

# Polynomial algorithm to compute the toughness of graphs with bounded treewidth

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# Toughness

## Definition

Let  $t$  be a positive real number. A graph  $G$  is called  **$t$ -tough**, if

$$c(G - S) \leq \frac{|S|}{t}$$

for any cutset  $S$  of  $G$ .

The toughness of  $G$ , denoted by  $\tau(G)$ , is the largest  $t$  for which  $G$  is  $t$ -tough, taking

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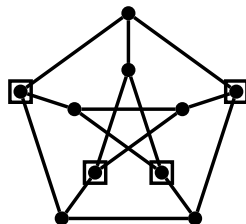
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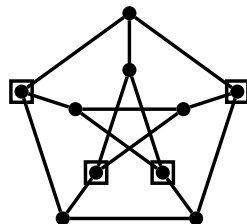
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A cycle is 1-tough.



# Toughness

In other words: for a non-complete, connected graph  $G$ ,

$$\tau(G) = \min_{S \text{ cutset}} \frac{|S|}{c(G - S)}.$$

$S$  is called a **tough set** if it gives the ratio  $\tau(G)$ .

## Observation

*For a non-complete, connected graph  $G$  on  $n$  vertices the toughness  $\tau(G)$  is a rational number  $\frac{p}{q}$  with  $1 \leq p, q \leq n$ .*

## Proof.

Clearly  $1 \leq |S| \leq n - 2$  and  $2 \leq c(G - S) \leq n - 1$ . □



# Complexity of toughness

Let  $t$  be an arbitrary positive rational number and consider the following problem.

## **$t$ -TOUGH**

*Instance:* A graph  $G$ ,

*Question:* Is it true that  $\tau(G) \geq t$ ?

**Theorem (Bauer, Hakimi, Schmeichel, 1990)**

*For any positive rational number  $t$ ,  $t$ -TOUGH is coNP-complete.*



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**Theorem (Bauer, van den Heuvel, Morgana, Schmeichel, 1998)**

*1-TOUGH is coNP-complete for  $r$ -regular graphs for all  $r \geq 3$ .*



# Complexity of toughness in special graph classes

## Theorem (Kratsch, Lehel, Müller, 1996)

*The problem **1-TOUGH** is coNP-complete for bipartite graphs.*

## Theorem (GY. K., K. Varga, 2022)

*For any positive rational number  $t \leq 1$  the problem **t-TOUGH** remains coNP-complete for bipartite graphs.*





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Deciding the toughness remains coNP-hard in many other special graph classes.

Famous open cases: planar graphs, chordal graphs



# Complexity of toughness in special graph classes

For some other special graph classes there are polynomial algorithms to compute the toughness:

- Claw-free graphs (Matthews, Sumner, 1984)
- Split graphs (Kratsch, Lehel, Müller, 1996; Woeginger, 1998)
- Interval graphs (Kratsch, Kloks, Müller, 1994)
- $2K_2$ -free graphs (Broersma, Patel, Pyatkin, 2014)
- some other more special classes ...



# Main result

## Theorem

*There exists an algorithm to compute the **toughness** of a graph  $G$  with running time  $\mathcal{O}(n^3 \cdot \text{tw}(G)^{2\text{tw}(G)})$ , where  $n$  is the number of vertices in  $G$  and  $\text{tw}(G)$  is the treewidth of  $G$ .*



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*Toughness is FPT parameterized with treewidth.*



# Tree decomposition

Let  $G = (V, E)$  be a graph. Let  $(X_t)_{t \in V(T)}$  be a family of vertex sets  $X_t \subseteq V$  (bags) indexed by the nodes of a tree  $T$ . The pair  $(T, \{X_t \mid t \in V(T)\})$  is a **tree decomposition** of  $G$  if it satisfies the following conditions:

- $\bigcup_{t \in V(T)} X_t = V$ ;
- for every edge  $e = vw \in E$  there is a  $t \in V(T)$  with  $v, w \in X_t$ ;
- if  $i, j, k \in V(T)$  and node  $j$  is on the path in  $T$  between nodes  $i$  and  $k$ , then  $X_i \cap X_k \subseteq X_j$ .

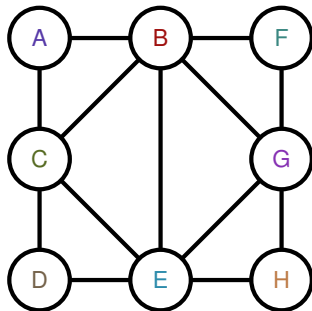
The **width** of the tree decomposition is  $\max_{t \in V(T)} |X_t| - 1$ .

The **treewidth** of a  $G$  is the minimum width of a tree decomposition of  $G$ .

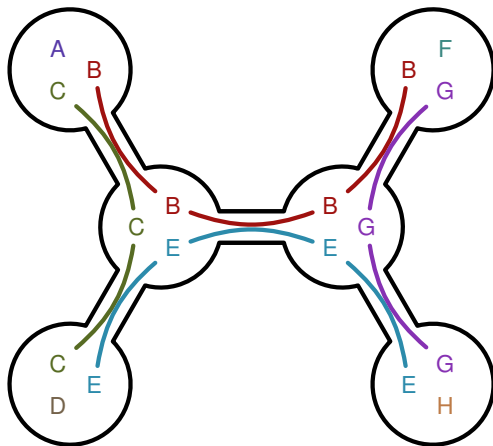


# Tree decomposition

Graph  $G$



Tree decomposition  $(T, X)$





# Bounded treewidth

## Theorem (Bodlaender, 1996)

*A tree decomposition with width  $\text{tw}(G)$  can be constructed in  $\text{tw}(G)^{\mathcal{O}(\text{tw}(G)^3)} \cdot n$  time.*



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Many NP-hard problems are FPT parameterized with treewidth, so they are solvable in polynomial time if the treewidth is bounded.



# Trees

Graphs with  $\text{tw}(G) = 1$  are **trees**.



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## Lemma

*The toughness of a tree is  $1/\Delta(G)$ .*

## Proof.

Every tough set is a single vertex with maximum degree. □



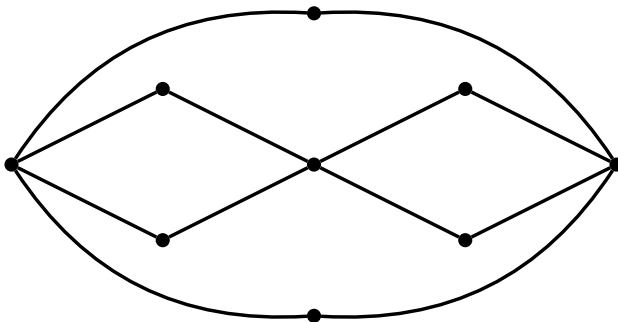
# Series parallel graphs

Graphs with  $\text{tw}(G) = 2$  are **series parallel graphs**.



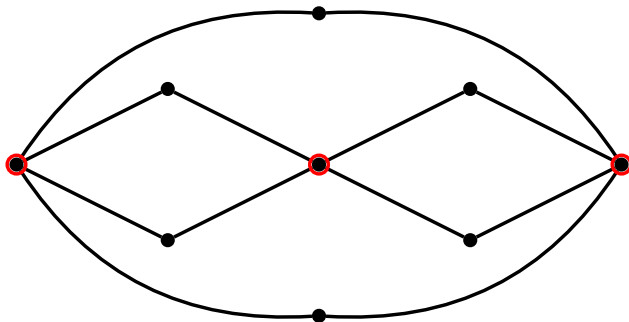
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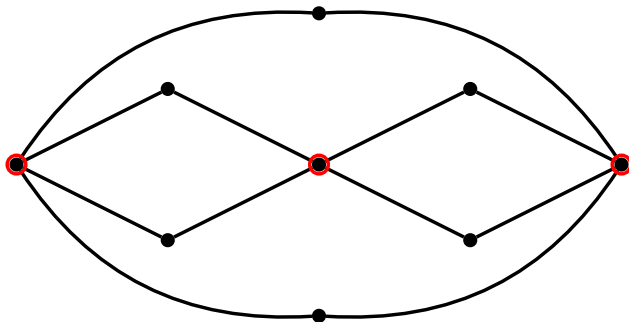
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A polynomial algorithm can be designed using dynamic programming on the series-parallel decomposition tree.





# Nice tree decomposition

A **rooted tree decomposition**  $(T, \{X_t : t \in T\})$  of a graph  $G$  is **nice** if every node  $t \in V(T) \setminus \text{root}$  is of one of the following four types:

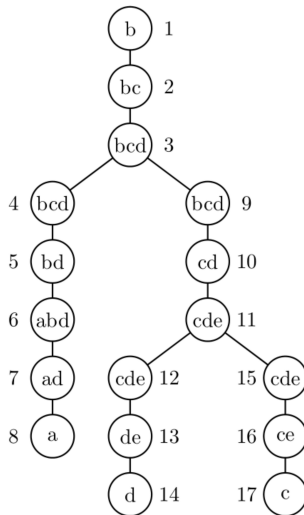
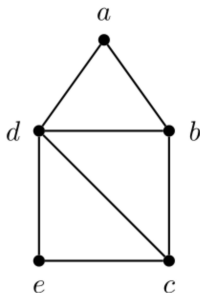
- **Leaf**: no children and  $|X_t| = 1$ .
- **Introduce**: a unique child  $t'$  and  $X_t = X_{t'} \cup \{v\}$  with  $v \notin X_{t'}$ .
- **Forget**: a unique child  $t'$  and  $X_t = X_{t'} \setminus \{v\}$  with  $v \in X_{t'}$ .
- **Join**: two children  $t_1$  and  $t_2$  with  $X_t = X_{t_1} = X_{t_2}$ .

## Theorem (Bodlaender, Kloks, 1996)

*A tree decomposition  $(T, \{X_t : t \in T\})$  of width  $\text{tw}(G)$  of an  $n$ -vertex graph  $G$  can be transformed in time  $\mathcal{O}(\text{tw}(G)^2 \cdot n)$  into a nice tree decomposition of  $G$  of width  $\text{tw}(G)$  and  $4n$  nodes.*



# Nice tree decomposition



# The algorithm

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- The total size of information for one vertex of  $t$  is  $\mathcal{O}(n \cdot \text{tw}(G)^{\text{tw}(G)})$ .



# The algorithm

- For the root  $r$  of the tree compute:

$$\tau(G) = \min \left\{ \frac{s}{\text{MNC}(r, s, Q, \mathcal{P})} \mid 0 \leq s < n; \text{MNC}(r, s, Q, \mathcal{P}) \geq 2 \right\}$$



# How to compute $\text{MNC}(t, s, Q, \mathcal{P})$ ?

- Leaf: trivial
- Forget: easy



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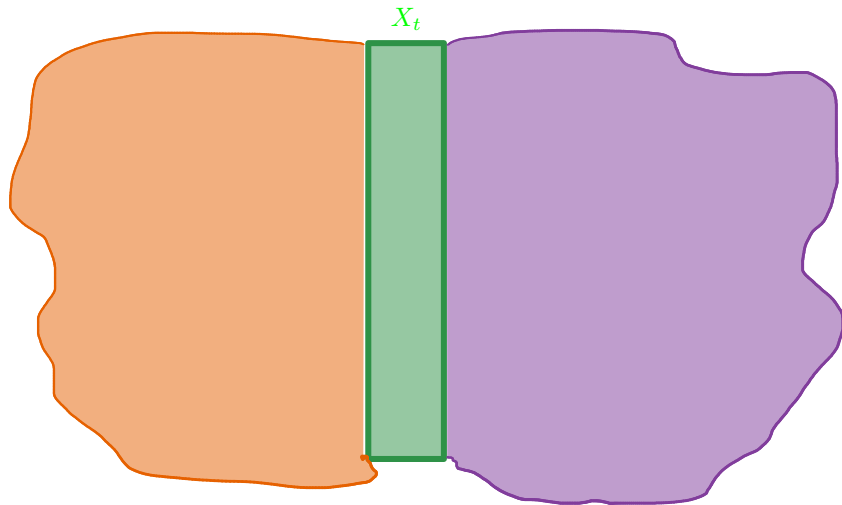


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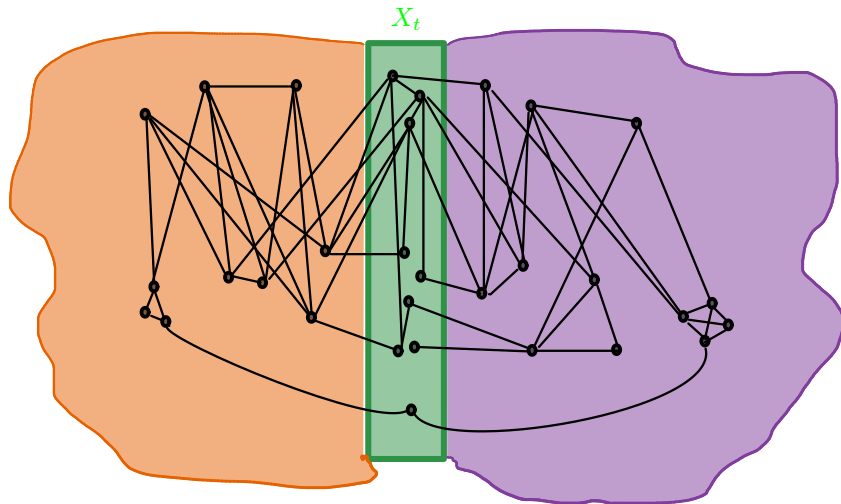
- Leaf: trivial
- Forget: easy
- Introduce: harder
- Join: hardest case



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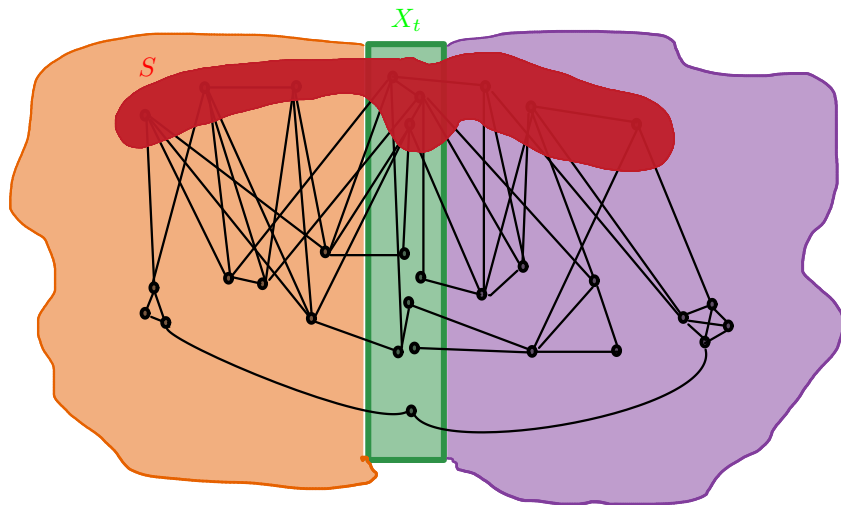


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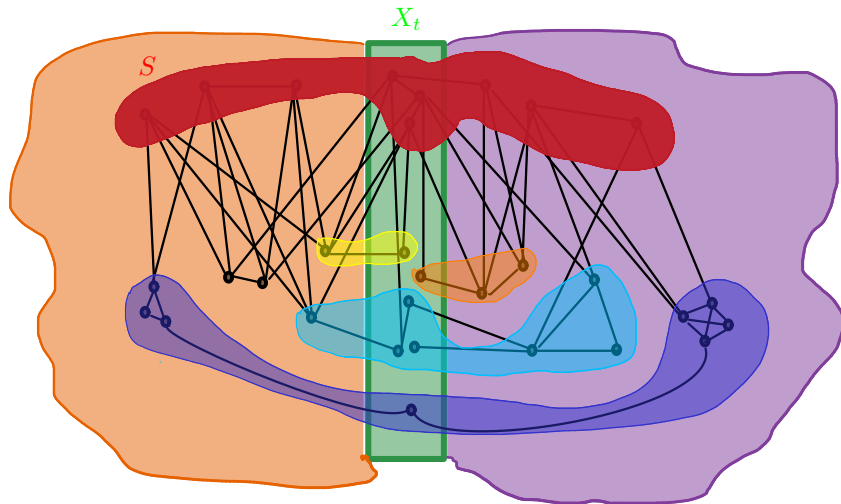




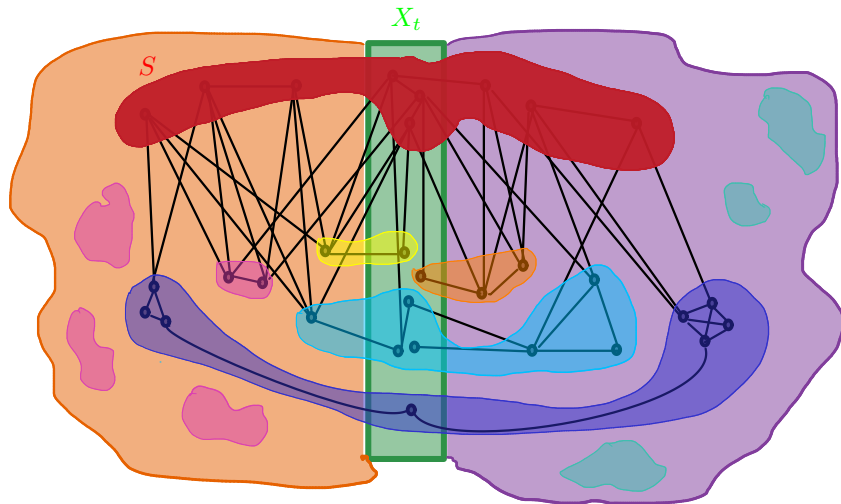
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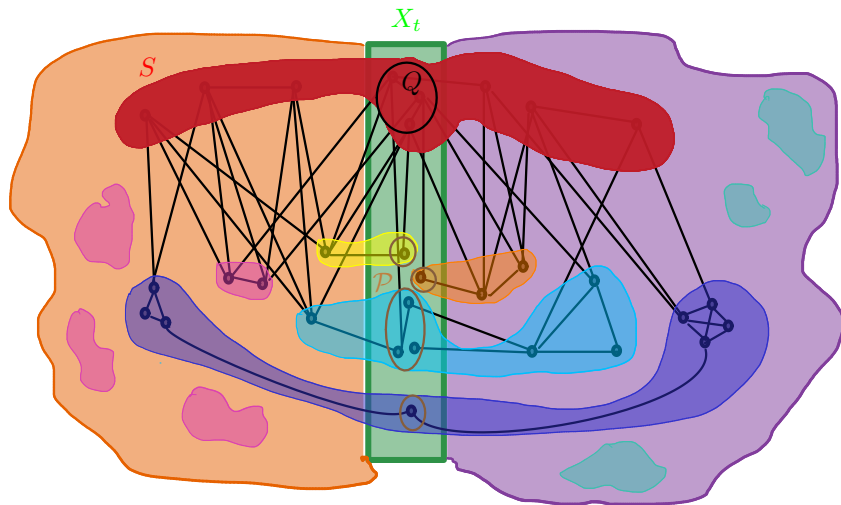
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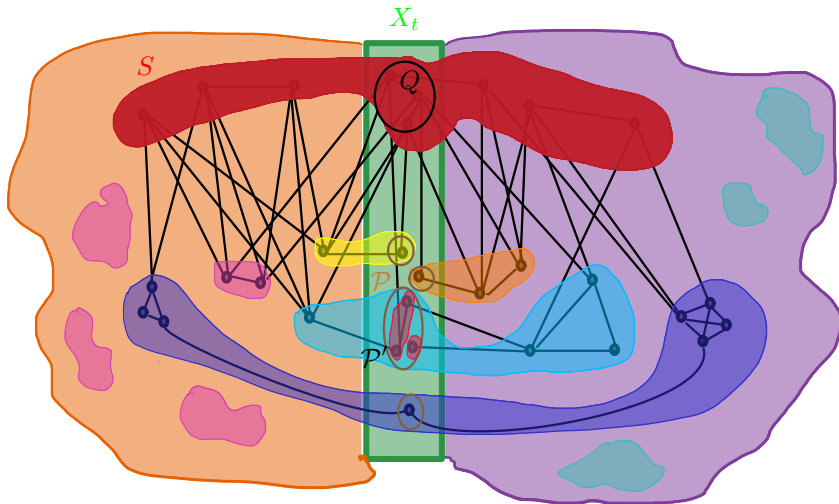
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# Running time

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I believe that the methods invented by Bodlaender, Cygan, Kratsch and Nederlof (2013) will work here, too.



# Open questions

## Question

*What is the complexity of  $t$ -TOUGH for*

- *chordal graphs?*
- *planar graphs?*



# The End

