Hierarchical Routing

Réseaux & Communication

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November 4, 2026



Roadmap

1. Introduction
2. Examples of clustering
   - Clustering of [3]
   - Clustering of [1]
3. Routing in a Clustering
4. References

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Idea

To reduce the **cost parameters** of the routing:

use of a **hierarchical division** of the network

Justification: Most of the communication is local, *i.e.*, between nodes at "relatively" small distances from each other

Cost Parameters

1. Length of addresses

   n nodes $\Rightarrow$ at least $\log(n)$ bits per address

   Maybe more, if information is encoded in addresses

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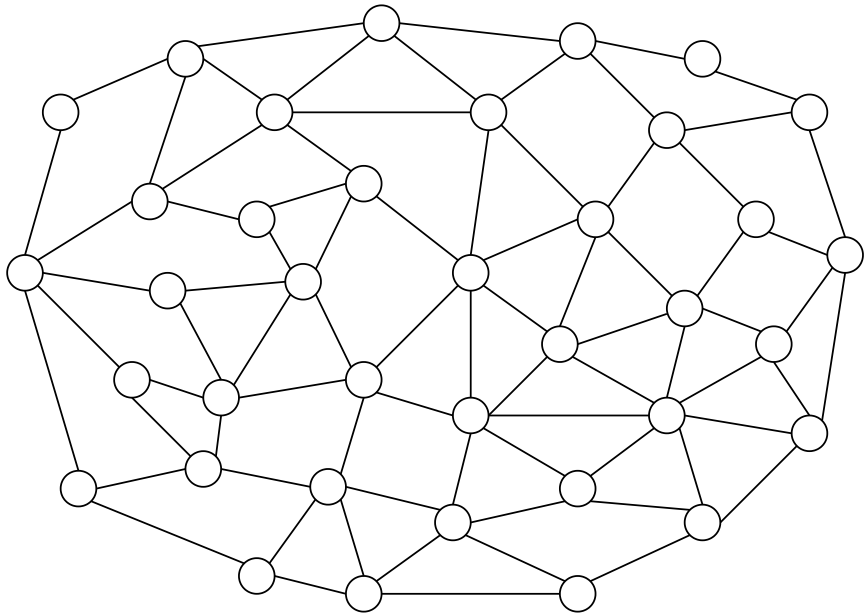
3. Cost of table lookups

   The cost (in time) of a single table lookups is likely as larger for a large routing table or for larger addresses

   The total table-lookup time for the delivery of a single message also depends on the number of times the tables must be accessed (number of hops)

Clustering

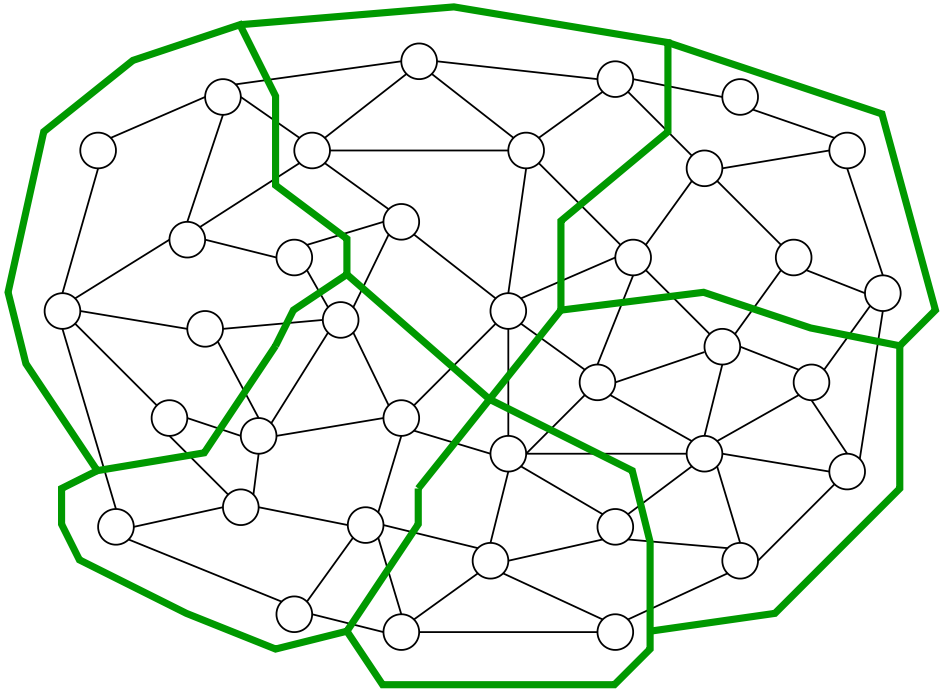
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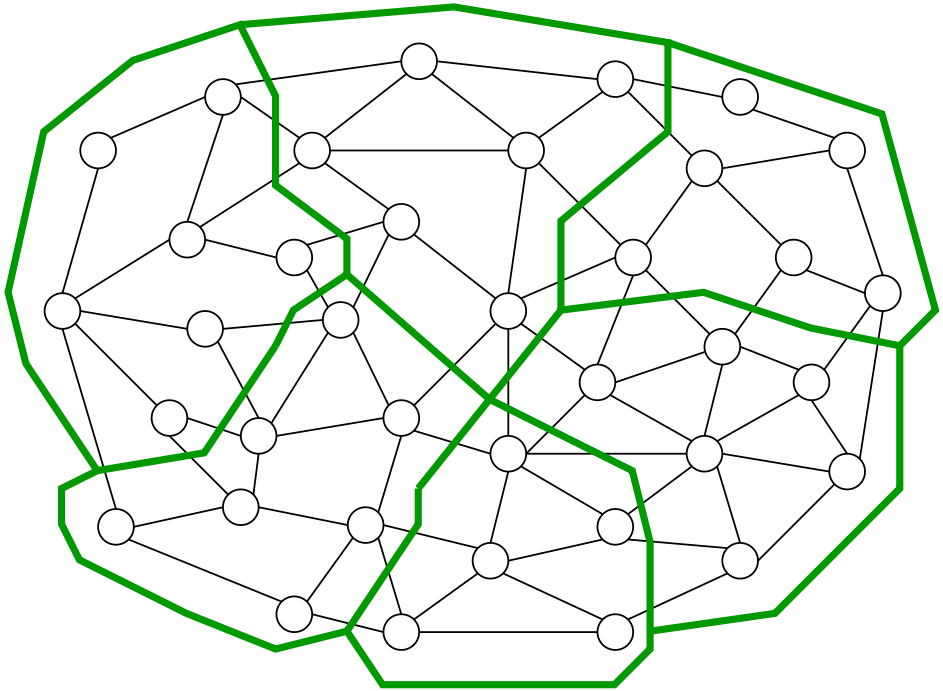
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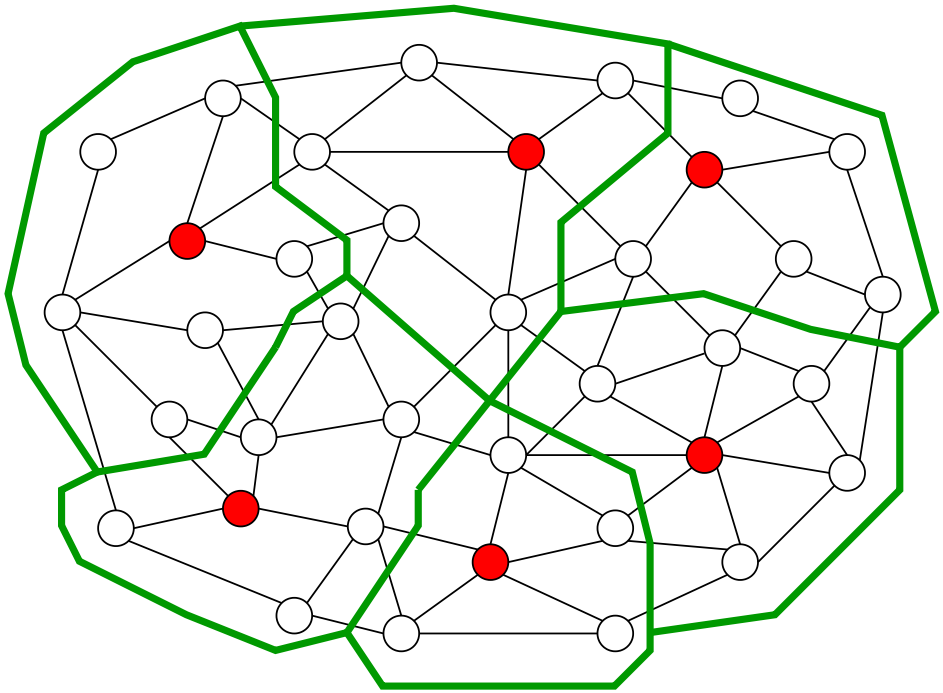
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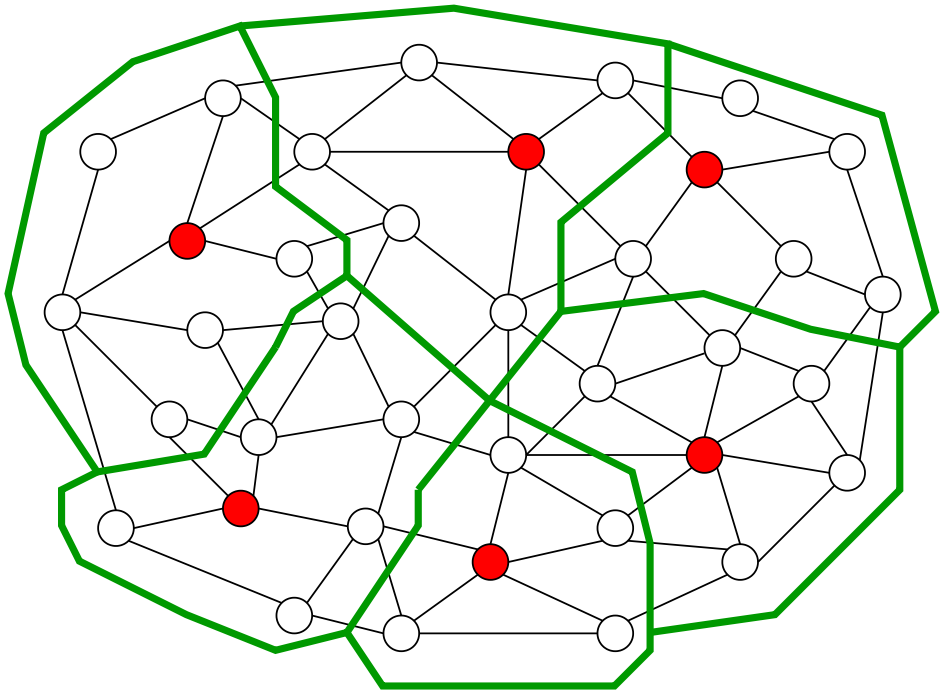
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- In each cluster, there is a designated node: the **clusterhead**



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Remark: Clustering maybe recursive, *i.e.*, each cluster may be partitioned into subclusters, and so on so forth, in order to obtain a multi-level division of the nodes.

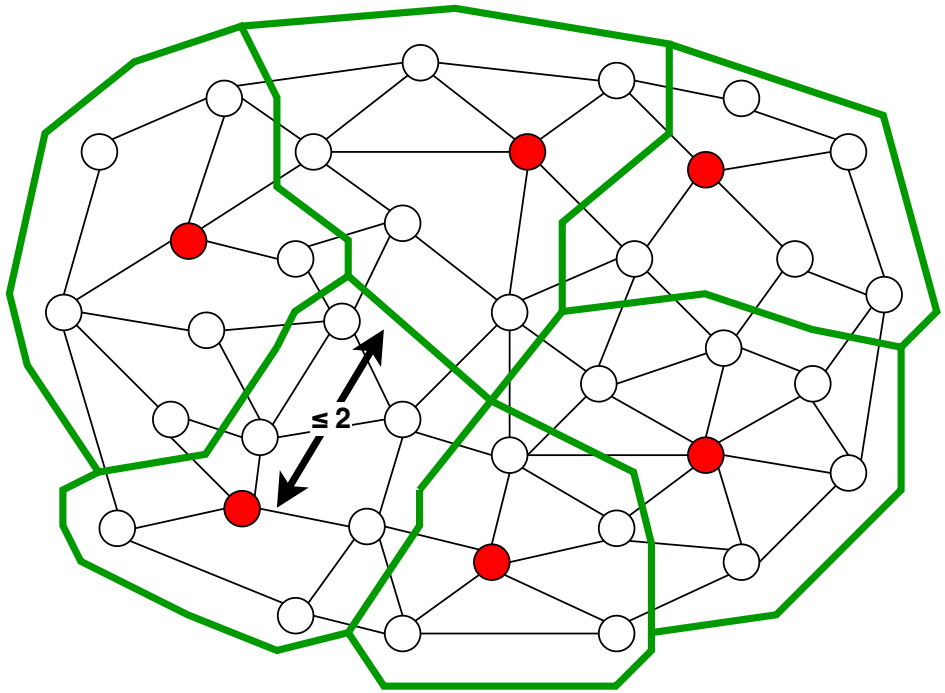
Here, we only consider 1-level division.

k-Clustering

A k-**clustering** is a clustering of the network where each cluster is of radius at most k

($k = 1 \Rightarrow$ dominating set)

In a cluster, each node is at distance at most k from its clusterhead.



2-clustering

A good k-Clustering?

Trivial solution: all nodes are clusterheads!

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Minimal k-clustering (*e.g.*, [2]).

Let C be the set of clusterheads of a minimal k-clustering.

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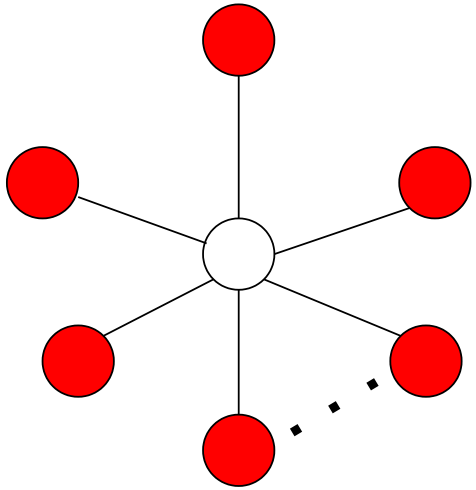
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minimal 1-clustering with $n-1$ clusters/clusterheads

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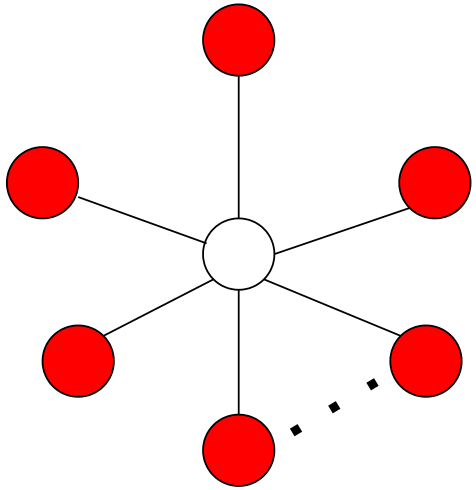
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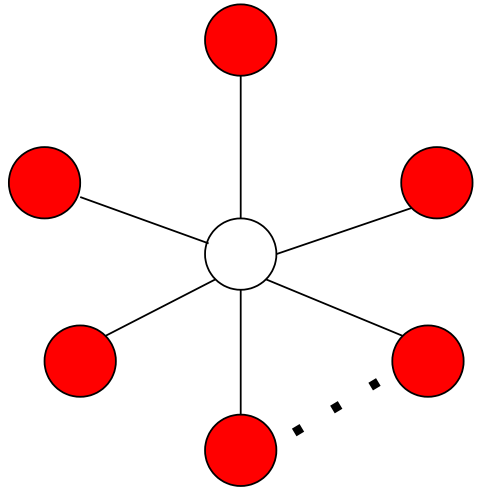
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Remark: a $(O(\frac{n}{k}))$ k-clustering can be made minimal to even more reduce the number of clusters [2].



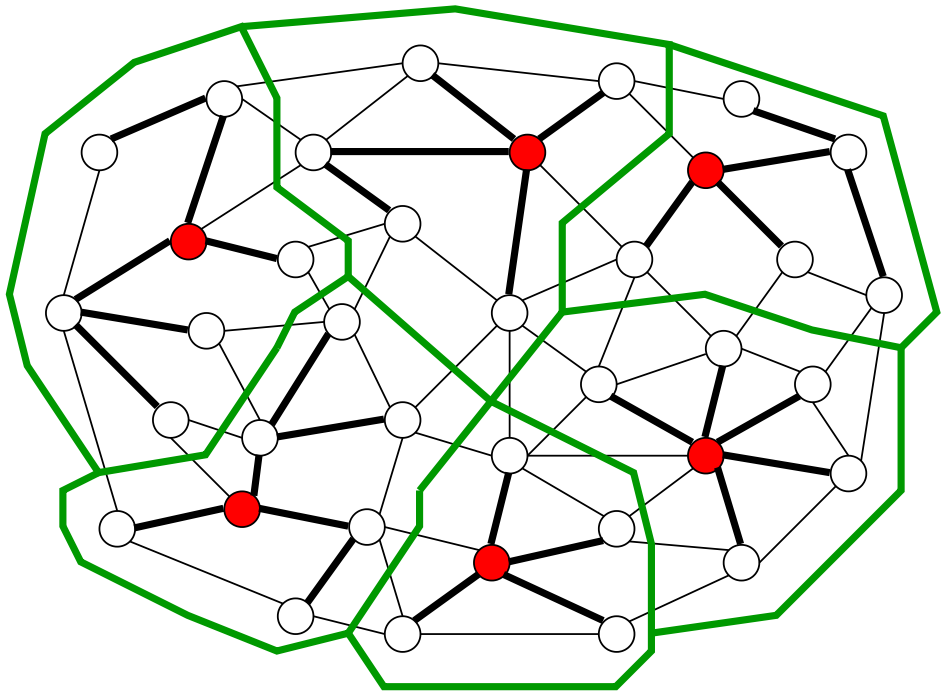
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Clustering

(Usual) Structure

Cluster = Tree rooted at its clusterhead

Spanning Forest



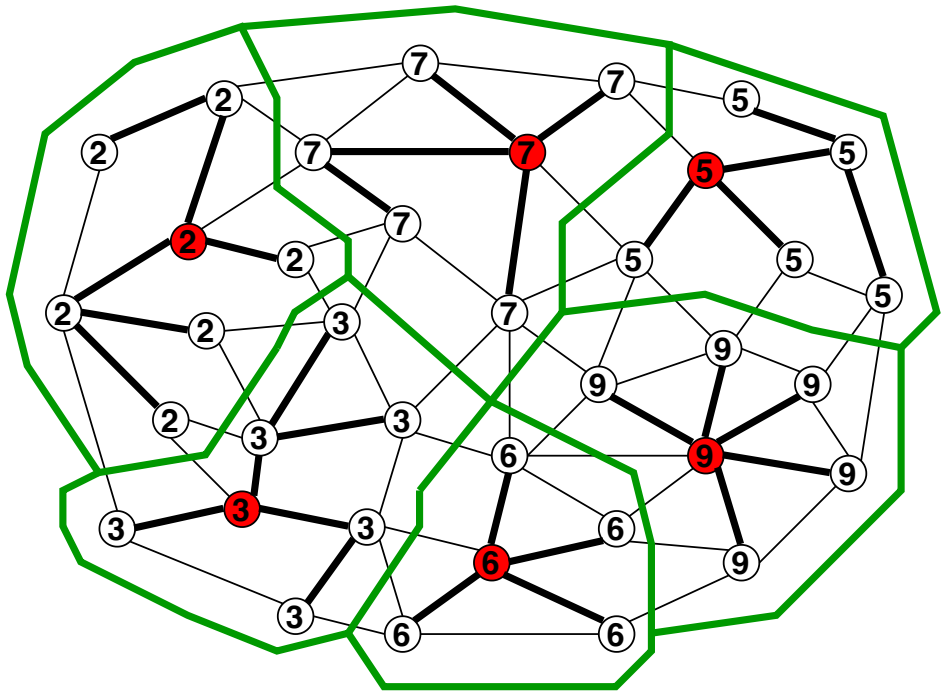
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(Usual) Structure

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Colored Trees, *e.g.*, with the clusterhead identifiers



Clustering-based Routing

A major application of k-clustering is in the implementation of an efficient routing scheme in a network.

To route a packet from p to q:

1. Route the packet from p to its clusterhead (intra-clustering)
2. Route the packet from the clusterhead of p to the clusterhead of q (inter-clustering)
3. Route the packet from the clusterhead of q to q (intra-clustering)

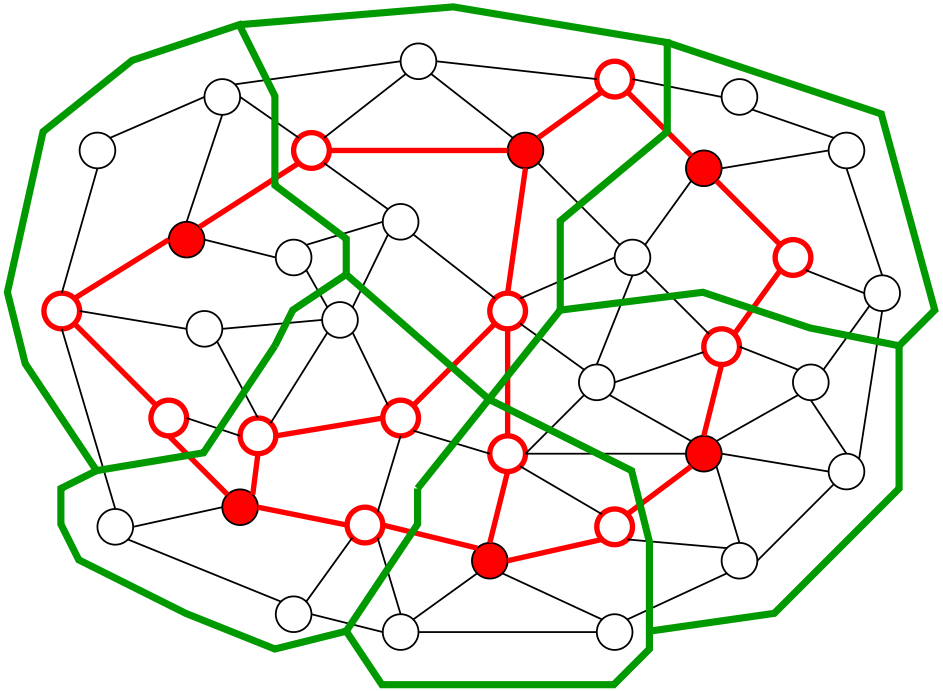
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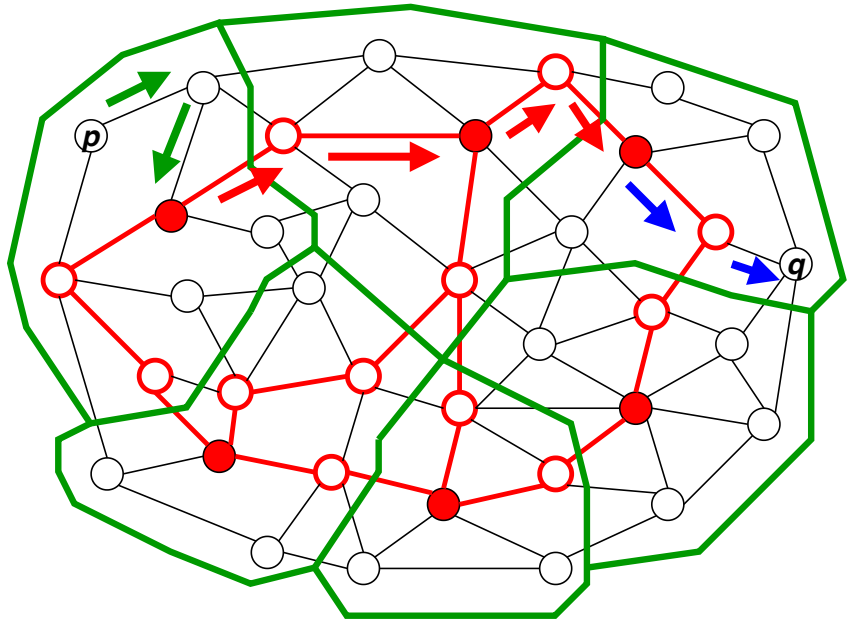
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- Depending of its color, a packet is either forwarded over a fixed channel or a more complex routing scheme
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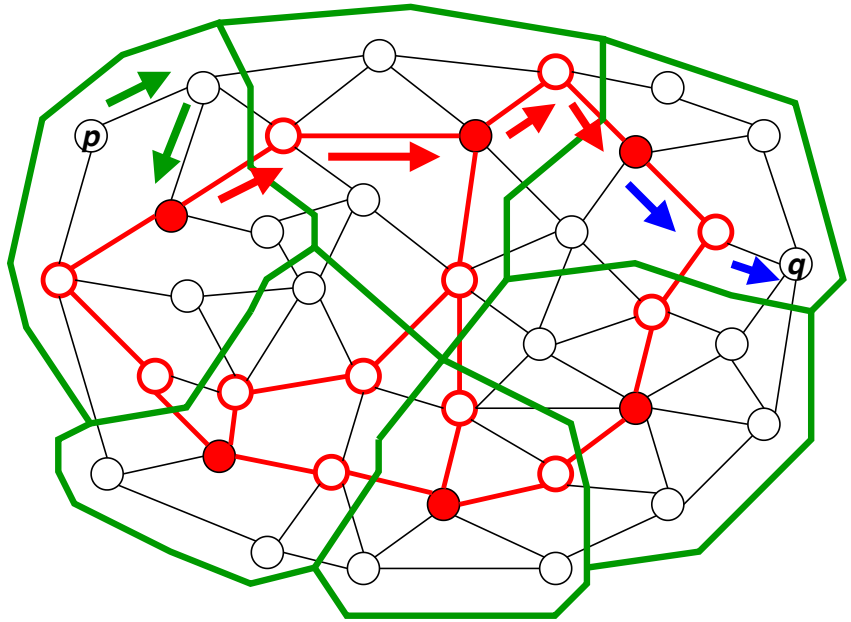
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Address: (Label of the clusterhead of the destination,destination label)



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Two examples

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Both computes $O(\frac{n}{k})$ clusters

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The second one also provides an approximation of the optimal clustering in case the network is a Unit (or Quasi Unit) Disk Graph

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Idea

Let $G = (V, E)$ be a connected graph of n nodes

A **k-dominating set** of G is a subset D of V such that $\forall p \in V$, $\exists q \in D$ such that $\|p, q\| \leq k$

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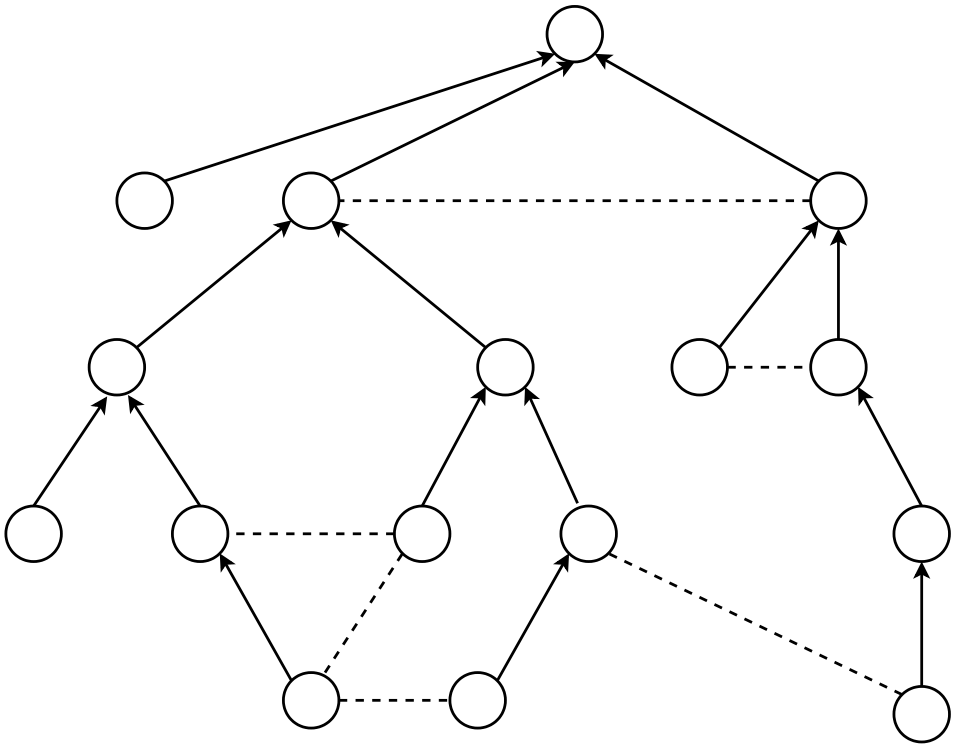
Theorem

For every $k \in \mathbb{N}$, there exists a k-dominating set D of G such that $|D| \leq \lceil \frac{n}{k+1} \rceil$

Proof of the Theorem

Some notations

Let $T = (V, E_T)$ be an arbitrary spanning tree of G rooted at node r

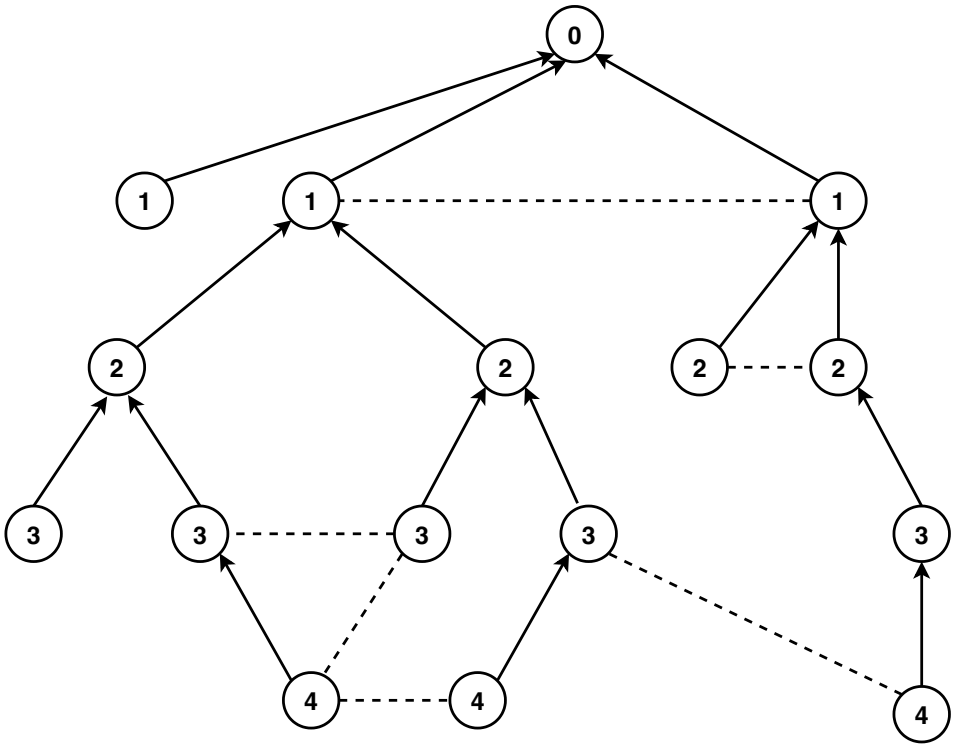


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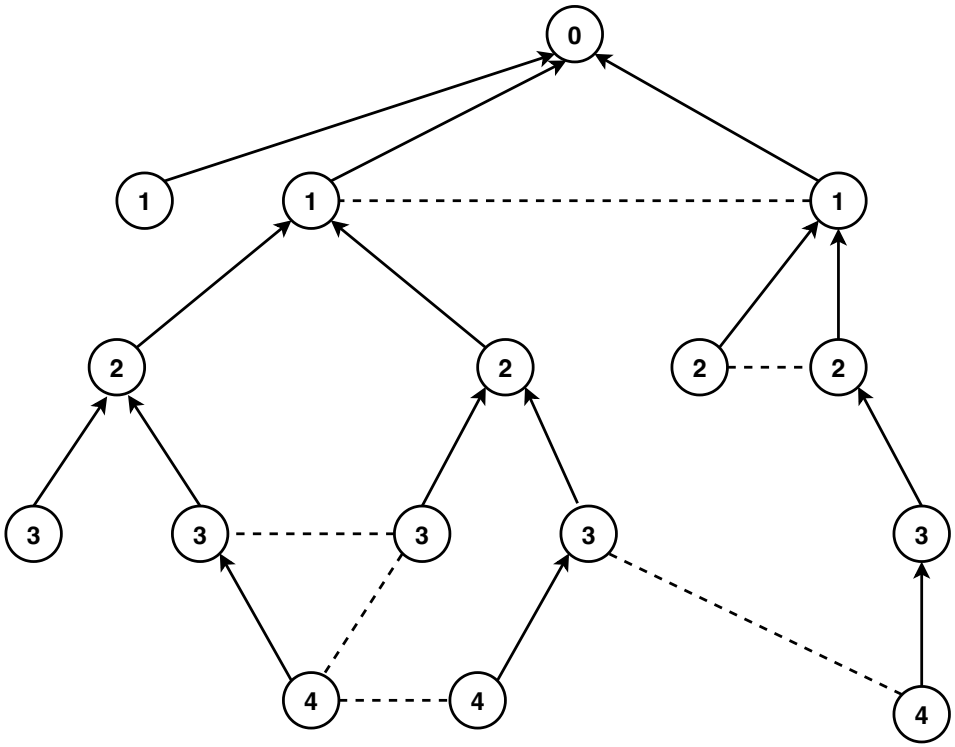
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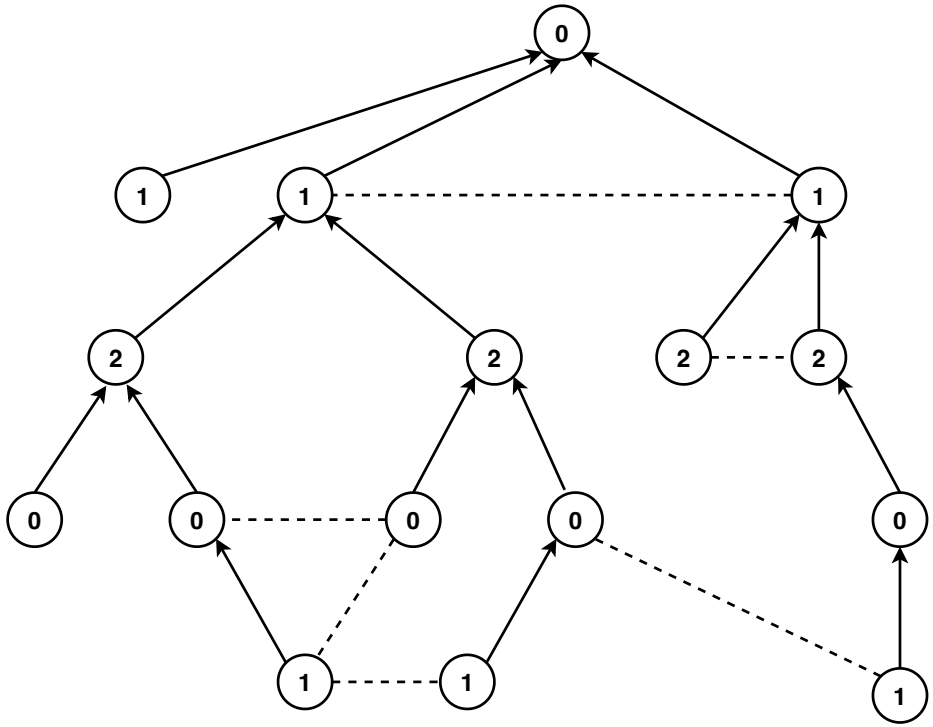
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For every $i \in [0..k]$, let

$$D_i = \{p \in V \mid L(p) \bmod (k+1) = i\}$$

(in the example, $k = 2$)



Proof of the Theorem

Easy cases: $n = 0$ or $k = 0$

1. If $n = 0$, then $\lceil \frac{n}{k+1} \rceil = 0 = |\emptyset|$ and $\emptyset$ is a k-dominating set of G.

2. If $k = 0$, then $D_0 = V$, and so D_0 is a k-dominating set of G. Moreover, $|D_0| = n = \lceil \frac{n}{k+1} \rceil$.

Proof of the Theorem

Case $n > 0$ and $k > 0$

3. Assume $k \geq H$. Then, D_0 only contains r and every other node is within distance k from r. So, D_0 is a k-dominating set of G whose size is $1 \leq \lceil \frac{n}{k+1} \rceil$.

Proof of the Theorem

Case $n > 0$ and $k > 0$

5. Assume $k \geq H$. Then, D_0 only contains r and every other node is within distance k from r. So, D_0 is a k-dominating set of G whose size is $1 \leq \lceil \frac{n}{k+1} \rceil$.

6. Assume $k < H$. Then, $\forall i \in [0..k], |D_i| > 0$.
   1. Assume that $\forall i \in [0..k-1], |D_i| = |D_{i+1}|$.

      Then, $\forall i \in [0..k]$, $|D_i| = \lceil \frac{n}{k+1} \rceil$.

      Let $v \notin D_0$. The level of v, $L(v)$, satisfies $L(v) = x(k+1) + y$, where $x \geq 0$ and $0 < y \leq k$.

      Let u be the ancestor of v such that $L(u) = L(v) - y$ (u exists because $y \leq L(v)$).

      By definition, $u \in D_0$ and $\|u, v\| \leq k$. Hence, D_0 is a k-dominating set of G such that $|D_0| = \lceil \frac{n}{k+1} \rceil$.

Proof of the Theorem

Case $n > 0$ and $k > 0$

6. 2. Assume that $\exists i \in [0..k-1], |D_i| \neq |D_{i+1}|$.

   Let $\min \in [0..k]$ such that $\forall i \in [0..k]$, $|D_{\min}| \leq |D_i|$.

   Then, $|D_{\min}| < \lceil \frac{n}{k+1} \rceil$.

   Let $D = D_{\min} \cup \{r\}$. Then, $|D| \leq \lceil \frac{n}{k+1} \rceil$.

   Let $v \notin D$.

   1. If $L(v) \leq k$, then v is at distance at most k from r and $r \in D$.
   2. If $L(v) > k$, then $L(v) = x(k+1) + y$ with $x > 0$, $0 \leq y \leq k$, and $y \neq \min$ (if $y = \min$, then $v \in D_{\min} \subseteq D$, a contradiction).

      If $y > \min$, then let u be the ancestor of v such that $L(u) = x(k+1) + \min$. Now, $0 \leq L(v) - L(u) = y - \min \leq k$.

      If $y < \min$, let u be the ancestor of v such that $L(u) = (x-1)(k+1) + \min$. Now, $0 \leq L(v) - L(u) = k + 1 + y - min \leq k$.

      By definition, $u \in D$ (more precisely, $u \in D_{\min}$) and $\|u, v\| \leq k$.

   Hence, D is a k-dominating set of G and $|D| \leq \lceil \frac{n}{k+1} \rceil$.

□

Solution

Let $mcd = \min\{|D_i| \mid i \in [0..k] \wedge D_i \neq \emptyset\}$: *mcd* is the minimum cardinal of a non-empty D-set.

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$D_x \cup \{r\}$ is a k-dominating set of G if size at most $\lceil \frac{n}{k+1} \rceil$.

Proof. $D_x \cup \{r\}$ corresponds to each set exhibited in Cases 2-6 of the theorem proof (*n.b.*, Case 1 is for the beauty of the art, but useless)

□

Distributed Computation

1. Leader Election

   ($O(mn)$ messages, $O(m)$ rounds, and $O(\delta_u + B)$ bits for messages and nodes, where B the number of bits to store an identifier)

2. BFS Spanning Tree Construction (with initialization and termination detection at the leader)

   ($O(D)$ rounds, $O(n.m)$ messages of $O(\log D)$ bits, and $O(\log D + \log \Delta)$ bits per node)

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3. Propagation of Information with Feedback in the tree:
   Broadcast $\Rightarrow$ Node $(k+1)$-Coloring,
   Feedback $\Rightarrow$ Computation of the D-set sizes

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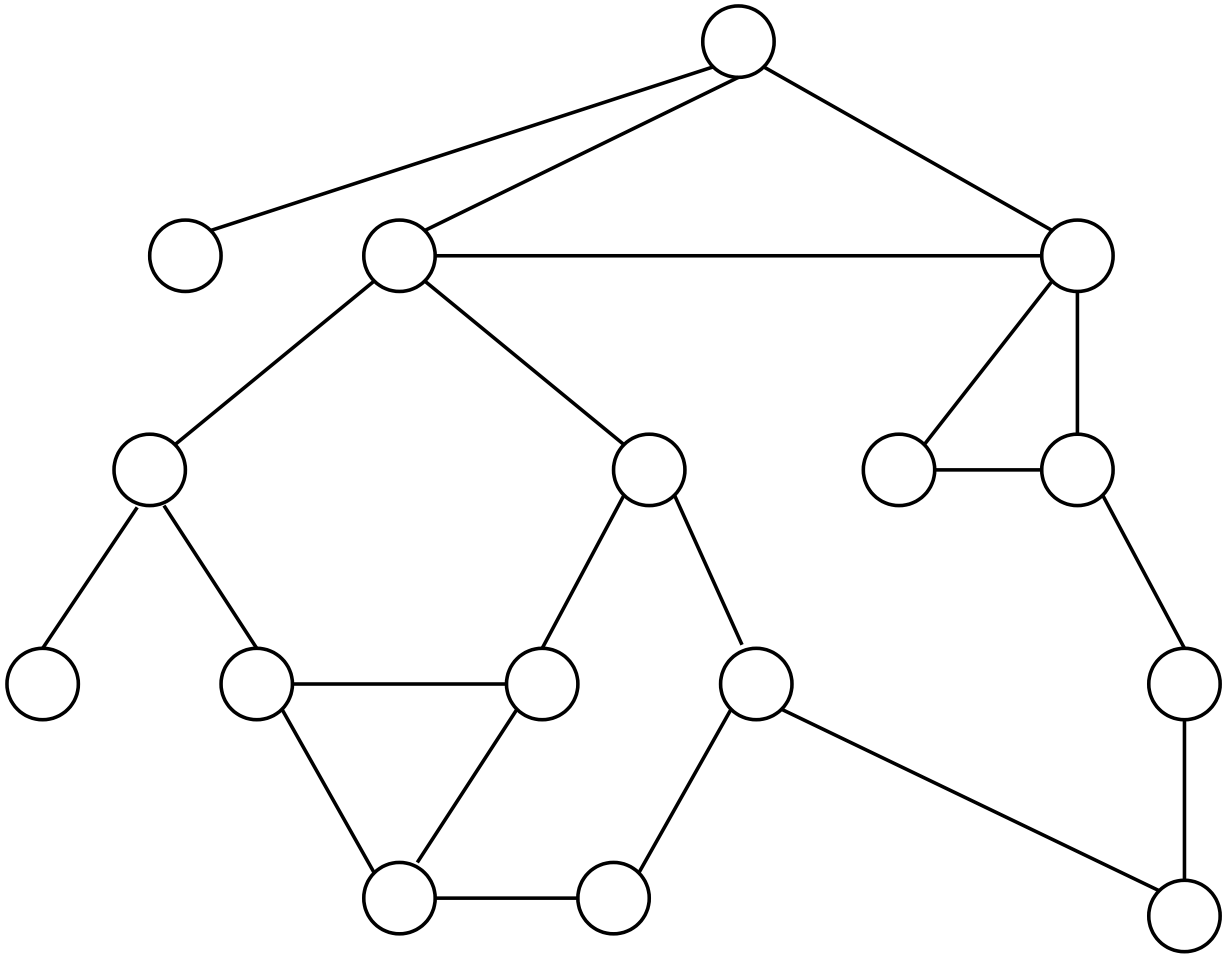
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4. Propagation of Information with Feedback in the tree:
   Broadcast $\Rightarrow$ clusterhead assignment and cluster coloring,
   Feedback $\Rightarrow$ Termination Detection at the Leader

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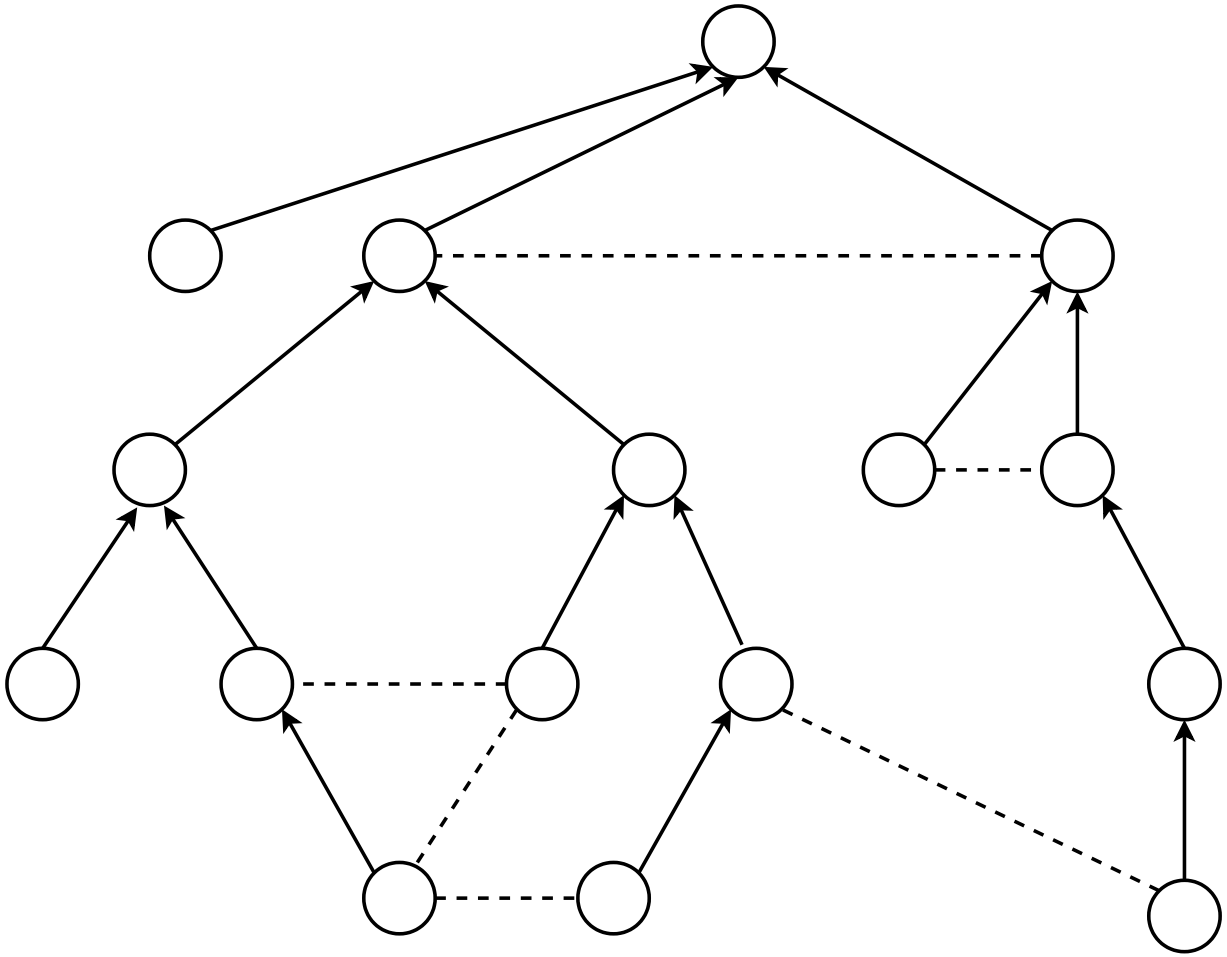
5. Routing Set-Up

Example with $k = 2$



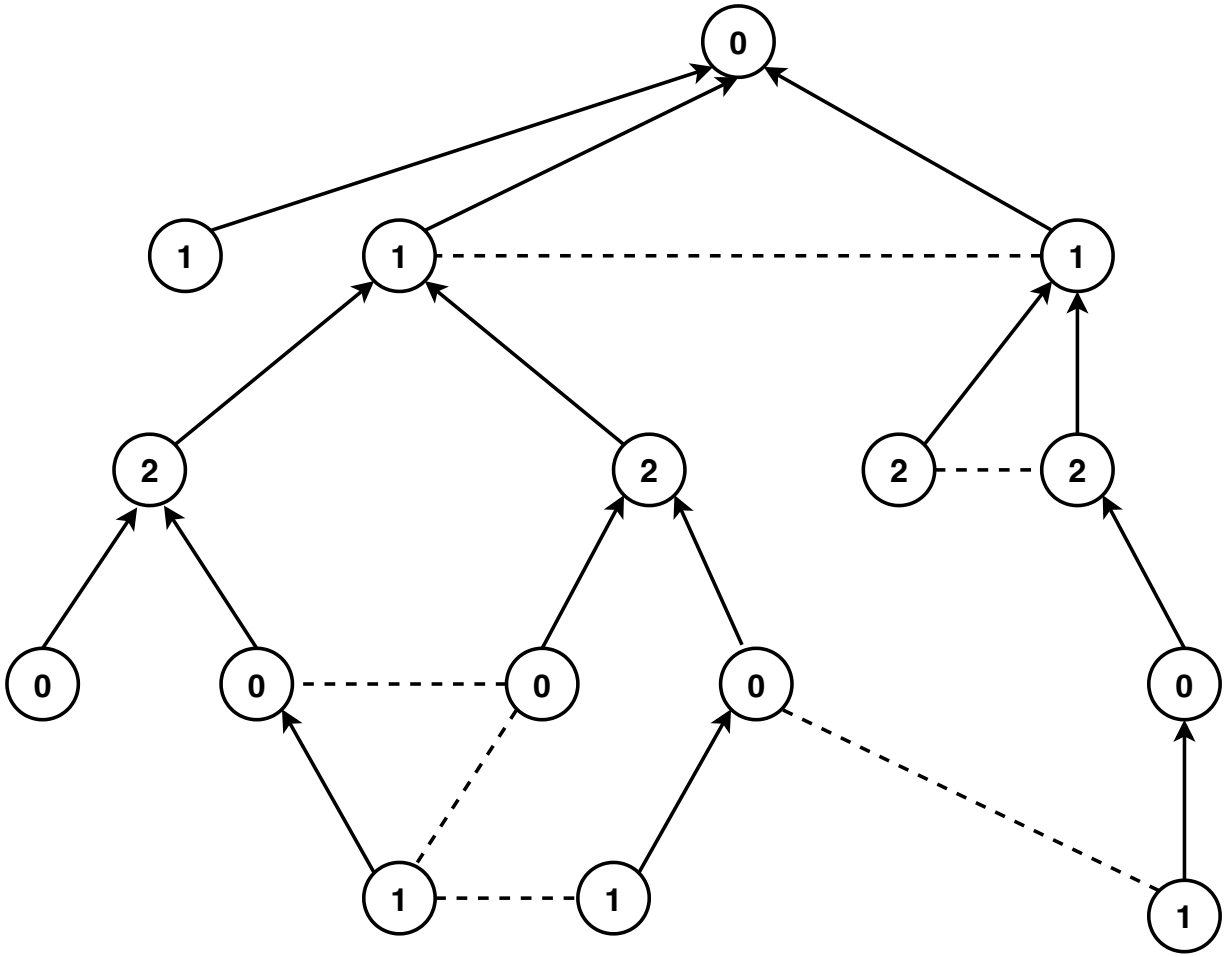
Network

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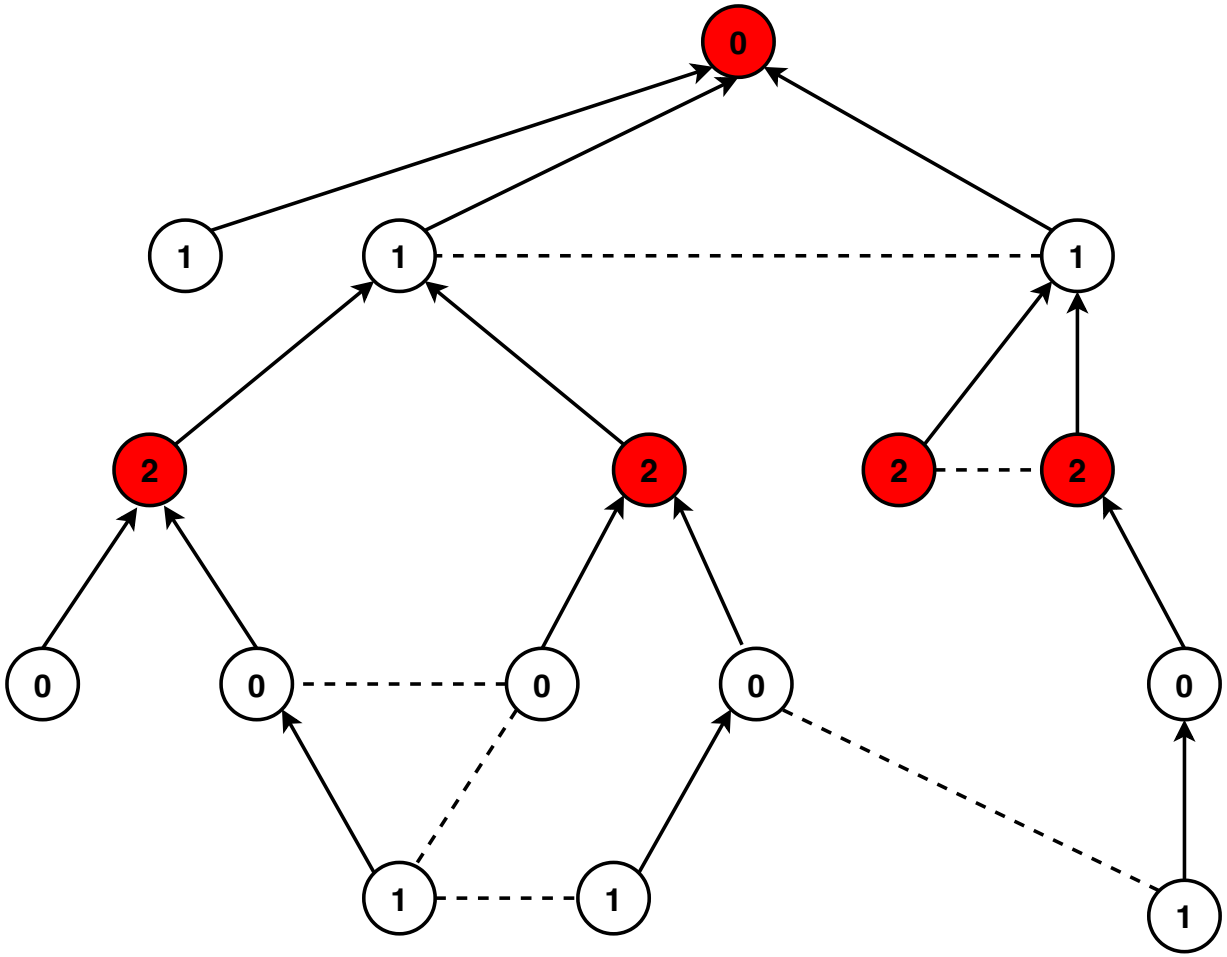
(BFS) spanning tree

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Node $(k+1)$-Coloring

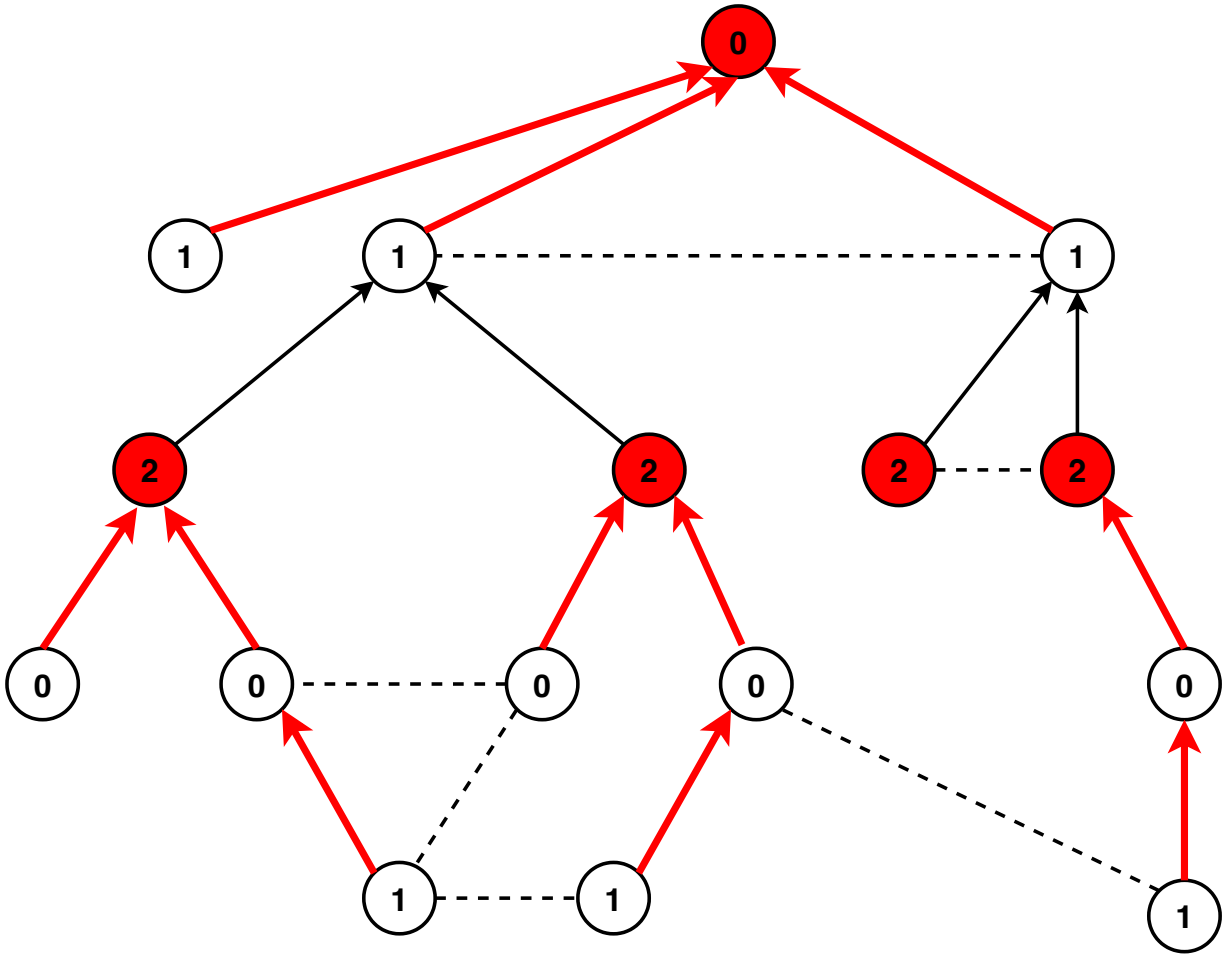
Example with $k = 2$



k-dominating set

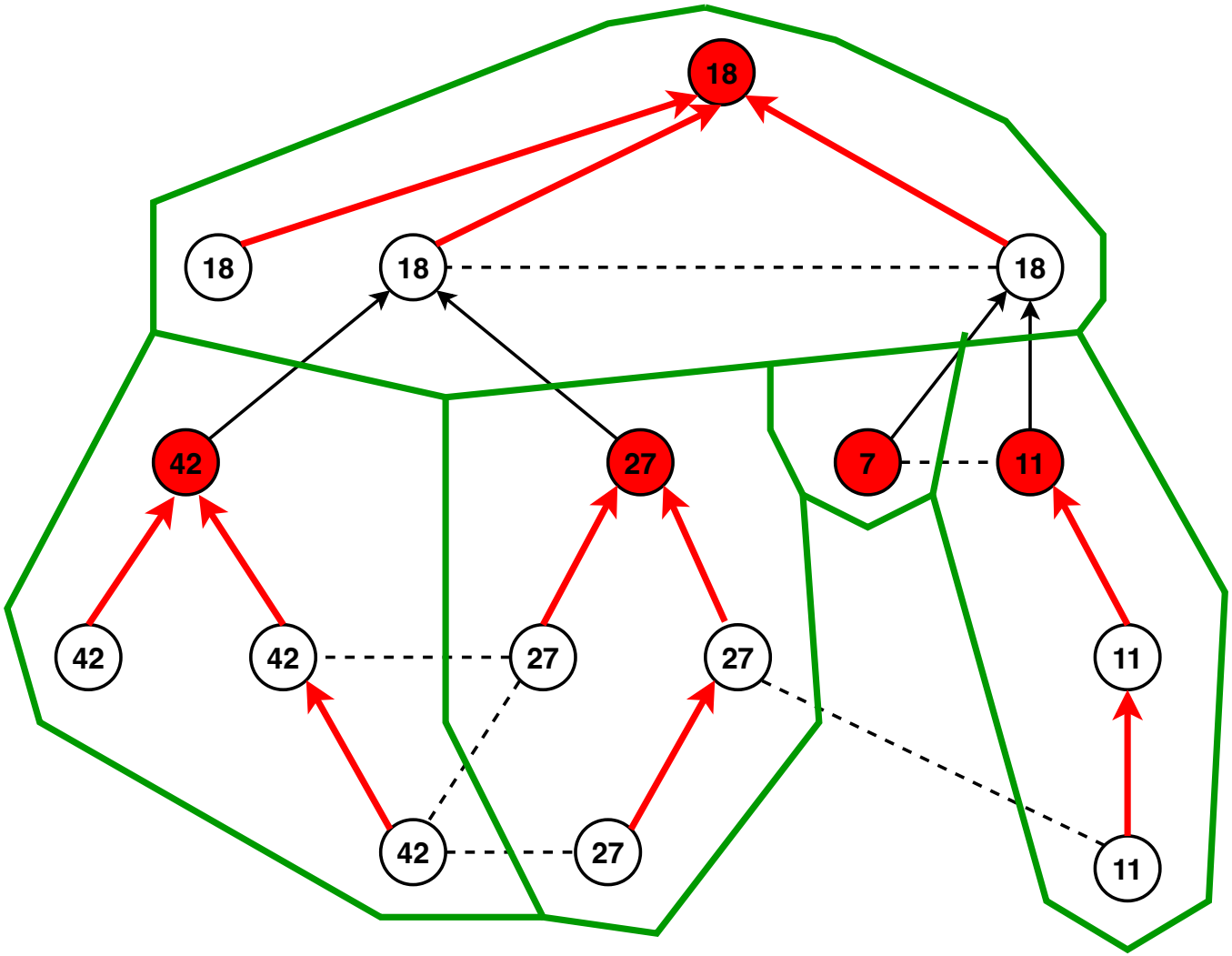
$$|D_0| = 6,\ |D_1| = 6,\ |D_2| = 4$$

Example with $k = 2$



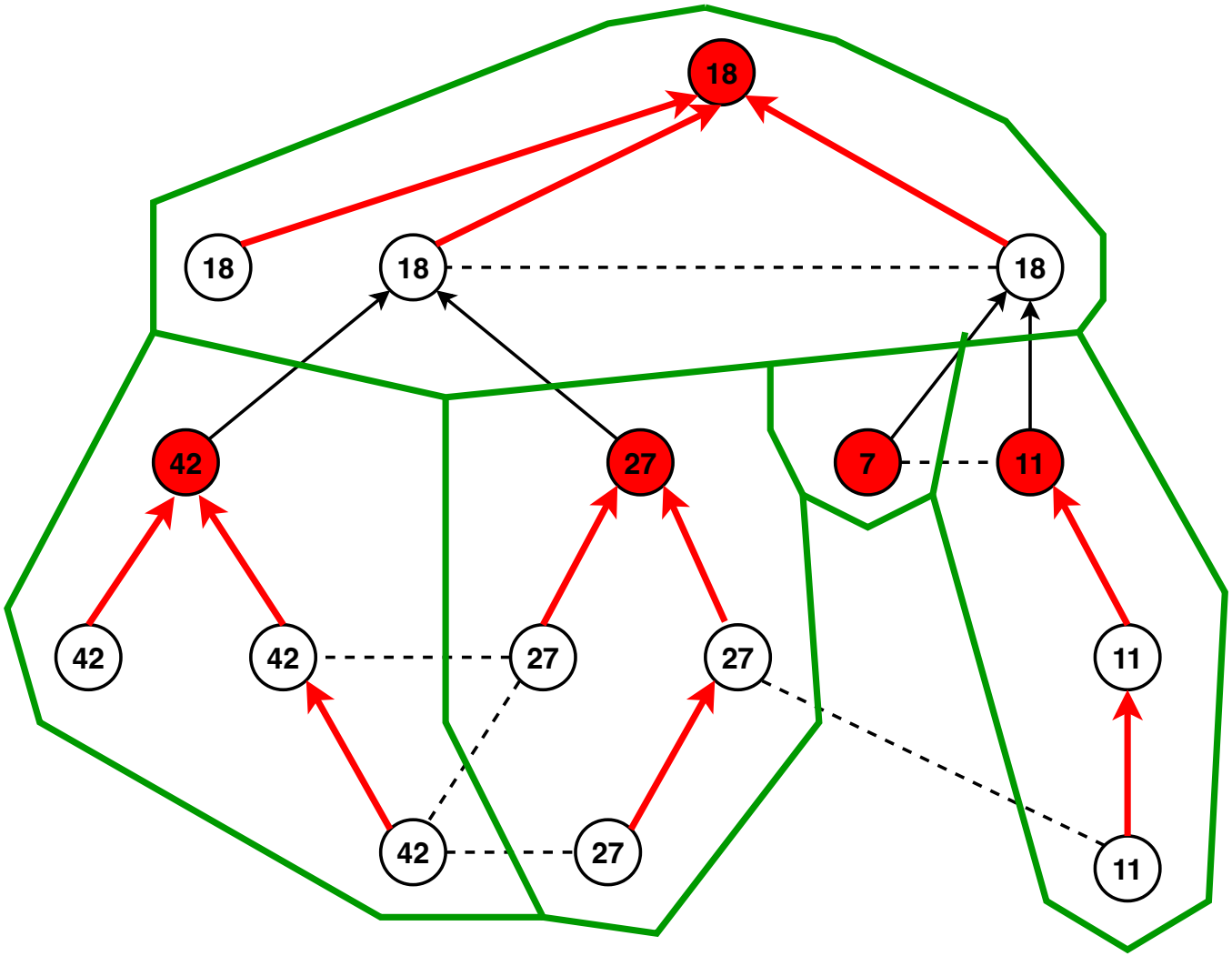
Clusters

Example with $k = 2$



Clustering

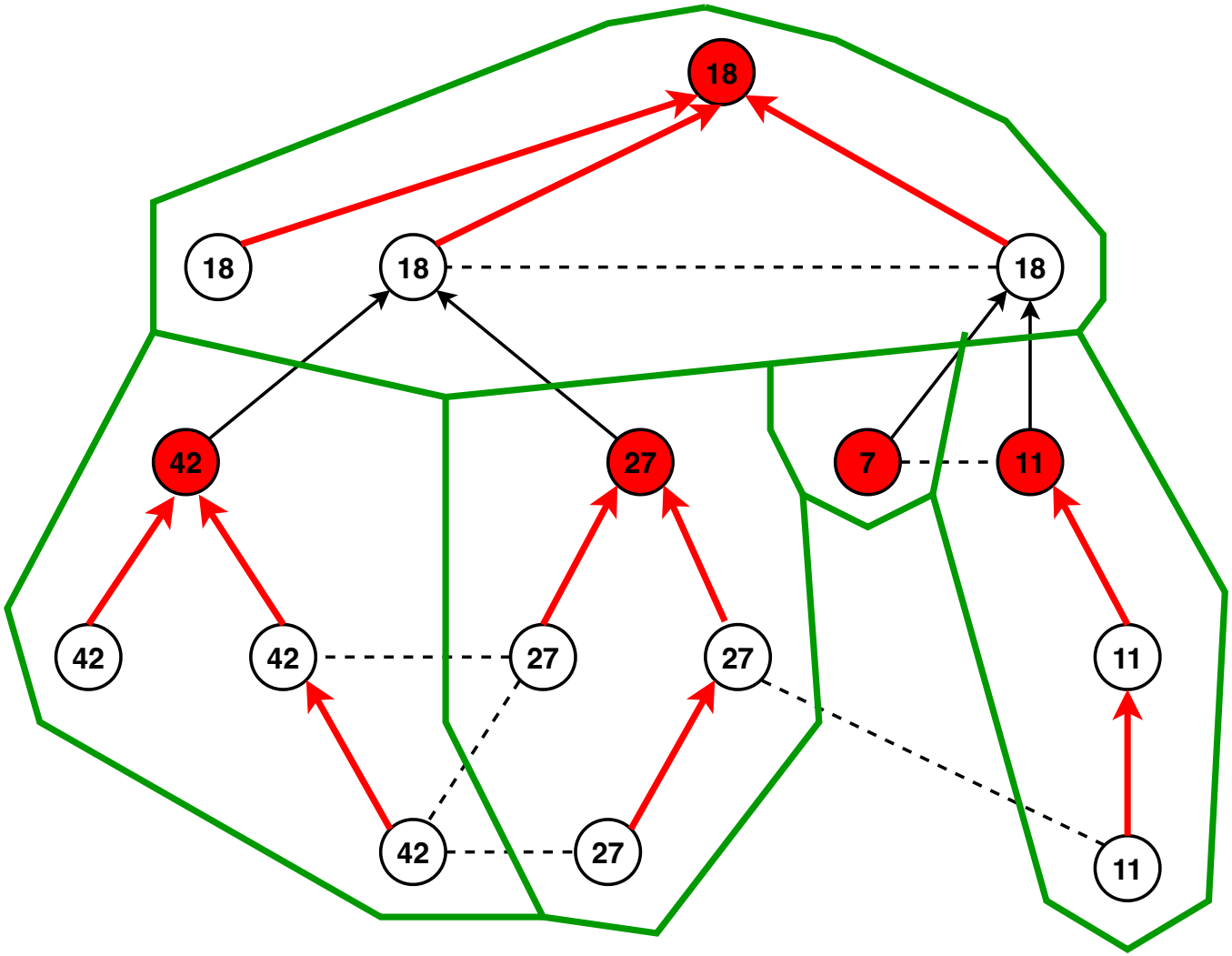
Example with $k = 2$



Clustering

#clusters $= 5 \leq \lceil \frac{16}{3} \rceil = 6$

Example with $k = 2$



Clustering

Remark: the spanning tree can be used for inter-cluster routing

Pros & Cons

Pros:

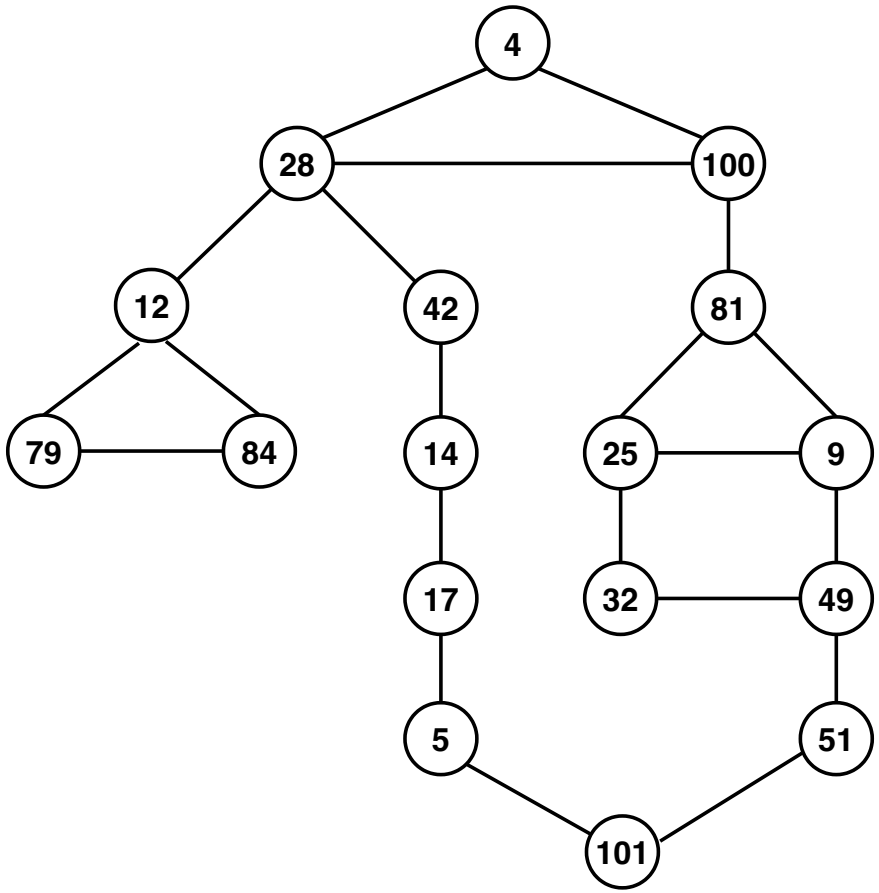
- Time-efficient computation
- BFS $\Rightarrow$ short paths

Cons:

- Memory requirement and message size: $\Omega(k \log n)$

Exercise

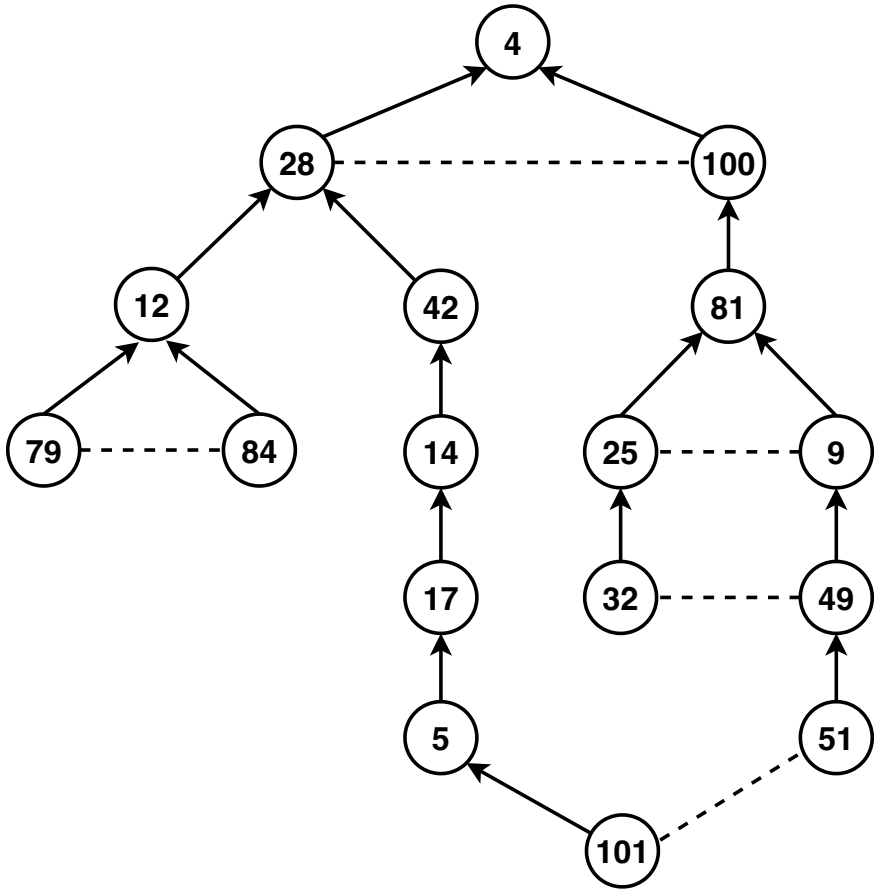
Compute a clustering using the algorithm of [3] with $k = 2$



A network

Exercise

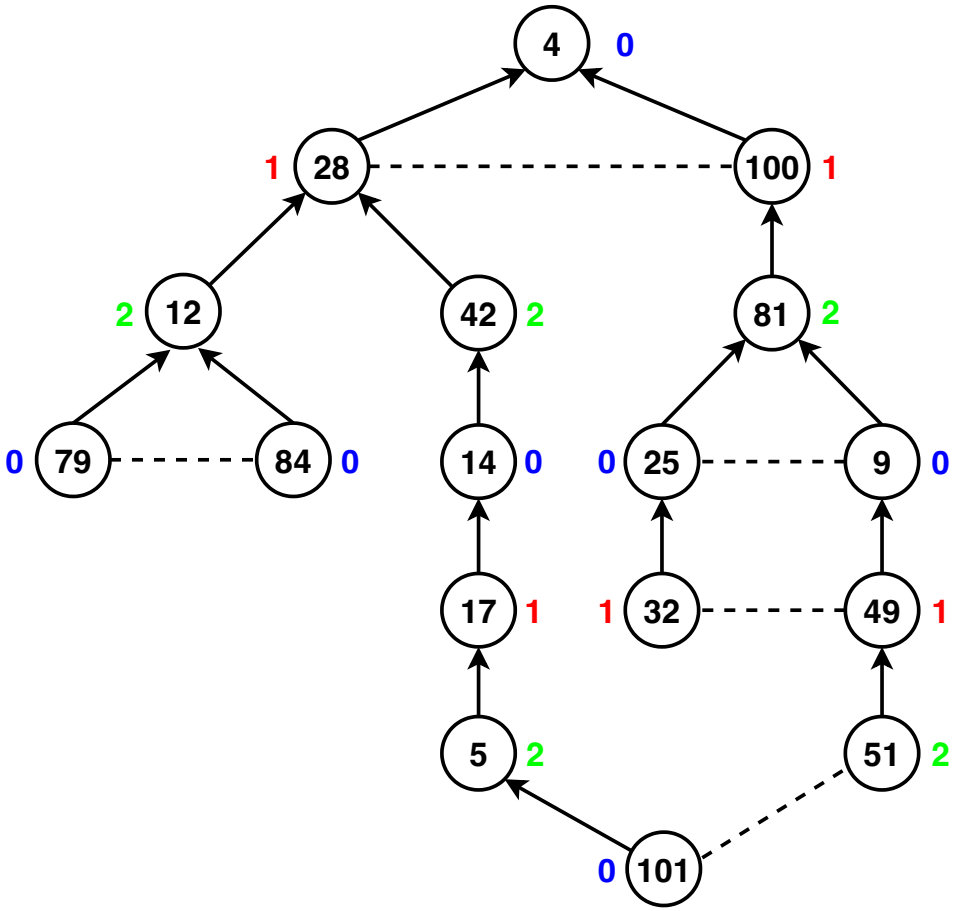
Compute a clustering using the algorithm of [3] with $k = 2$



(BFS) spanning tree

Exercise

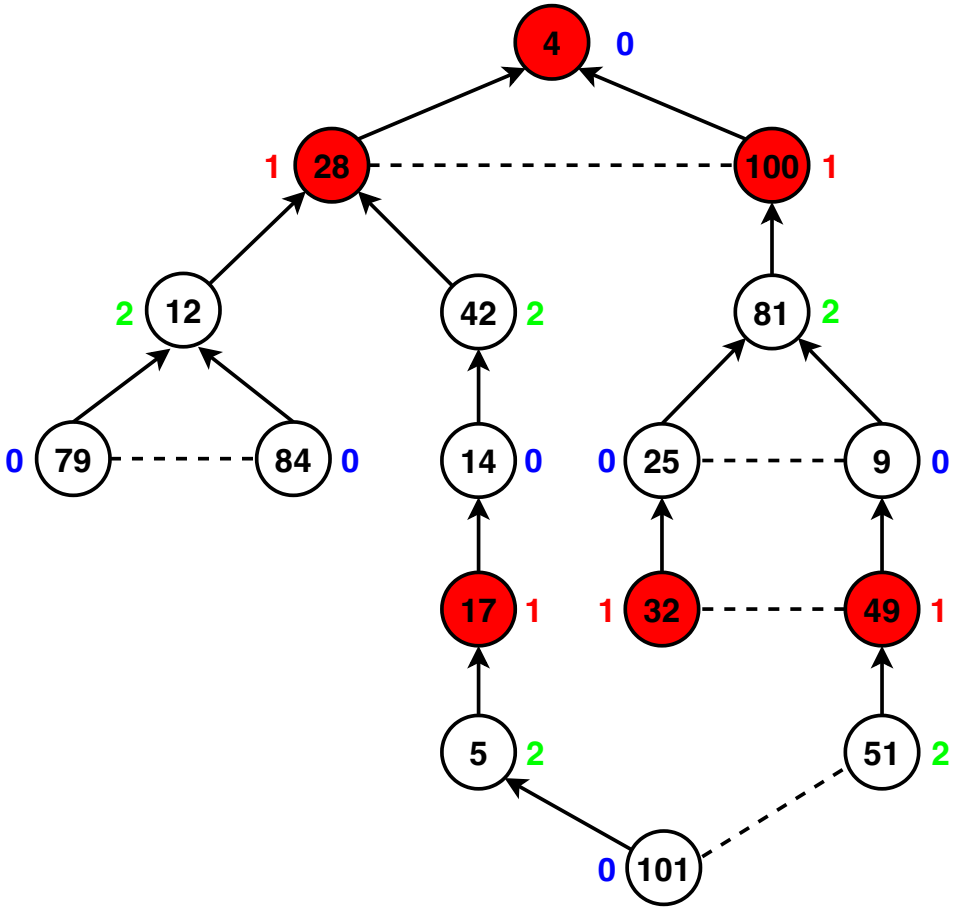
Compute a clustering using the algorithm of [3] with $k = 2$



Coloring

Exercise

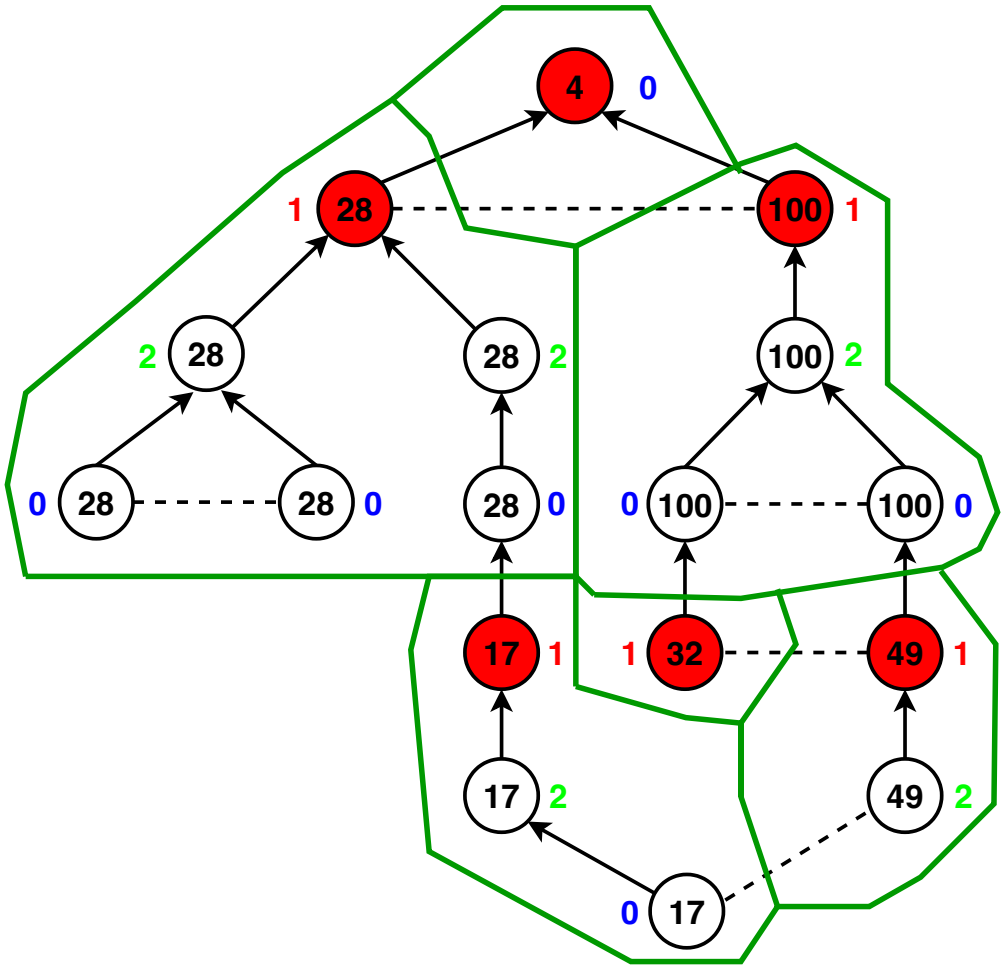
Compute a clustering using the algorithm of [3] with $k = 2$



Color choice: $|D_0| = 7$, $|D_1| = 5$, $|D_2| = 5$

Exercise

Compute a clustering using the algorithm of [3] with $k = 2$

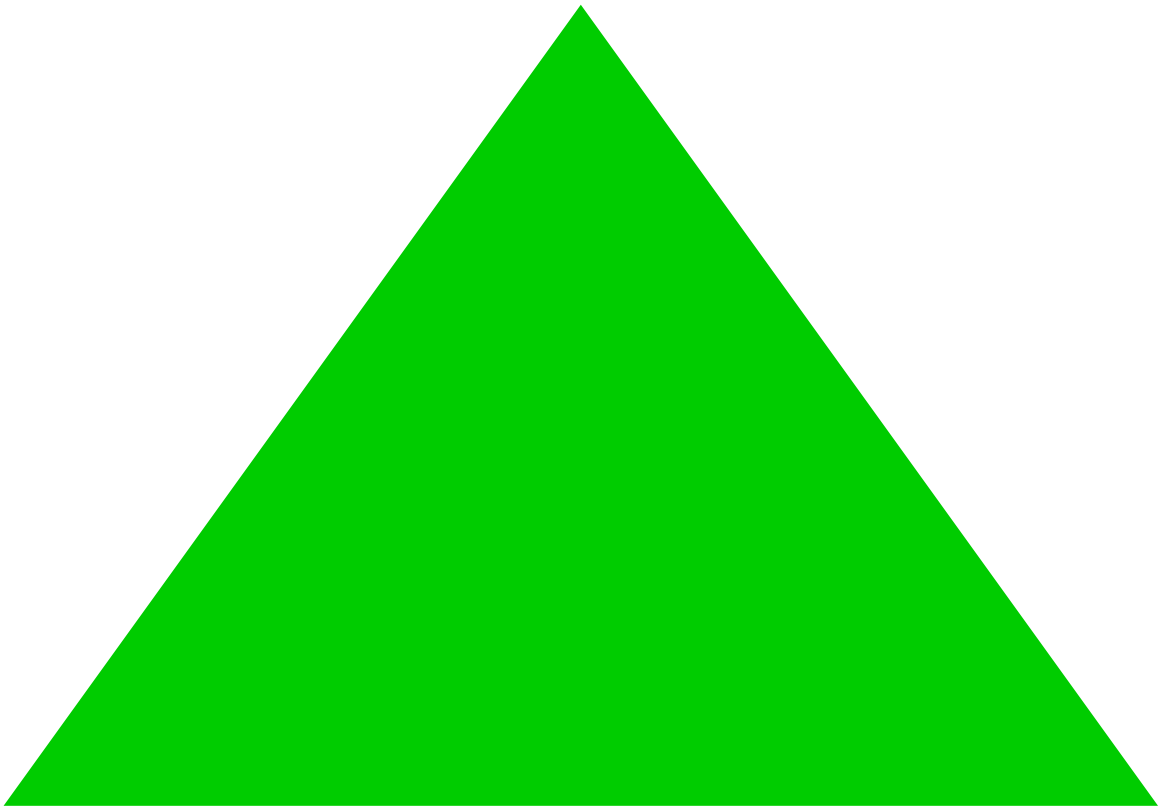


Clustering, **#clusters** $= 6 \leq \lceil \frac{17}{3} \rceil = 6$

Roadmap

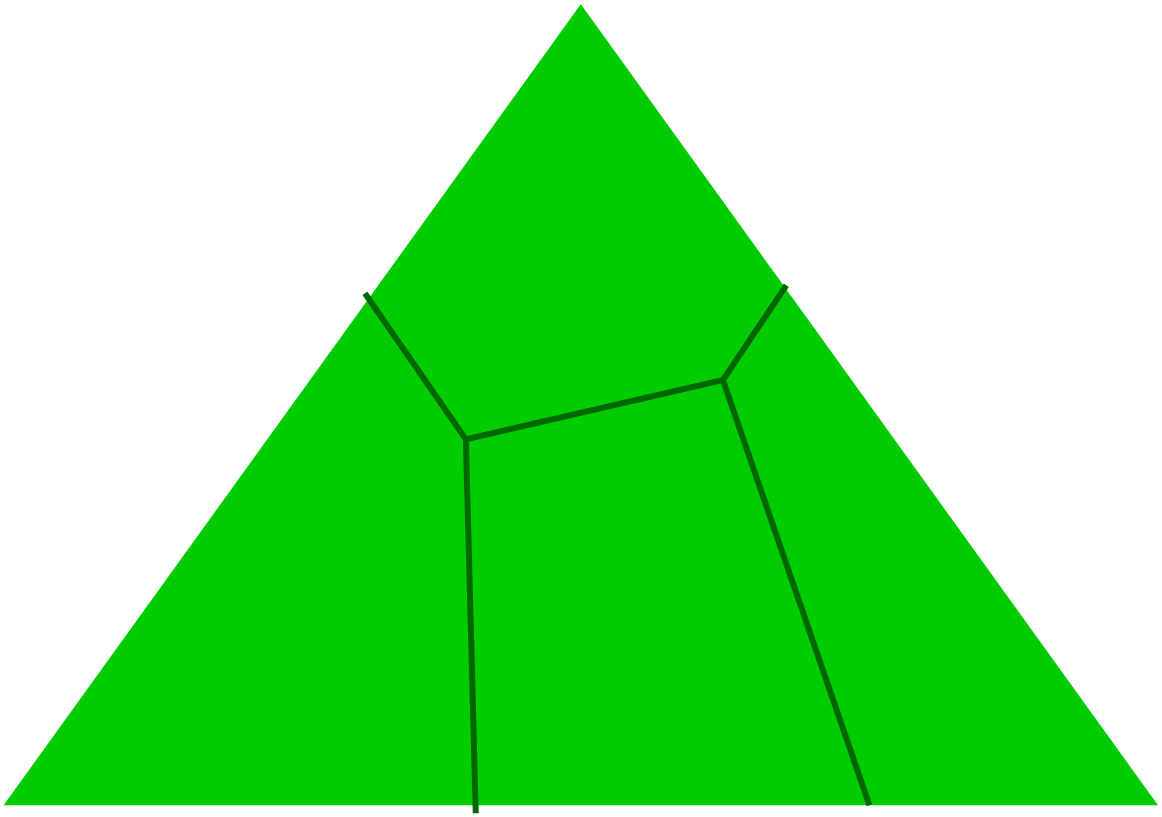
1. Introduction
2. Examples of clustering
   - Clustering of [3]
   - Clustering of [1]
3. Routing in a Clustering
4. References

Idea



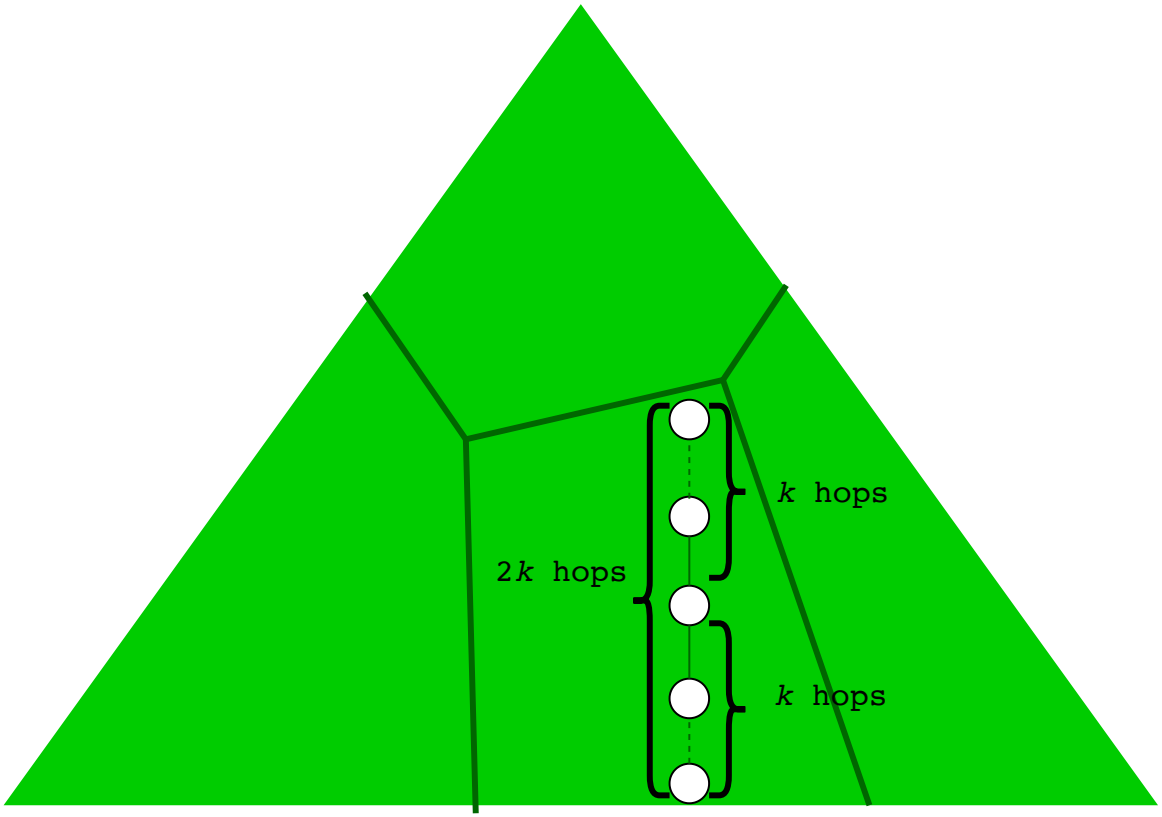
A (spanning) in-tree T

Idea



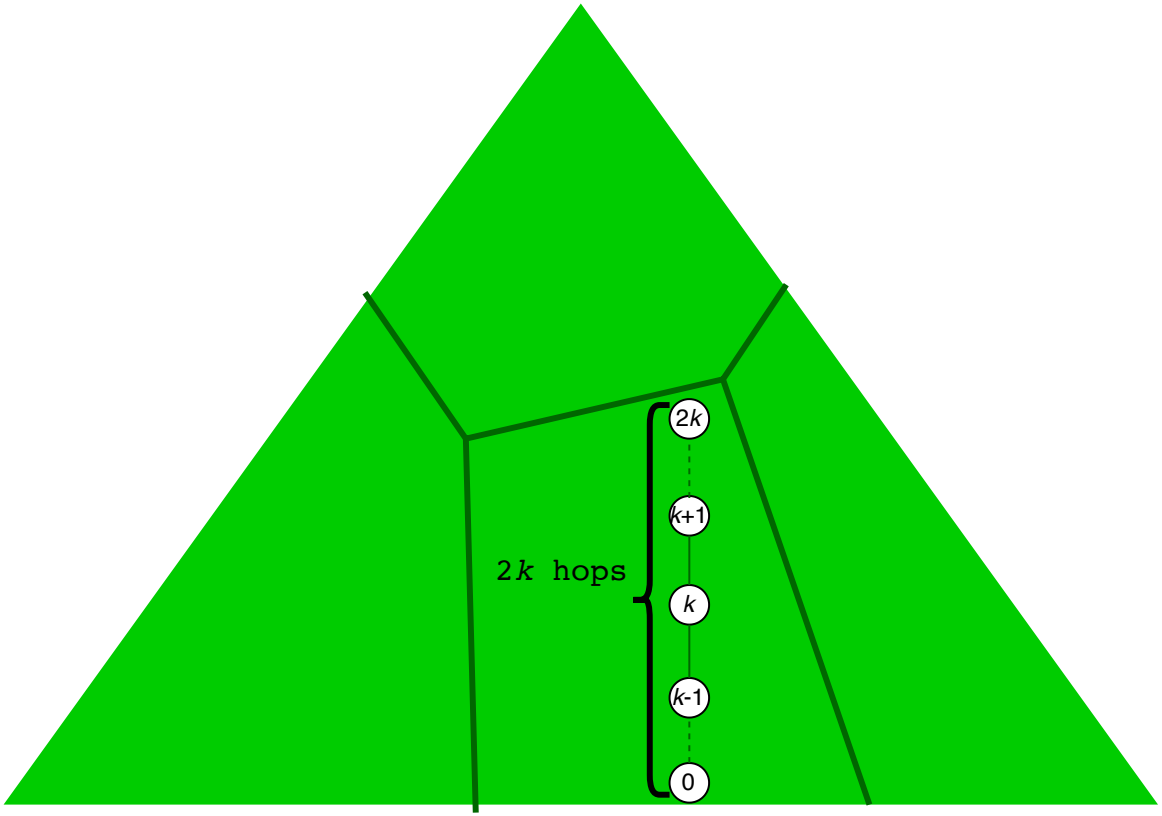
A (spanning) in-tree T with clusters

Idea



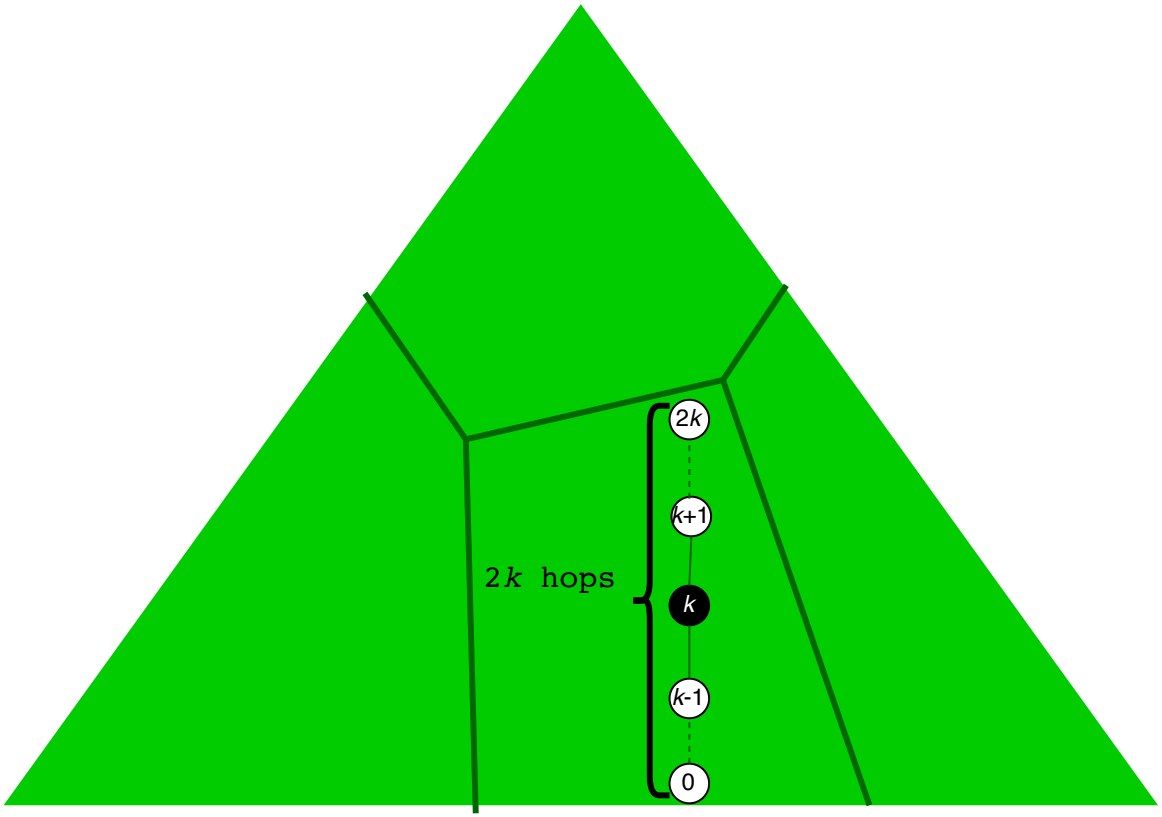
Build paths of 2k hops ($2k + 1$ nodes)

Idea



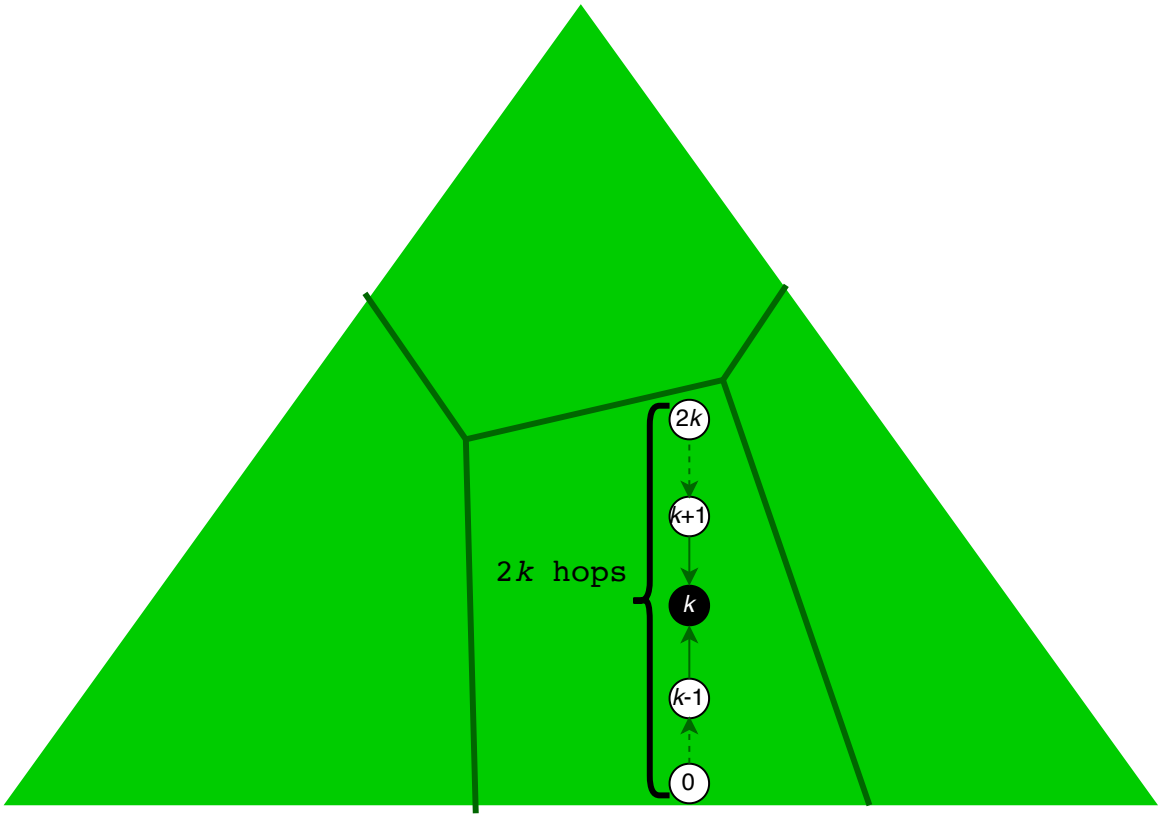
Node numbering

Idea



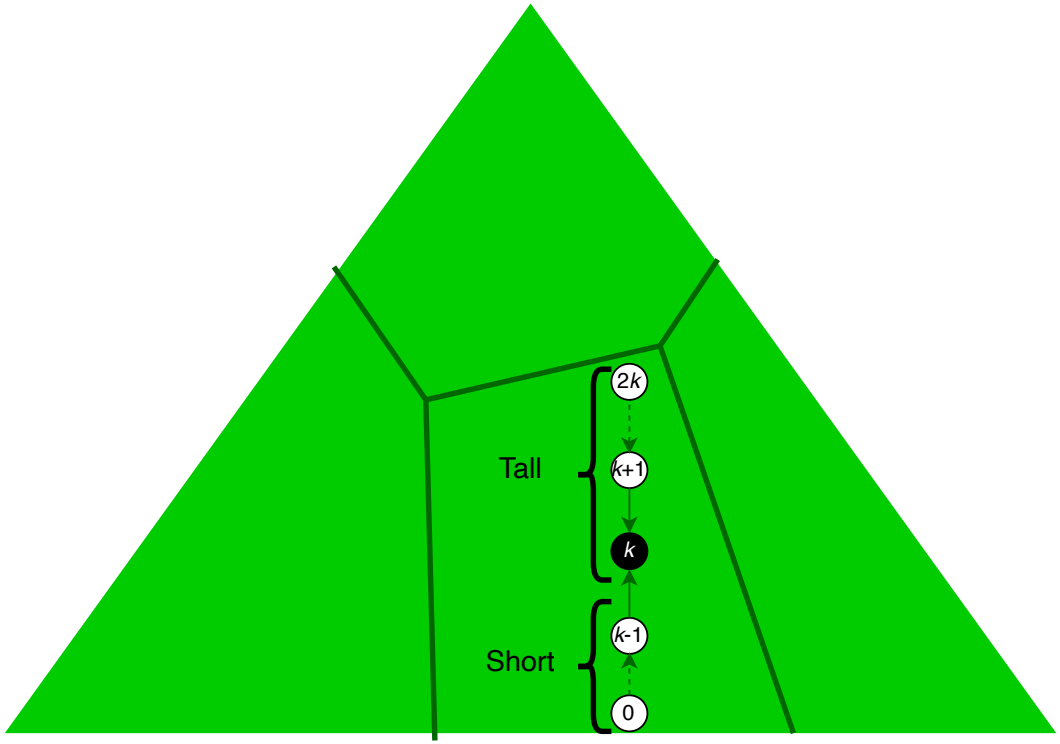
Node labeled k: clusterhead

Idea



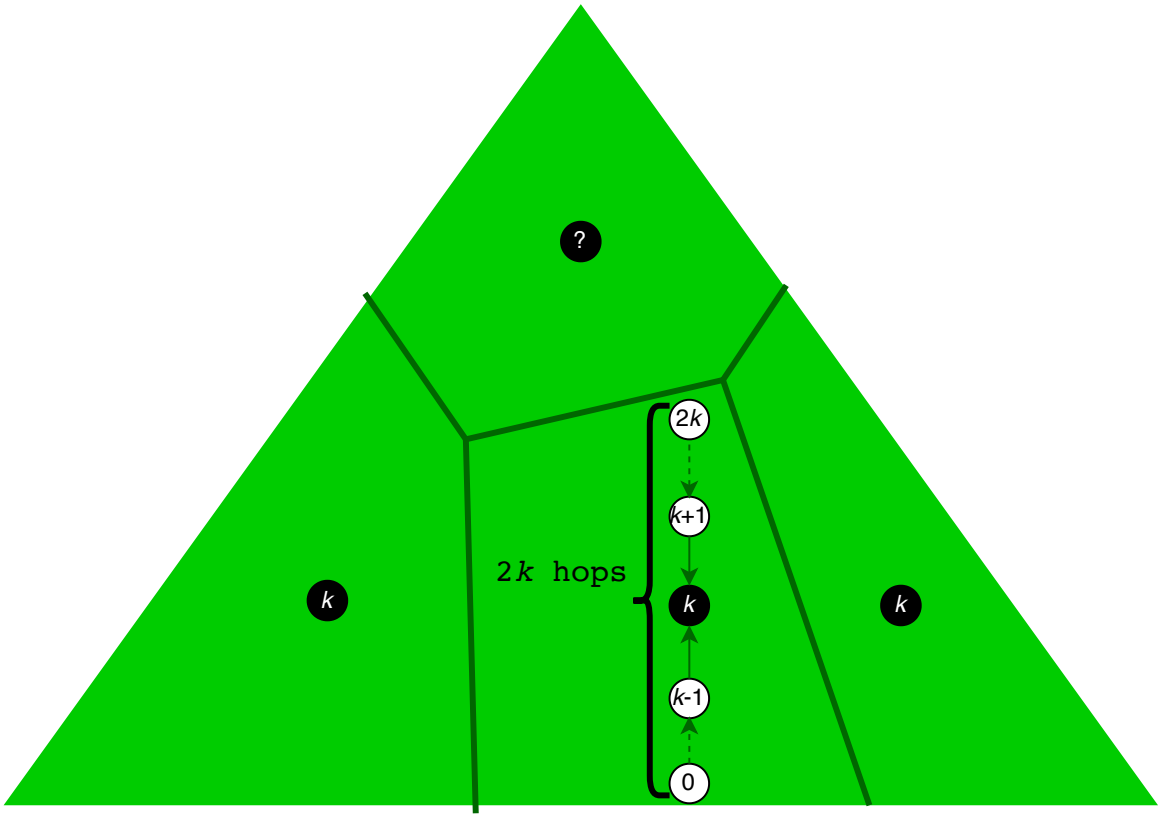
Parent Link

Idea



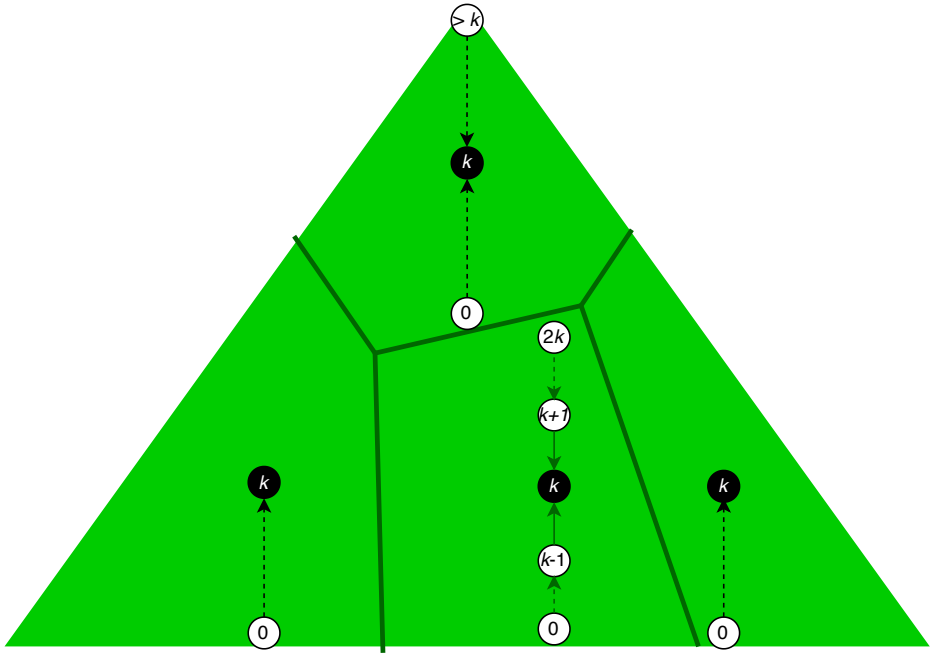
Tall and short nodes

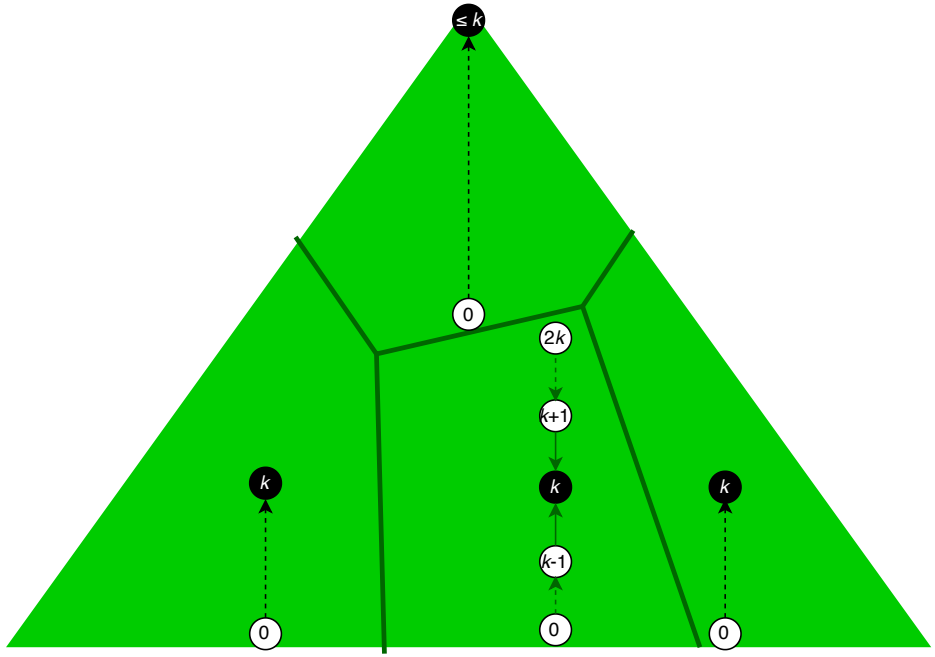
Idea



Issue: cluster of the root

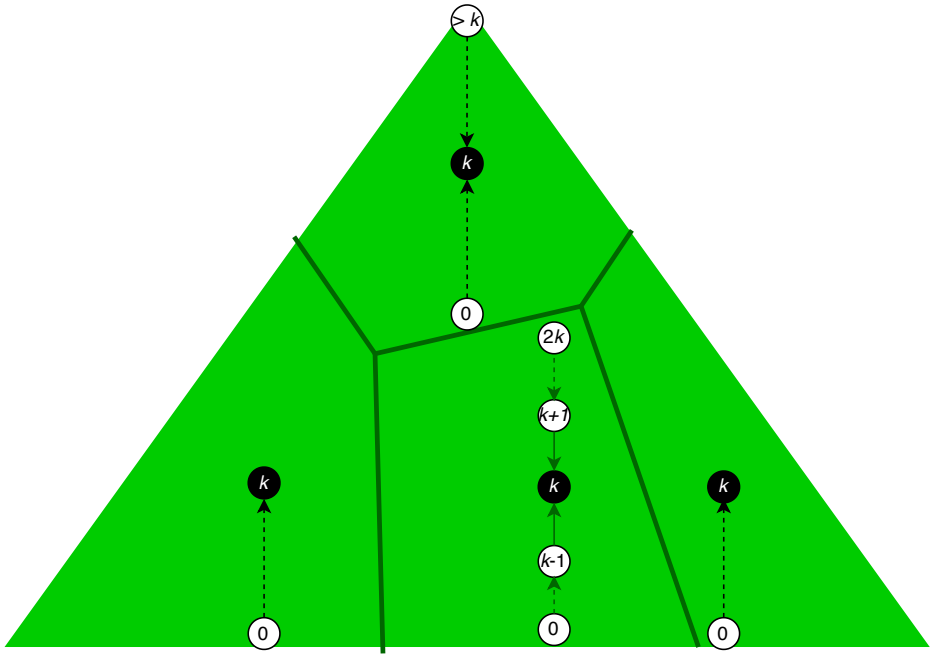
Idea

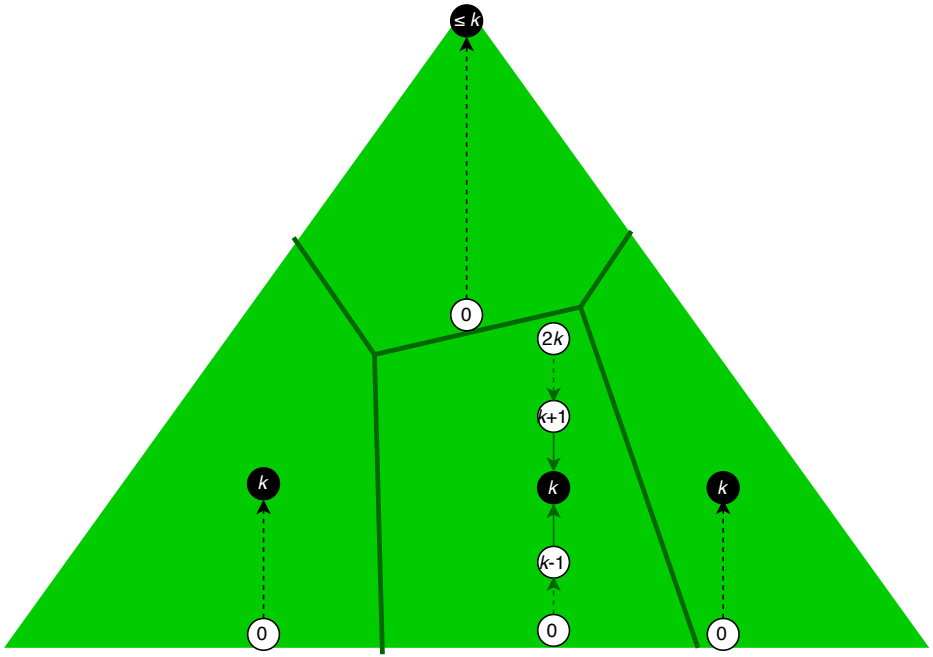




2 cases

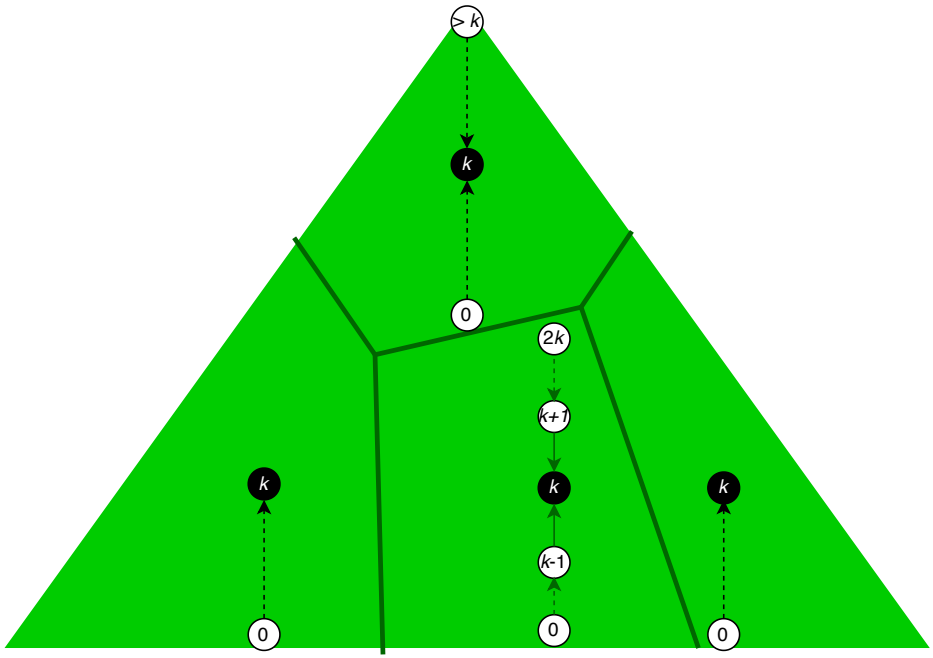
Number of clusterheads

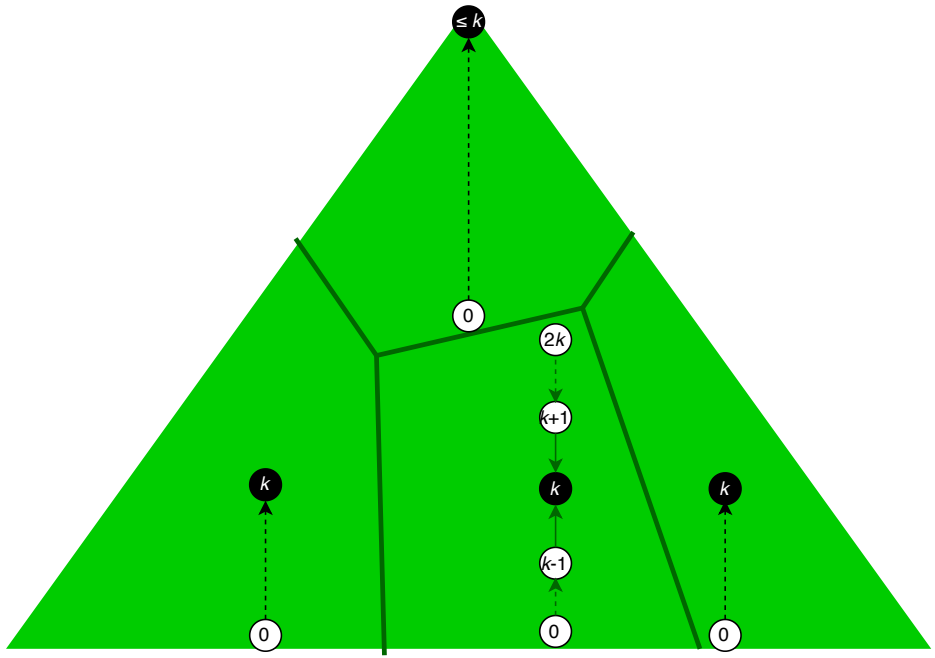




1. Non-root cluster: at least $k + 1$ nodes

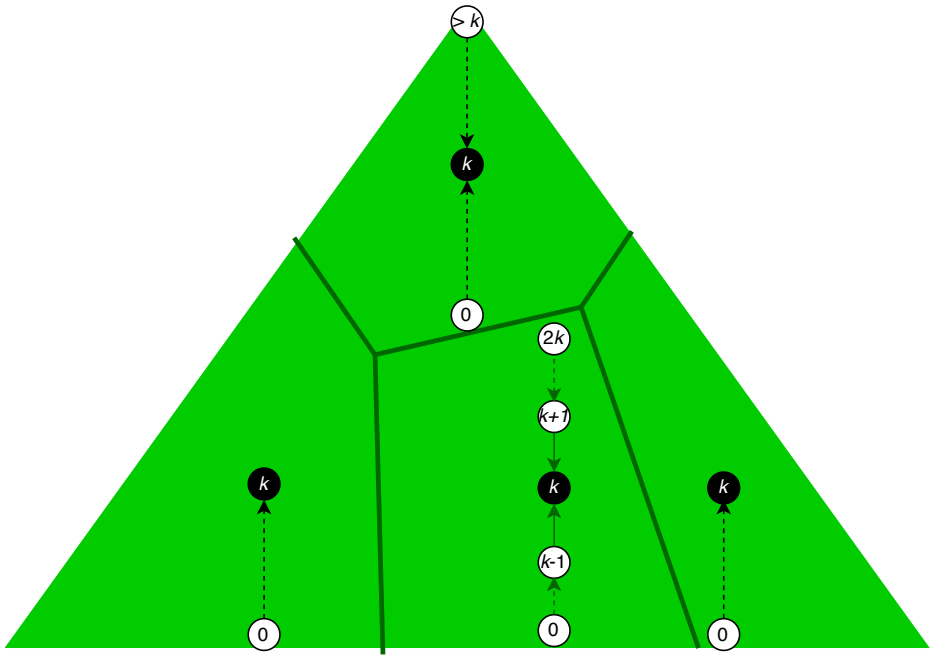
Number of clusterheads

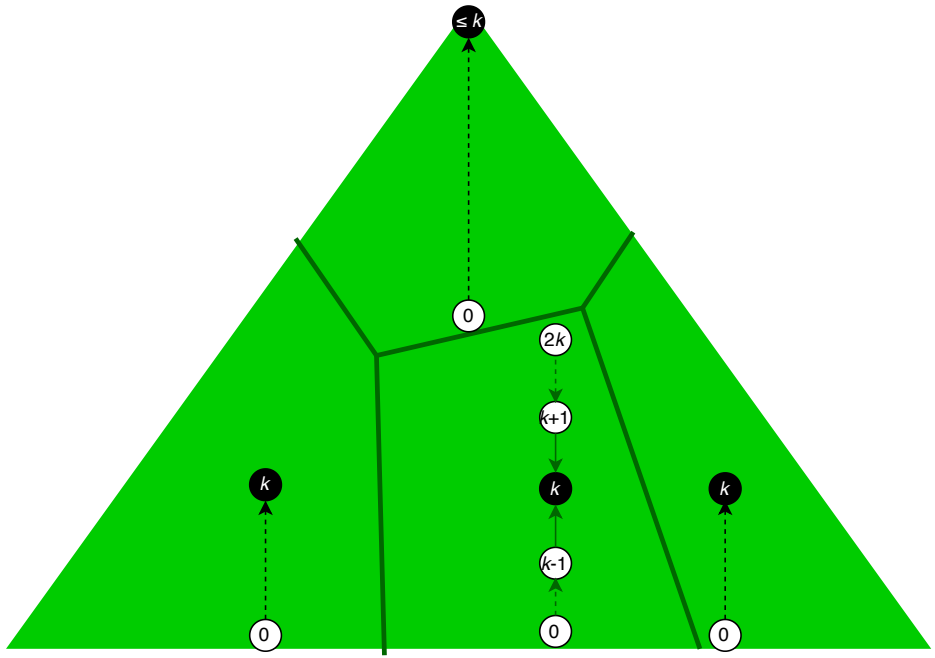




1. Non-root cluster: at least $k+1$ nodes
2. 1 cluster with the root + at most $\lfloor \frac{n-1}{k+1} \rfloor$ non-root cluster

Number of clusterheads





1. Non-root cluster: at least $k+1$ nodes
2. 1 cluster with the root + at most $\lfloor \frac{n-1}{k+1} \rfloor$ non-root cluster
3. $\sharp\text{Clusters} \leq 1 + \lfloor \frac{n-1}{k+1} \rfloor = \lfloor \frac{n+k}{k+1} \rfloor \leq \lceil \frac{n}{k+1} \rceil$

Node Numbering: function α

For every node p, $\alpha(p) \in [0..2k]$

- $\mathit{maxShort}(p) = \max(\{\alpha(q) \mid q \in \mathit{Children}(p) \wedge \alpha(q) < k\} \cup \{-1\})$
- $\mathit{minTall}(p) = \min(\{\alpha(q) \mid q \in \mathit{Children}(p) \wedge \alpha(q) \geq k\} \cup \{2k+1\})$

if $\mathit{maxShort}(p) + \mathit{minTall}(p) + 2 \leq 2k$ **then**
    $\alpha(p) = \mathit{minTall}(p) + 1$
else
    $\alpha(p) = \mathit{maxShort}(p) + 1$
end if

Tree Structure of clusters

$\alpha(p) = k$ or (p is the root and $\alpha(p) \leq k$): clusterhead (root of the cluster)

If p is not a clusterhead:

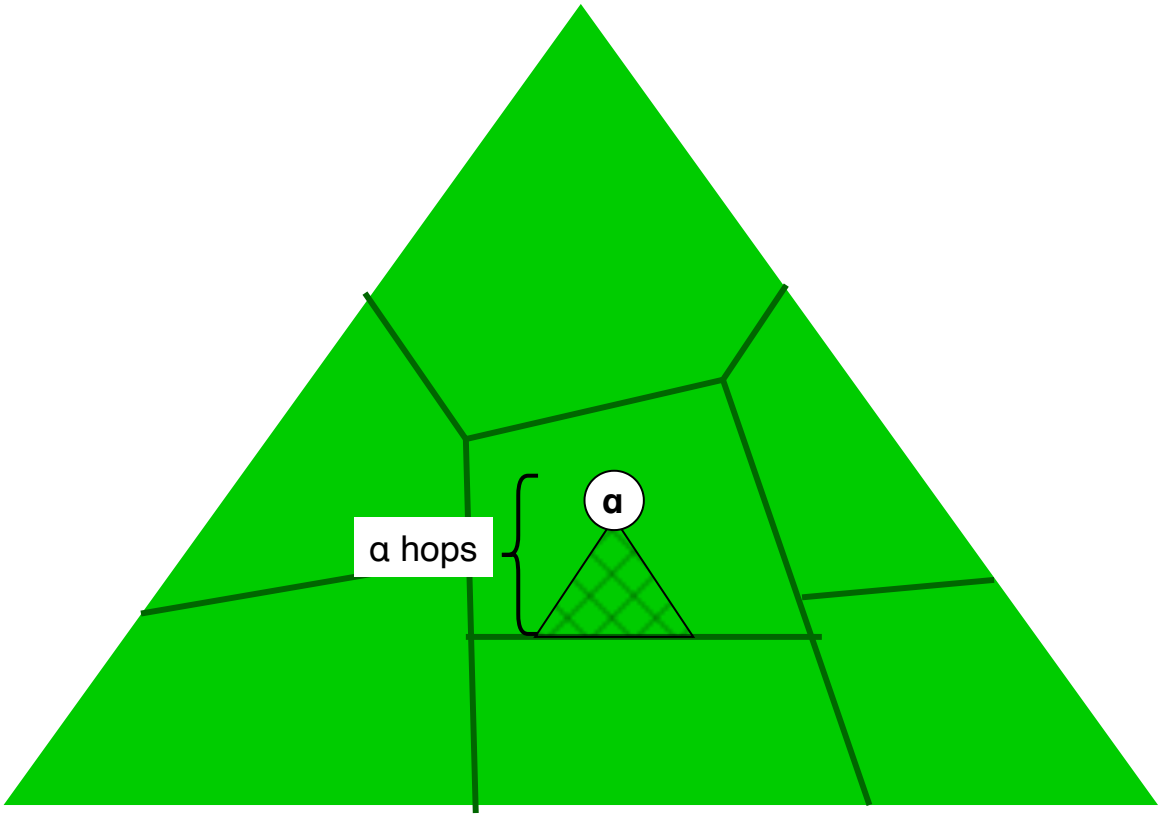
$\alpha(p) < k$: parent in the cluster := parent in the tree

$\alpha(p) > k$: parent in the cluster := a child q with $\alpha(q) = \mathit{minTall}(p)$

(ties can be broken arbitrarily, *e.g.*, using identifiers)

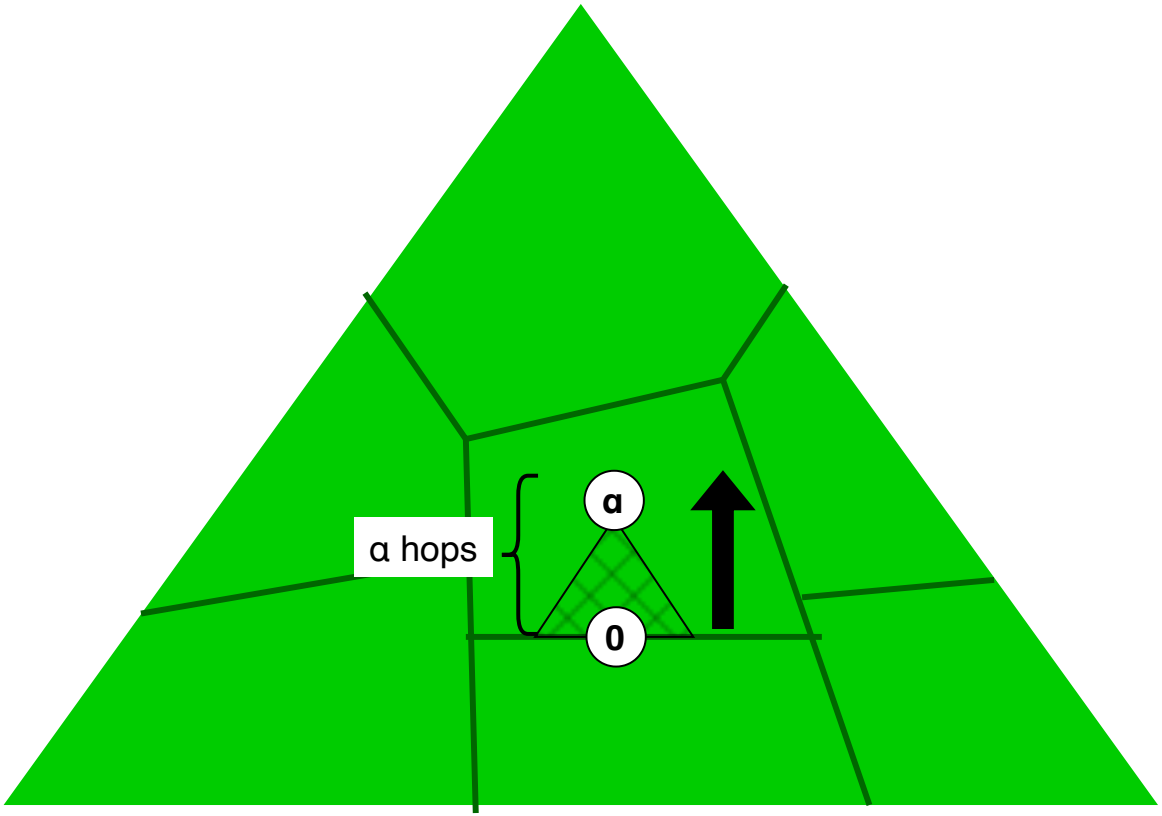
Meaning of α

$\alpha(p)$ is the distance from p to q where q is the furthest process in $T(p)$ that is in the same cluster as p



Meaning of α

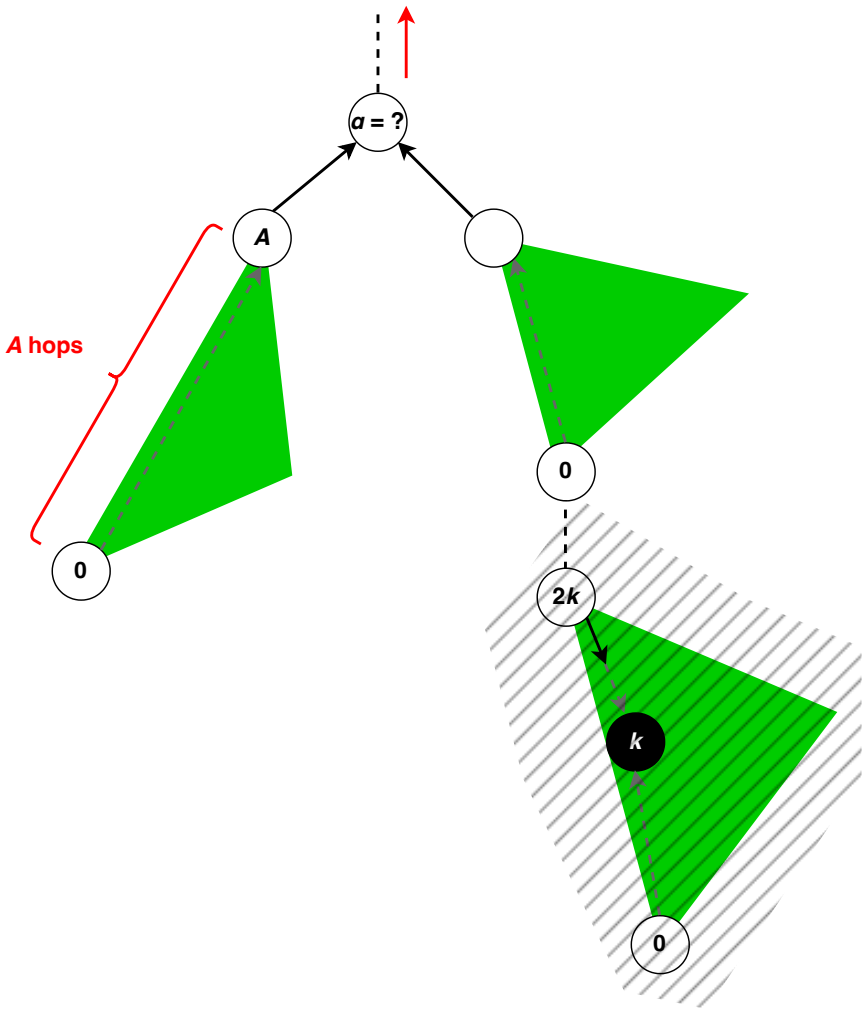
$\alpha(p)$ is the distance from p to q where q is the furthest process in $T(p)$ that is in the same cluster as p



Bottom-up Computation

Computation of α: special cases

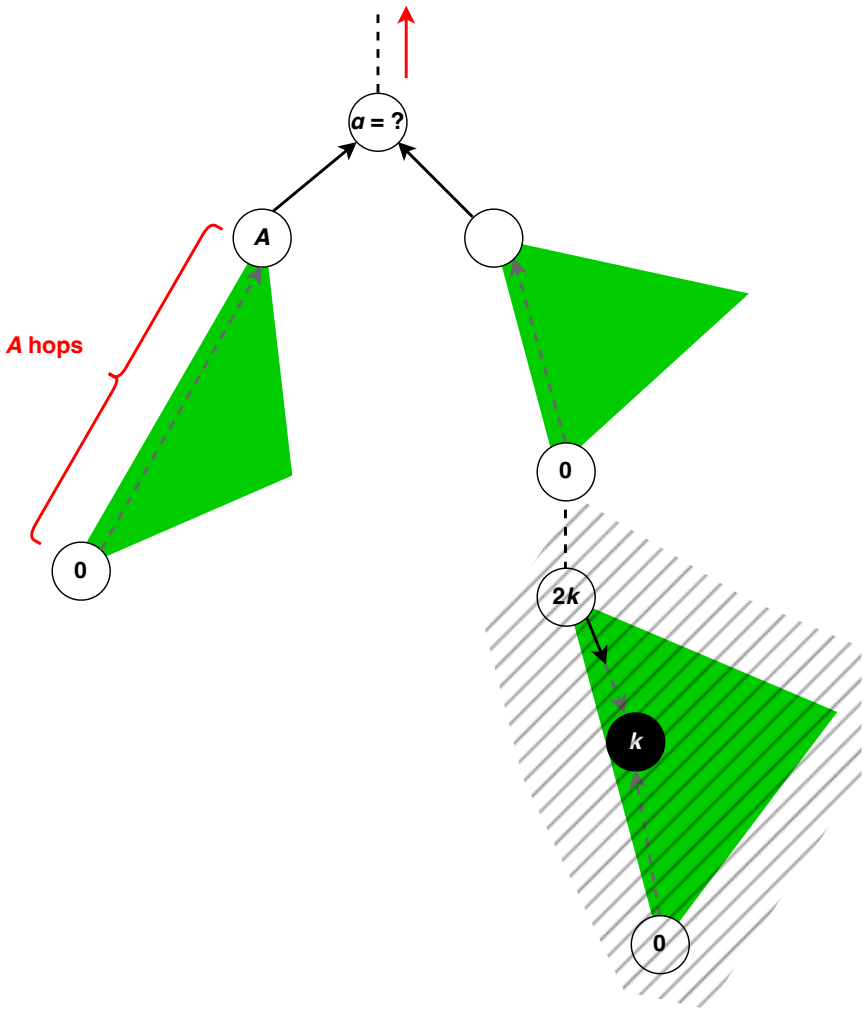
No tall children



- $A = maxShort(p)$
- $-1 \leq A < k$
- $minTall(p) = 2k + 1$

Computation of α: special cases

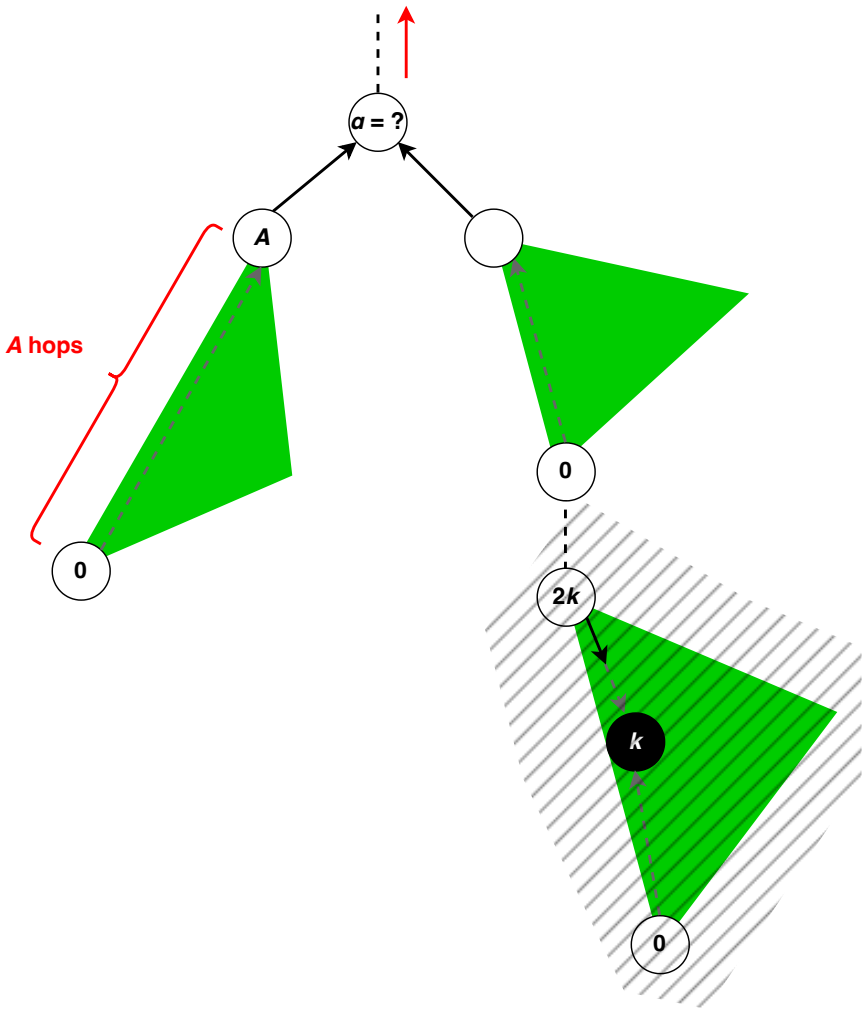
No tall children



- $A = maxShort(p)$
- $-1 \leq A < k$
- $minTall(p) = 2k + 1$
- $maxShort(p) + minTall(p) + 2 > 2k$
  $\rightarrow \alpha := A + 1$

Computation of α: special cases

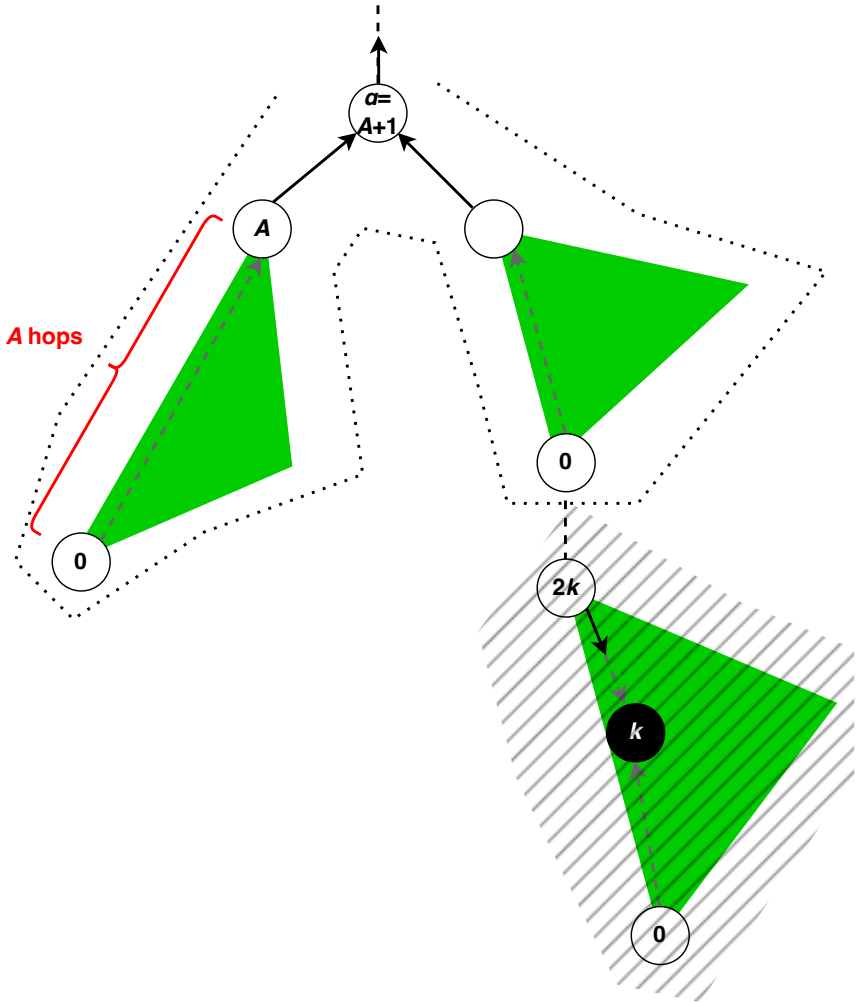
No tall children



- $A = maxShort(p)$
- $-1 \leq A < k$
- $minTall(p) = 2k + 1$
- $maxShort(p) + minTall(p) + 2 > 2k$
  $\rightarrow \alpha := A + 1$
- Two cases:
  - **There are short children:**
  - **There are no short children:**
    (the node is a leaf)

Computation of α: special cases

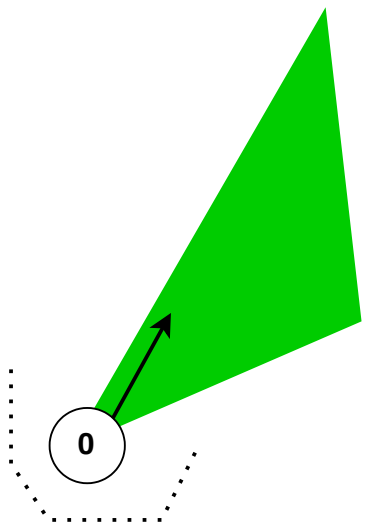
No tall children



- $A = maxShort(p)$
- $-1 \leq A < k$
- $minTall(p) = 2k + 1$
- $maxShort(p) + minTall(p) + 2 > 2k$
  $\rightarrow \alpha := A + 1$
- Two cases:
  - **There are short children:** $0 \leq A < k$
  - **There are no short children:**
    (the node is a leaf)

Computation of α: special cases

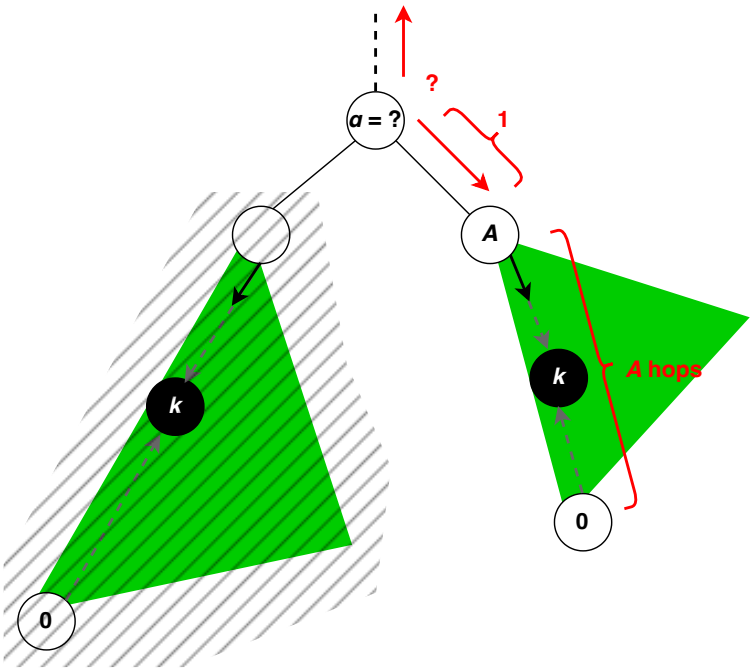
No tall children



- $A = maxShort(p)$
- $-1 \leq A < k$
- $minTall(p) = 2k + 1$
- $maxShort(p) + minTall(p) + 2 > 2k$
  $\rightarrow \alpha := A + 1$
- Two cases:
  - **There are short children:** $0 \leq A < k$
  - **There are no short children:**
    (the node is a leaf)
    $\alpha = -1 + 1 = 0$

Computation of α: special cases

Tall children, but no short



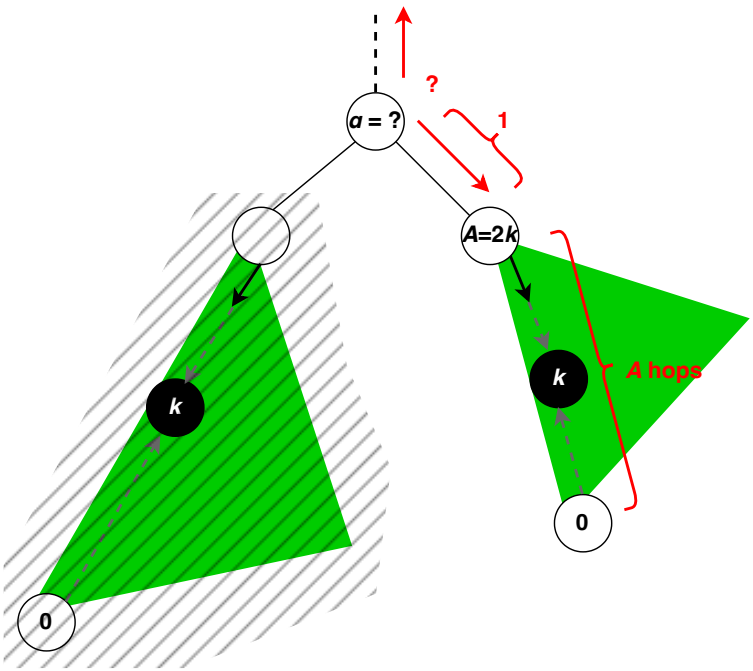
$maxShort(p) = -1$

$minTall(p) = A$

Two cases: either $A = 2k$ or $A < 2k$

Computation of α: special cases

Tall children, but no short



$maxShort(p) = -1$

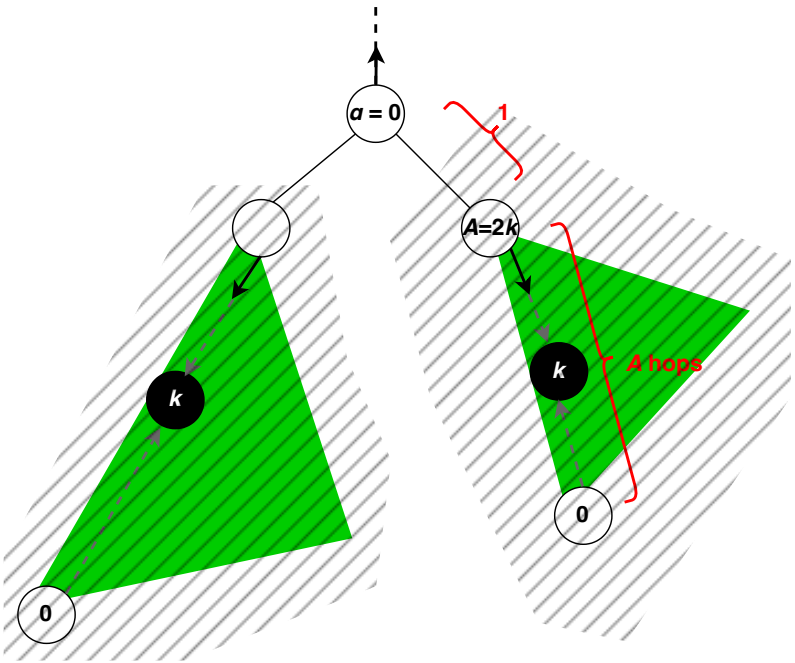
$minTall(p) = A$

Two cases: either $A = 2k$ or $A < 2k$

- $A = 2k$:

Computation of α: special cases

Tall children, but no short



$maxShort(p) = -1$

$minTall(p) = A$

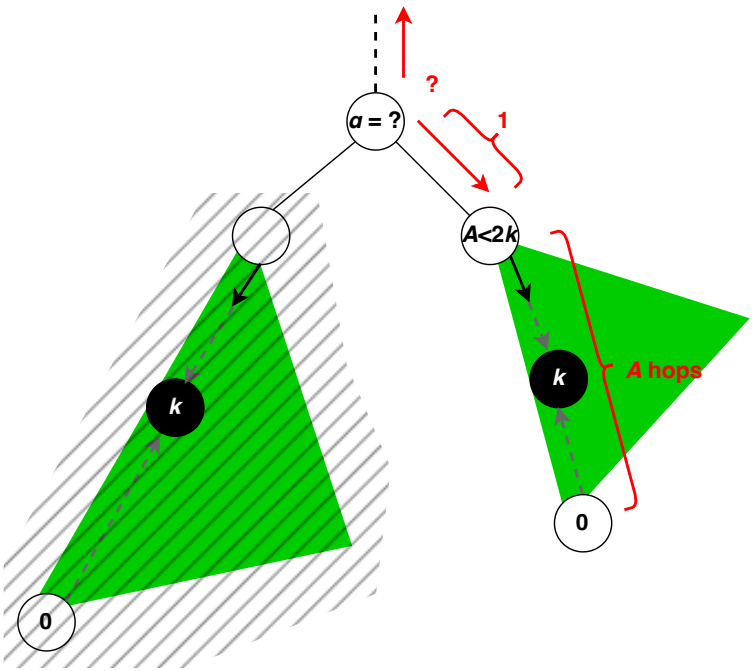
Two cases: either $A = 2k$ or $A < 2k$

- $A = 2k$:

  $maxShort(p)+minTall(p)+2=2k+1>2k$
  $\rightarrow \alpha := maxShort(p) + 1 = 0$

Computation of α: special cases

Tall children, but no short



$maxShort(p) = -1$

$minTall(p) = A$

Two cases: either $A = 2k$ or $A < 2k$

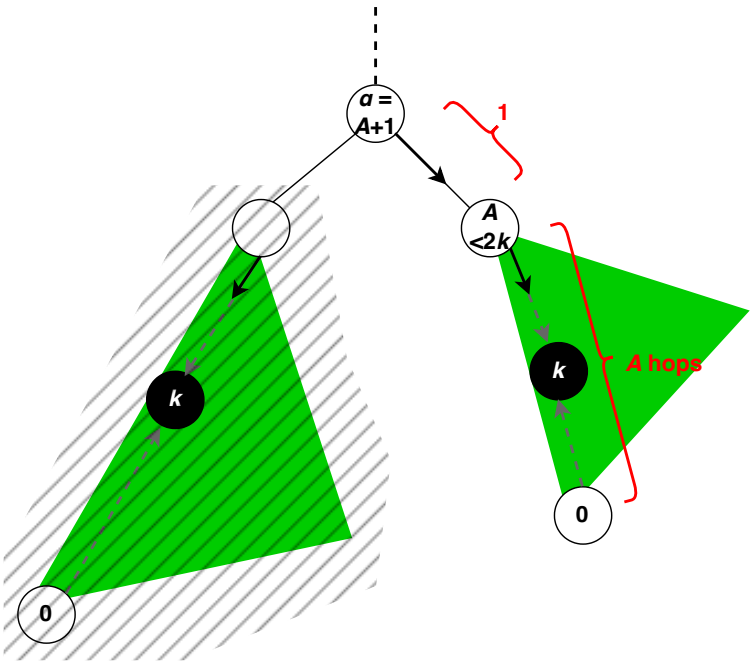
- $A = 2k$:

  $maxShort(p)+minTall(p)+2=2k+1>2k$
  $\rightarrow \alpha := maxShort(p) + 1 = 0$

- $A < 2k$:

Computation of α: special cases

Tall children, but no short



$maxShort(p) = -1$

$minTall(p) = A$

Two cases: either $A = 2k$ or $A < 2k$

- $A = 2k$:

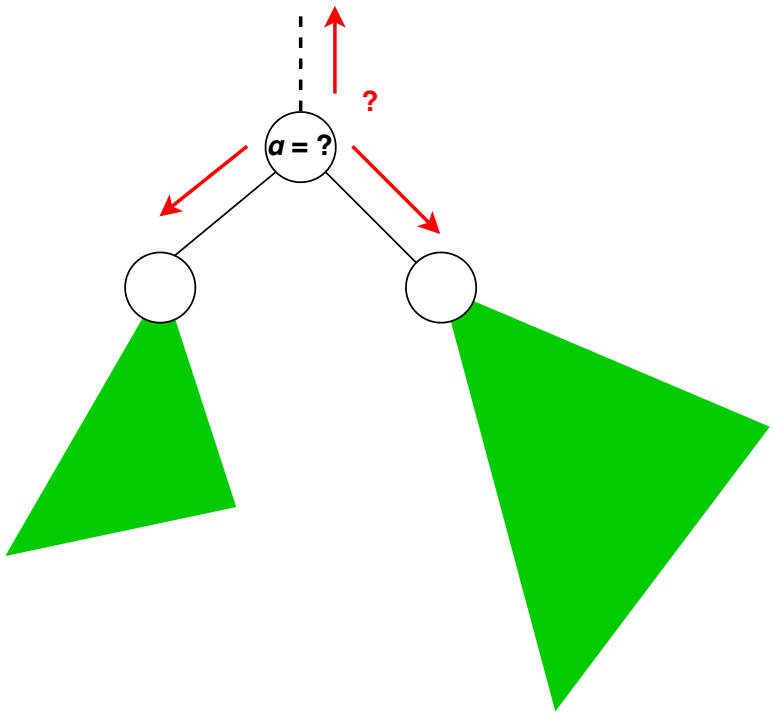
  $maxShort(p)+minTall(p)+2=2k+1>2k$
  $\rightarrow \alpha := maxShort(p) + 1 = 0$

- $A < 2k$:

  $maxShort(p) + minTall(p) + 2 \leq 2k$
  $\rightarrow \alpha := A + 1$

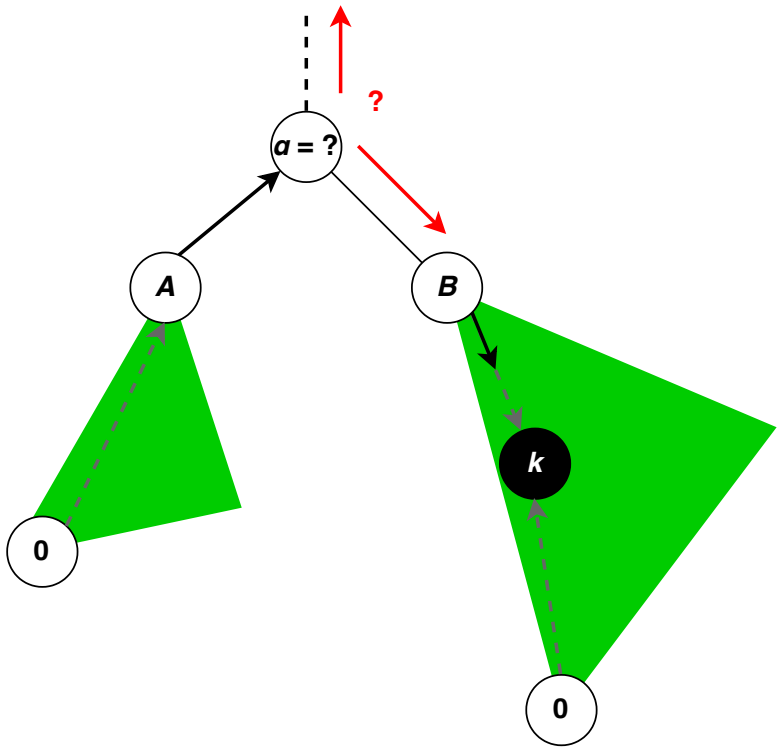
Computation of α: general case

Tall and short children



Computation of α: general case

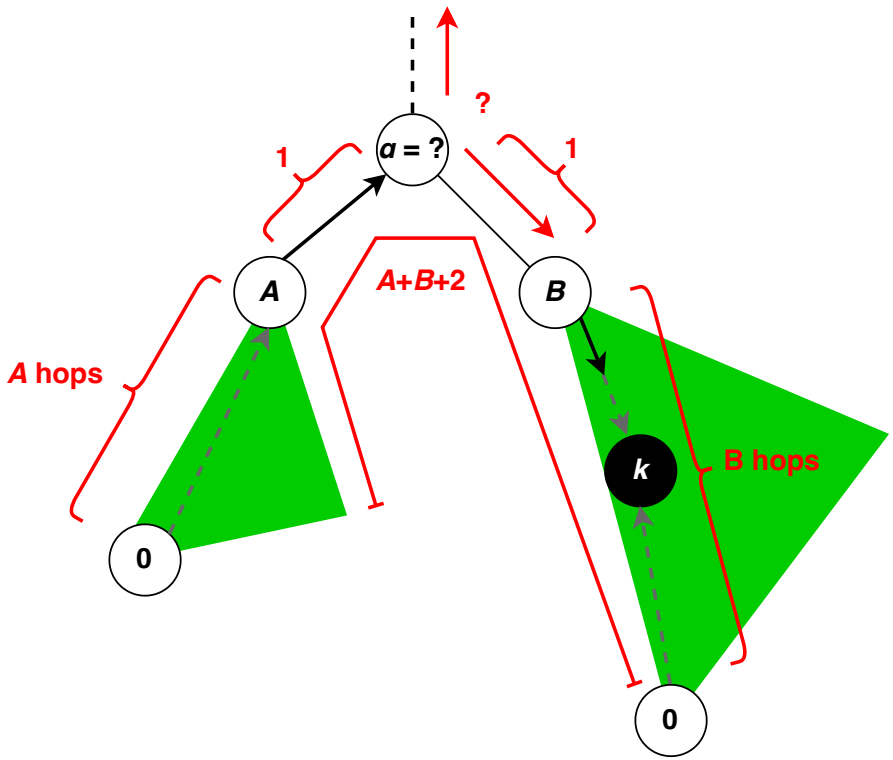
Tall and short children



- $A = maxShort(p)$
- $B = minTall(p)$
- $0 \leq A < k \leq B \leq 2k$

Computation of α: general case

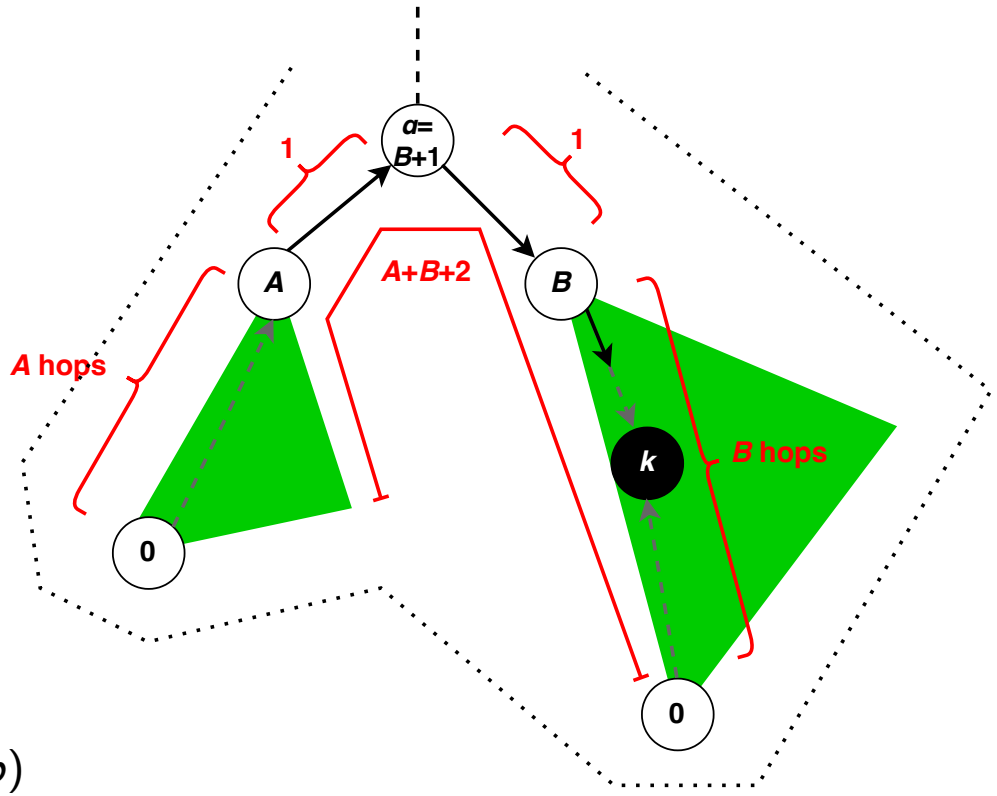
Tall and short children



- $A = maxShort(p)$
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- $0 \leq A < k \leq B \leq 2k$

Computation of α: general case

Tall and short children

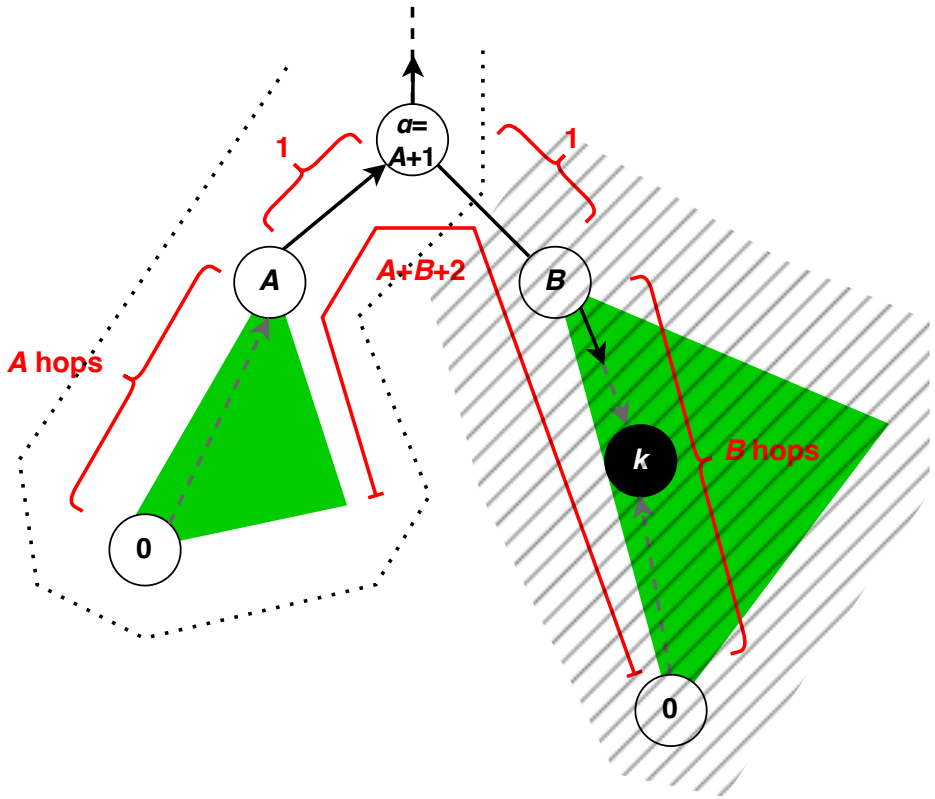


- $A = maxShort(p)$
- $B = minTall(p)$
- $0 \leq A < k \leq B \leq 2k$

Case $\boldsymbol{A + B + 2 \leq 2k}$: $\alpha := B + 1$

Computation of α: general case

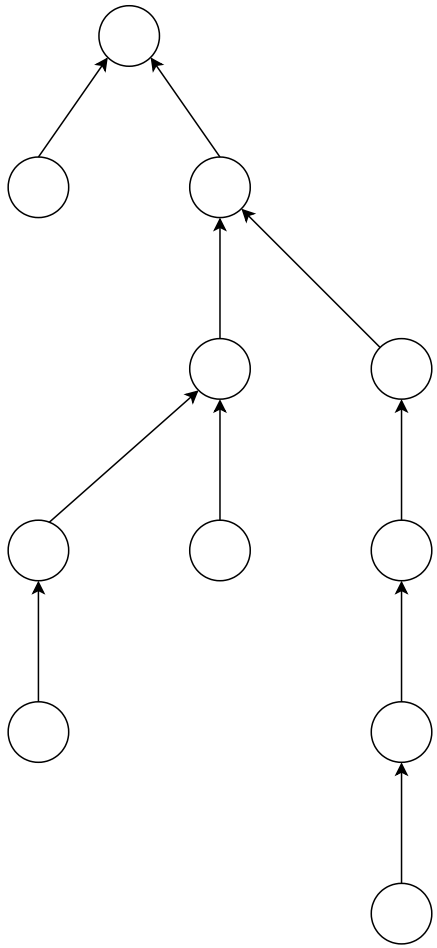
Tall and short children



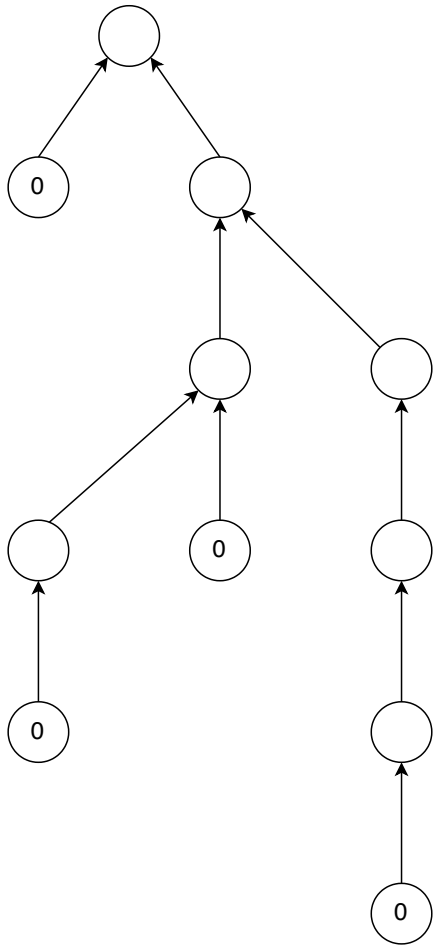
- $A = maxShort(p)$
- $B = minTall(p)$
- $0 \leq A < k \leq B \leq 2k$

Case $\boldsymbol{A + B + 2 > 2k}$: $\alpha := A + 1$

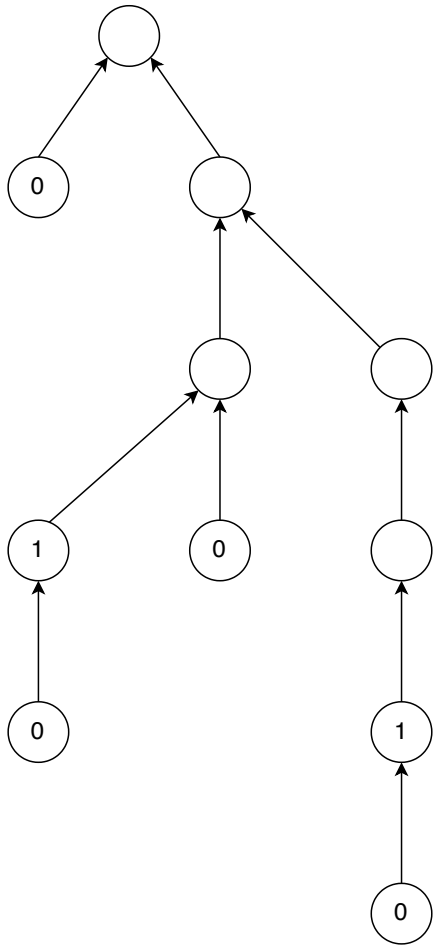
Example with $k = 2$



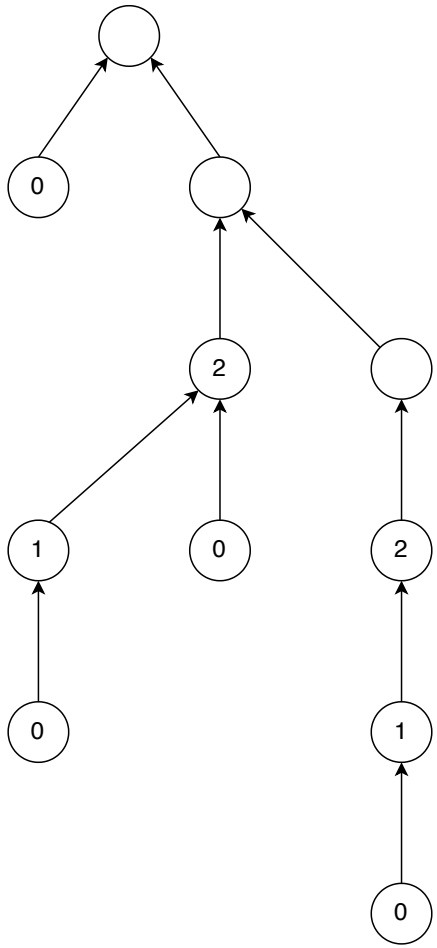
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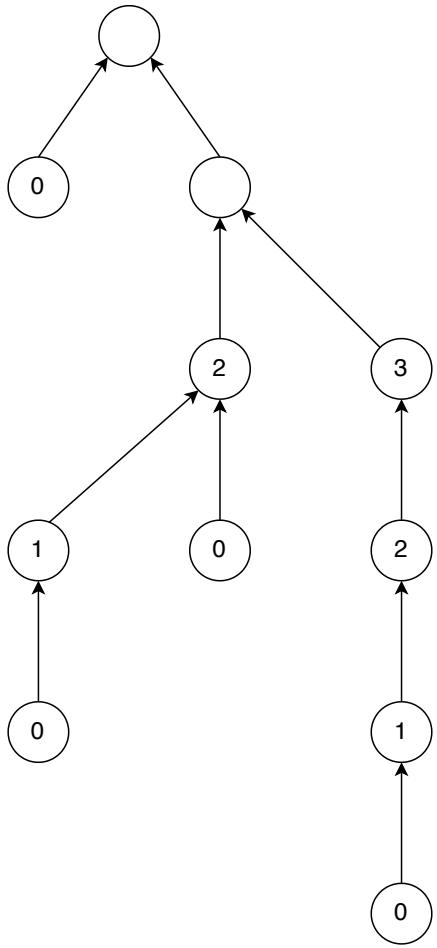
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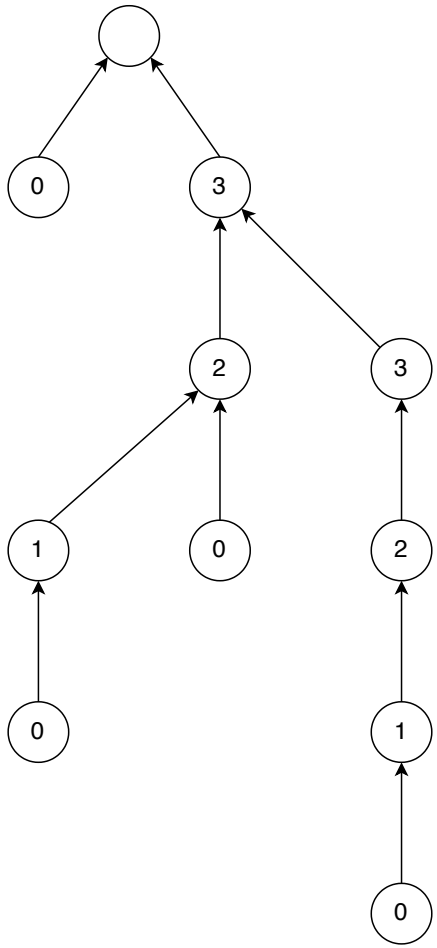
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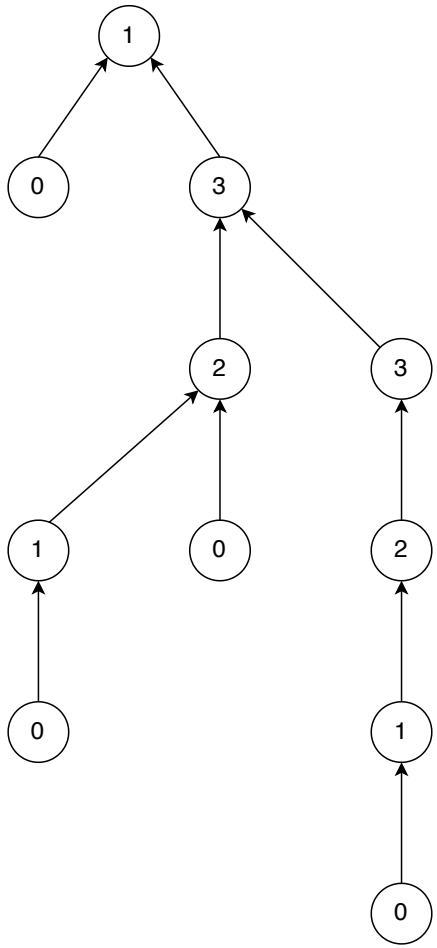
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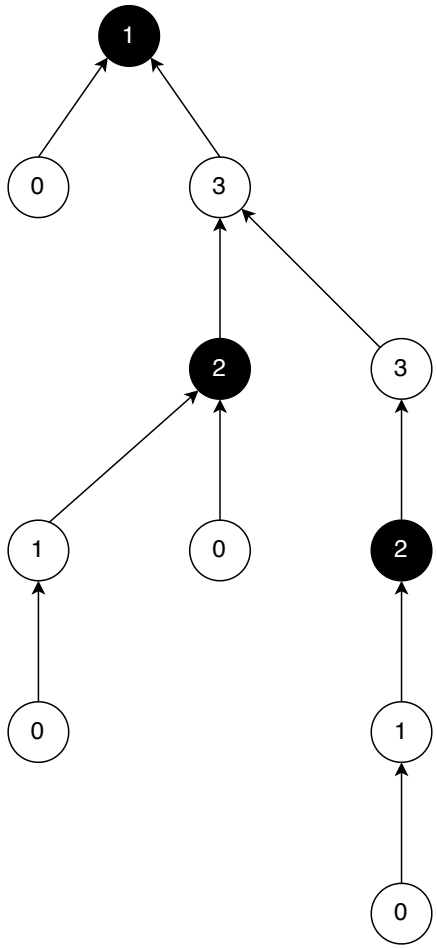
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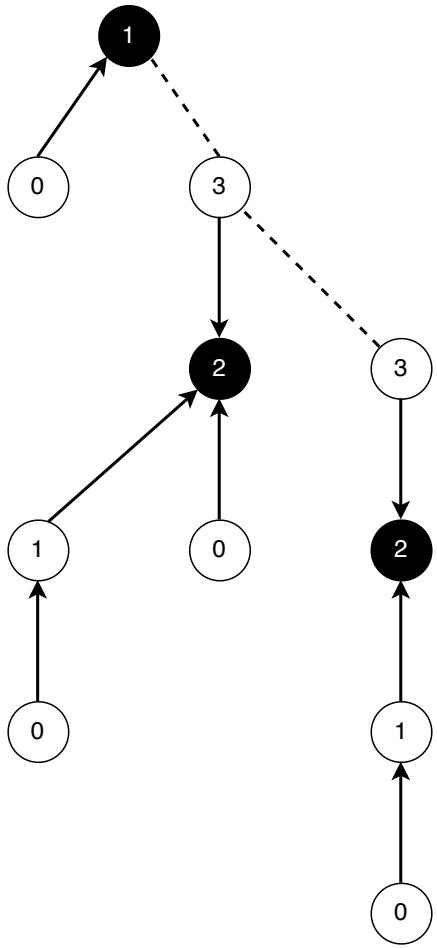
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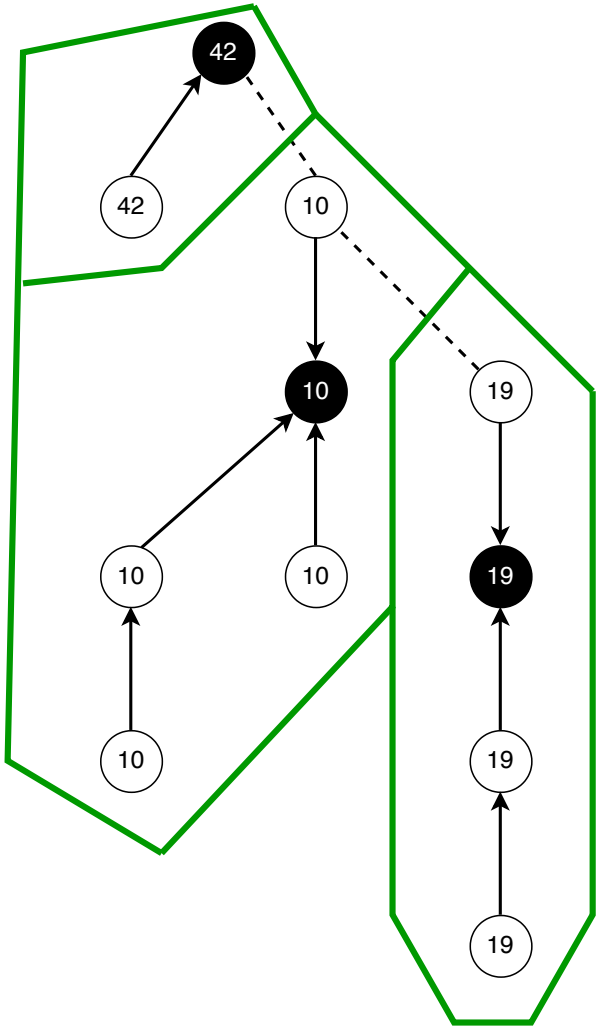
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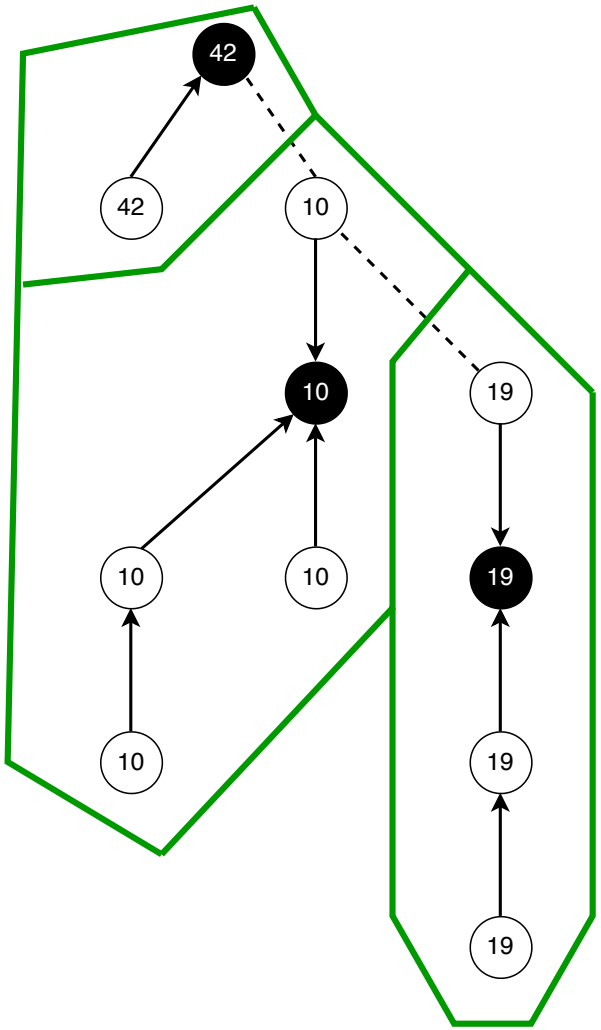
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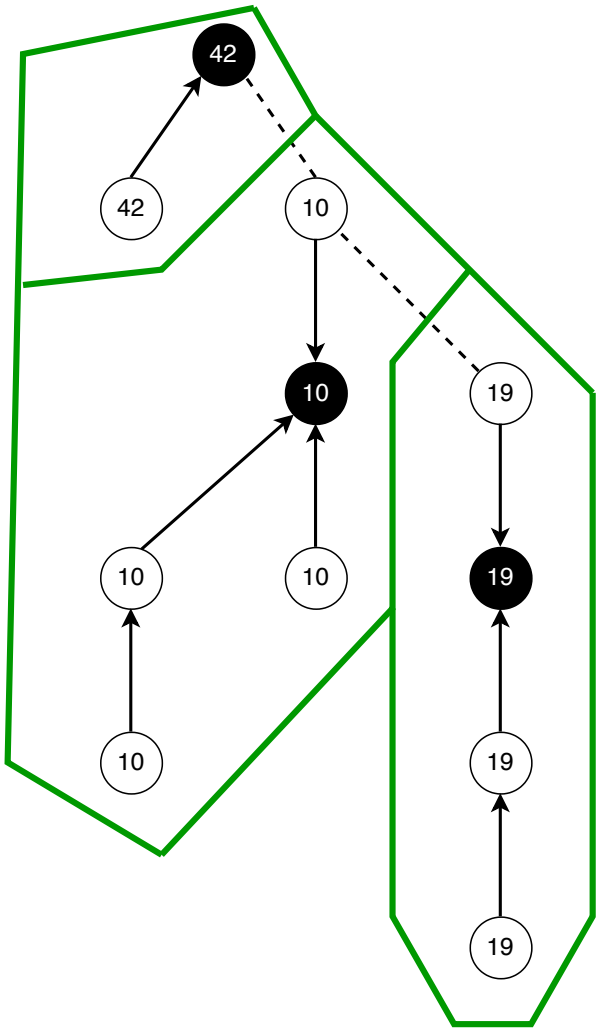


Example with $k = 2$



#clusters $= 3 \leq \lceil \frac{11}{3} \rceil = 4$

Example with $k = 2$



Remark: the spanning tree can be used for inter-cluster routing

Topologies

- Optimal in trees [1]
- Arbitrary Connected Networks: at most $\lceil \frac{n}{k+1} \rceil$ clusters
- Unit Disk Graph (UDG): $7.2552k + O(1)$-approximation of the optimal
- Quasi Unit Disk Graph (QUDG): $7.2552\lambda^2 k + O(\lambda)$-approximation of the optimal

UDG and QUDG are models for Wireless Sensor Networks Topologies

Distributed Computation

In Arbitrary Connected Networks

1. **Leader Election**

   ($O(mn)$ messages, $O(m)$ rounds, and $O(\delta_u + B)$ bits for messages and nodes, where B the number of bits to store an identifier)

2. **BFS Spanning Tree Construction** (with initialization and termination detection at the leader)

   ($O(D)$ rounds, $O(n.m)$ messages of $O(\log D)$ bits, and $O(\log D + \log \Delta)$ bits per node)

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   Feedback $\Rightarrow$ Bottom-up computation of α + Parent pointer assignment

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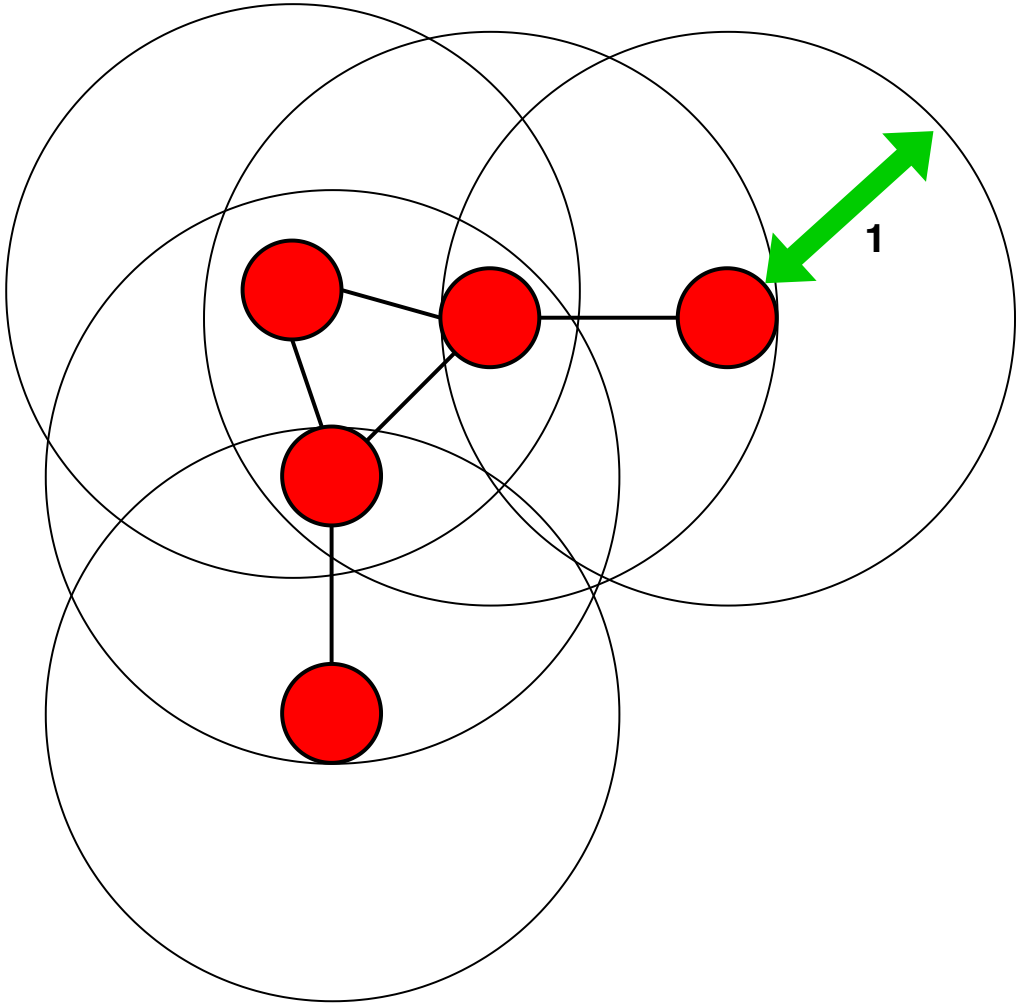
4. Propagation of Information with Feedback in the tree:
   Broadcast + Feedback $\Rightarrow$ cluster coloring,
   Feedback $\Rightarrow$ Termination Detection at the Leader

   ($O(D)$ rounds, $O(n)$ messages of $O(\log k + \log B)$ bits, and $O(\log B)$ bits per node)

5. Routing Set-Up

UDG: Unit Disk Graphs

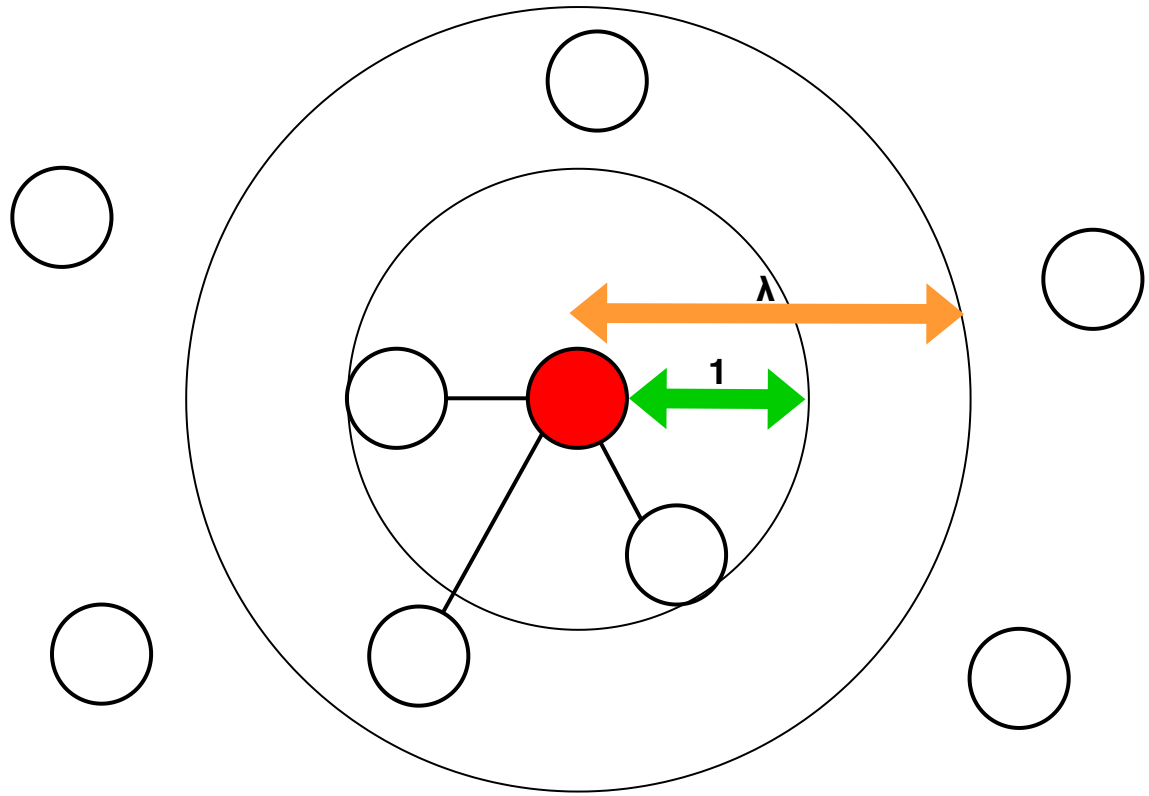
Nodes u and v are neighbors $\equiv \|u, v\| \leq 1$



QUDG: Quasi Unit Disk Graphs

Let $\lambda \geq 1$

- $\|u, v\| \leq 1 \Rightarrow u$ and v are neighbors
- u and v are neighbors $\Rightarrow \|u, v\| \leq \lambda$



Approximation of the optimal in UDGs

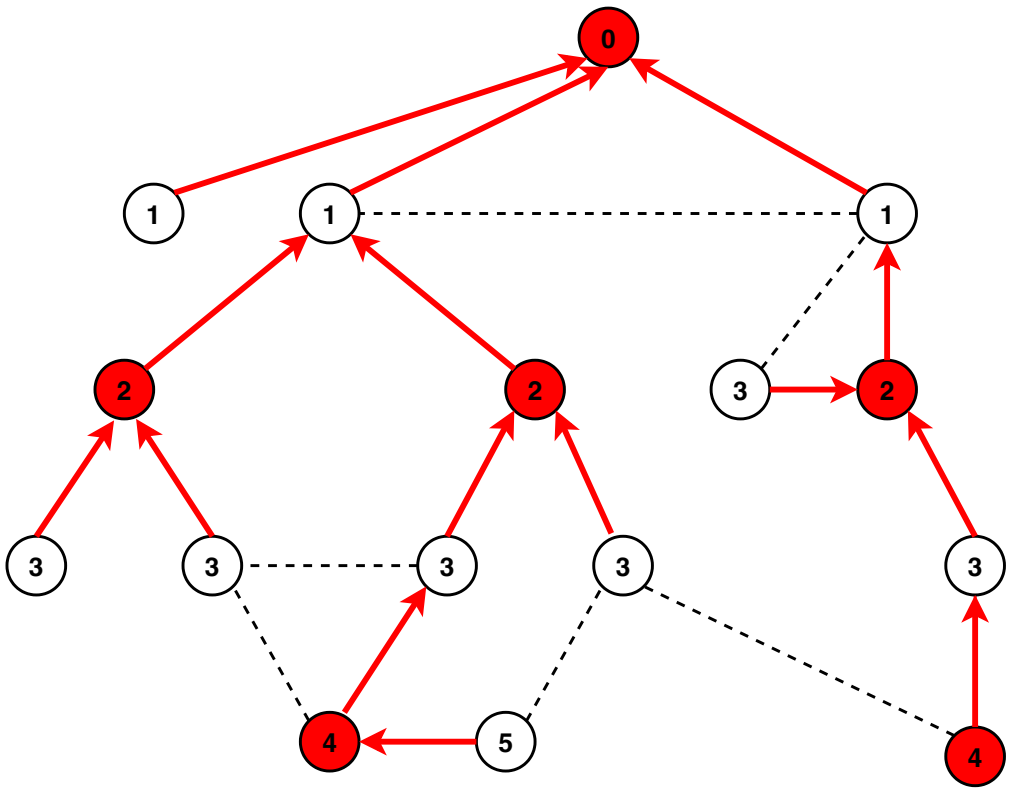
(at most M times the size of the optimal one)

Computing α on a *MIS* Tree of $G = (V, E)$

A spanning tree of G whose nodes at even level form a **maximal independent set**

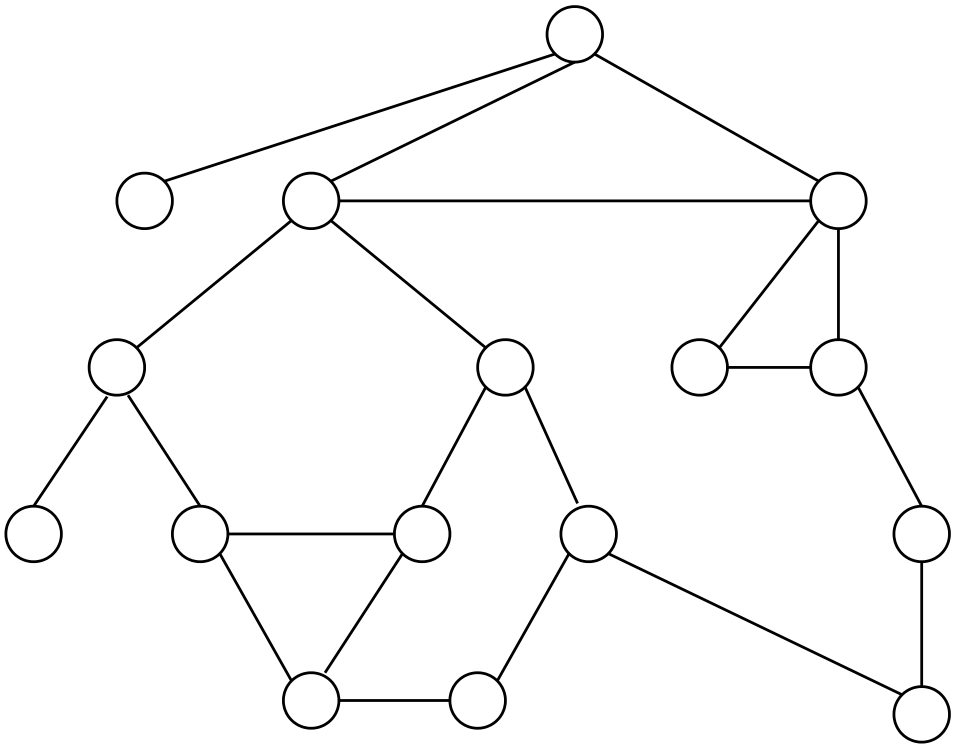
$S \subseteq V$ is a maximal independent set if

- S is independent: no two distinct nodes of S are neighbors in G
- S is maximal (by inclusion): no proper superset of S is independent



Approximation of the optimal in UDGs

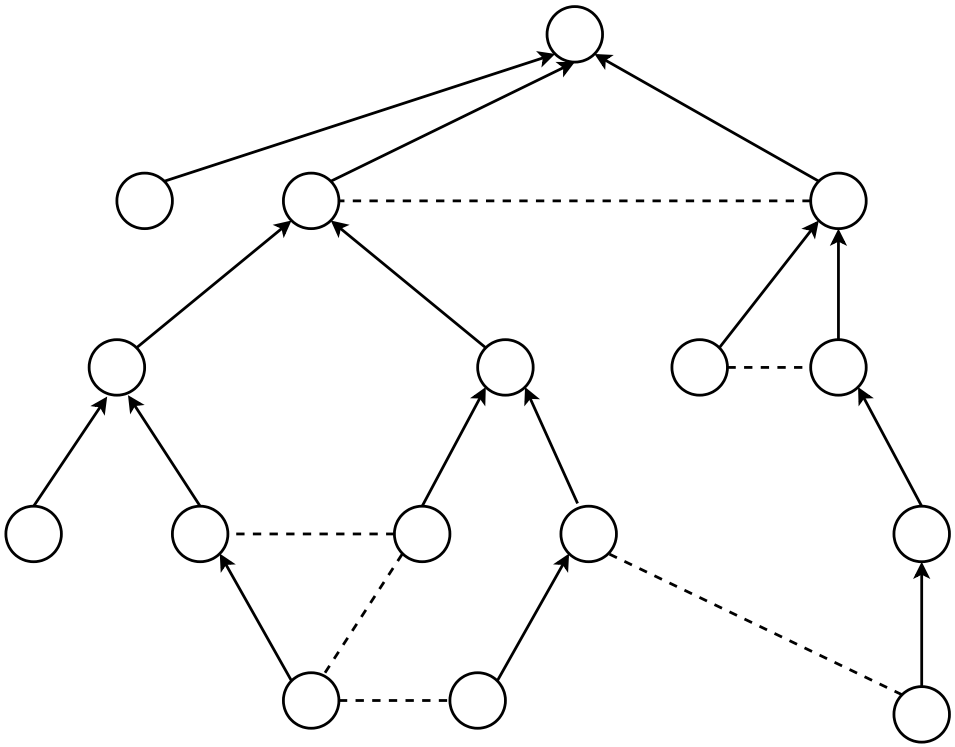
Compute a *MIS* Tree



A network

Approximation of the optimal in UDGs

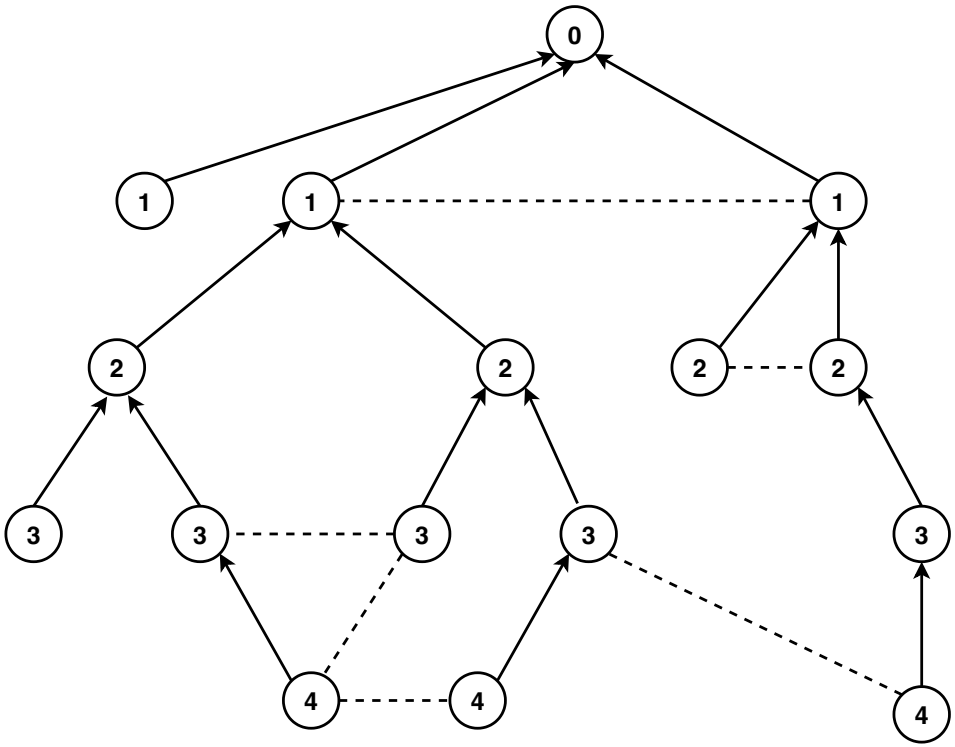
Compute a *MIS* Tree



BFS Spanning Tree

Approximation of the optimal in UDGs

Compute a *MIS* Tree



BFS Spanning Tree with levels

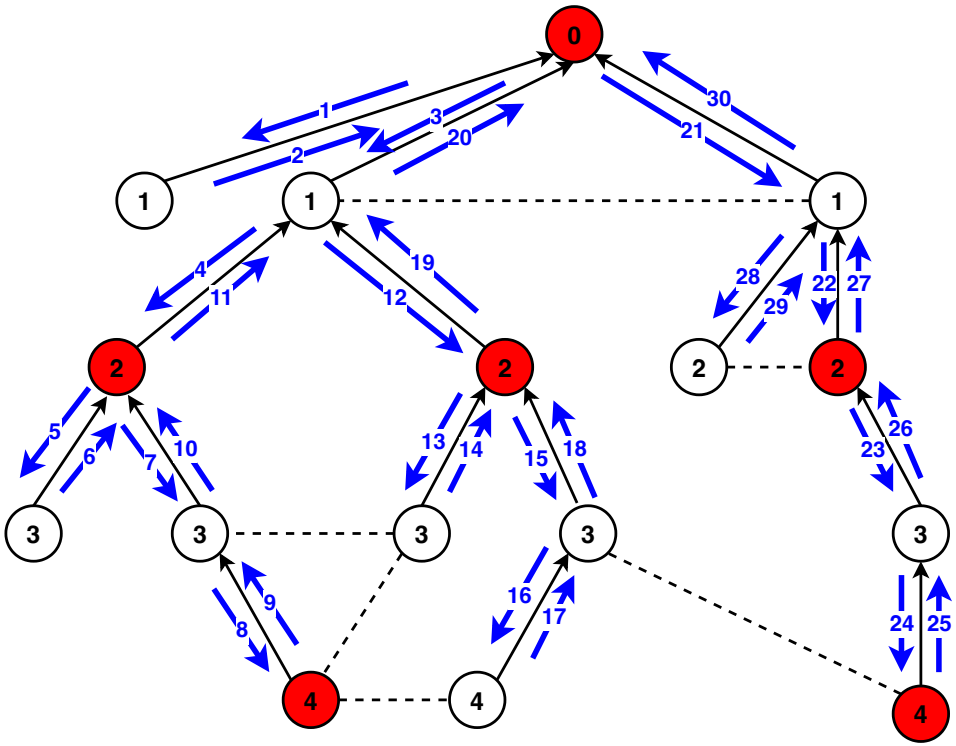
Approximation of the optimal in UDGs

Compute a *MIS* Tree

- $p.status \in \{In, Out\}$, initially *Out*
- DFS traversal of the spanning tree

At the 1st visit of node p:

if every neighbor q of p satisfies $q.status = Out$ **then**
    $p.status \leftarrow In$
else
    $p.status \leftarrow Out$
end if



BFS Spanning Tree with *MIS*

Approximation of the optimal in UDGs

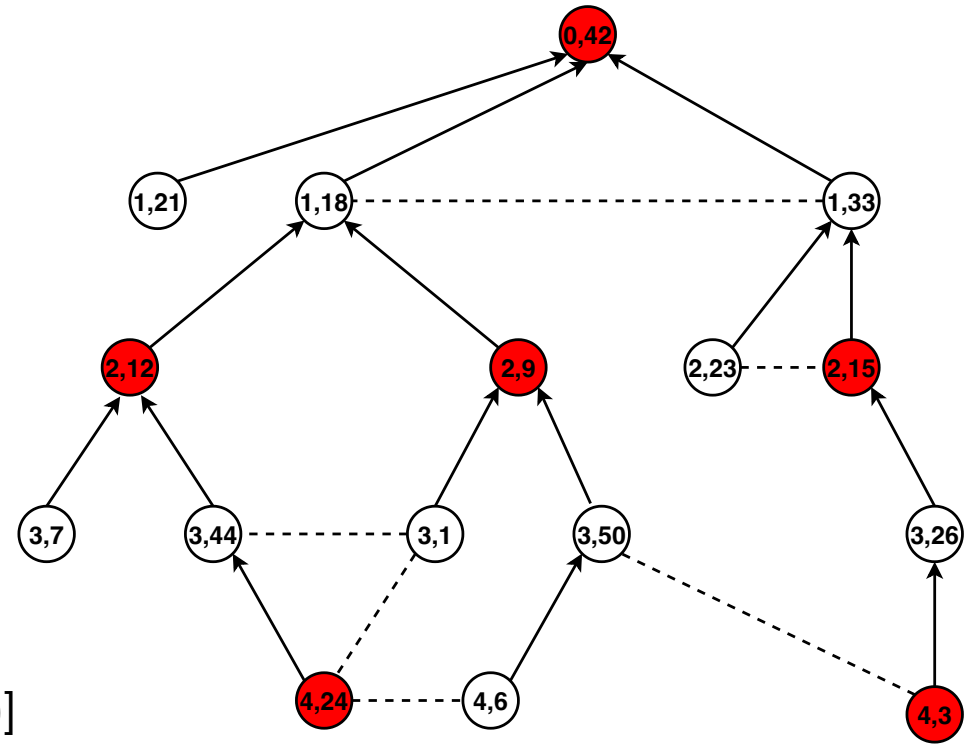
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- **Key:** pair level/identifier
- **Total order on keys:**
  $(a, b) \prec (c, d) \equiv [a < c \vee (a = c \wedge b < d)]$



BFS Spanning Tree and *MIS* with keys

Approximation of the optimal in UDGs

Compute a *MIS* Tree

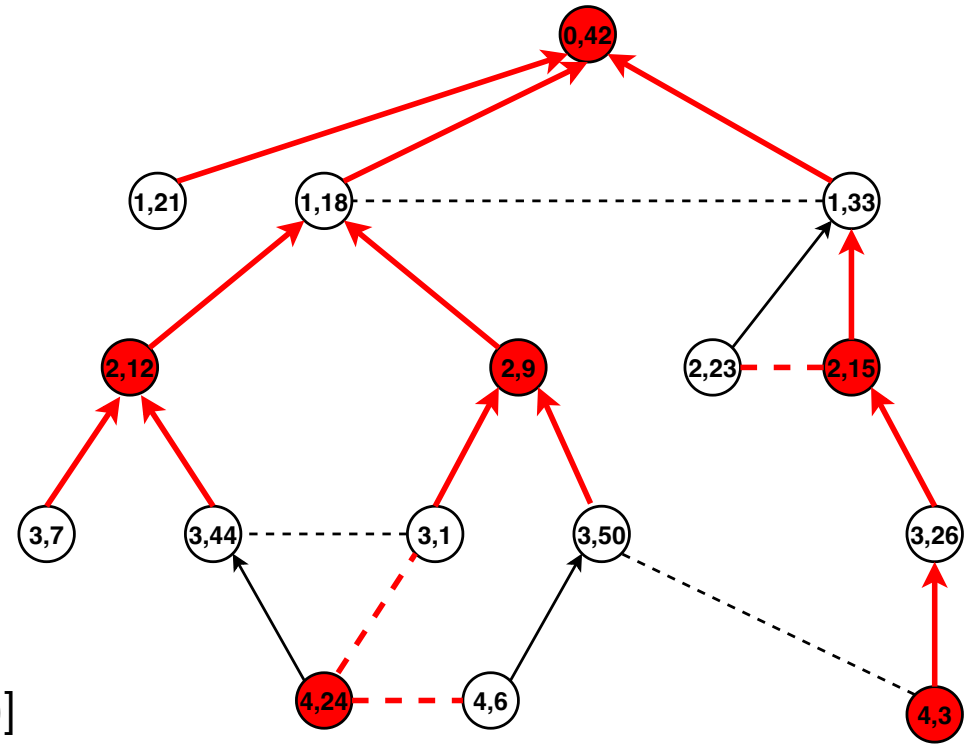
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- **Parent of non-root nodes:** the neighbor with a different status of smallest key



MIS tree

Approximation of the optimal in UDGs

Compute a *MIS* Tree

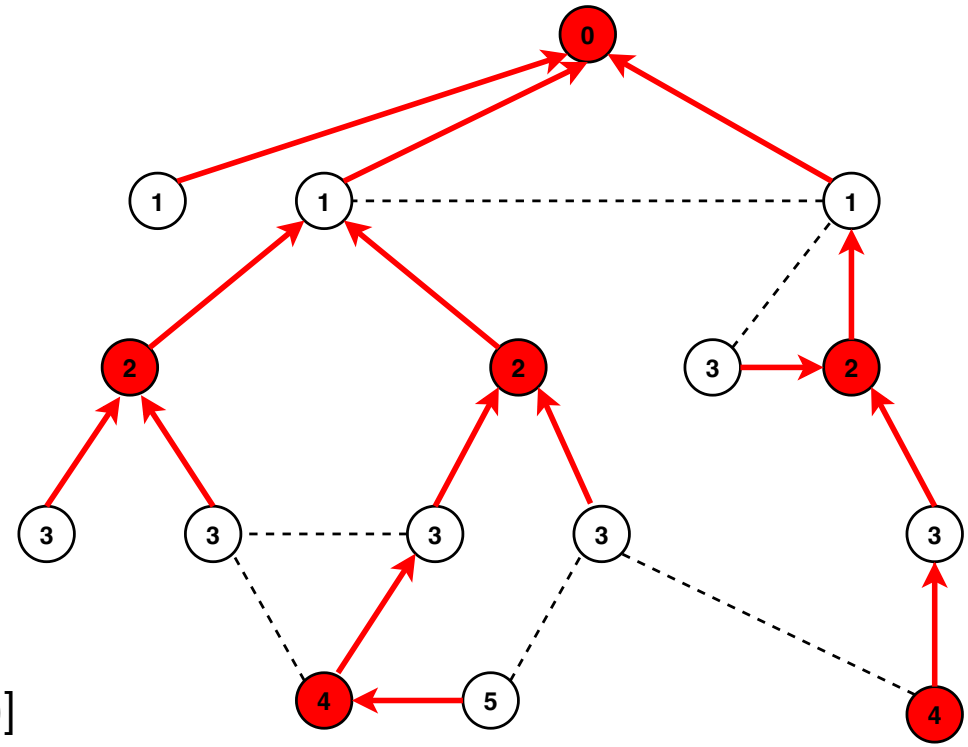
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MIS tree with levels

Approximation of the optimal in UDGs

Distributed computation of a *MIS* Tree

1. Leader Election

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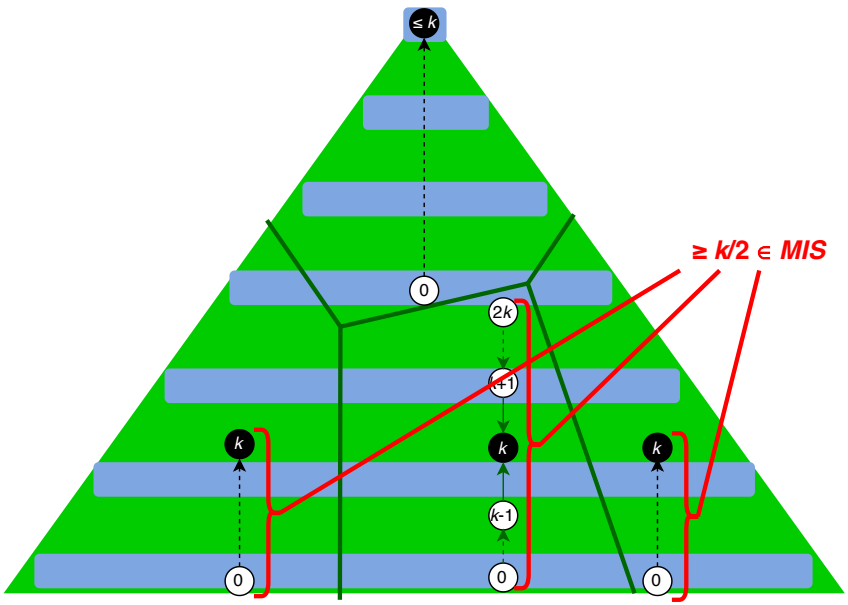
   ($O(D)$ rounds, $O(m)$ messages of $O(\log k + \log B)$ bits, and $O(\Delta(\log k + \log B))$ bits per node)

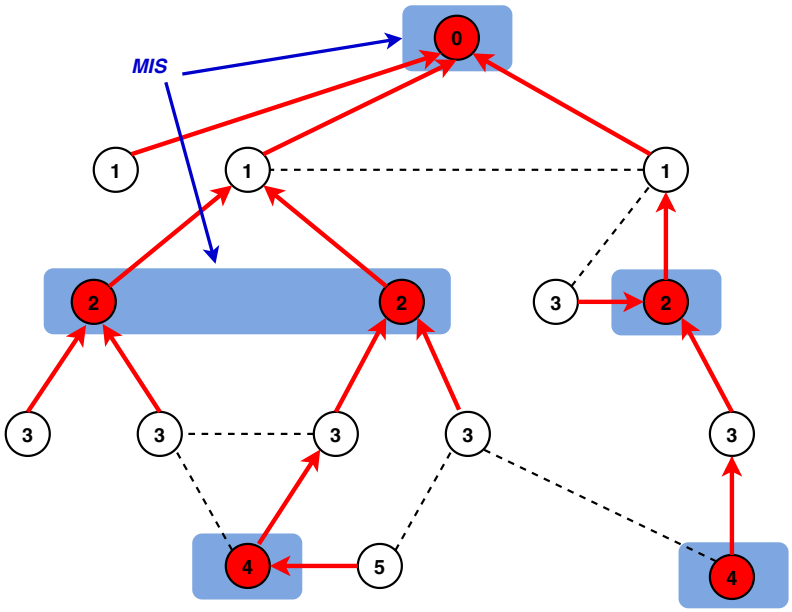
5. Clustering computation on the *MIS* tree

Approximation of the optimal in UDGs

k-clustering *vs.* *MIS*

Clr: partition into clusters



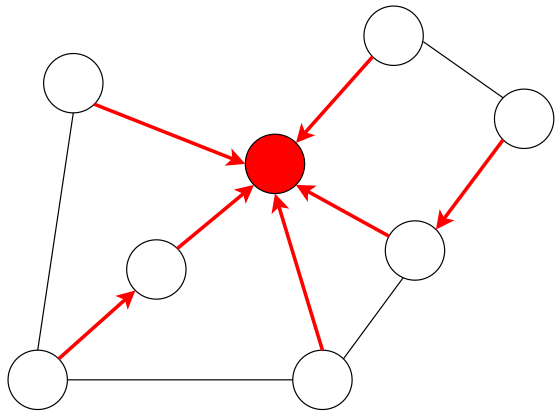


$$(|Clr| - 1).\frac{k}{2} \leq |MIS| - 1$$

Approximation of the optimal in UDGs

Independent Set *vs.* Optimal k-clustering Clr_{opt}

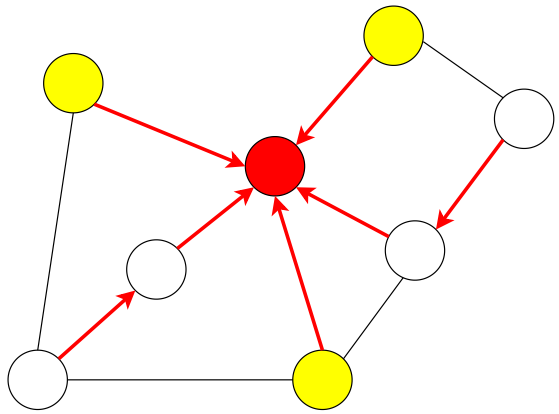
- Let C be any cluster of Clr_{opt}



Approximation of the optimal in UDGs

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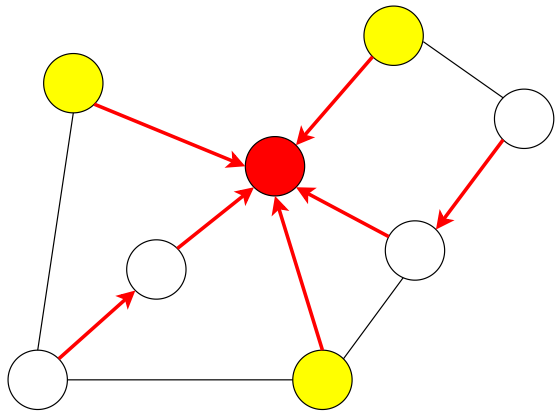
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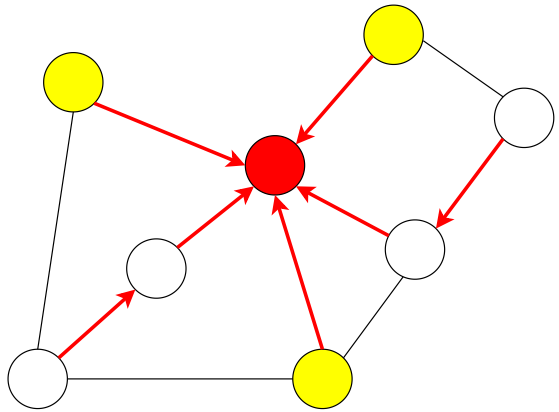
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- **UDG:**
  $\forall \boldsymbol{p}, \boldsymbol{q} \in \boldsymbol{I}, \boldsymbol{p} \neq \boldsymbol{q} \Rightarrow \|\boldsymbol{p}, \boldsymbol{q}\| > \mathbf{1}$



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Theorem (Folkman & Graham, [4])

Let X be a compact convex region. Let $Y \subseteq X$ s.t. $\forall p, q \in Y, (p \neq q \Rightarrow \|p, q\| \geq 1)$.

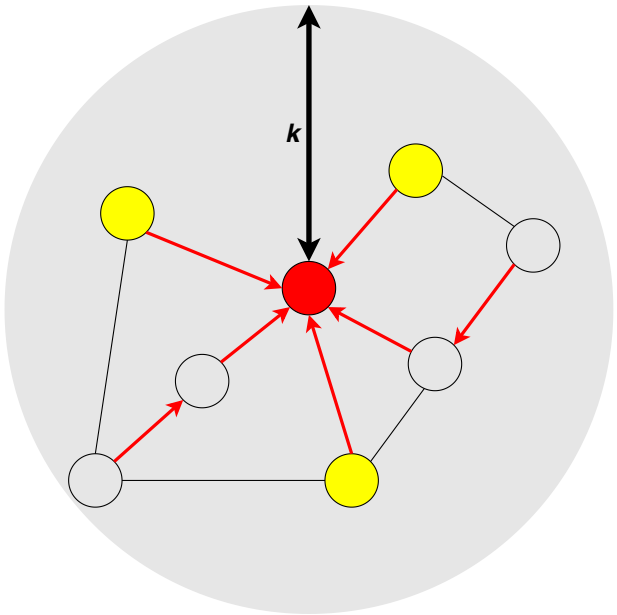
$$|Y| \leq \lfloor \frac{2.A(X)}{\sqrt{3}} + \frac{P(X)}{2} + 1 \rfloor$$

where $A(X)$ and $P(X)$ are respectively the area and perimeter of X.

Approximation of the optimal in UDGs

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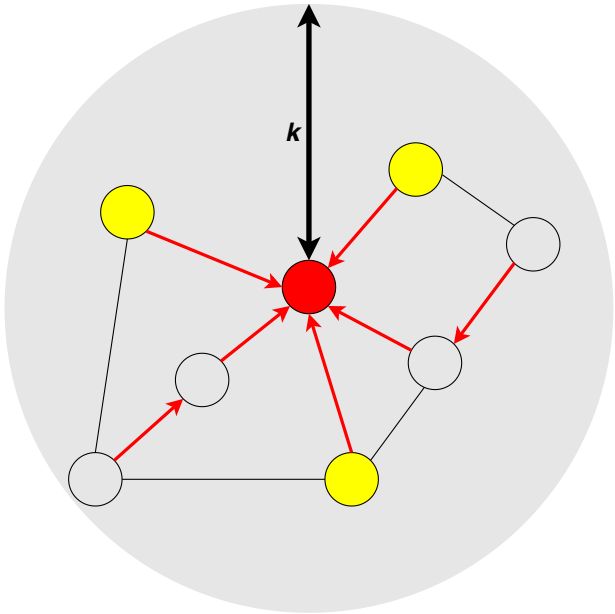
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$A(X) = \pi.k^2$, $P(X) = 2\pi.k$

- $|I \cap C| \leq \lfloor \frac{2.\pi.k^2}{\sqrt{3}} + \pi.k + 1 \rfloor$
- $|I| \leq \lfloor \frac{2.\pi.k^2}{\sqrt{3}} + \pi.k + 1 \rfloor . |Clr_{opt}|$

Result

- $|MIS| \leq \lfloor \frac{2.\pi.k^2}{\sqrt{3}} + \pi.k + 1 \rfloor . |Clr_{opt}|$
- $(|Clr| - 1).\frac{k}{2} \leq |MIS| - 1$

$\Rightarrow |Clr| \leq 1 - \frac{2}{k} + (\frac{4\pi.k}{\sqrt{3}} + 2\pi + \frac{2}{k}).|Clr_{opt}|$

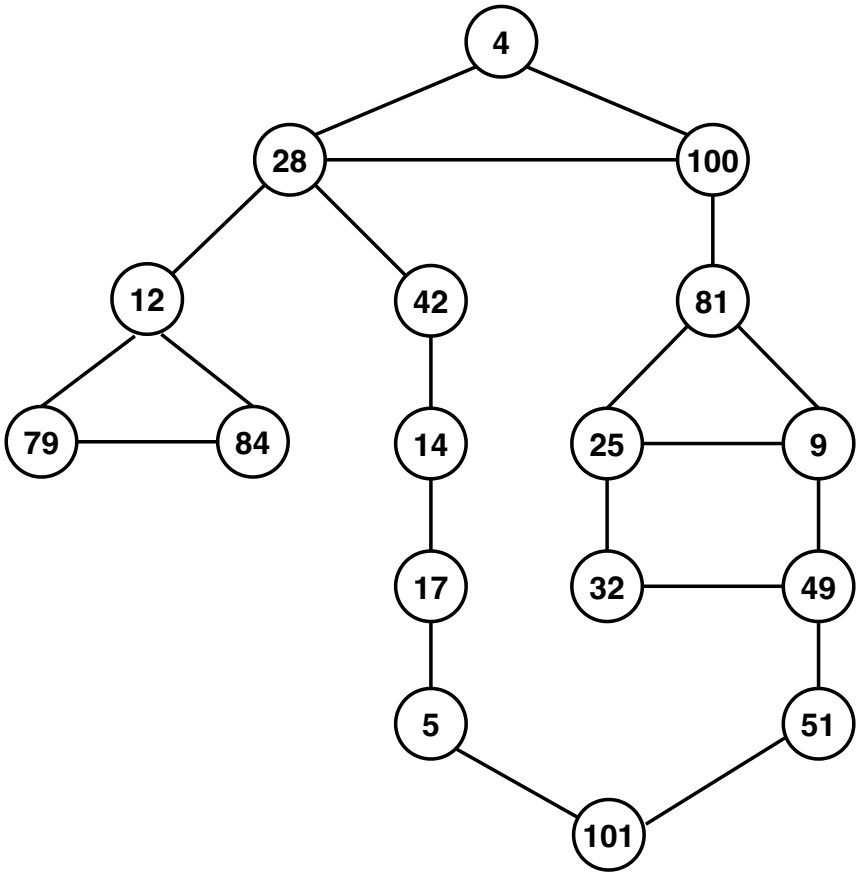
$$\Rightarrow \mathbf{7,2552k + O(1) - competitive\ in\ UDG}$$

Generalization to QUDG (disc of radius $k\lambda$):

$$\mathbf{7,2552\lambda^2 k + O(1) - competitive\ in\ QUDG}$$

Exercise

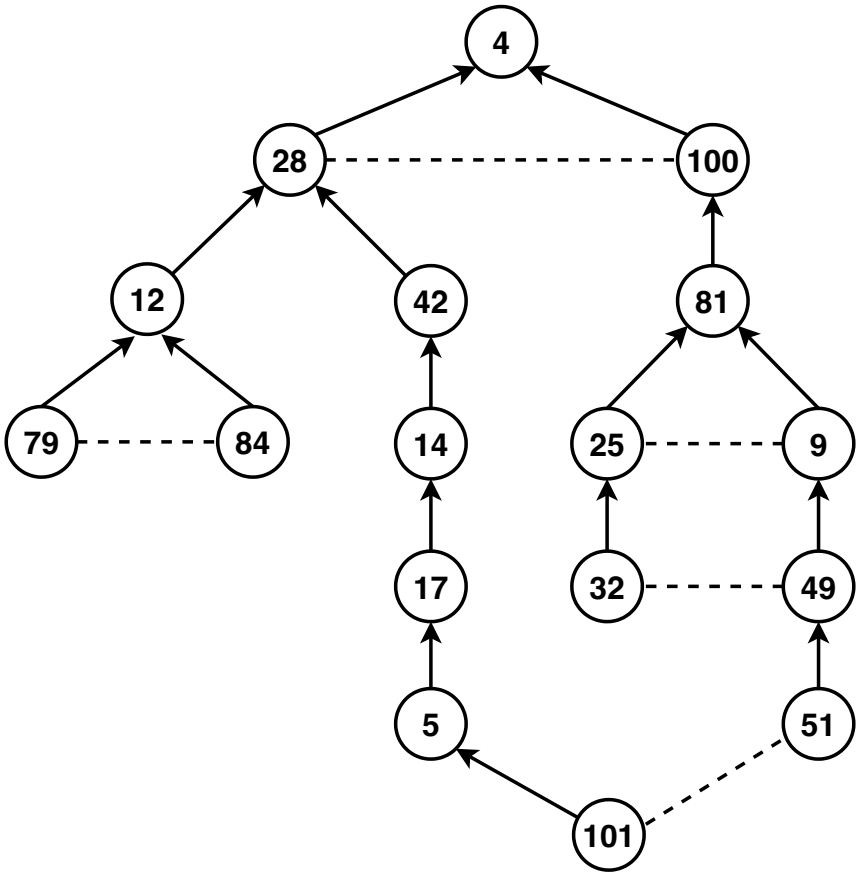
Compute a clustering using the algorithm of [1] **with a *MIS* tree** and $k = 2$



A network

Exercise

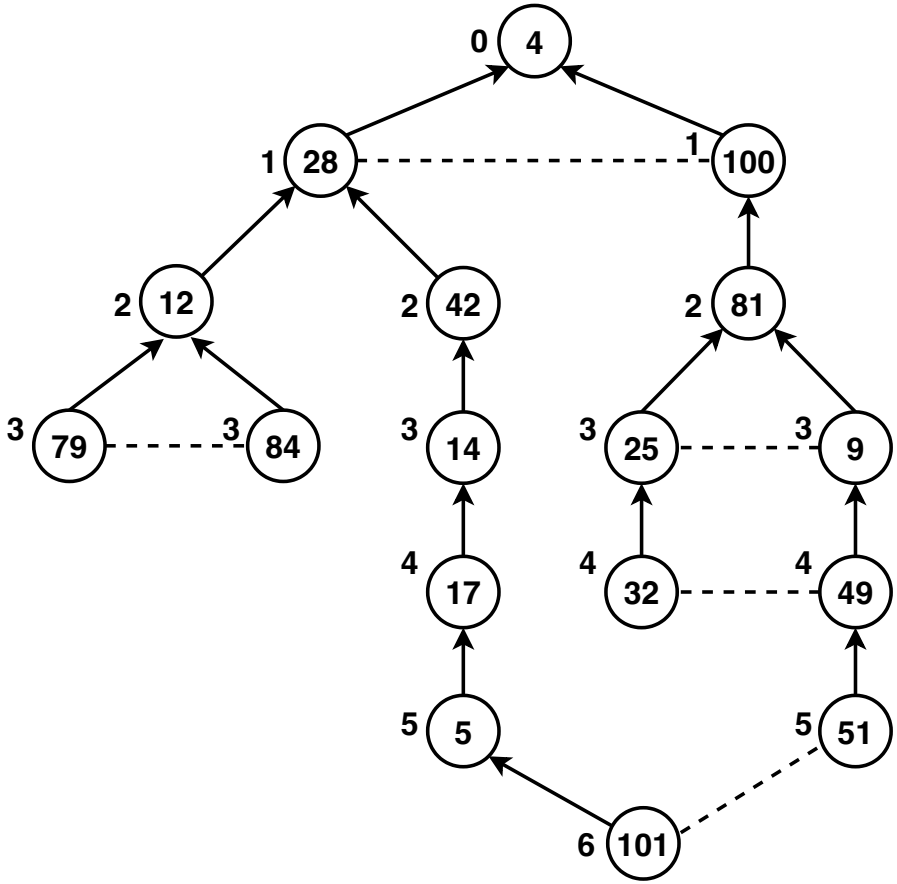
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BFS spanning tree

Exercise

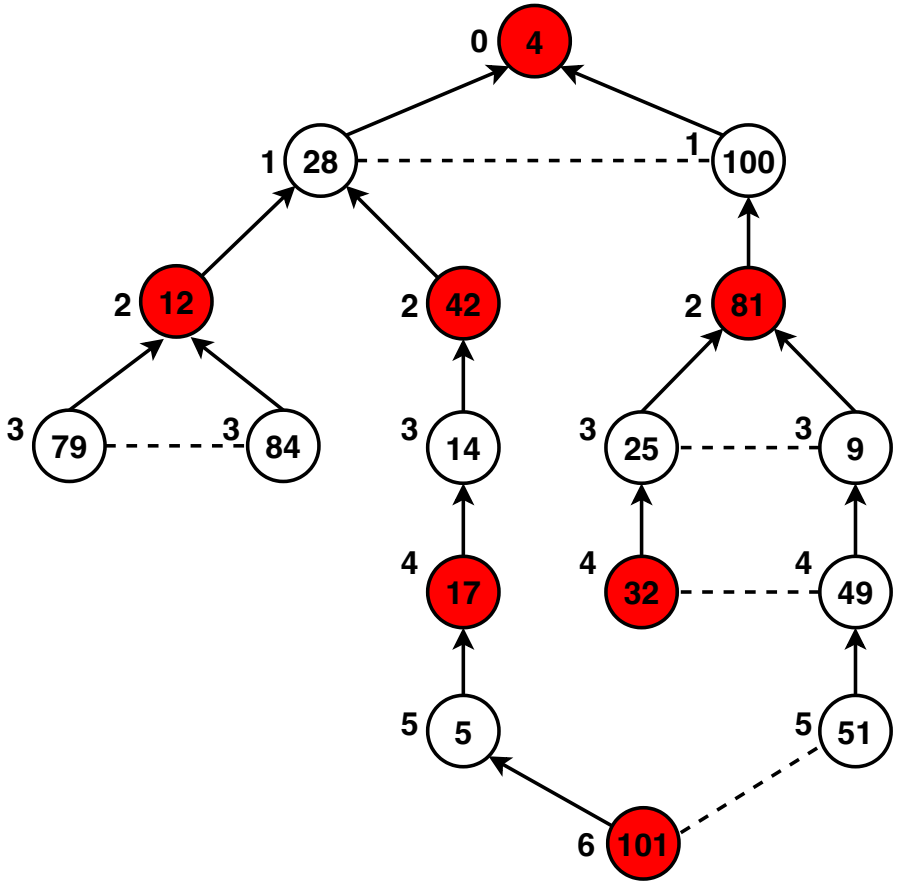
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BFS spanning tree with levels

Exercise

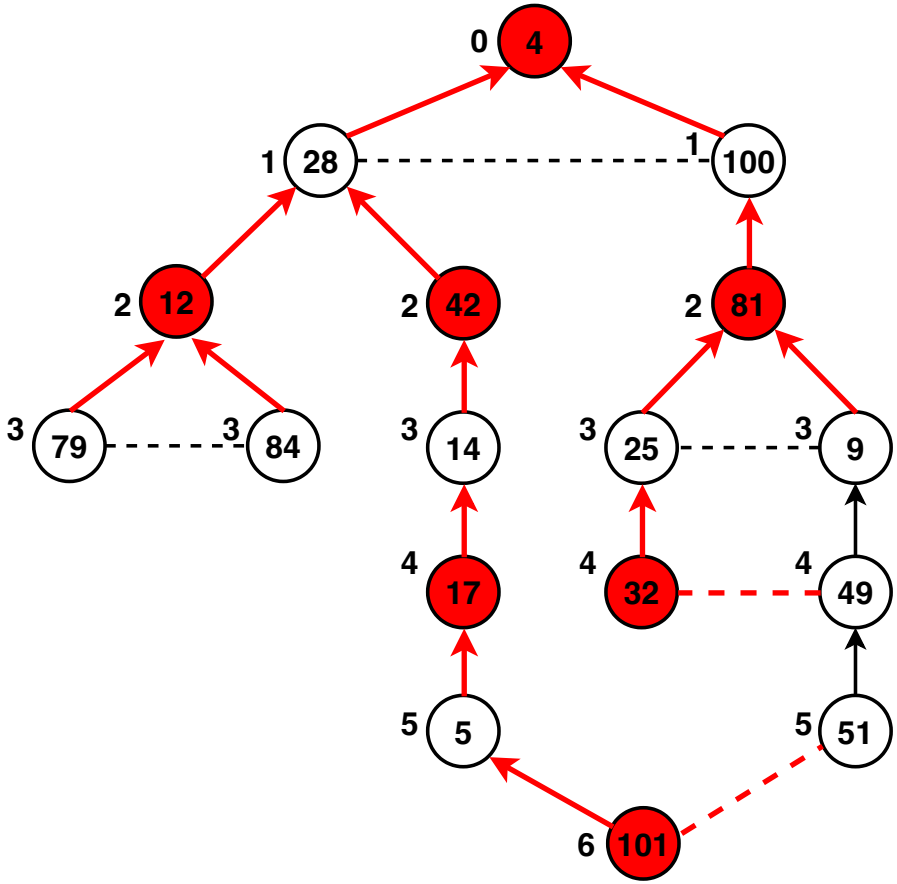
Compute a clustering using the algorithm of [1] **with a *MIS* tree** and $k = 2$



BFS spanning tree with levels + Independent Set

Exercise

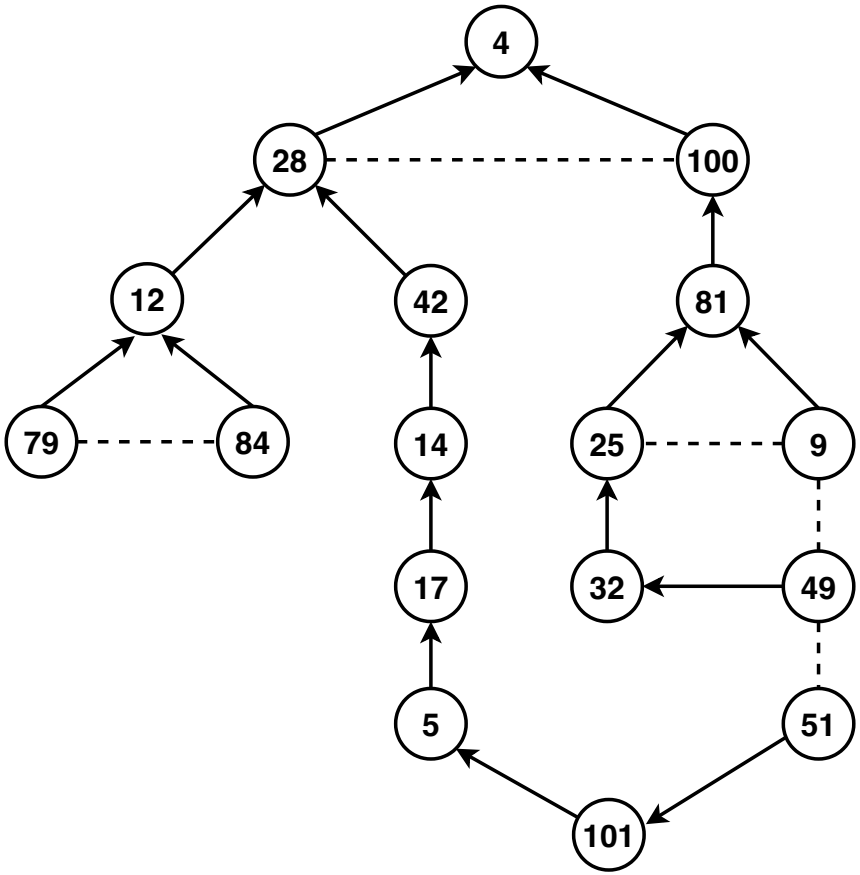
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BFS spanning tree with levels + Independent Set + *MIS* tree

Exercise

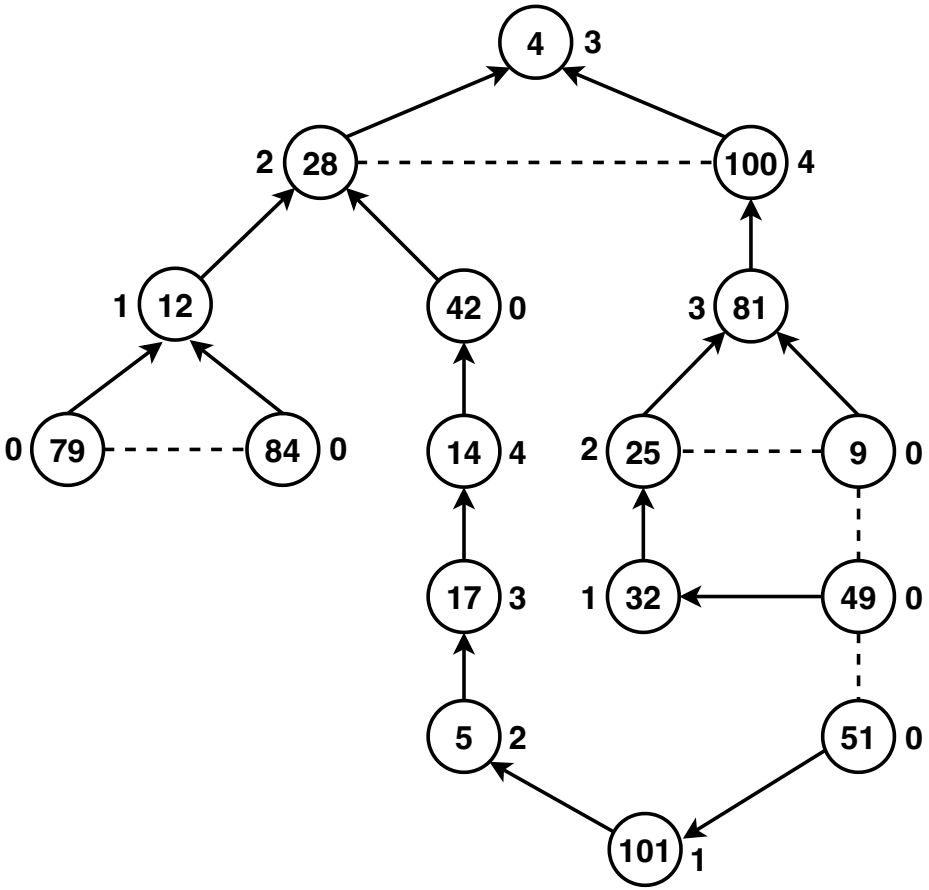
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MIS tree

Exercise

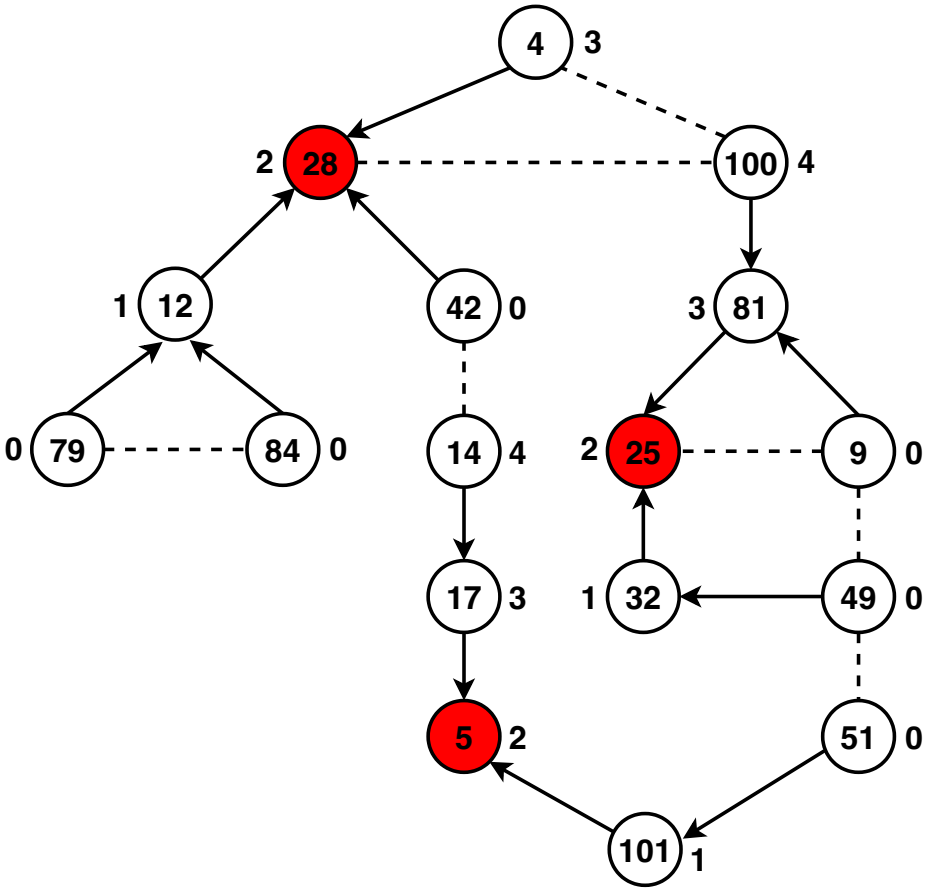
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α-computation

Exercise

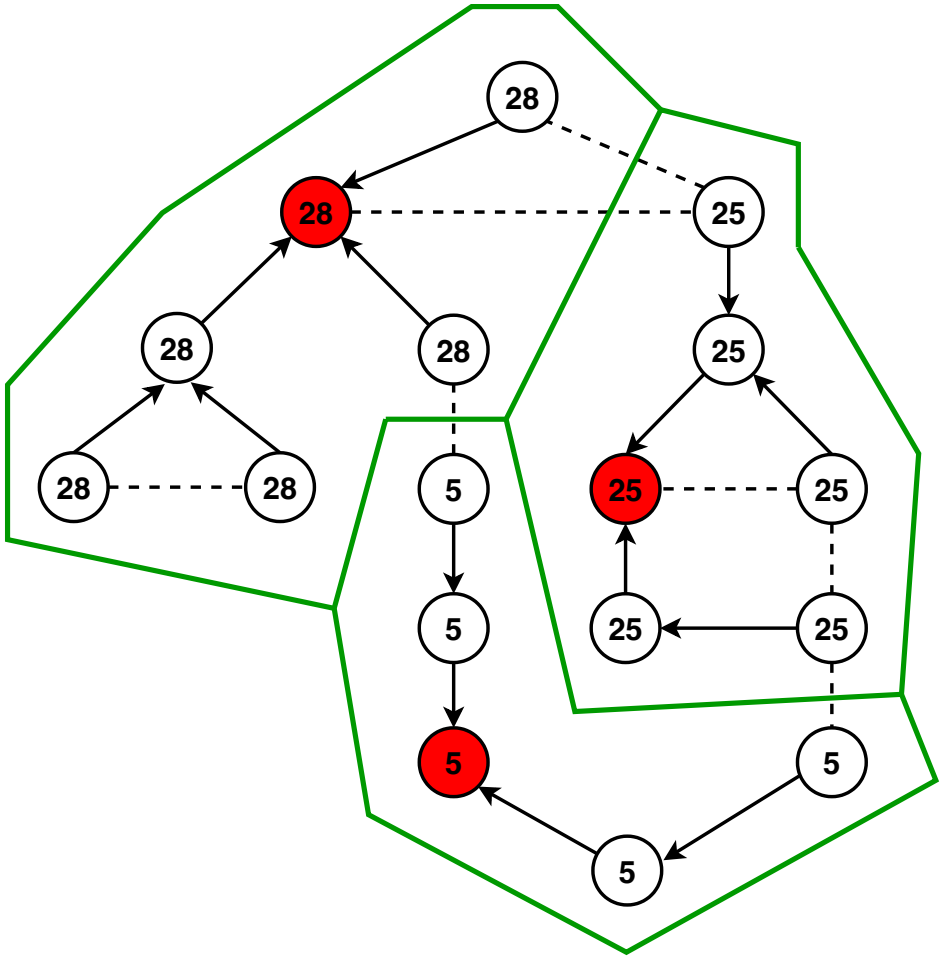
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Spanning forest

Exercise

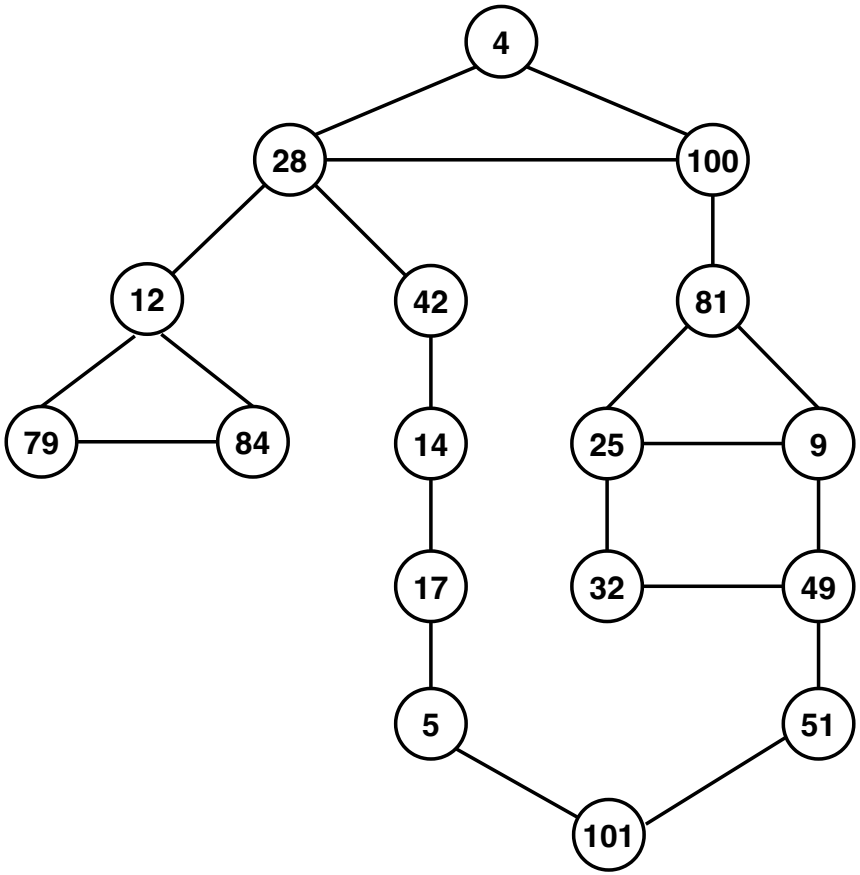
Compute a clustering using the algorithm of [1] **with a *MIS* tree** and $k = 2$



Clustering, **#clusters** $= 3 \leq \lceil \frac{17}{3} \rceil = 6$

Exercise

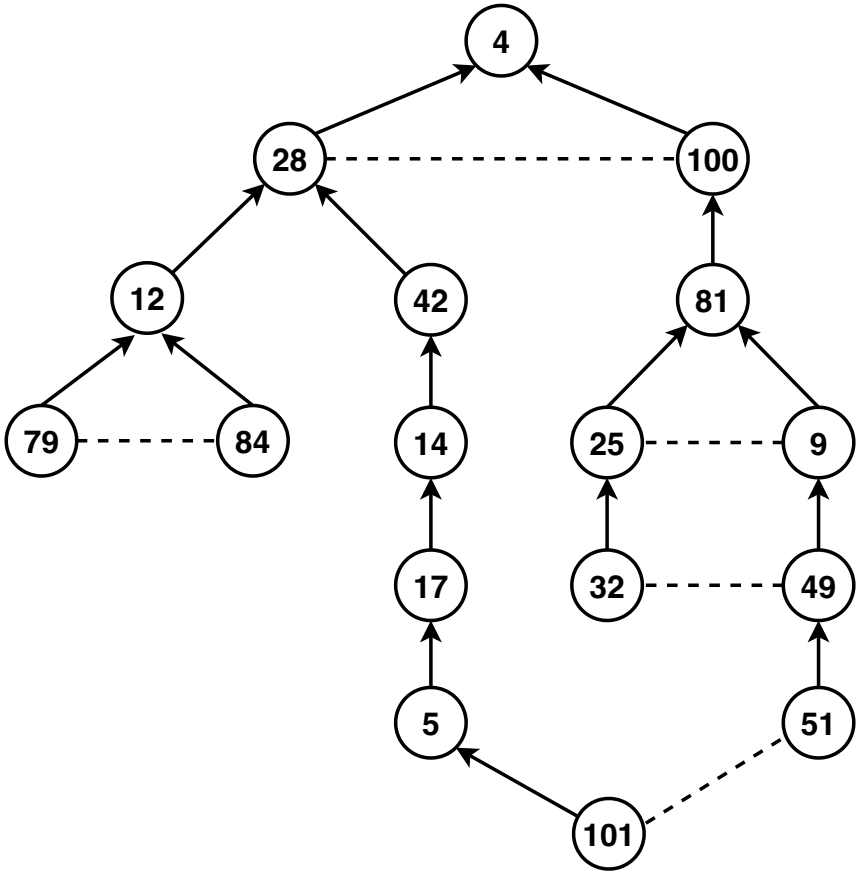
Compute a clustering using the algorithm of [1] **without a *MIS* tree** and with $k = 2$



A network

Exercise

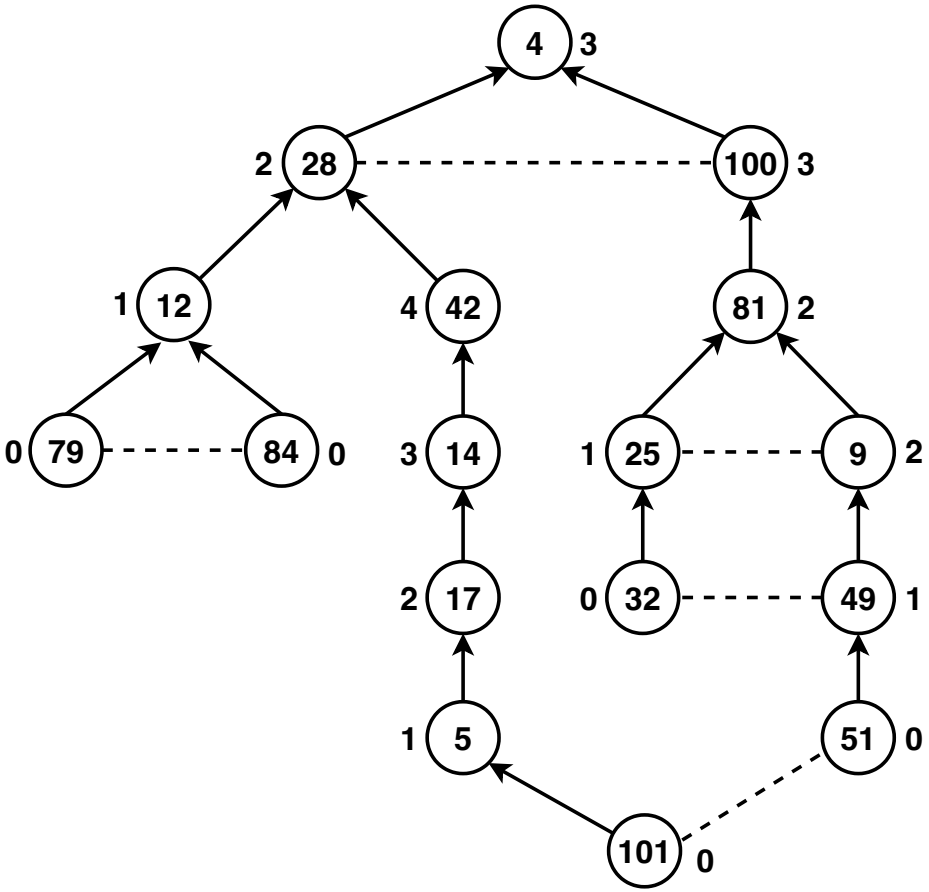
Compute a clustering using the algorithm of [1] **without a *MIS* tree** and with $k = 2$



(BFS) spanning tree

Exercise

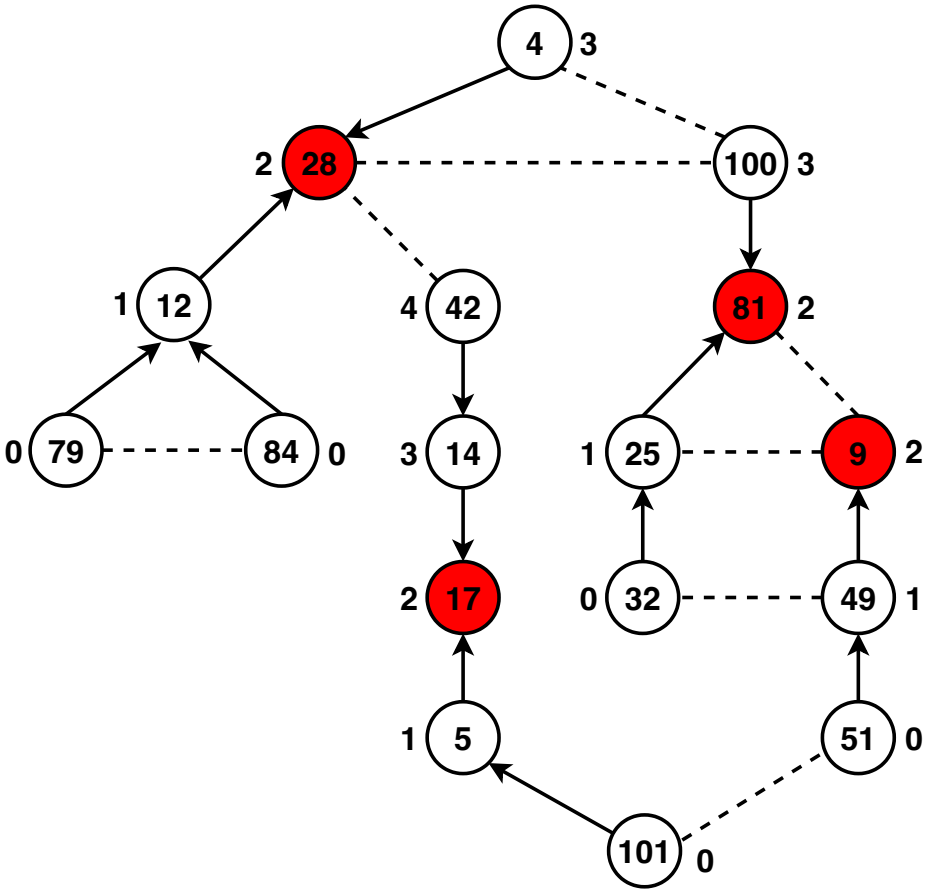
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α-computation

Exercise

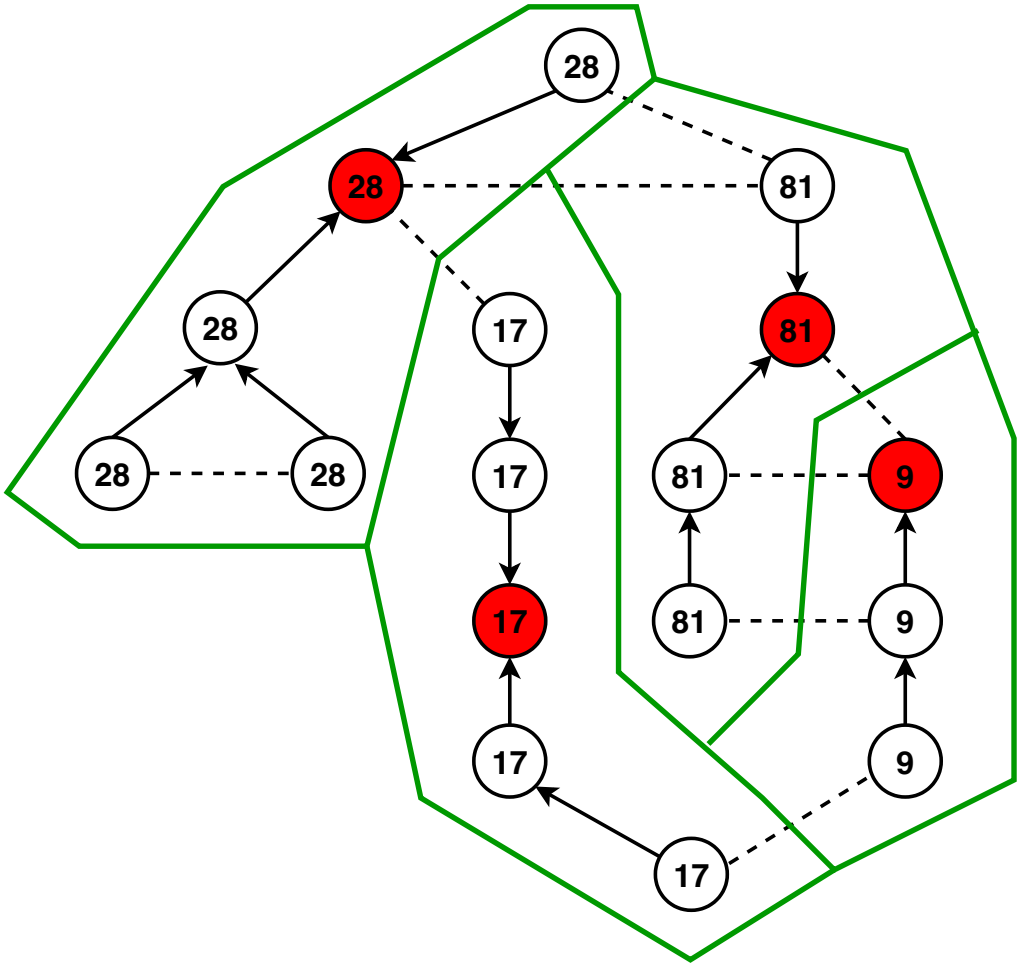
Compute a clustering using the algorithm of [1] **without a *MIS* tree** and with $k = 2$



Spanning forest

Exercise

Compute a clustering using the algorithm of [1] **without a *MIS* tree** and with $k = 2$



Clustering, **#clusters** $= 4 \leq \lceil \frac{17}{3} \rceil = 6$

Roadmap

1. Introduction
2. Examples of clustering
   - Clustering of [3]
   - Clustering of [1]
3. Routing in a Clustering
4. References

Routing from p to q

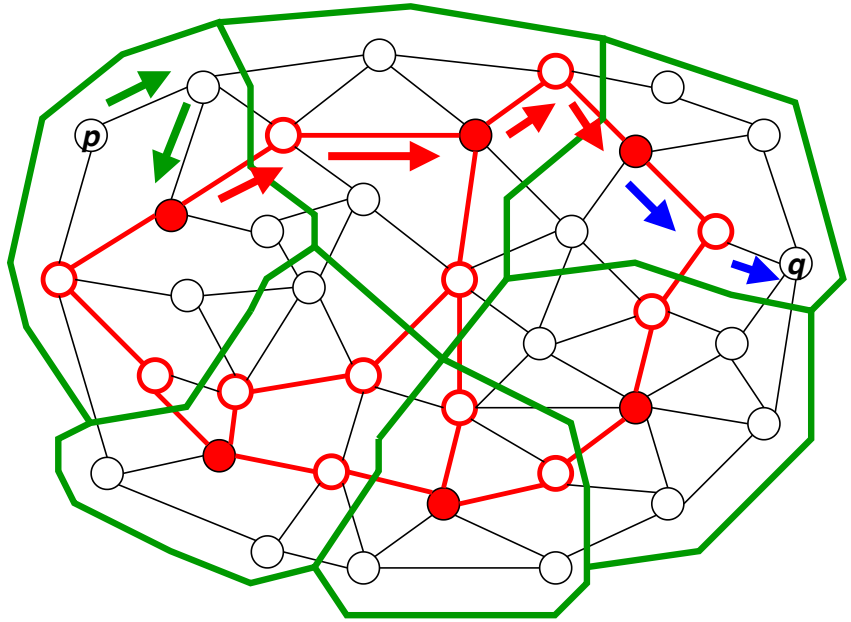
Address: (label of q's clusterhead,label of q)

Phases: 3 phases distinguished using **3 colors**.

1. Route the packet from p to its clusterhead (intra-clustering)
2. Route the packet from the clusterhead of p to the clusterhead of q (inter-clustering)
3. Route the packet from the clusterhead of q to q (intra-clustering)

Different routing methods can be used for each phase, *e.g.*,

1. Parent links to route from p to its clusterhead
2. Shortest paths[1] to route from the clusterhead of p to the clusterhead of q
3. Interval routing from the clusterhead of q to q



[1] in the network or in the tree computed during the clustering computation

Routing from p to q

Address: (label of q's clusterhead,label of q)

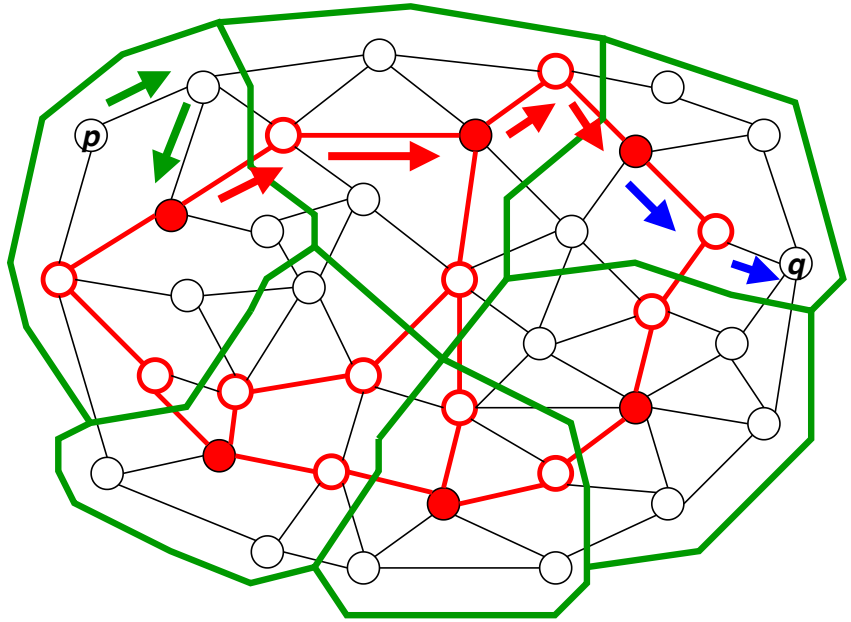
Phases: 3 phases distinguished using **3 colors**.

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Different routing methods can be used for each phase, *e.g.*,

1. Parent links to route from p to its clusterhead
2. Shortest paths[1] to route from the clusterhead of p to the clusterhead of q
3. Interval routing from the clusterhead of q to q

$\Rightarrow$ **Address: (ID of q's clusterhead,label of q in the ILS of its cluster)**



[1] in the network or in the tree computed during the clustering computation

Shortest Paths for inter-cluster routing

GLOBAL:

1. The leader launches a PIF to collect the identifiers of all clusterheads
2. The leader launches a PIF to broadcast the identifiers of all clusterheads
3. For each node p and each clusterhead q:
   - $Par_p[q] \leftarrow p$
   - if $p \neq q$, $D_p[q] \leftarrow \infty$, else $D_p[q] = 0$

LOCAL:

1. Each node p creates cells $Par_p[q]$ and $D_p[q]$ for its clusterhead q and

   informs its neighbors of the existence of q
2. $Par_p[q] \leftarrow p$ and

   if $p \neq q$, $D_p[q] \leftarrow \infty$, else $D_p[q] = 0$
3. If p learns the existence of some clusterhead r, p creates cells $Par_p[r]$ and $D_p[r]$, initialized to p and ∞ resp.

4. Bellman-Ford: for each node p, each clusterhead q, and every neighbor v of p
   - if $D_p[q] > D_v[q] + 1$, then $D_p[q] \leftarrow D_v[q] + 1$; $Par_p[q] \leftarrow v$

See [6] for further details.

Using the Spanning Tree for inter-cluster routing

Thank to the cluster colors, local detection of tree edge $\{p, q\}$ between two different clusters

1. Add $\{p, q\}$ to the subgraph connecting clusterheads
2. Add the cluster path from p to its clusterhead to the subgraph connecting clusterheads
3. Add the cluster path from q to its clusterhead to the subgraph connecting clusterheads

We obtain a tree

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We obtain a tree

Apply the tree labeling scheme or an PLS on the tree to allow routing between clusters

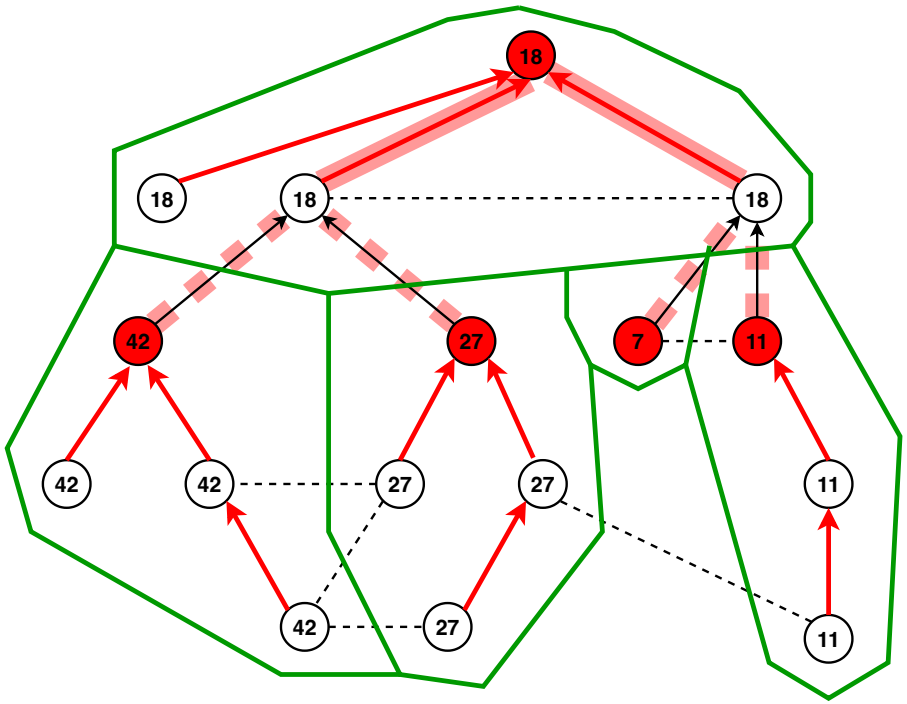
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Clustering of [3]

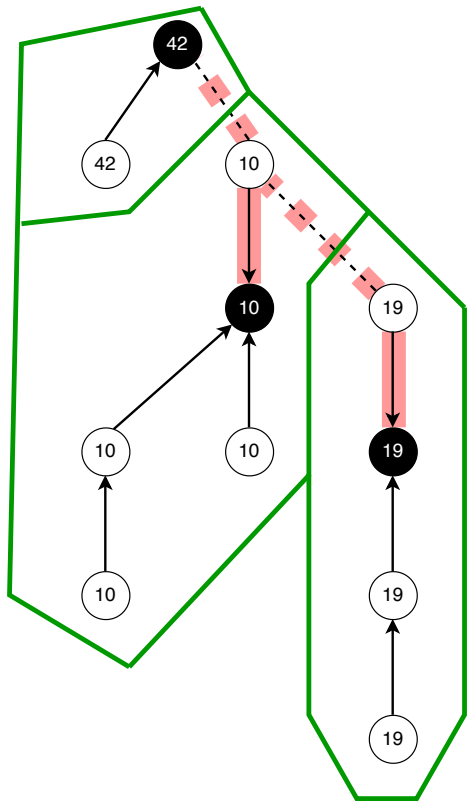
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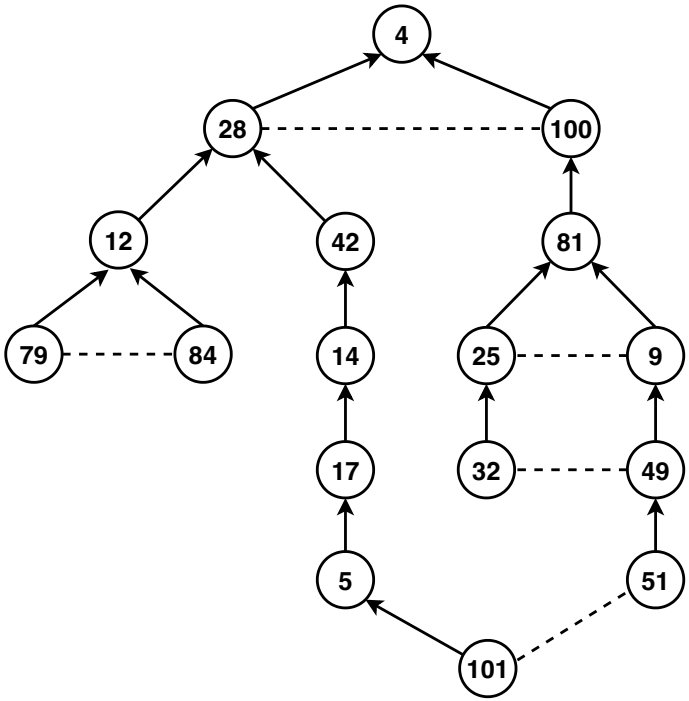
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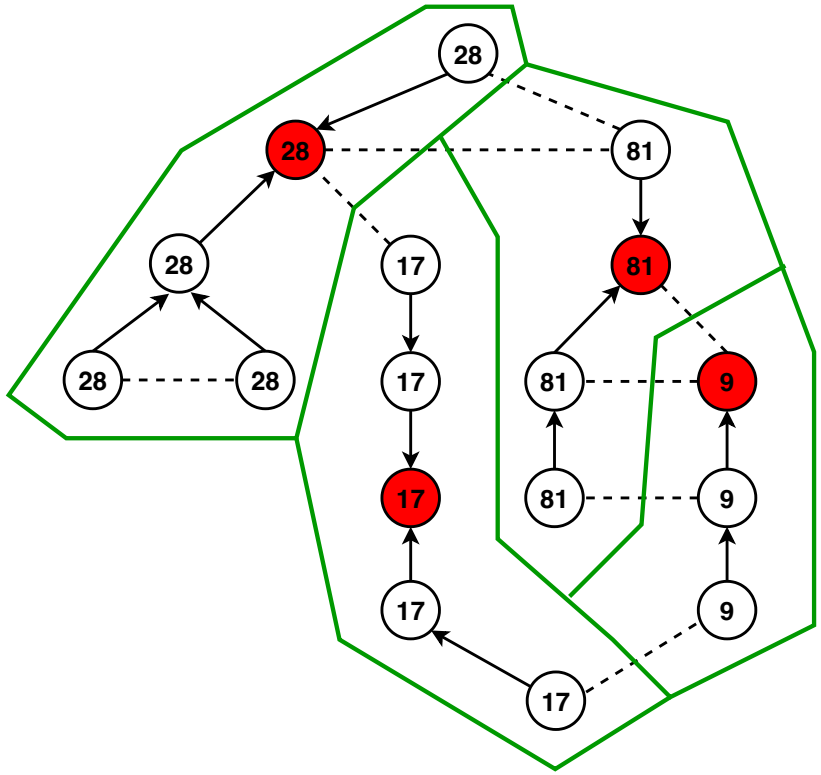


Clustering of [1]

Exercise: subgraph connecting clusterheads?

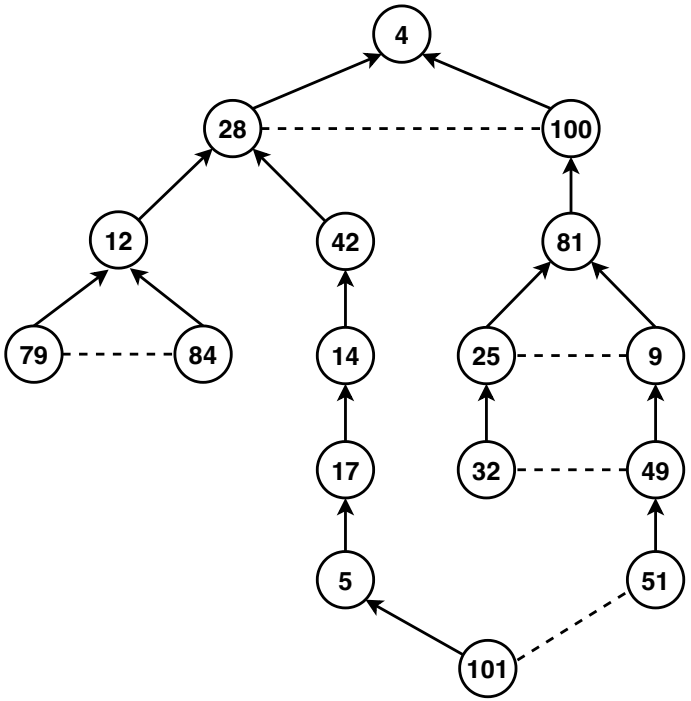


Spanning tree

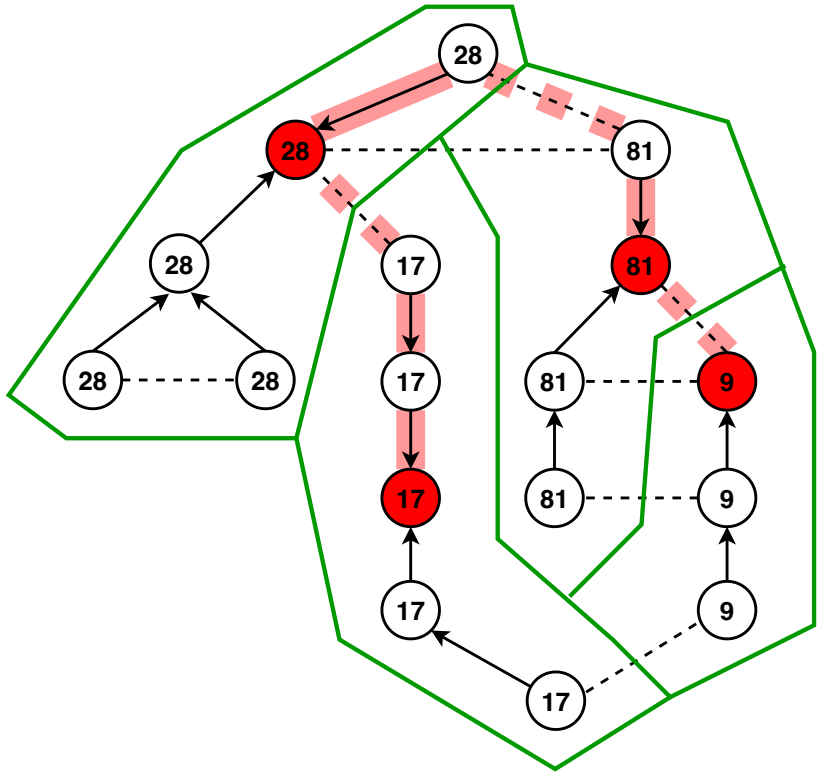


Clustering

Exercise: subgraph connecting clusterheads?



Spanning tree



Clustering

Subgraph connecting clusterheads

Optimal subgraph for the Inter-cluster Routing
Steiner Tree

Non-negative edge weights.

Subset of nodes: set of terminals

Compute a tree of minimum weight that

- contains all terminals (but may include additional vertices) and
- minimizes the total weight of its edges.

Here, terminals are clusterheads

However, $\mathcal{NP}$-hard [5]

But approximation algorithms exist!

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References

[1] A. K. Datta, S. Devismes, K. Heurtefeux, L. L. Larmore, and Y. Rivierre.
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Theor. Comput. Sci., 626:110–133, 2016.

[2] A. K. Datta, S. Devismes, and L. L. Larmore.
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[3] A. K. Datta, L. L. Larmore, S. Devismes, K. Heurtefeux, and Y. Rivierre.
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