

# Distributed Computing

## 12 - Sleeping Model

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INSTITUT  
DE RECHERCHE  
EN INFORMATIQUE  
FONDAMENTALE

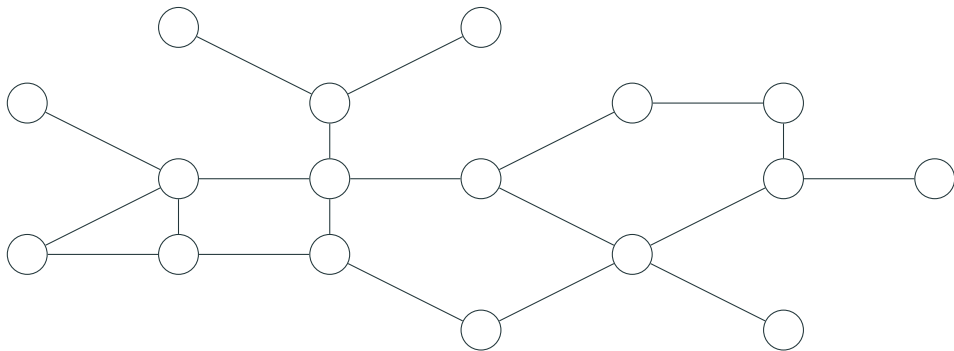


## The Awaken Complexity

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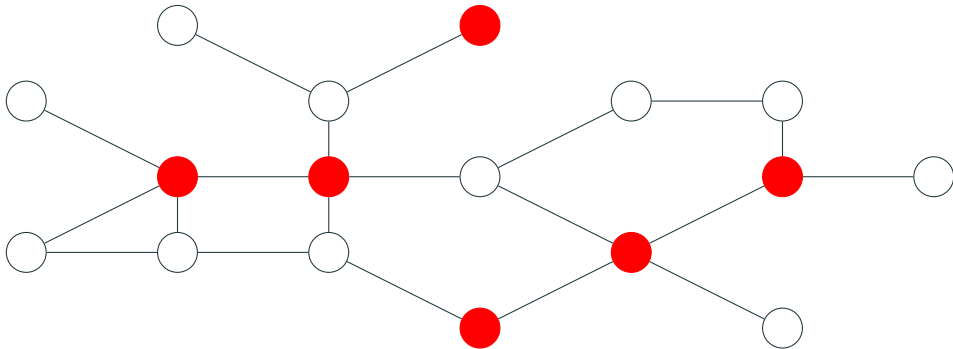
## Distributed Sleeping Model

- LOCAL model
- At each round, a node decides if it is active or not
- A node communicates only with its active neighbors
- Complexity : maximal number of awaken rounds for a single node



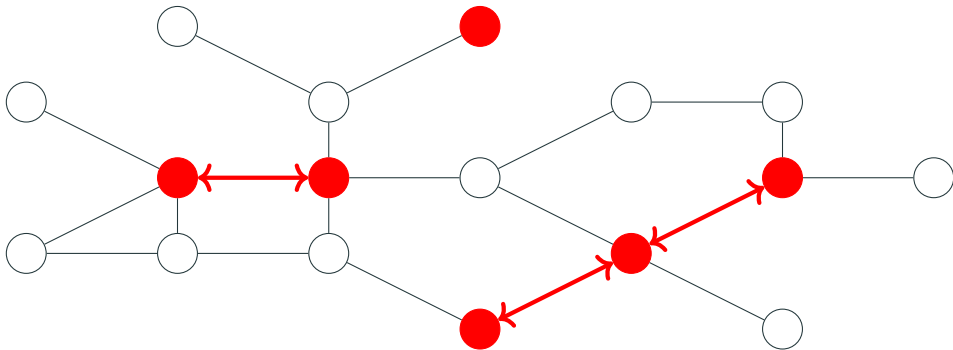
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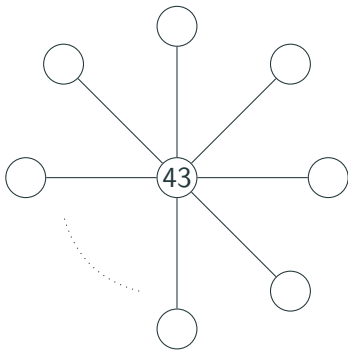


# Distributed Sleeping Model

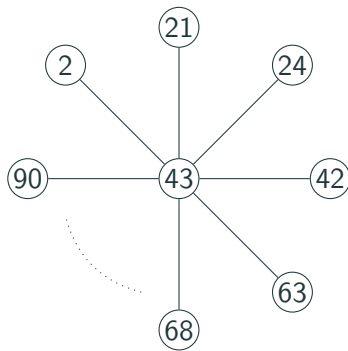
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## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds

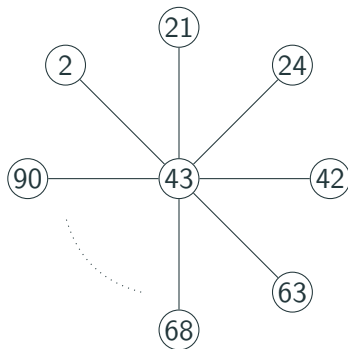


## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



- Round 0 : Learn the identifiers of my neighbors

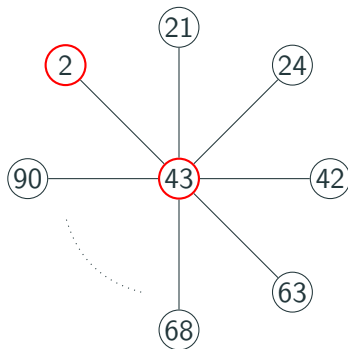
## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up

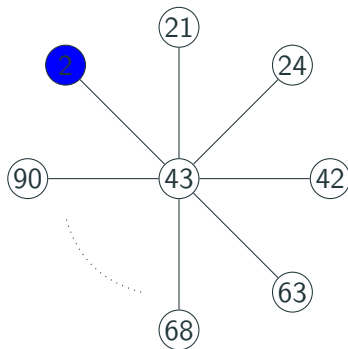


## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



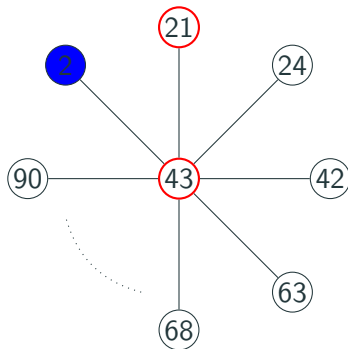
- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 2 : Node 2 chooses its color

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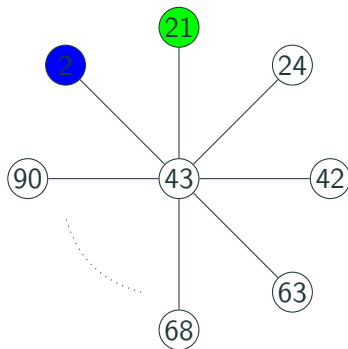
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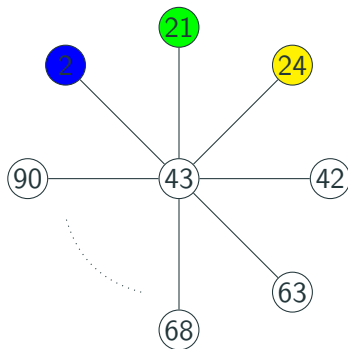
- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 21 : Node 21 chooses its color

## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



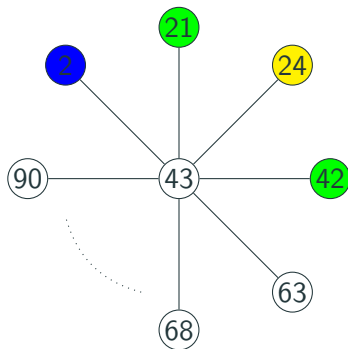
- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 21 : Node 21 chooses its color

## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



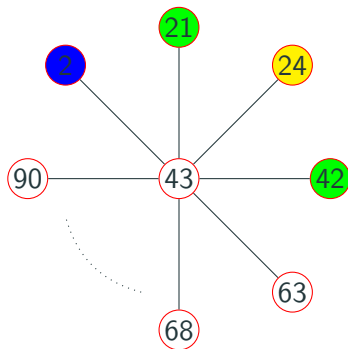
- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 24 : Node 24 chooses its color

## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



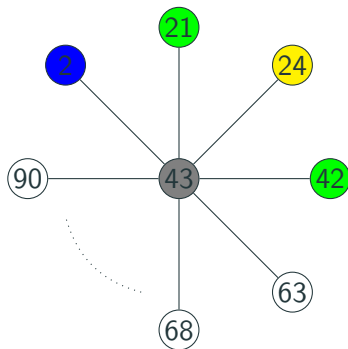
- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 42 : Node 42 chooses its color

## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



- Round 0 : Learn the identifiers of my neighbors
- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 43 :  $u$  learns colors of nodes 2, 21, 24, 42 and chooses its color

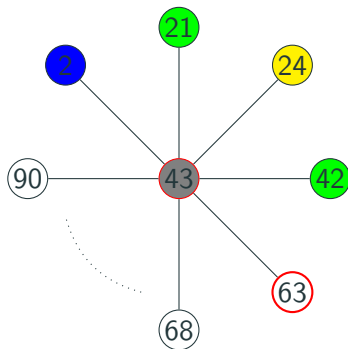
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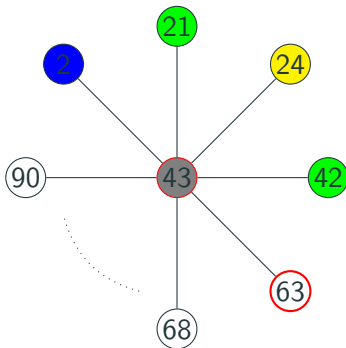


## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



- Round 0 : Learn the identifiers of my neighbors
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- Round 63 : Node 63 learns color of  $u$  and chooses its color

## $\Delta + 1$ -Coloring in $O(\Delta)$ awaken rounds



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- For each  $i \in N(u)_{\leq 1}$ , round  $i$  : Wake up
- Round 63 : Node 63 learns color of  $u$  and chooses its color

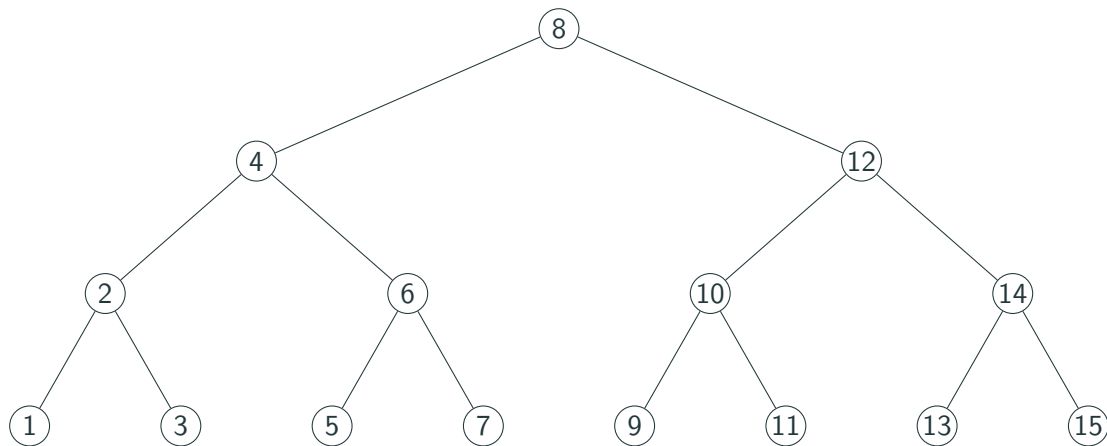
Drawback : The round complexity is  $O(M)$ ,  $M$  being the maximal identifier.

## Reduce $K$ colors in $\log K$ awake rounds

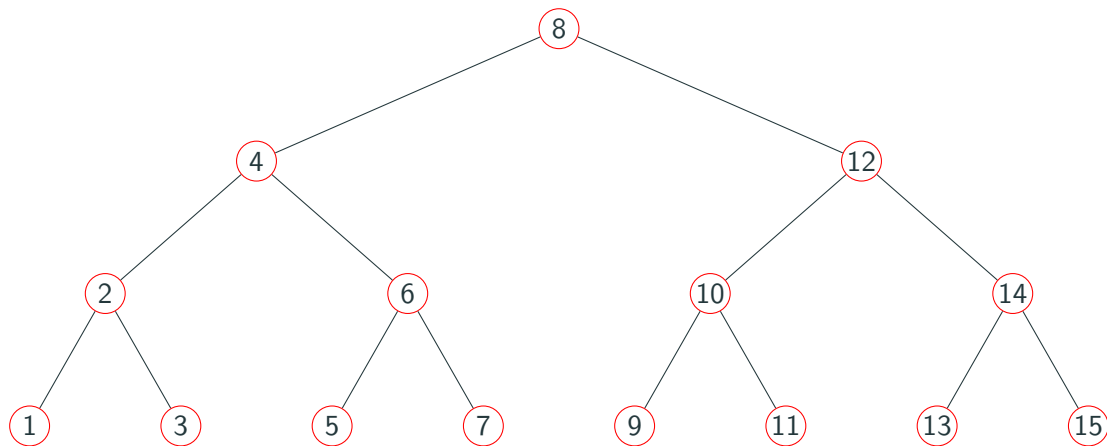
### Barenboim and Maimon (2021)

Given a  $K$ -coloring of the graph, we can compute a  $(\Delta + 1)$ -coloring in  $O(\log K)$  awaken rounds and  $O(K)$  rounds in the Sleeping LOCAL model.

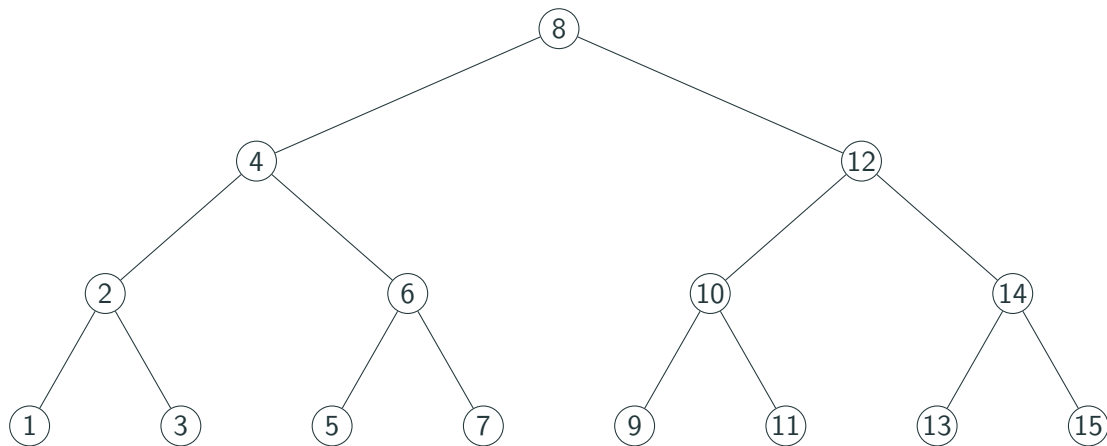
## Reduce $K$ colors in $\log K$ awake rounds



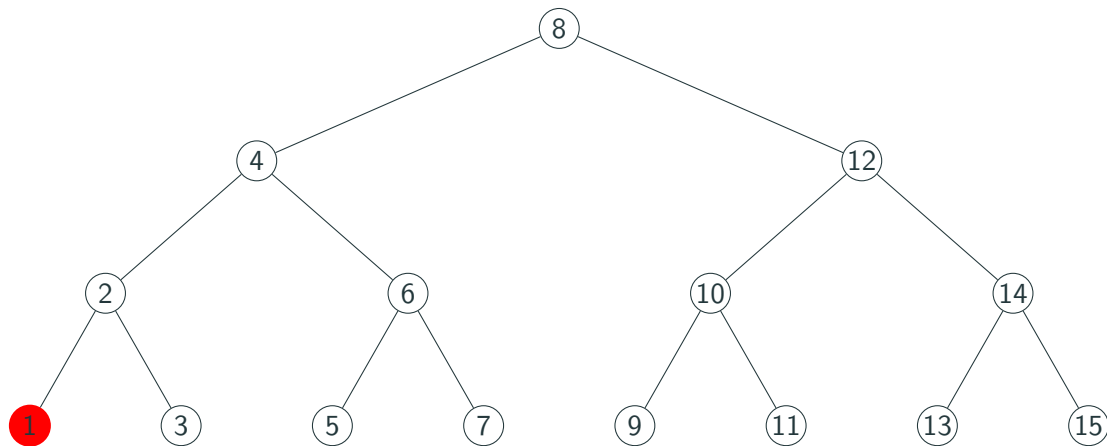
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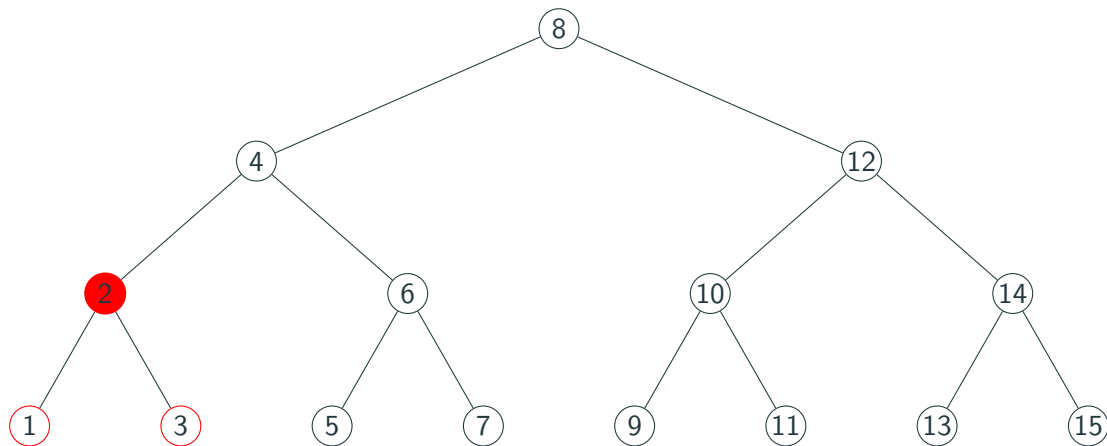
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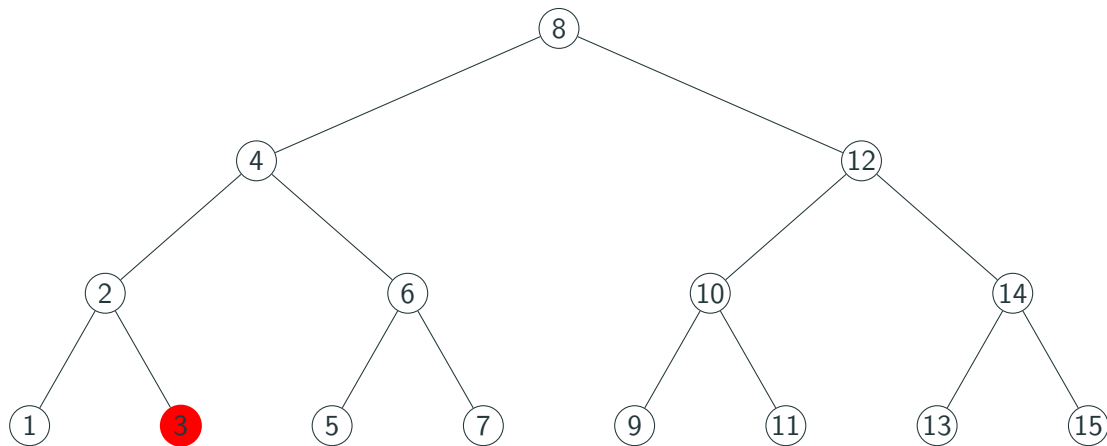


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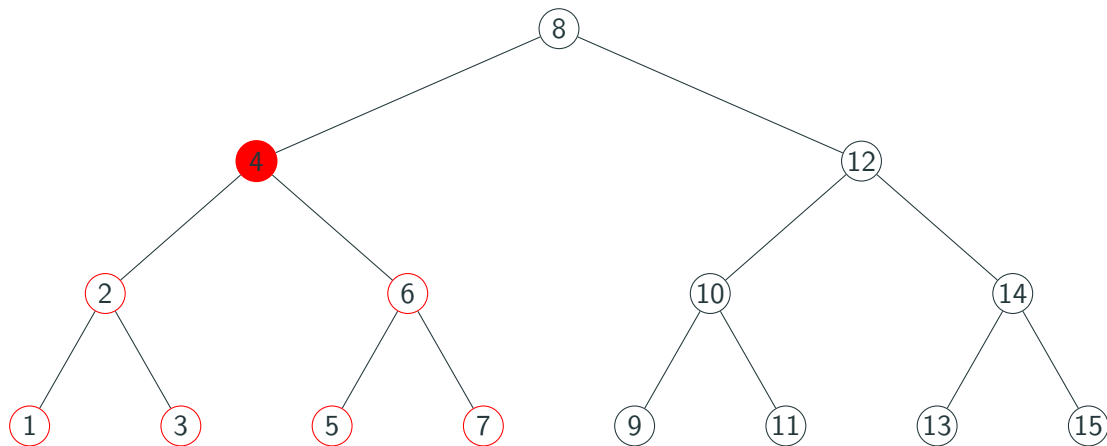




## Reduce $K$ colors in $\log K$ awake rounds



## Reduce $K$ colors in $\log K$ awake rounds



## The $\log n$ Complexity

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# Full Knowledge of the Graph

## Barenboim and Maimon (2021)

Any graph problem can be solved in  $O(\log n)$  awaken rounds in the Sleeping LOCAL model.

This algorithm takes  $O(\text{poly } M)$  rounds.

# Full Knowledge of the Graph

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Distributed Layered Tree (DLT) - Oriented Spanning Tree such as :

- Each vertex has a label
- The label of a vertex is bigger than its parent's
- Each vertex knows the label of its neighbours in the tree

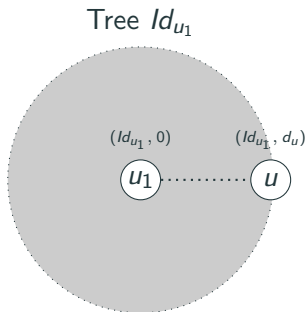
## Constant Coordination

Broadcast and Convergecast can be done in  $O(1)$  rounds in a DLT.

### Barenboim and Maimon (2021)

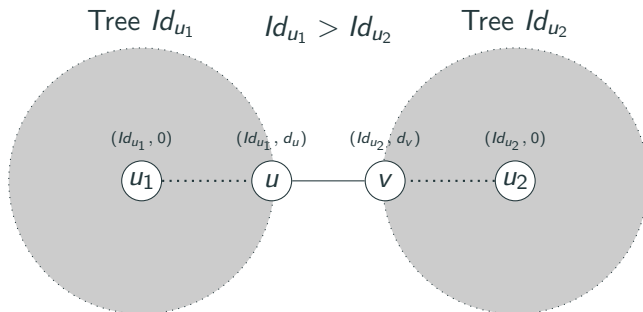
A DLT can be built in  $O(\log n)$  awaken rounds in the Sleeping LOCAL model.

# Building a DLT



- Labels are of the form  $(a, b)$ , ordered lexicographically.
- At the beginning, all nodes have label  $(Id(u), 0)$ .
- At the beginning of each expand step, all nodes of a subtree  $T$  are of the form  $(L(T), b)$ .

# Building a DLT

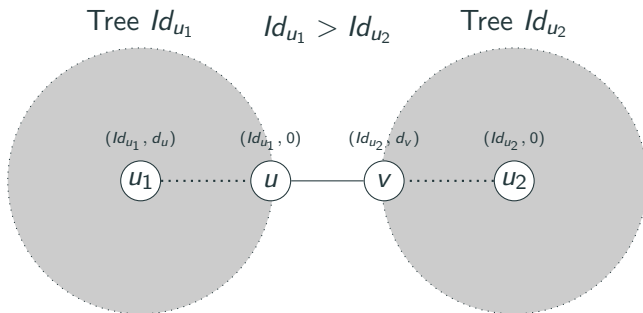


■ Repeat  $\log n$  times :

1. Select a neighbour Tree  $T'$  with smaller label ( $Id_{u_1} > Id_{u_2}$ ).

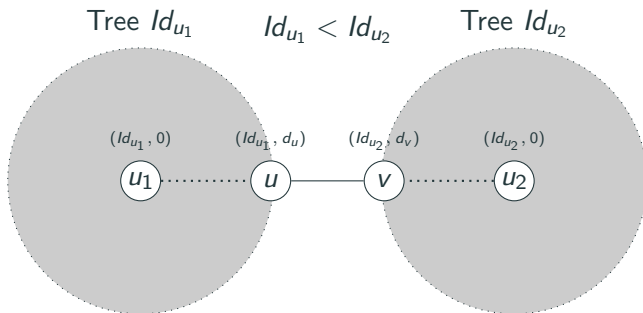


# Building a DLT

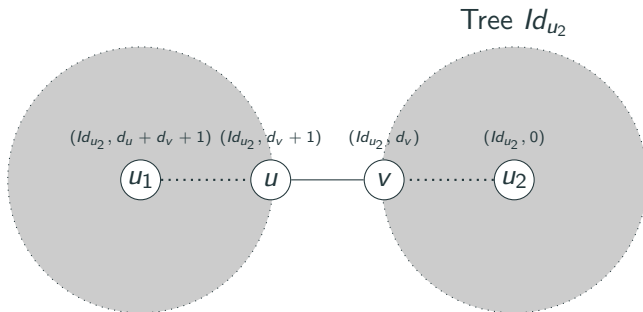


- Repeat  $\log n$  times :
2. Merge  $T$  and  $T'$ , using an edge  $(u, v)$ .

## Building a DLT



- Repeat  $\log n$  times :
3. If  $T$  could not choose a neighbour and was not selected  
 $T$  chooses a tree  $T'$  to join using an edge  $(u, v)$ .  
This forms a star of trees around  $T' \Rightarrow O(1)$  merge rounds.



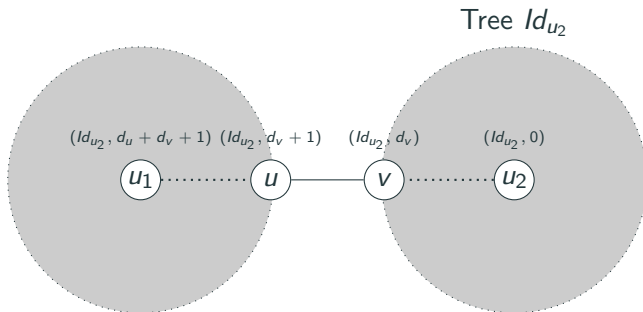
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# Building a DLT



- Repeat  $\log n$  times :
4. All nodes learn their new neighbours in the tree.
  5. Convergeicast to gather the new structure of the component  $C$  to the root  $r$ .
  6. Broadcast a new labelling  $(L(r), dist(r))$ .

## Sleeping Lower Bound

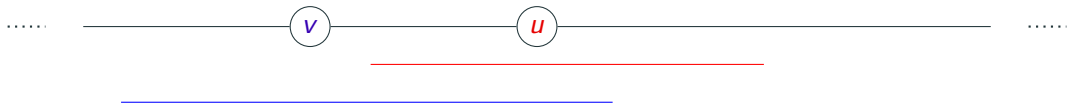
**Augustine et. al (2022)**

Any algorithm to solve 2-coloring with probability exceeding  $1/8$  on a ring network requires  $\Omega(\log n)$  awake time.

# Sleeping Lower Bound

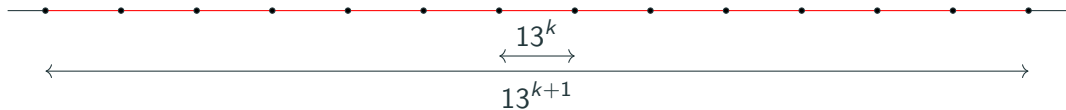
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- After  $k$  rounds, a node knows about some segment that includes itself
- No node  $v$  on the left of  $u$  in the path can know more than  $u$  on its right

## Sleeping Lower Bound



By induction : For any  $k$ , for any segment  $I$  of  $13^k$  nodes, there exists, with probability  $\mathcal{P} > 1/2$ , a node  $u \in I$  who knows less than  $I$  after  $k$  rounds.

## Sleeping Lower Bound



By induction : For any  $k$ , for any segment  $I$  of  $13^k$  nodes, there exists, with probability  $\mathcal{P} > 1/2$ , a node  $u \in I$  who knows less than  $I$  after  $k$  rounds.

- Probability that it is true on 5 of the 13 subsegments is at least  $5/6$
- Probability that  $B, C$  or  $D$  wakes up before  $A$  and  $E$  is at least  $1/2$



## $(\Delta + 1)$ -Coloring

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## Trade-Off

Find the possible trade-off between awaken and usual rounds to resolve a problem.

$(\Delta + 1)$ -coloring of paths :

Awaken rounds	Rounds

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$(\Delta + 1)$ -coloring of paths :

Awaken rounds	Rounds
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$O(\log^* n)$	$O(\log^* n)$
$3 + k$	$O(\log^{(2k)} M)$

## Trade-Off

Find the possible trade-off between awaken and usual rounds to resolve a problem.

$(\Delta + 1)$ -coloring :

Awaken rounds	Rounds
$O(\Delta)$	$O(M)$
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$O(\log^* n + \log \Delta)$	$O(\log^* n + \text{poly } \Delta)$

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**Linial (1992)**

There exists an algorithm that solves  $O(\Delta^2)$ -coloring with round-complexity  $O(\log^* n)$ .

## Trade-Off

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$O(\log^* n + \log \Delta)$	$O(\log^* n + \text{poly } \Delta)$

**Balliu, Fraigniaud, Olivetti, R.**

There exists an algorithm that solves  $(\Delta + 1)$ -coloring with  $O(\sqrt{\log n} \cdot \log^* n)$  awake-complexity and round-complexity  $\text{poly}(M)$ .



## Uniquely-labeled BFS-clustering :

- Two functions  $(\ell, \delta)$  assigning a pair  $(\ell(v), \delta(v)) \in \mathbb{N} \times \mathbb{N}$  to each node  $v \in V$
- $\forall i > 0, V_i = \{v \in V \mid \ell(v) = i\}$  and  $G_i = G_{V_i}$ . If  $G_i$  is non-empty :
  - $G_i$  is connected
  - There is a unique node  $u$  of  $G_i$  with  $\delta(u) = 0$
  - $\forall v \in V_i, \delta(v)$  is the distance from  $u$  to  $v$  in  $G_i$ .
- Each  $G_i$  is a cluster.

## Virtual Graph $H = (V_H, E_H)$ :

- $V_H = \{\text{clusters } C \text{ induced by labels}\}$
- $CC' \in E_H \Leftrightarrow \exists u \in C \text{ and } v \in C' \text{ such that } uv \in E$ .

## Colored BFS-clustering :

- Two functions  $(\gamma, \delta)$  assigning a pair  $(\gamma(v), \delta(v)) \in \mathbb{N} \times \mathbb{N}$  to each node  $v \in V$
- $\forall i > 0, V_i = \{v \in V \mid \gamma(v) = i\}$  and  $G_i = G_{V_i}$ . If  $G_i$  is non-empty :
  - For any connected component  $C_i$  of  $G_i$
  - There is a unique node  $u$  of  $C_i$  with  $\delta(u) = 0$
  - $\forall v \in C_i, \delta(v)$  is the distance from  $u$  to  $v$  in  $C_i$ .
- Each  $C_i$  is a cluster.

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## From Clusters to $(\Delta + 1)$ -coloring

### $(\Delta + 1)$ -coloring

- $(\gamma, \delta)$  : a colored BFS-clustering of  $G$
- Assume each node  $v$  of  $G$  knows  $\gamma(v)$  and  $\delta(v)$
- $c = \max_{v \in V} \gamma(v)$

$(\Delta + 1)$ -coloring can be solved by a distributed algorithm with awake complexity  $O(\log c)$ , and round complexity  $O(c \cdot n)$ .

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$(\Delta + 1)$ -coloring can be solved by a distributed algorithm with awake complexity  $O(\log c)$ , and round complexity  $O(c \cdot n)$ .

Idea :

- For each  $i \leq c$  :
  - If cluster has color in the Binary subtree of  $i$  :
    - Wake up, gather colors from awaken neighboring clusters
    - If cluster color is  $i$ , compute and broadcast your output colors
    - Gather output colors from neighbor clusters of color  $i$

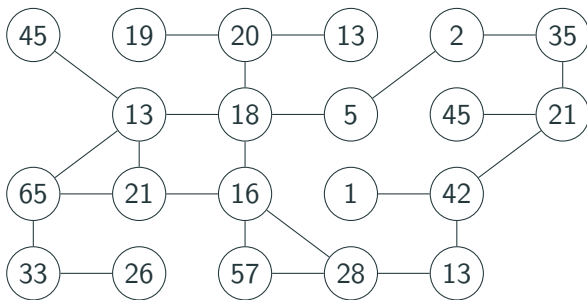
# Clustering a Graph

## One-round Cluster Reduction

$\exists a > 0 : \forall b > 0, \exists$  algorithm  $\mathcal{A}$ , with awake complexity  $O(\log^* n)$  and round complexity  $O(n^4)$  which computes a colored BFS-clustering  $(\gamma, \delta)$  such that :

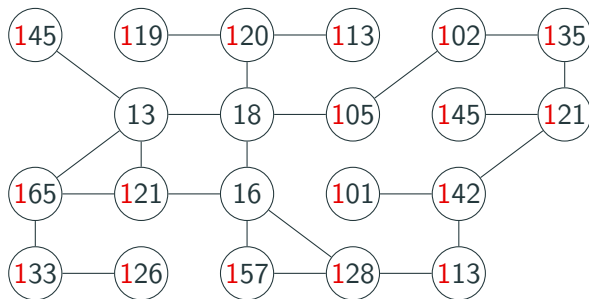
- $(\gamma, \delta)$  restricted to  $\{v \in V \mid \gamma(v) \in \{1, \dots, a \cdot b^2\}\}$  is a colored BFS-clustering
- $\forall v \in V$  with  $\gamma(v) \in \{1, \dots, a \cdot b^2\}$ ,  $\delta(v) = 0$  (i.e.,  $v$  is alone in its cluster) ;
- $(\gamma, \delta)$  restricted to  $\{v \in V \mid \gamma(v) > a \cdot b^2\}$  is a uniquely-labeled BFS-clustering with at most  $n/b$  clusters.

# Clustering a Graph



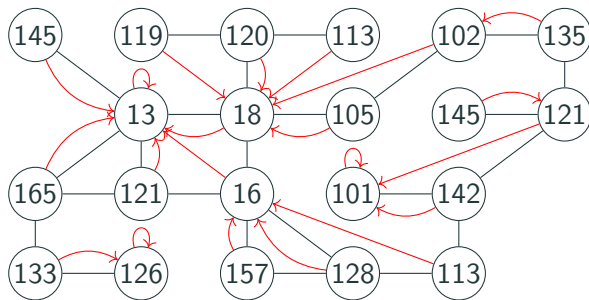
1. Compute a distance-2  $k$ -coloring in  $O(\log^* n)$  rounds ( $k \in O(n^4)$ )

## Clustering a Graph



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2. Add  $k$  to the colors of nodes of degree  $\leq b$

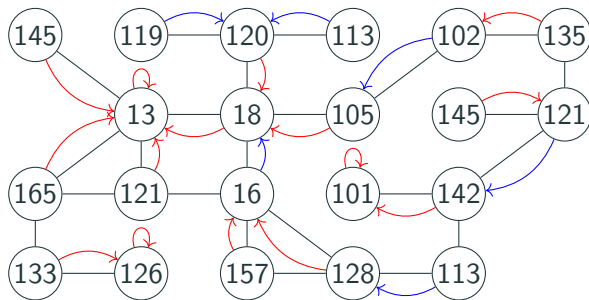
## Clustering a Graph



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2. Add  $k$  to the colors of nodes of degree  $\leq b$
3. Choose as parent :
  - yourself if your color is the smallest at distance 2
  - your smaller neighbor if one of them has smaller color than yours
  - your distance-2 smaller neighbor otherwise

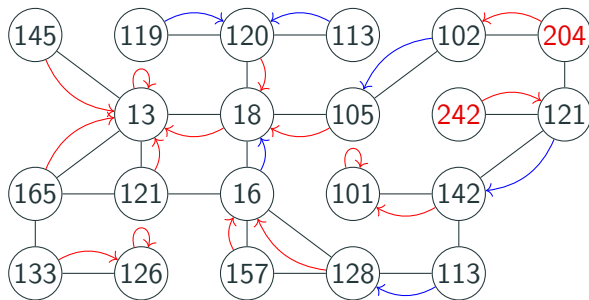


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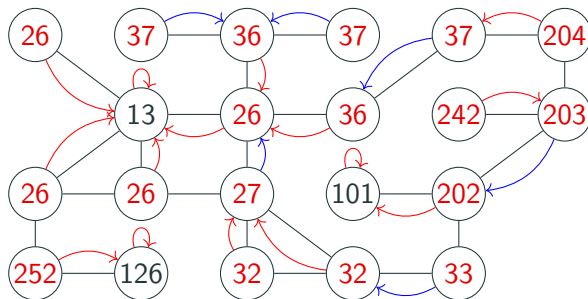
3. Choose as parent :
  - yourself if your color is the smallest at distance 2
  - your smaller neighbor if one of them has smaller color than yours
  - your distance-2 smaller neighbor otherwise
4. In the third case, choose a neighbor with same parent (blue case)

# Clustering a Graph



5. Compute a new color :
  - 2 times the color of your parent if you are not your parent
  - Add 1 if you are in the blue case
  - 0 if you are your own parent

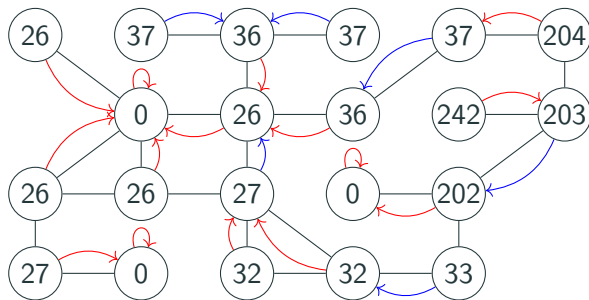
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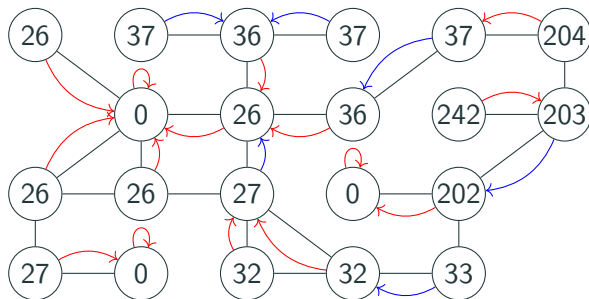
## Clustering a Graph



### 5. Compute a new color :

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## Clustering a Graph

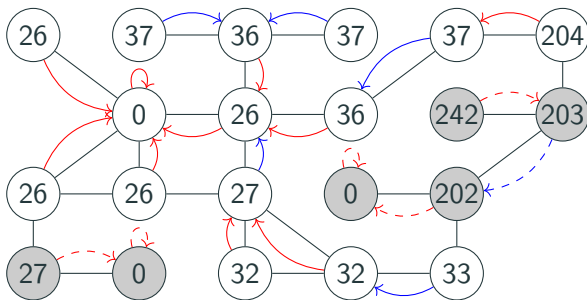


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- Add 1 if you are in the blue case
- 0 if you are your own parent

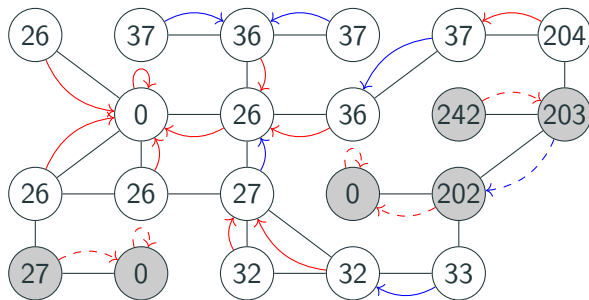
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## Clustering a Graph



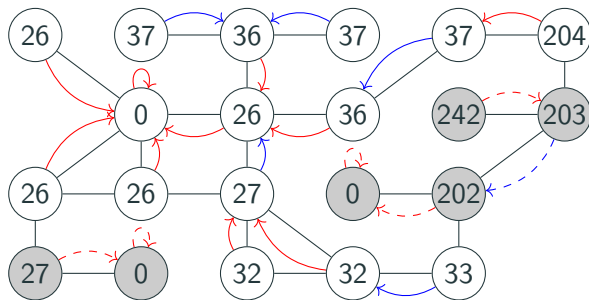
5. You have built DLTs
6. Perform a convergecast and broadcast  
Nodes compute their distance to their root
7. Keep only DLTs with roots of degree  $> b$

## Clustering a Graph



7. Keep only DLTs with roots of degree  $> b$
8.  $U$  is the set of nodes not in a DLT :
  - They compute a  $O(b^2)$ -coloring of  $G_U$  in  $O(\log^* n)$  rounds
  - They have colors from 1 to  $a \cdot b^2$  ( $a$  is the constant of Linial's algorithm)
  - Nodes of  $U$  form clusters of size 1 with this new color

# Clustering a Graph



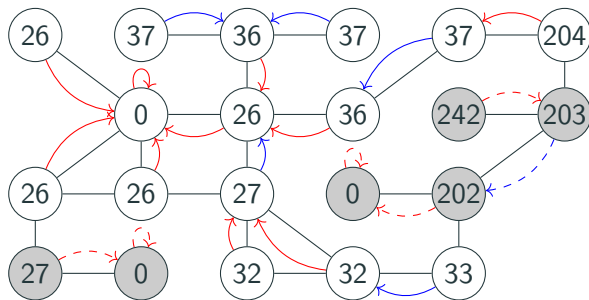
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9. Other nodes set  $\gamma(u) = \ell(v) + a \cdot b^2$  and  $\delta(u) = \text{dist}(u, v)$ ,  $v$  being the root of  $u$



## Clustering a Graph



**Observations :** Subgraph induced by  $\{v \in V \mid \gamma(v) > a \cdot b^2\}$  has at most  $n/b$  clusters

- Each root has more than  $b$  neighbors, all in its cluster (roots are distance-2 minimas)
- For each root, we can charge  $b$  nodes to itself
- Each node is charged at most once
- $\Rightarrow$  We have at most  $n/b$  roots of degree  $> b$

# Cluster Graph Simulation

## LOCAL Simulation on Clusters

- $H$  : the virtual graph induced by some uniquely-labeled BFS-clustering  $(\ell, \delta)$  of  $G$ .
- $\mathcal{A}$  : a distributed algorithm running on  $H$  with  $\alpha$  awake rounds and round complexity  $\varrho$ .
- $\forall u \in V_H$ ,  $\text{input}(u)$  is the input of  $u$  in  $H$ , and  $\text{output}(u)$  the output of  $\mathcal{A}$  at vertex  $u$ .
- Assume each node  $v$  of  $G$  knows  $\text{input}(\ell(v))$

It is possible to run  $\mathcal{A}$  on  $G$  such that every node  $v$  of  $G$  computes  $\text{output}(\ell(v))$  with awake complexity at most  $7 \cdot \alpha$  and round complexity  $O(\varrho \cdot n)$ .

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Idea :

- The root of each cluster (i.e. with  $\delta = 0$ ) simulates  $\mathcal{A}$
- Each time a node awakes in  $H$ , the cluster performs some broadcasts and convergecasts
- In each round, 7 awaken activations are needed, and  $\forall u \in V, \delta(u) \leq n$

## Building Incrementally the Colored Clustering

Input :  $G = (V, E)$

1.  $H_0 : \ell_0(v) = \text{ID}(v)$  and  $\delta_0(v) = 0$
2. While  $H_{i-1} \neq \emptyset$   
    Use clustering algorithm on  $H_{i-1}$   
    to compute  $H_i$

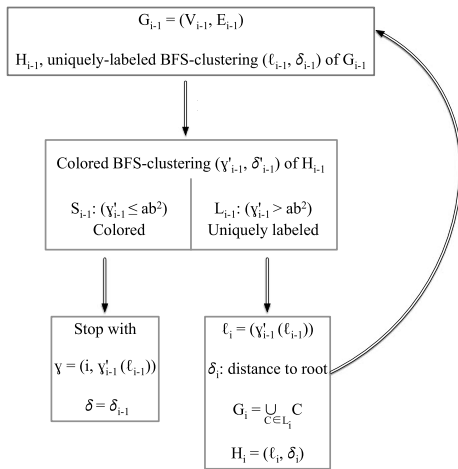
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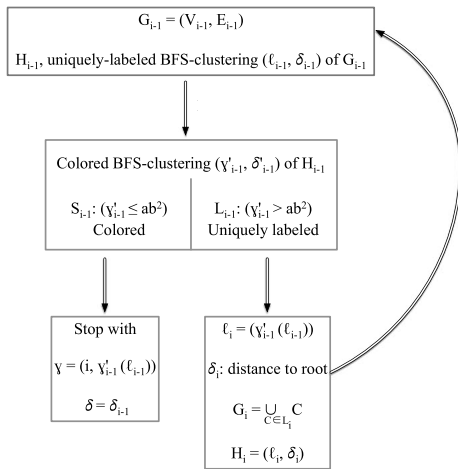
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After each iteration, we go from  $N$  to  
(up to)  $N/b$  vertices

$\Rightarrow k$  such that  $b^k > n$  is enough

$\Rightarrow$  Clusters use  $kab^2$  colors



## Putting Things Together

We have :

- Two parameters  $b$  and  $k$  with  $b^k \geq n$
- We compute a  $(\gamma, \delta)$  colored BFS-clustering in  $O(k \log^* n)$  rounds
- $\gamma$  has maximal value  $kab^2$  ( $a$  is a constant)
- From the BFS-clustering, we get a  $(\Delta + 1)$ -coloring in  $O(\log(kb^2))$

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- **Conclusion** :  $k = 2\sqrt{\log n}$  and  $b = 2\sqrt{\log n}$  work

We get a  $(\Delta + 1)$ -coloring algorithm with  $O(\sqrt{\log n} \log^* n)$  awake complexity

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