Technische Universität Darmstadt





TK1: Distributed Systems Programming & Algorithms

Chapter 3: Distributed Algorithms

Section 5: Local Algorithms

Lecturer: Prof. Dr. Max Mühlhäuser

Contributors: Michael Stein (michael.stein@tk.informatik.tu-darmstadt.de),

Immanuel Schweizer (schweizer@tk.informatik.tu-darmstadt.de)

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OUTLINE

- Motivation
- Local Algorithms in Theoretical Context
- Theory Meets Practice...
- Summary



Introduction



• The size of the Internet increases rapidly!

2015: 15 Billion (Cisco, 2011)

2020: 20 Billion (IMS Research, 2010)





- Internet of Things
 - RFID
 - Sensors



One fundamental issue: Scalability



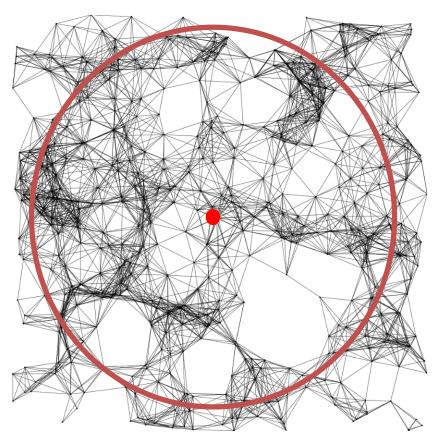
Reducing the Scope of Knowledge



 Observation: Considering scalability, having a global view on the network is too expensive



- Intuition: Limiting interactions to the local neighborhood of individual nodes improves scalability
- Other desired properties (e.g., robustness)
- Problem: Trade-Off (Performance vs. Scope)
- What can be done based only on local knowledge?





Fundamental Questions



- What is a local algorithm?
- What can be computed locally?





OUTLINE

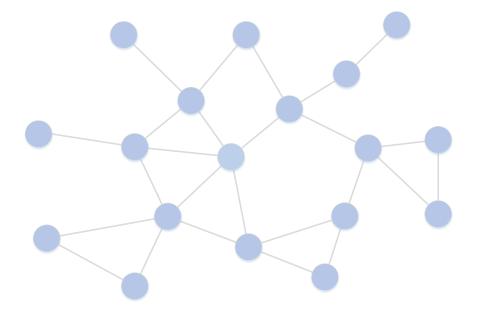
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Graph Definitions



- We use a graph G = (V,E) do model communication topologies
- Vertices (V) (also called nodes)
 - Denote networking devices
 - |V| = n
- Edges (E)
 - Denote connectivity between devices
 - Might be directed or undirected





In Theory... The *LOCAL* Model



- Wide-spread deterministic model of distributed computing (Initially proposed by Linial [1992])
- The nodes perform globally synchronized communication rounds

Initially, each node has only task-specific input and no knowledge about other nodes in the network.

In each round, each node performs the following three consecutive steps:

- 1. Local computation
- 2. Send a message to each neighbor
- 3. Receive one message from each neighbor

After k rounds, the algorithm terminates and each node computes and announces its local output.



Properties of the *LOCAL* model



- This model focuses on the locality in distributed processing [Linial 1992] and abstracts away restricting factors [Peleg 2000]
 - Computation is free
 - Message size is unbounded
- Observation: In k synchronous communication rounds, information may travel at most k hops through the network
- Bounding k, we are able to ensure that nodes have only restricted knowledge on the network!
- In the following, we focus on algorithms where k is a constant that is independent from the network size [Suomela 2013]
- Such algorithms run in constant time independently from the network size!



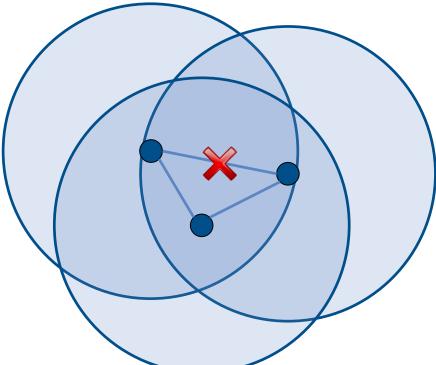
kTC: An Example of an algorithm in the *LOCAL* Model



- Related Work: Local Algorithm kTC [Schweizer et al. 2012]
 - Round 1: Each node broadcasts its ID
 - Round 2: Each node broadcasts its 1-hop neighborhood

■ After Round 2: Remove the longest edge in each triangle (e.g., based on

signal strength)





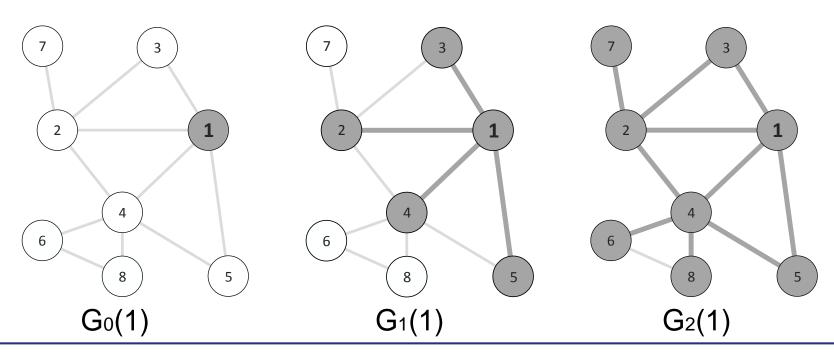
Definition: k-hop Neighborhood



When d(u,v) gives the hop count between u and v, the k-hop neighbors set of a node v is given by:

$$B_k(v) = \{u \in V_T : \min\{d(u, v), d(v, u)\} \le k\}$$

■ G_k(v) is the subgraph of G that can be constructed in the *LOCAL* model by v in k rounds





Limitations of the *LOCAL* Model



- What can be achieved in this model?
- In particular, there are two arguments that can be used to show that a certain problem can not be solved in this model [Suomela 2013]
 - Inherently non-local Problem
 - Impossibility of Symmetry Breaking



Inherently Non-Local Problems

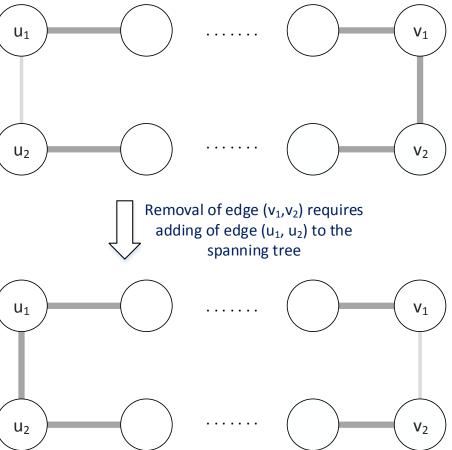


■ A problem is inherently non-local if the output of a node v depends on the initial input of a node outside of $G_k(v)$

Example: Creating a spanning tree is inherently non-local[Suomela 2013]



■ The decision if (u₁,u₂) should be added to the tree depends on whether (v₁,v₂) is contained in the tree!

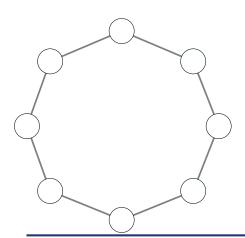




Impossibility of Symmetry Breaking



- A distributed algorithm in the LOCAL model is equivalent to a function that maps local neighborhoods G_k to local outputs [Fraigniaud et al. 2013]
- As the model is deterministic, two nodes with an equal view of the graph produce the same output!
- Example: Impossibility to assign addresses in an n-cycle



Each node assigns itself the same address!



Results for the *LOCAL* Model



- Many papers investigate usage of the model in the context of graph-theoretical problems
- Possibility results regarding this model are rather pessimistic: Many problems can neither be solved exactly nor approximated in the LOCAL model in constant time ⊗
- A comprehensive collection of results is given in the survey by Suomela [2013]



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Theory Meets Practice



■ How do the *LOCAL* model and corresponding results map to the real world?

Problems:

- The *LOCAL* model is a synchronized round-based model. However, communication in practice is carried out asynchronously!
- Nodes do not have access to information that may be available in practice (environment-specific knowledge, locally observable message flow, etc.)
- The *LOCAL* model terminates after k rounds. What about algorithms that do not terminate at all?
- So... what is a local algorithm in practice? And how can we check the practical relevance of the theoretical results?
- → Just take a look at practical algorithms that claim to be local!



Applications of Local Algorithms



- In the following application domains, there do exist algorithms that are considered to be local in corresponding publications
 - Addressing
 - Clustering
 - Connectivity
 - Load Balancing
 - Localization
 - Routing
 - Service Discovery

- Service Placement
- Sleep Control
- Video Streaming
- Graph-Theoretical Problems
- Topology Control
- Topology Mismatching

Next, we will give an exemplary insight into two of these domains

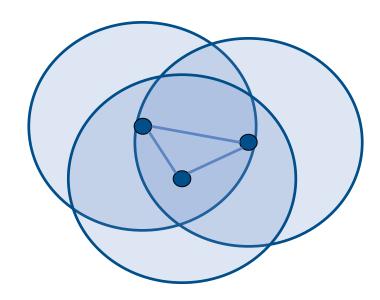


Excursus on Wireless Sensor Networks



- Large networks of small, strictly hardware-limited devices
 (e.g., only a few kBytes of RAM...)
- The devices are battery-powered
- Wireless communication interface with maximum transmission range
- The nodes aim at sensing their environment and send the data to a central base station
- → How to do routing?







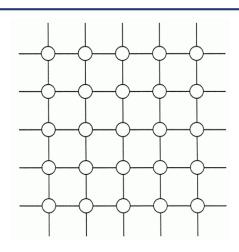
Greedy Routing



- Greedy forwarding
 - Each node knows its own coordinates
 - Source knows coordinates of destination



- Take the "closest" neighbor
- Close might be defined different to enhance the algorithm
- Works well on some networks, e.g., grid networks





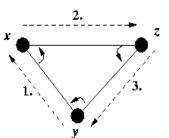
Greedy Perimeter Stateless Routing[Karp and Kung 2000]

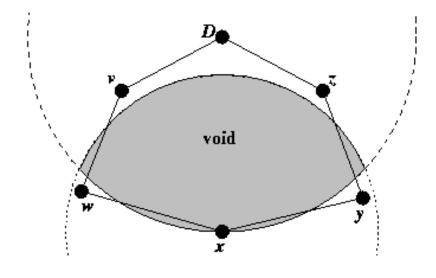


- Doesn't work with voids in the network
 - Use greedy when possible, change to right hand rule otherwise

Right-hand rule

When arriving at a node x from node y, the next edge traversed is the next one sequentially counterclockwise from the edge (x,y)





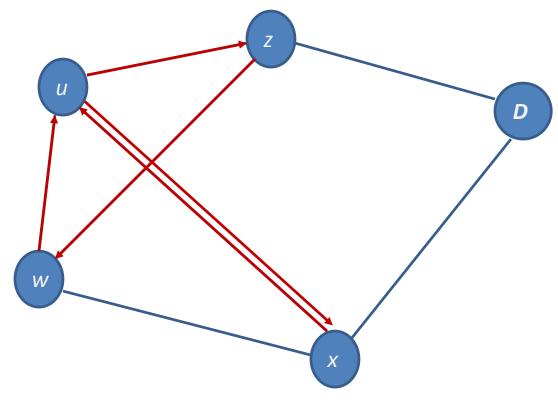


Greedy Perimeter Stateless Routing[Karp and Kung 2000]



- Works very well in random graphs
 - But: No edges are allowed to cross, otherwise it runs into a loop

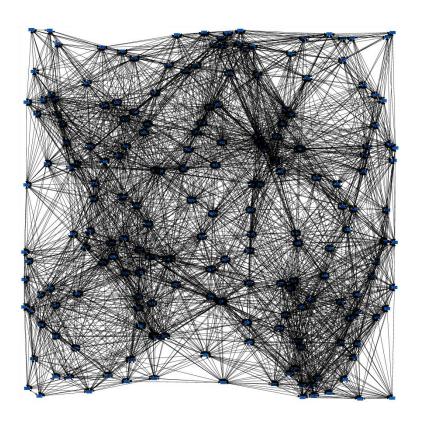
- Start with x to u
 - X-u-z-w-u-x
- Solutions:
 - No crossing heuristic
 - Remove an edge that is encountered again
 - Might partition the network
 - Planar graphs
 - No crossings allowed
 - Topology Control!

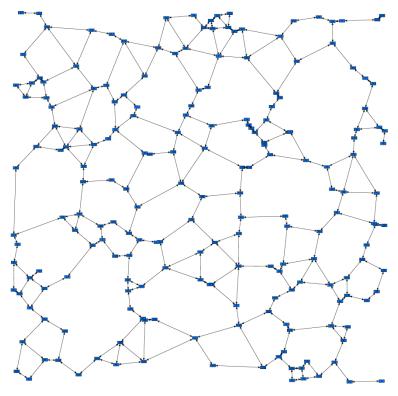




Topology Control









Topology Control



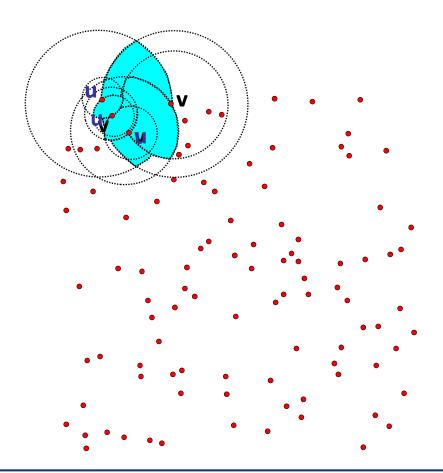
- Make the topology "better" without:
 - Partitioning the network
 - Making the paths much longer
- What does "better" mean
 - Planarity as , for example, required by Greedy Perimeter Stateless Routing
 - Less logical neighbors
 - Allows for shrinking the transmission power of nodes
 - Less physical neighbors
 - Less interference



Relative Neighborhood Graph (RNG)



Edge (u, v) exists if the intersection of the disks centered at u and v is free of other nodes.

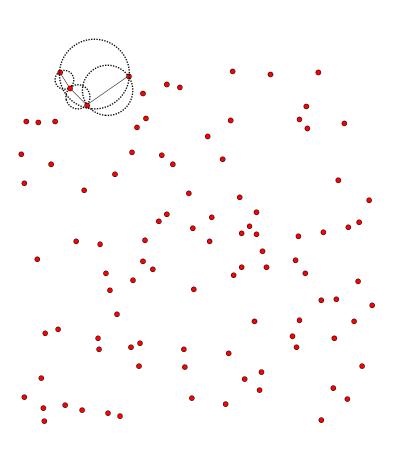


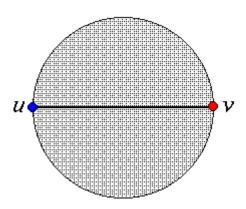


Gabriel Graph (GG)



■ Edge (u,v) exists iff disk(u,v) is free of other nodes







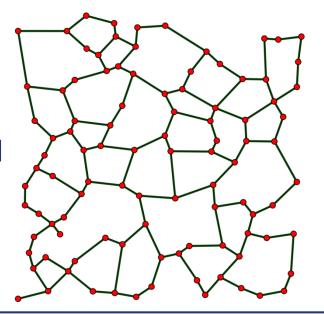
Topology Control

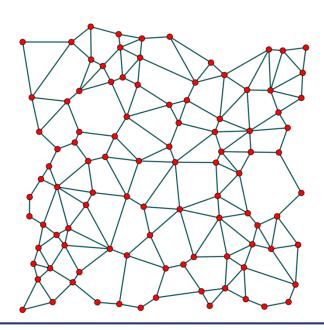


- Easy to calculate locally
 - Only 1-hop information about position needed
- The resulting graph is planar and sparse
 - |E| ≤ 3 |V|

■ RNG ⊂ GG

Also: connected







Relation of These Algorithms to the *LOCAL* Model?



- Intuitively, these algorithms are local
 - GPSR selects the next hop to the target coordinates based on the coordinates of the 1-hop neighbors
 - For GG and RNG, the edge removal decision depends on node coordinates in the 1-hop neighborhood
- What about the *LOCAL* model
 - GPSR is absolutely non-local according to the *LOCAL* model because data may be forwarded an arbitrary number of hops through the network
 - GG and RNG may indeed be implemented in the *LOCAL* model



Key Observations

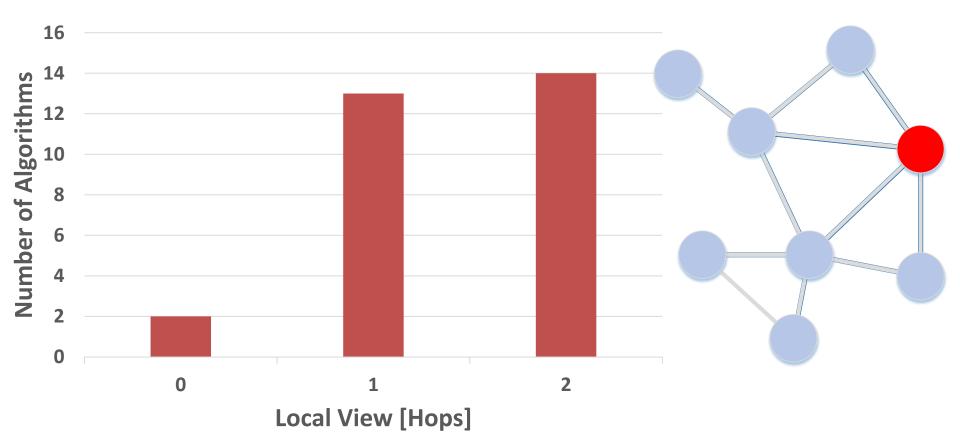


- Many problems that are not solvable in the theoretical model can be heuristically tackled in practice by local algorithms!
- Most importantly: Most of the investigated algorithms (e.g., GPSR) are not covered by the LOCAL model!



Another Key Observation: Size of the "Local View"





A local view of only two hops is sufficient to tackle many problems!



How to Define Locality in Practice?



- The algorithms claiming to be local differ from each other with respect to their degree of locality
- Hence, it is not meaningful to give a broad definition that covers all of them
- Open research question: Is it possible to define classes of locality?

Strictly local algorithms (hard requirements)

Distributed algorithms (low requirements)



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Summary and Outlook



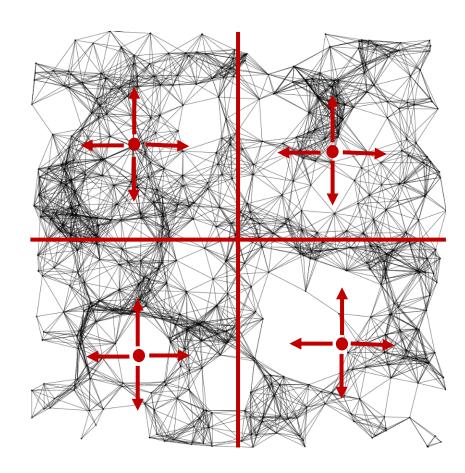
- What is a local algorithm?
- Well, that depends on the chosen perspective...
- Local algorithms in theory...
- ... and in practice



Outlook: Local/Global Trade-off



- Obviously, there is a trade-off between the performance of an algorithm and its degree of locality
- How can this trade-off be controlled?
 - Increasing k
 - Divide the network into regions and apply centralized algorithms?





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