

Treewidth Computation and Courcelle's Theorem in the CONGEST Model

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Algorithmic Meta Theorems

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*Families of **properties**, often defined by logical conditions, can be **efficiently solved** on **certain classes** of graphs.*

- ✦ **We focus in properties expressible in Monadic Second Order Logic**
- ✦ **We focus in graphs of bounded treewidth**
- ✦ **$\tilde{O}(D)$ rounds**

Monadic Second-Order Logic (MSO)

A property $\mathcal{P}$ is MSO-expressible if it can be defined using quantification over **vertices/edges** and over monadic **sets of vertices/edges**.

2-VERTEX COVER

Given a graph $G = (V, E)$ a vertex cover is a set $U \subseteq V$ such each each $uv \in E$ satisfies that $u \in S$ or $v \in S$

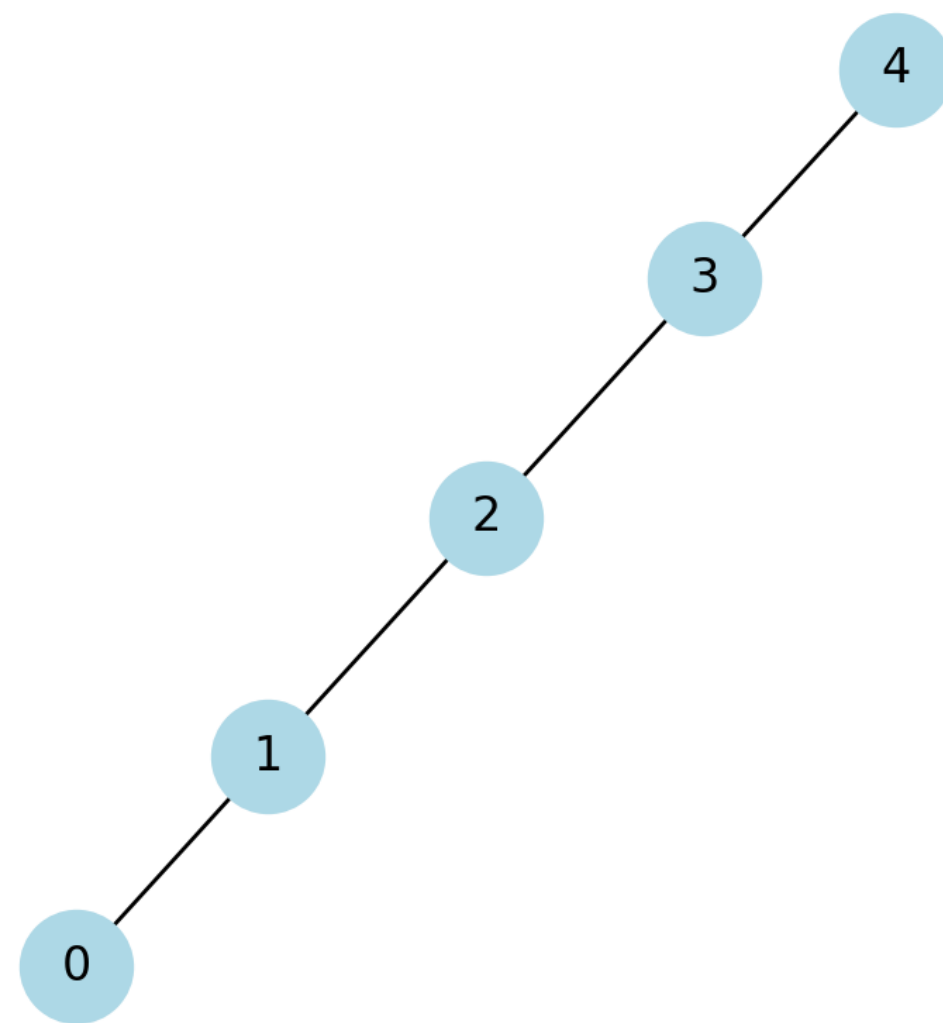
$$(\text{MSO}) \exists X \subseteq V, \exists x_1 \in V, \exists x_2 \in V, \forall y \in S, (y = x_1 \vee y = x_2)$$

$$\forall u \in V, \forall v \in V, (uv \in E \implies u \in X \vee v \in X)$$

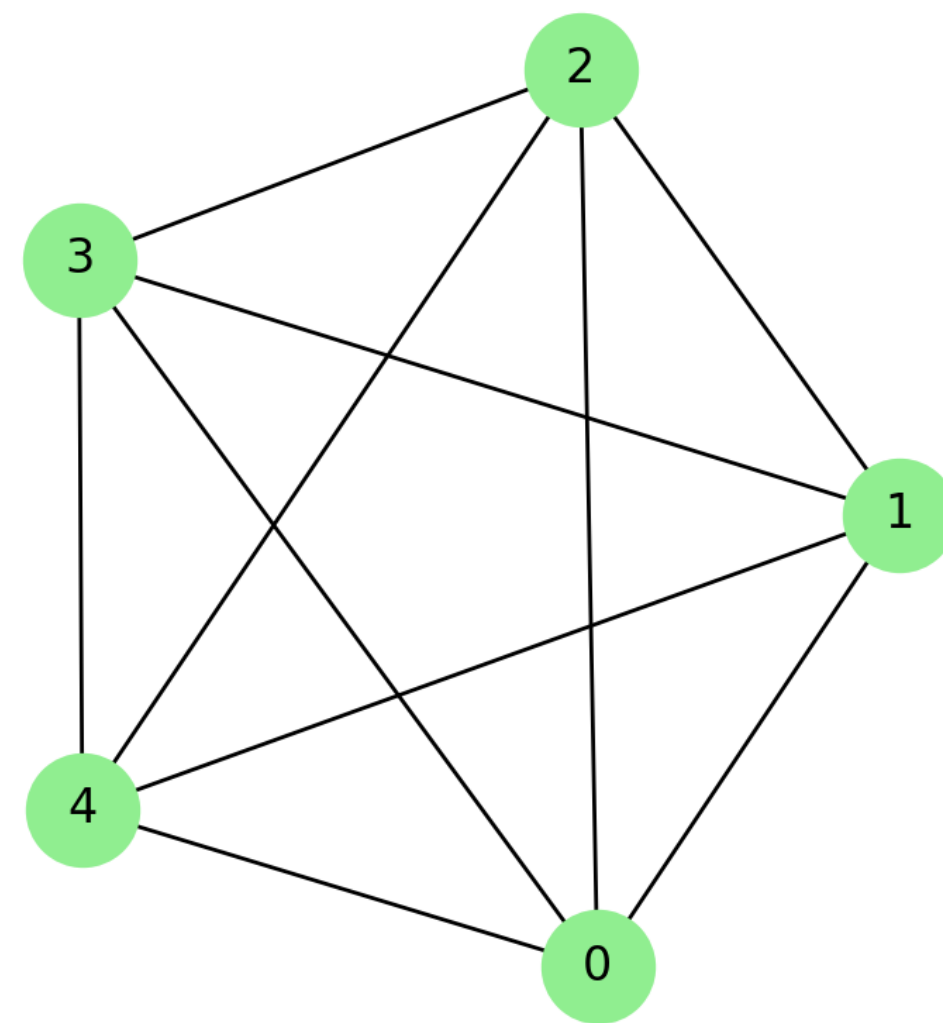
Optimization problem: Given G , find a vertex cover S of *minimum size*

Treewidth of a graph

- **Treewidth** is a **parameter** of a graph
- Characterizes how close a graph is to being a tree.



Treewidth 1



Treewidth $|V| - 1$

Class of k -bounded treewidth



Graphs with treewidth $\leq k$

Courcelle's Theorem

Courcelle's Theorem (Courcelle, 90'. Informal). Let φ be an MSO property and $k \in \mathbb{N}$. There exists an algorithm such that, for any graph G with $tw(G) \leq k$, decides if $G \models \varphi$ in linear time in the size of G .

Optimization version

Theorem (Borie, Parker and Tovey, 92'; Courcelle and Mosbah, 93'. Informal).

Given a graph optimization problem $\mathcal{P}$ of an MSO property and a class $\mathcal{G}$ of k -bounded treewidth graphs, there exists an algorithm such that, for any $G \in \mathcal{G}$, solves $\mathcal{P}(G)$ in linear time

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➔ **Optimization version of many NP problems solved in linear time in k -bounded treewidth graphs.**

Maximum Independent Set, Minimum Vertex Cover, etc...

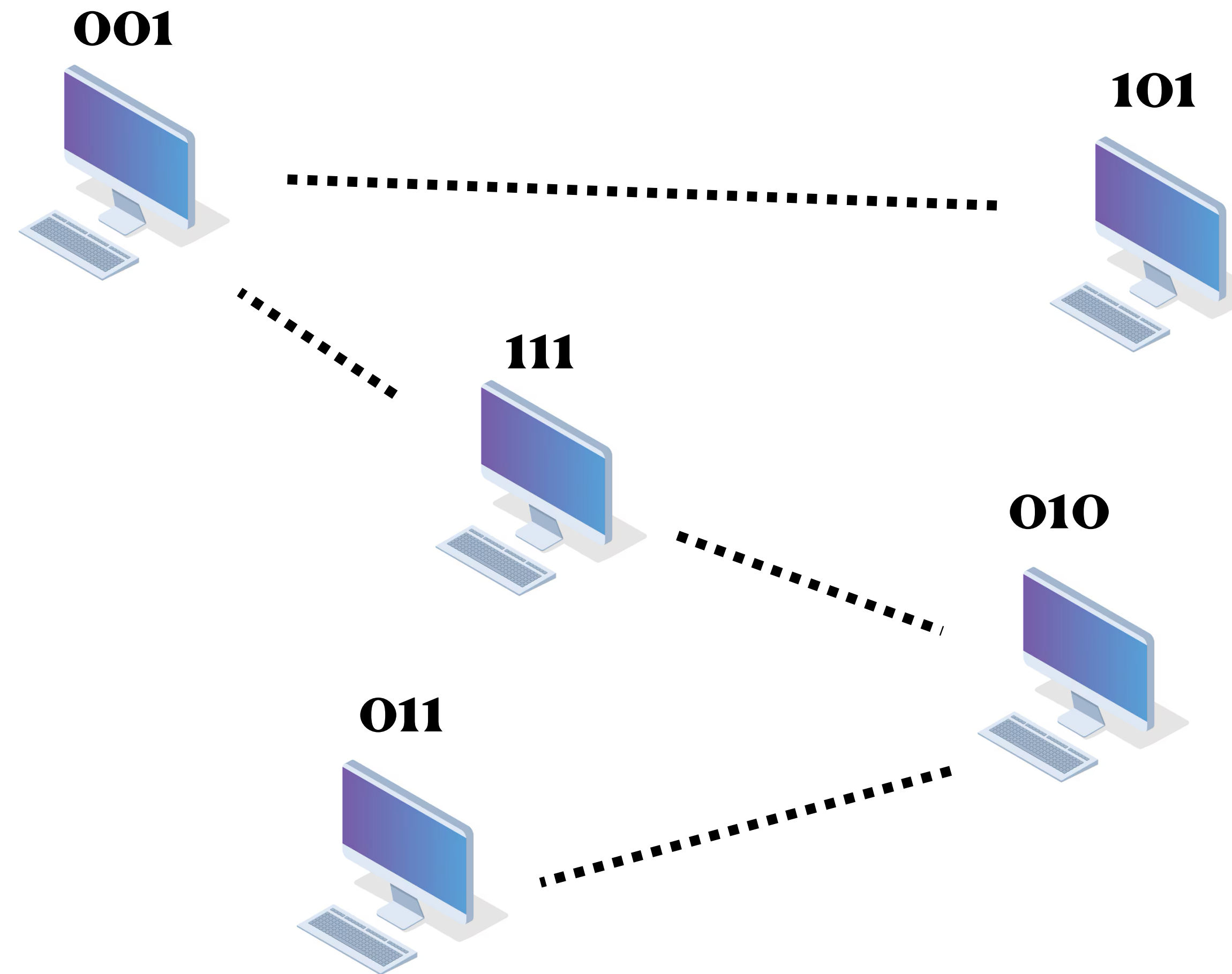
➔ **Linear in $|G|$ but super-exponential dependance in k**

Distributed Computing

CONGEST Model

Distributed network represented as n -node graph $G = (V, E)$

- Each node has a unique **ID**.
- Nodes interact with their neighbors in **synchronous rounds**.
- Message size of $\mathcal{O}(\log n)$ bits.
- **Round complexity:** Maximum number of rounds a node in G sends a message to a neighbor.



Recent years

- ❖ Special interest in **sparse classes of graphs**
- ❖ Mainly **planar graphs**

Problem	General	Planar Graphs	Reference
Reachability	$\Omega(D + \sqrt{n}/\log n)$	$\tilde{\mathcal{O}}(D)$	Parter, 2020.
MST	$\Omega(D + \sqrt{n})$	$\tilde{\mathcal{O}}(D)$	Ghaffari and Haeupler, 2016.
Max-Flow	$\Omega(D + \sqrt{n})$	$\tilde{\mathcal{O}}(D)$	Y. Abd-Elhaleem et al., 2025

MSO properties?

❖ Many of them are **hard to solve** in CONGEST.

Problem	General	Reference
Maximum IS	$\Omega(n^2/\log^2 n)$	Censor-Hillel, 2017.
Maximum Matching	$\tilde{\mathcal{O}}(\mu(G)^2)$	Ben-Basat et al et al. 2018
Minimum Vertex Cover	$\mathcal{O}(n^2/\log n)$	Folklore

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Can we improve them in bounded treewidth graph?

Our results

(Decision version)

Theorem (J. Li, P. Montealegre, I. Todinca, B.J.). Given a fixed integer k and an MSO-property φ , there is a deterministic CONGEST algorithm that, given an n -node graph with diameter D and treewidth at most k , decides $G \models \varphi$ in $\tilde{\mathcal{O}}(D)$ rounds.

(Optimization version)

Theorem (J. Li, P. Montealegre, I. Todinca, B.J.). Given a fixed integer k and an optimization problem $\mathcal{P}$ from an MSO-property φ , there is a deterministic CONGEST algorithm that, given an n -node graph with diameter D and treewidth at most k , computes $\mathcal{P}(G)$ in $\tilde{\mathcal{O}}(D)$ rounds.

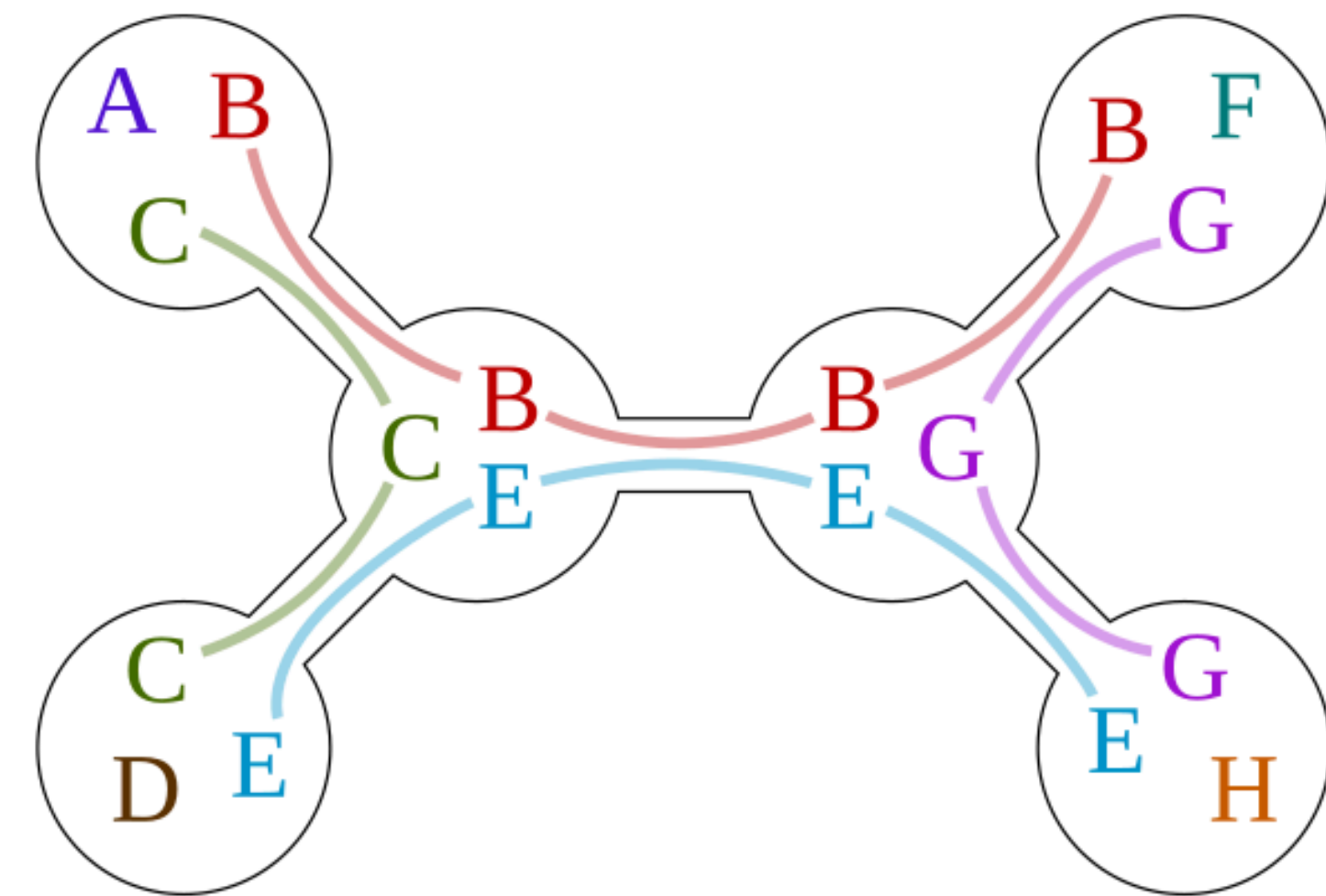
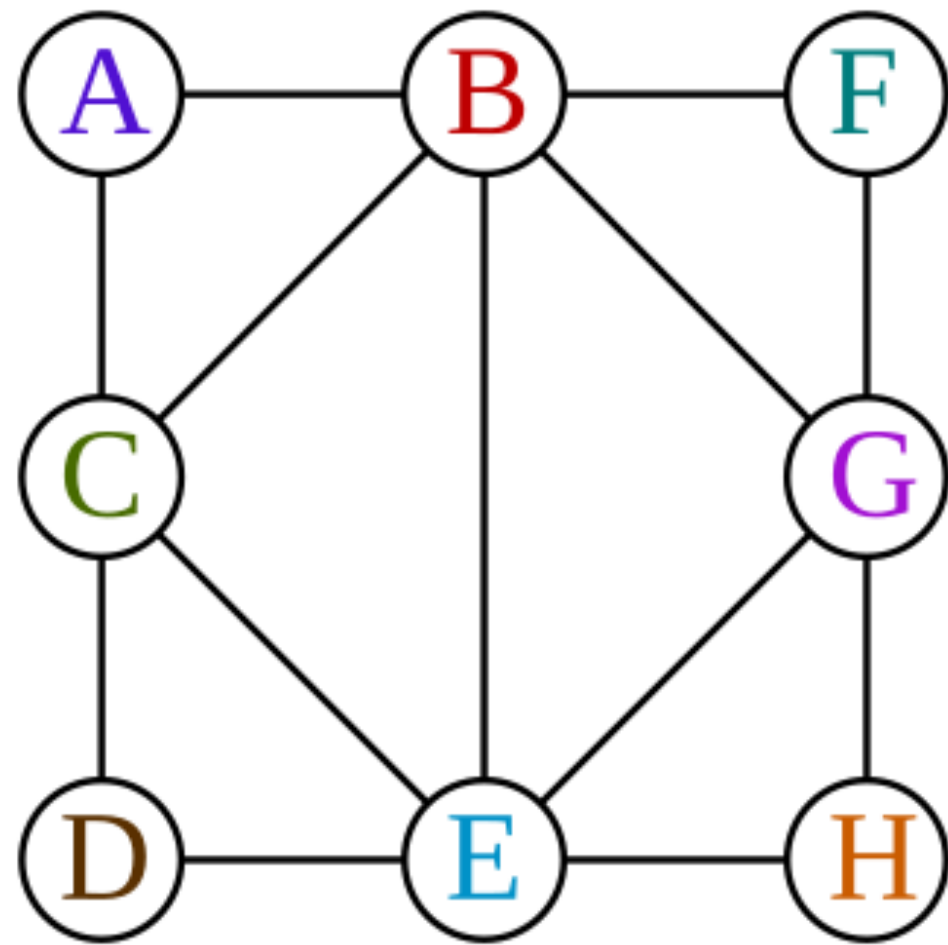
Applications

Problem	General	Bounded treewidth
Maximum IS	$\Omega(n^2/\log^2 n)$	$\tilde{\mathcal{O}}(D)$
Maximum Matching	$\tilde{\mathcal{O}}(\mu(G)^2)$	$\tilde{\mathcal{O}}(D)$
Minimum Vertex Cover	$\mathcal{O}(n^2/\log n)$	$\tilde{\mathcal{O}}(D)$

... Minimum Dominating Set, Maximum Clique, etc...

High Level Idea

TREE DECOMPOSITION: Pair (X, T) . T is a tree, $V(T) = X$ and $X \subseteq \mathcal{P}(V)$



- (1) Union of bags equals V
- (2) Every edge is contained in a bag
- (3) For every node v , graph induced by bags containing v is connected

Courcelle's idea

Theorem. Let G be an n -node graph with $tw(G) \leq k$. Then G admits a tree decomposition (X, T) such that

- T has depth $O(\log n)$ and,
- For every node $t \in X$, the bag B_t has size $O(k)$.

ALGORITHM

(I) Compute a tree decomposition (X, T)

(II) Apply dynamic programming over the tree decomposition

Distributed implementation

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**(II) Apply dynamic
programming over the tree
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Distributed implementation

(I) Compute a tree decomposition (X, T)



How to represent a tree decomposition?



How to build a tree decomposition?

(II) Apply dynamic programming over the tree decomposition

Distributed implementation

(I) Compute a tree decomposition (X,T)



How to represent a tree decomposition?



How to build a tree decomposition?

(II) Apply dynamic programming over the tree decomposition



How to apply dynamic programming

In $\tilde{O}(D)$ rounds?

Computing a tree decomposition

Definition. Given an n -node graph G a separator set S of G is a subset of nodes such that the size of each connected component of $G - S$ is at most $2n/3$

One Separator

Lemma (J. Li, P. Montealegre, I. Todinca, B.J.) There exists a CONGEST algorithm such that given a graph G with $tw(G) = k$ and any connected subgraph H , computes in $\tilde{\mathcal{O}}(D)$ rounds, a separator set of size at most k

Computing a tree decomposition

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Disjoint Separators

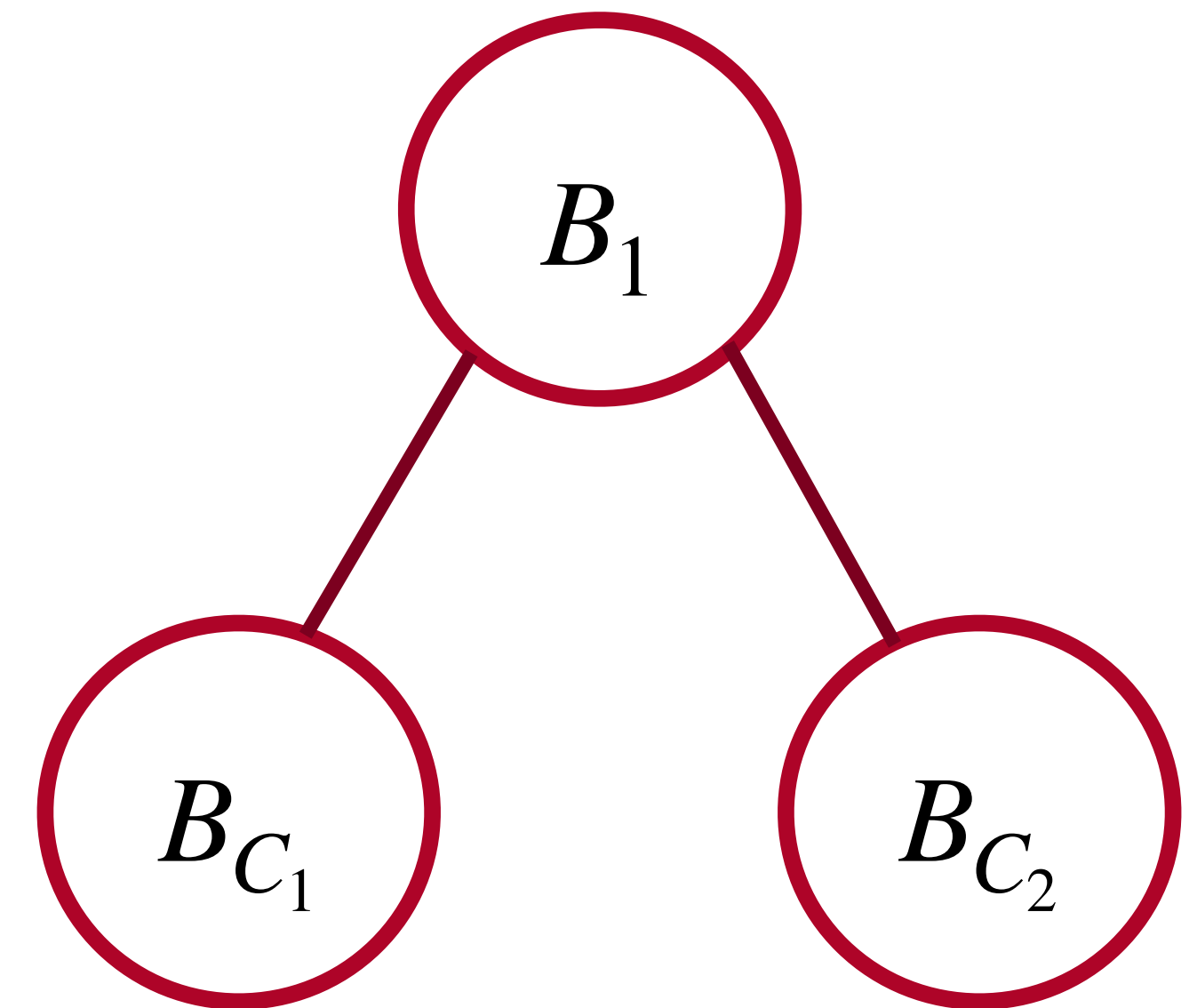
Lemma (J. Li, P. Montealegre, I. Todinca, B.J.) There exists a CONGEST algorithm such that given a graph G with $tw(G) = k$ and any collection $H_1, \dots, H_j$ of disjoint and connected subgraphs, computes in $\tilde{\mathcal{O}}(D)$ rounds, in parallel, a separator set of size at most k

Distributed computation

Recurse in instances (U, X) with $U \subseteq V, X = N(U)$.

Start with $(V, \emptyset)$

- ▶ $S_V = \text{separator}(G)$
- ▶ $B_1 = S_V$ **(root bag)**
- ▶ **In parallel in each c.c. C_i of $G - B_1$**
 - $S_{C_i} = \text{separator}(C_i)$
 - $B_{C_i} = S_{C_i} \cup N(C_i)$ (**)
 - **Continue recursively in c.c. of $C_i - S_{C_i}$**

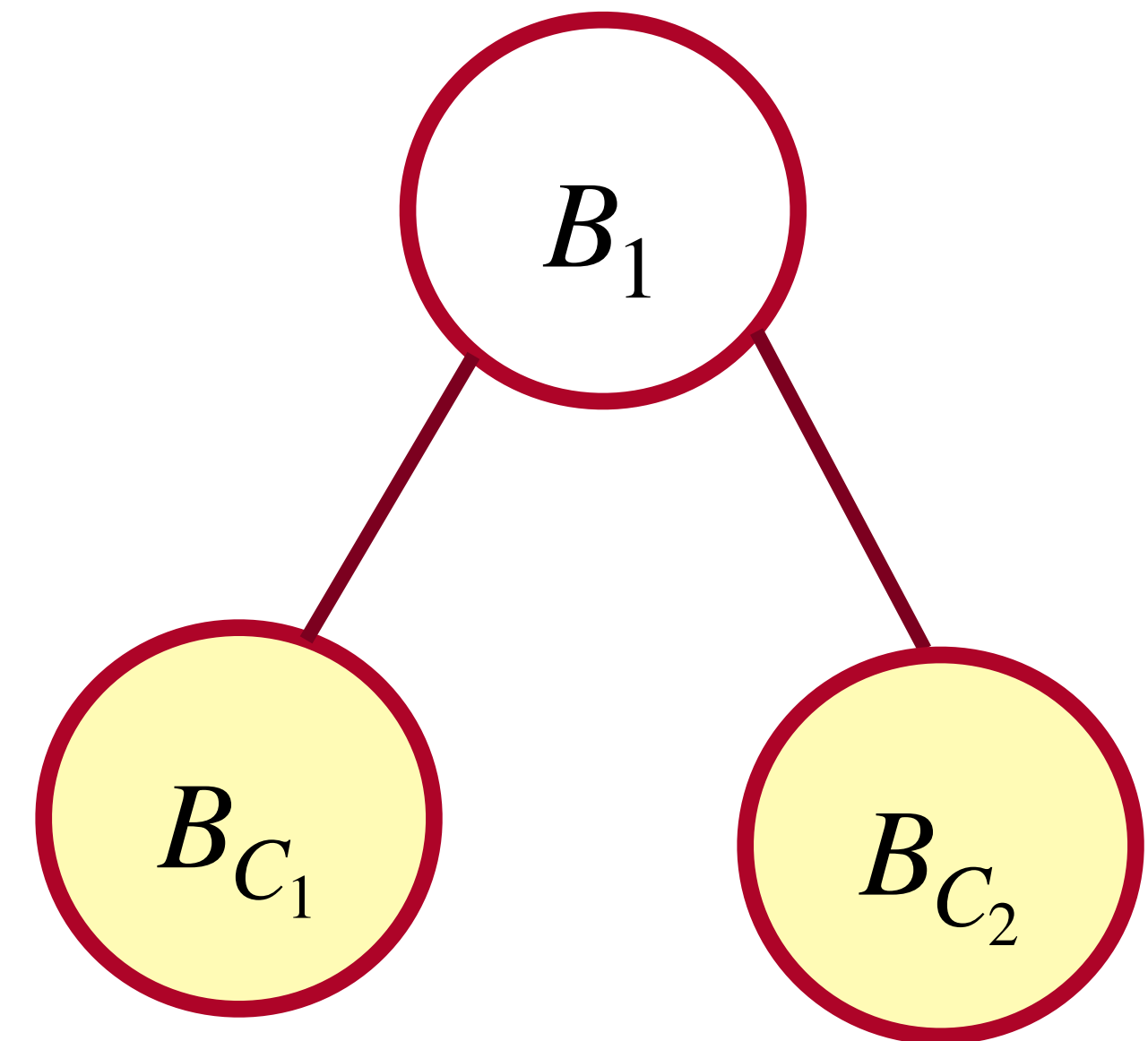


Properties

- ✦ **Size $\mathcal{O}(k)$ of each bag**
- ✦ **Depth $\mathcal{O}(\log n)$**
- ✦ **Inherently we obtain a distributed representation in each node**
- ✦ **Each bag has a leader**

Dynamic programming: Maximum Independent Set

(1) Leaf bags compute all possible solutions greedily.

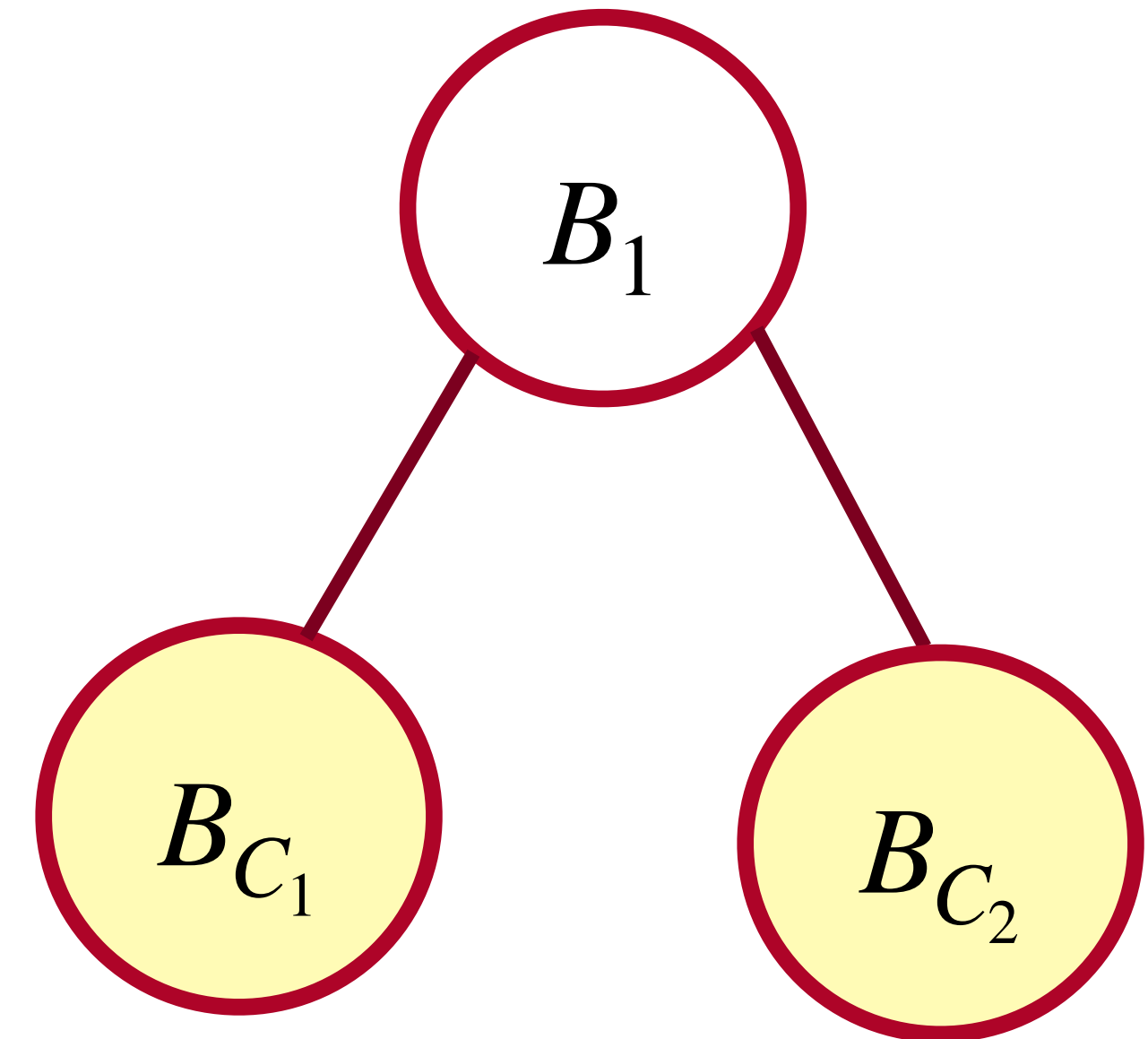


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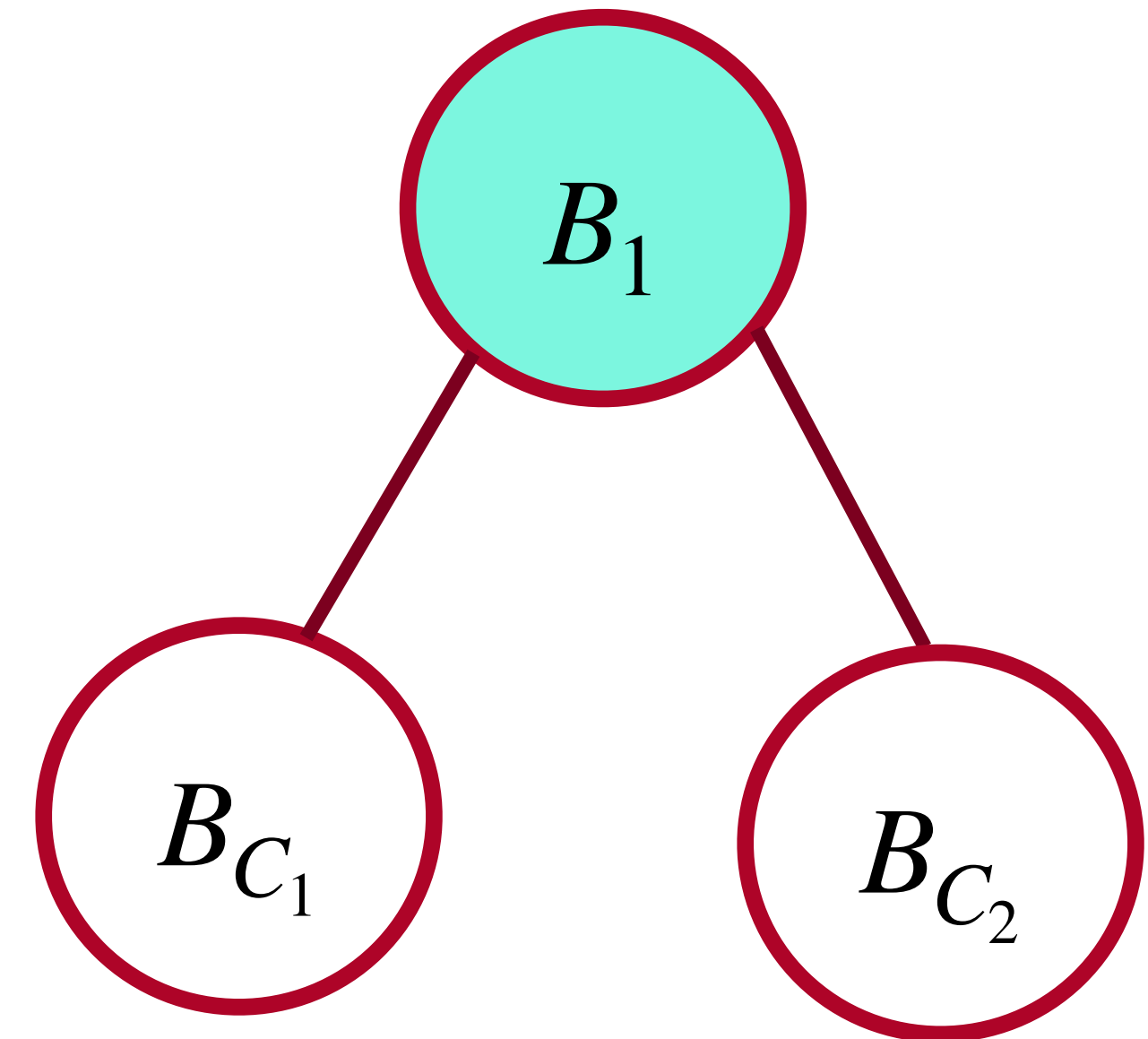
$$OPT(X_{C_i}) = \left| \text{MaxIS} \left(G \left[X_{C_i} \right] \right) \right|$$

$2^{O(k)}$ rounds



Dynamic programming: Maximum Independent Set

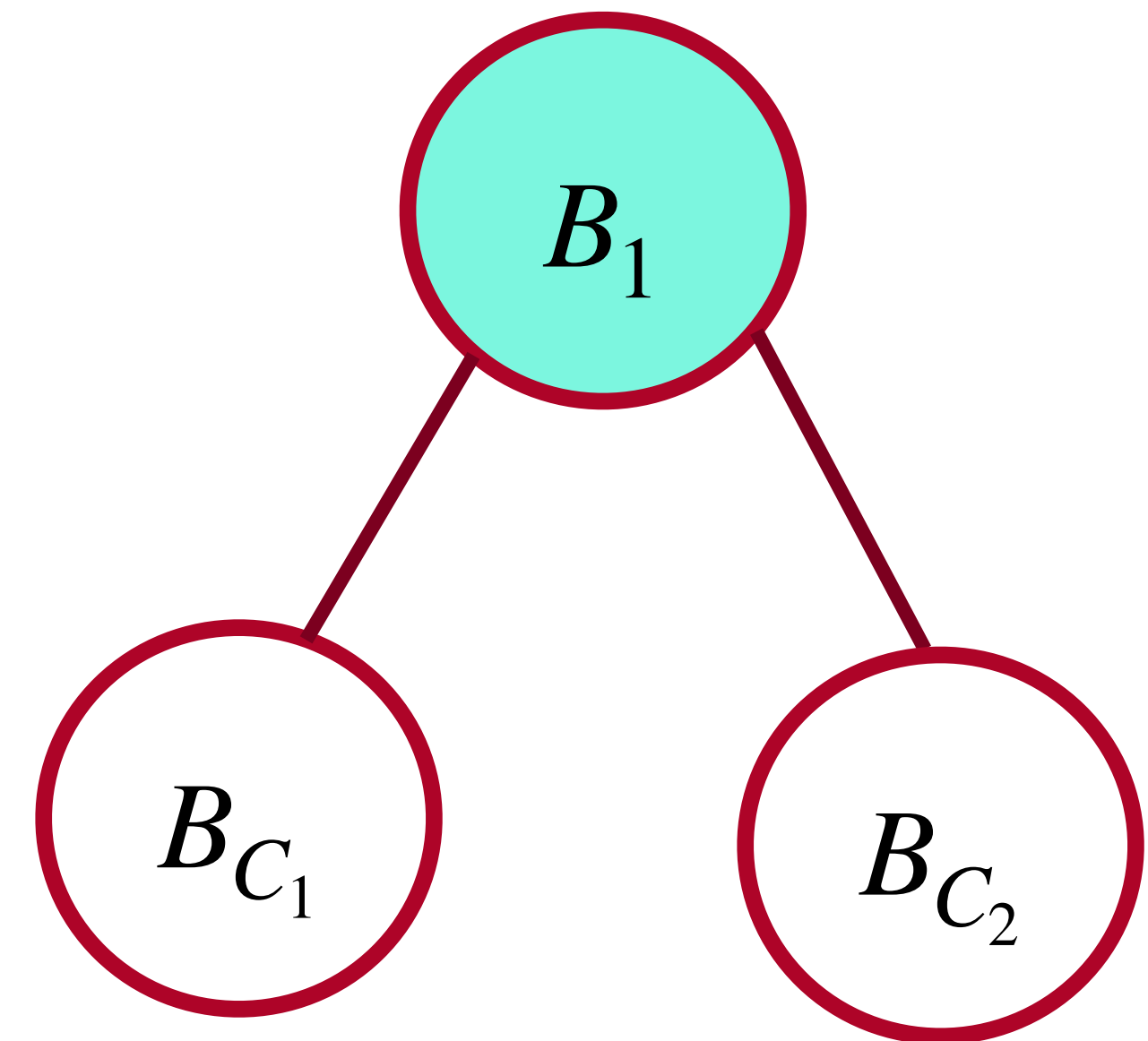
(2) Non-leaf bag: Extend the solution of its children



Dynamic programming: Maximum Independent Set

(2) Non-leaf bag: Extend the solution of its children

“Compute *commutative and associative* function of local input”



Great tool: Partwise aggregation problems

$\tilde{O}(k \cdot D)$ rounds

Summary

- ➡ Deterministic implementation of Courcelle's Theorem in CONGEST.
- ➡ Improve to $\tilde{\mathcal{O}}(D)$ rounds several problems **with the same algorithm**
- ➡ Decision, optimization and **counting** version implemented.



Paper