. if a satisfying assignment contains a smaller satisfying assignment, then their difference is also satisfying.

If property 1 is violated:

$$R(0, 0, 0, 0, 0, 0, 0, 0) = 1$$

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Maximum Independent Set

\Rightarrow can be expressed!

Hardness results: case 2

Definition: R is weakly separable if

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2. if a satisfying assignment contains a smaller satisfying assignment, then their difference is also satisfying.

If property 2 is violated:

$$R(0, 0, 0, 0, 0, 0, 0, 0) = 1$$

$$R(1, 1, 1, 1, 1, 0, 0, 0) = 1$$

$$R(0, 0, 0, 1, 1, 0, 0, 0) = 1$$

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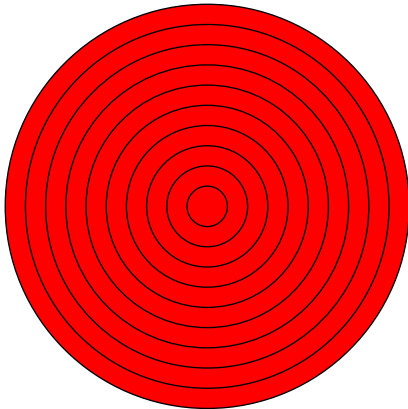
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Lemma: The problem is W[1]-complete for the constraint \rightarrow .

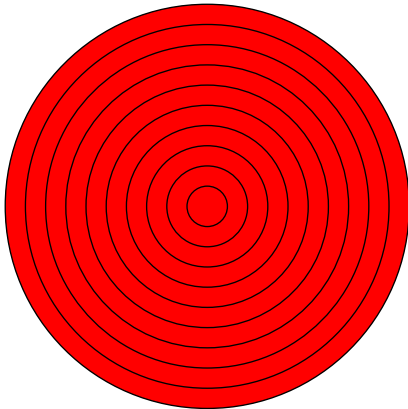
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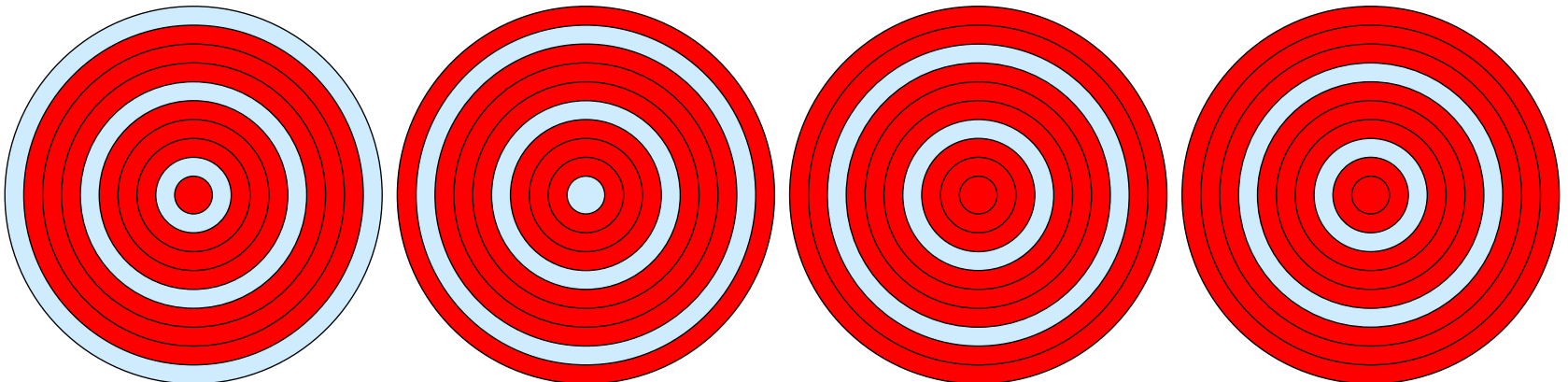


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There are $k + 1$ ways of doing this.

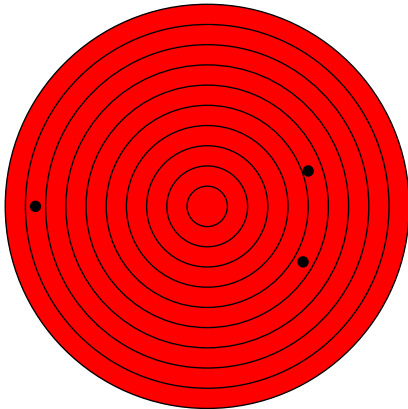
One of them will not hurt the solution.

Example with $k = 3$:



Planar formulae

If the primal graph of the formula is **planar**, then the layering method of Baker can be used.

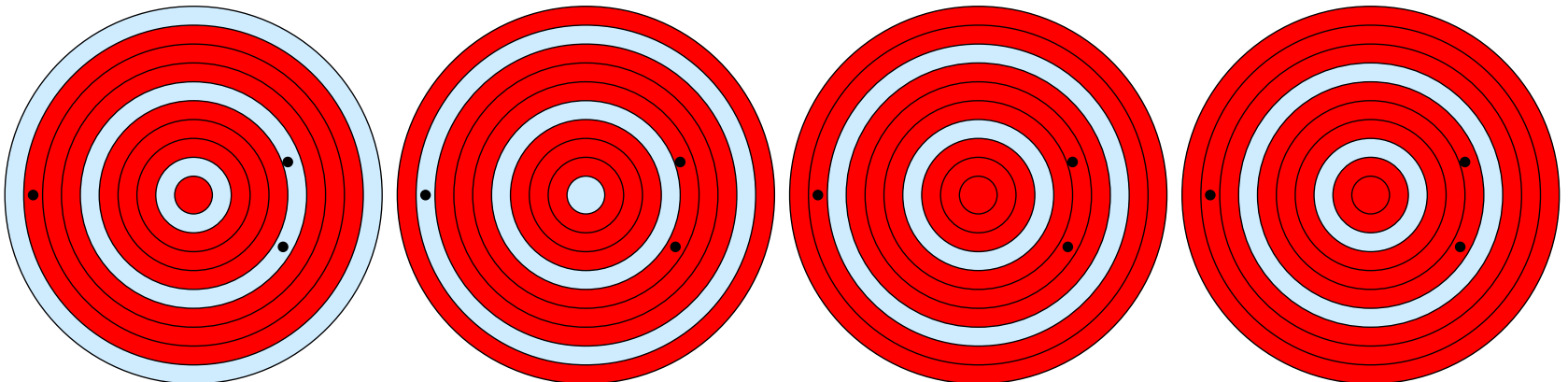


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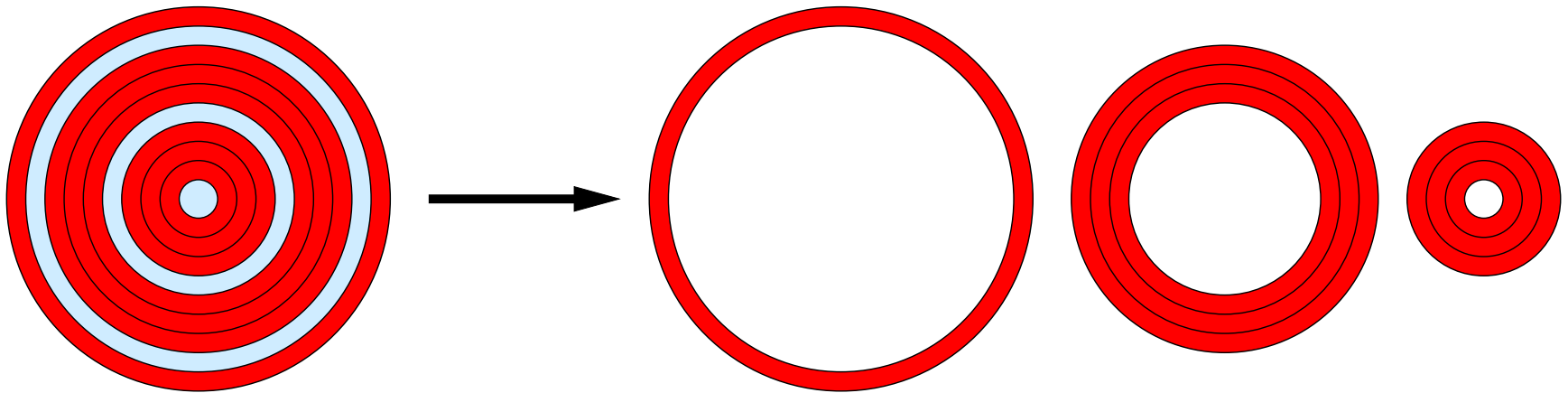
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Planar formulae (cont.)

If we delete every $(k + 1)$ th layer, then the remaining formula has only k layers:

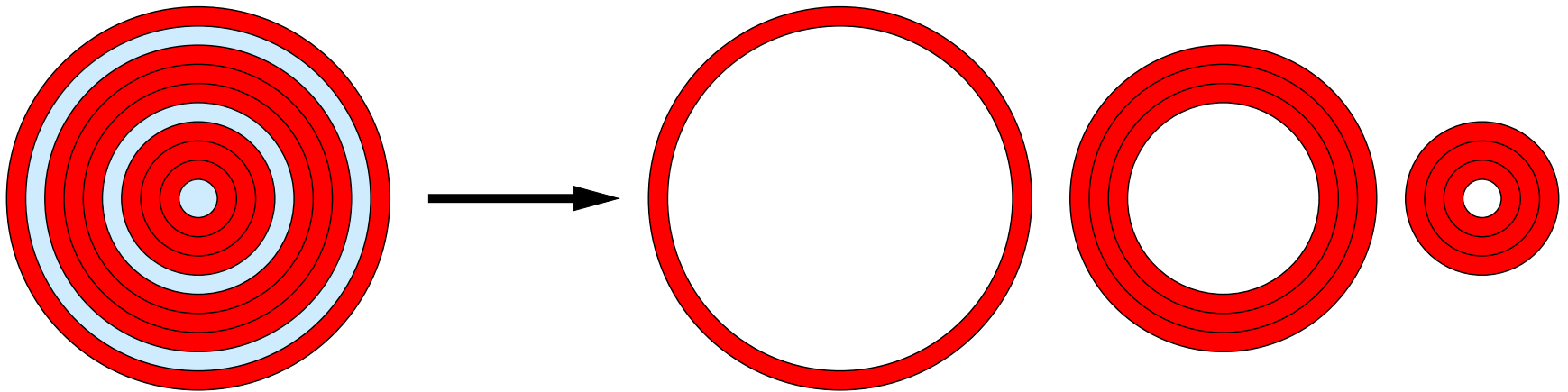


Lemma (Bodlaender): The treewidth of a k -layered graph is at most $3k - 1$.

If the primal graph has bounded treewidth, then the problem can be solved in linear time using standard techniques.

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If the primal graph has bounded treewidth, then the problem can be solved in linear time using standard techniques.

Incidence graph: bipartite graph, vertices are the clauses and the variable, edge means “appears in”.

Theorem: Linear time alg. if the incidence graph of the formula is planar.

Summary

- ⑥ Parameterized version of \mathcal{R} -SAT
- ⑥ FPT or $W[1]$ -complete depending on weak separability
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Thank you for your attention!
Questions?