



1st Workshop on CyberSecurity

A Self-Stabilizing Algorithm for Maximal *p*-Star Decomposition of General Graphs

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Université Claude Bernard (



Team: Graphs, Algorithms and Multi-Agents (GrAMA)

Outlines

- I. Introduction
- II. Graph decomposition and self-stabilization
- III. System Model
- IV. Proposed self-stabilizing Algorithm for star decomposition
- V. Proofs of proposed algorithm (Correcteness and convergence and complexity)
- VI. Conclusion and Futur work

Introduction



The decomposition problem is a way for partitioning a network into small components that satisfy some specific properties (topology, number of nodes, density, etc.).

Introduction



Self-stabilizing behavior of a system

Some self-stabilizing algorithms for graph decompositions

- F. Belkouch et al. in [IJPDC 02] considered a particular graph decomposition problem that consists in partitioning a graph of k² nodes into k partitions of order k.
- E. Caron et al. in [Euro-Par 09], C. Johnen et al. in [OPODIS 06], Bein et al. [ISPAN 05] focused on decomposing graphs into clusters.
- B. Neggazi et al. in [SSS 12] considered decomposition of graphs into triangles.

Star decomposition problem

This type of decomposition describes a graph as the union of disjoint stars.





Star decomposition problem

A **uniform** decomposition into stars is one in which all stars have equal size.

A p-star has one center node and p leaves where $p \ge 1$.



A p-star decomposition subdivides a graph into p-stars

Variant of generalized matchings and general graph factor problems that were proved to be NP-Complete [D. Kirkpatrick et al. in STOC 78], Journ. Comp. 83]

p-Star Decomposition of General Graphs



Graph G = (V,E) *p=3*

p-Star Decomposition of General Graphs



Graph G = (V,E) *p=3*

p-Star Decomposition of General Graphs Star 2 Star 1 Star 3 Star 4 Graph G = (V,E)p=3

Maximal *p*-star Decomposition

p-Star Decomposition vs Master-Slaves paradigm

This decomposition offers similar paradigm as the Master-Slaves paradigm used in :

- ✤ Grid [M. Mezmaz PDP 07].
- ✤ P2P infrastructures [A. Bendjoudi Int. J. Grid Util. Comput 09].



Contribution

The purpose of this work is to

- Develop a distributed and self-stabilizing algorithm for decomposing a graph into p-stars.
- * Operate with an unfair **Distributed Scheduler**.
- Suppose only local knowledge (Distance-1 knowledge).

A self-stabilizing system, regardless of its initial configuration, converges in finite time, without any external intervention. [E.W. Dijkstra 74]



p(v) is true -> v is enabled -> Move

Two types of schedulers (daemons) :

- ✤ central (serial).
- Distributed.
 Special case : Synchronous

Fairness:

- ✤ Fair.
- Unfair (adversarial).

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NB. This work assumes the most general scheduler.

Complexity :

- * Moves
- Steps
- * Rounds

Graph G = (V, E),

Assume that each node "v "has "id" (locally distinct).

We denote : - N(v) open neighborhood,

- d(v) degree of a node v,
- **p** is a positive integer.

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Definition. A p-star Decomposition D of a graph G = (V,E) is a set of subgraphs of the form $S_i = (V_i, E_i)$ such that the sets $V_i \subseteq V$ are disjoint and each S_i is a p-star.

D is maximal if the subgraph induced by the nodes of **G** not contained in **D** does not contain a p-star as a subgraph.

Impossibility of finding a deterministic self-stabilizing algorithm for maximal matching in anonymous graph under a distributed scheduler. [F. Manne et al. TCS 2009] p-star decomposition is a
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Impossibility result remains valid for p-star decomposition for all $p \ge 1$

p-star decomposition algorithm requires a **mechanism for symmetry breaking**

General idea

STEP 1 : The node v with the smallest identifier having at least p neighbors becomes master.

- > **STEP 2** : The p neighbors v_1, \ldots, v_p of v with the smallest identifiers become slaves of v.
- The previous steps are repeated for the subgraph of G consisting of all nodes except v, v₁, ..., v_p.

The challenge is to design an efficient distributed version of this algorithm under an unfair distributed scheduler.

Let X be a set and p is a positive integer.

Two operators :

$$X^{p} = \begin{cases} \phi & \text{if } |X|
$$\min X = \begin{cases} null & \text{if } |X| = \phi \end{cases}$$$$

 $\begin{bmatrix} -\\ \end{bmatrix}$ the smallest element of X otherwise

If identifier of v is smaller than identifier of u then we note v < u.

We define that $\forall v \in V : v < null$

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Each node v maintains two variables:

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SMSD uses the following code permitting a node v to compute its new values of s_{new} and m_{new} .

If
$$(\min M(v) < v \lor S(v)^p = \phi)$$
 then
 $v.s_{new} \coloneqq \phi$; $v.m_{new} \coloneqq \min M(v)$;
else
 $v.s_{new} \coloneqq S(v)^p$; $v.m_{new} \coloneqq null$;

Algorithm 1: Star Decomposition (SMSD)

Nodes: v is the current node

$$v.m \neq v.m_{new} \lor v.s \neq v.s_{new} \rightarrow v.m \coloneqq v.m_{new}; v.s \coloneqq v.s_{new}; [R]$$

Example of executing Algorithm SMSD under the synchronous scheduler.

Graph G is a complete graph Let p = 3

Initial configuration





m = null





Example of executing Algorithm SMSD under the synchronous scheduler.

Graph G is a complete graph Let **p** =**3**

Final configuration



Correctness proof

Lemma 1. In a configuration with no node is enabled, the following properties hold for each $v \in V$:

(a) If
$$v.s \neq \phi$$
 then $v.s \subseteq N(v)$ and $|v.s| = p$ and $v.m = null$.

(b) If $v.m \neq null$ then $v.m \in N(v)$.

(c) If $v \in w.s$ then v.m = w and $v.s = \phi$.

Correctness proof

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(b) If $v.m \neq null$ then $v.m \in N(v)$.

(c) If $v \in w.s$ then v.m = w and $v.s = \phi$.

Lemma 2. In a configuration with no enabled node the stars induced by all nodes v with $v.s \neq \phi$ form a maximal p-star decomposition of G.

The time complexity of the algorithm is measured in rounds.

A round under an unfair distributed scheduler may consist of an **infinite** number of **moves**.

Not sufficient to prove that the algorithm stabilizes after a finite number of rounds.

Theorem 1 : SMSD requires a finite number of moves.

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Theorem 1 : SMSD requires a finite number of moves.

For each node v : we distinguish

- *m-move* if v executes rule R and assigns a new value to *v.m*
- *s-move* if v executes rule R and assigns a new value to v.s

Note: A move can be a *m-move* and a *s-move* at the same time.

Lemma 3. Let $v \in V$ and "e" an execution of Algorithm SMSD such that no node u with u < v makes an s-move in e. Then v makes at most d(v)+2 s-moves in e.

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Lemma 4. The total number of s-moves in any execution of Algorithm SMSD is finite.

Lemma 5. Let Δ be the maximum node degree in the graph G. The total number of m-moves in any execution of Algorithm SMSD is at most $\Delta C + n$, here C denotes the total number of s-moves during the execution.

Theorem1. Algorithm SMSD is a self-stabilizing algorithm for

computing a maximal p-star decomposition.

The complexity ??

Lemma 6. After round r_0 and in all following rounds, each node $v \in V$ satisfies the following properties.

(a)
$$v.m = null$$
 or $v.m \in N(v)$.

(b) If $v.s \neq \phi$ then $|v.s| = p \land v.s \subseteq N(v) \land d(v) \ge p \land v.m = null$.

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Lemma 7. After round r_1 and in all following rounds, each node $v \in V$ with v.m = u satisfies $d(u) \ge p$ and $v.s = \phi$.

Lemma 8. Let v^* be the smallest node in G such that $d(v^*) \ge p$. Then, (a) After round r_2 and in all following rounds, $v^*.m = null$ and $v^*.s = N(v^*)^p$ (b) Let be $S^* = (v^* \cup v^*.s)$. After round r_3 and in all following rounds, $v.m \notin S^*$ and $v.s \cap S^* = \phi$ for all $v \in V(G) \setminus S^*$.

Lemma 8. Let v^* be the smallest node in G such that $d(v^*) \ge p$. Then, (a) After round r_2 and in all following rounds, $v^*.m = null$ and $v^*.s = N(v^*)^p$ (b) Let be $S^* = (v^* \cup v^*.s)$. After round r_3 and in all following rounds, $v.m \notin S^*$ and $v.s \cap S^* = \phi$ for all $v \in V(G) \setminus S^*$.

Lemma 9. Algorithm SMSD stabilizes after at most $2\left\lfloor \frac{n}{p+1} \right\rfloor + 2$ rounds.

Theorem 2. Algorithm SMSD is self-stabilizing algorithm for maximal *p*-star decomposition and converges after at most $2\left\lfloor \frac{n}{p+1} \right\rfloor + 2$ rounds under the unfair distributed scheduler using $O(p \log n)$ memory.

- First self-stabilizing algorithm for graph decomposition into disjoint p-stars (SMSD).
- > SMSD operates under the unfair distributed scheduler and stabilizes after at most $2\left|\frac{n}{p+1}\right|+2$ rounds.

- First self-stabilizing algorithm for graph decomposition into disjoint p-stars (SMSD).
- > SMSD operates under the unfair distributed scheduler and stabilizes after at most $2\left|\frac{n}{p+1}\right|+2$ rounds.
- The proposed algorithm generalizes maximal matching algorithms where p = 1. The time complexity in rounds of SMSD has the same order as the best known self-stabilizing algorithm for maximal matching under the synchronous scheduler or the distributed scheduler.

- > SMSD requires at most $O(\frac{n^2}{p})$ moves using the synchronous scheduler.
- > The exact move complexity of the algorithm under the unfair distributed scheduler is unknown.

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As future works, we aim to

- > Bound moves complexity of SMSD.
- > Generalize SMSD to weighted graphs.



End