## Analysis of Link Reversal Routing Algorithms for Mobile Ad Hoc Networks

Seminar of Distributed Computing WS 04/05 ETH Zurich, 1.2.2005

> Nicolas Born nborn@student.ethz.ch

### Paper

Analysis of Link Reversal Routing
 Algorithms for Mobile Ad Hoc Networks
 Costas Busch, Srikanth Surapaneni, Srikanta Tirthapura;
 SPAA 2003

## Overview

Link Reversal Routing Algorithms
 Full Reversal
 Partial Reversal

#### Equivalence of Executions

- Performance Analysis
- Results

#### Conclusion

## Link Reversal Routing Algorithms

Introduced by Gafni and Bertsekas (1981)
Routing in mobile ad hoc networks
Adaptive, self-stabilizing

 Contribution of the paper: first performance analysis

#### **Model** Link Reversal Routing Algorithms

Ad-Hoc Network

Network connectivity is assumed

- Each node has an unique id
- Suited for networks with "average mobility"



## **Underlying Communication Graph**

Link Reversal Routing Algorithms

 Convert the ad-hoc network to a destination oriented graph



## Notation

Link Reversal Routing Algorithms

#### Destination

- Good nodes: nodes with at least one directed path to the destination
- Bad nodes: nodes with no directed path to the destination
- Sinks: nodes with only incoming links



#### **Routing** Link Reversal Routing Algorithms

 When a node receives a packet, it forwards the packet on any outgoing link. The packet will eventually reach the destination.



## **Route Maintenance**

Link Reversal Routing Algorithms

- If a node loses its route to the destination, the algorithm reacts by performing link reversals.
- Node finds out that it has become a sink -> it reverses the directions of some or all incoming links.



#### Work and Time Link Reversal Routing Algorithms

# Work: number of reversals until stabilization.

Time: number of parallel time steps until stabilization.

## Overview

Link Reversal Routing Algorithms OFull Reversal
OPartial Reversal
Equivalence of Executions
Performance Analysis
Results
Conclusion

## **Full Reversal Algorithm**

 When a node becomes a sink, it reverses the directions of all its links.



## Implementation

Full Reversal Algorithm

 Idea: analogy to a river. Water flows from bigger height to lower height.

• => Implemented with heights • Height of node  $v_i$ :  $h_i$ •  $h_d = 0$ •  $N_i$ : neighborhood of  $v_i$ • Height of  $v_i$  after reversal: max{ $h_j | v_j \in N_i$ } + 1



#### **Example** Full Reversal Algorithm



## Overview

Link Reversal Routing Algorithms OFull Reversal
Partial Reversal
Equivalence of Executions
Performance Analysis
Results

Conclusion

## **Partial Reversal Algorithm**

- If a node v becomes a sink, it reverses the links to those neighbors that have not reversed their links into v.
- If every neighbor node has a reversed link to v, it reverses every link.

## Implementation

**Partial Reversal Algorithm** 

- Also implemented using heights
   Height of node v<sub>i</sub>: h<sub>i</sub>
   h<sub>d</sub> = o
   Height of v<sub>i</sub> after reversal: min{ h<sub>j</sub> | v<sub>j</sub> ∈ N<sub>i</sub>} + 1
- Every node v keeps a list of its neighboring nodes that have reversed their links into v.

#### **Example** Partial Reversal Algorithm







Node that reverses

Reversals: 5 Time: 4

## Overview

Link Reversal Routing Algorithms
 OFull Reversal
 OPartial Reversal

- Equivalence of Executions
- Performance Analysis
- Results
- Conclusion

## **Equivalence of Executions**

There are many different reversal schedules.

 Goal: show that any two executions of a deterministic reversal algorithm starting from the same initial state are equivalent.

## **Dependency Graph**

**Equivalence of Executions** 

- Execution  $R=r_{\eta},...,r_{k}$
- Directed edge from  $r_i$  to  $r_j$ , iff
  - $v_i$  is neighbor of  $v_i$
  - $r_i$  is first reversal of  $v_i$  after  $r_i$  in execution R



## Main Theorem

**Equivalence of Executions** 

Two executions are equivalent, if they have the same dependency graph.

 Theorem: Any two executions of a deterministic reversal algorithm starting from the same initial state are equivalent.

## Conclusions

**Equivalence of Executions** 

- For all executions of a deterministic reversal algorithm starting from the same initial state:
  - OFinal state is the same
  - ONumber of reversals of each node is the same
- The depth of the dependency graph is a lower bound for the time complexity of execution of a deterministic reversal algorithm.

## Overview

 Link Reversal Routing Algorithms OFull Reversal
 OPartial Reversal
 Equivalence of Executions

- Performance Analysis
- Results
- Conclusion

## **Full Reversal Algorithm**

**Performance Analysis** 

Goal: lower and upper bound on the performance of the full reversal algorithm



#### Question Full Reversal Algorithm

 For any reversal algorithm starting from any initial state, a good node never reverses till stabilization.



But how many times do the bad nodes reverse?Idea: Group the bad nodes in layers!

#### **Layers** Full Reversal Algorithm

Bad node *v* is in layer *i,* iff

 $\bigcirc$  there is an incoming link to v from a node in layer *i-1*, or

 $\bigcirc$  there is an outgoing link from v to a node in layer *i*.



## **Schematic View**

Full Reversal Algorithm



#### Execution E<sub>1</sub> (Step 1) Full Reversal Algorithm

 There exists an execution E<sub>1</sub> which brings the system from state / to state /', such that every bad node reverses exactly one time.



# **Execution** *E*<sub>1</sub>(Step 2) Full Reversal Algorithm



# Execution E<sub>1</sub>(Step 3) Full Reversal Algorithm



# Execution E<sub>1</sub>(Step 4) Full Reversal Algorithm



# Execution E<sub>1</sub>(Step 5) Full Reversal Algorithm



## End of Execution E<sub>1</sub>

#### Full Reversal Algorithm



## After Execution E,

Full Reversal Algorithm

 At the end of this execution, all the bad nodes of layer 1 have become good, while all the bad nodes in the other layers stay bad.



## Lemma

Full Reversal Algorithm

Lemma: At the end of an execution E<sub>i</sub>, all the bad nodes of layer *i* become good, while all the bad nodes in layers *j>i*, remain bad.

#### **Proof** Full Reversal Algorithm

Any bad node not adjacent to a good node will remain in the same (bad) node-state after execution E<sub>i</sub>.

ONode-state: directions of its incident links



Each neighbor node is bad in state /  $\Rightarrow$  Each of them reverses in  $E_i$ 

 $\Rightarrow$  *v* also reverses in *E<sub>i</sub>* 

 $\Rightarrow$  Reversals leave the directions the same

#### **Proof** Full Reversal Algorithm

#### Proof:

○Bad nodes of layer *i* become good:



Layer i





#### **Proof** Full Reversal Algorithm

#### $\bigcirc$ Bad nodes in layers *j*>*i* remain bad.



## Lemma

Full Reversal Algorithm

Lemma: Layer *j+1* becomes layer *j* after execution *E<sub>i</sub>* (in the new state).

Proof:

- ○All bad nodes of layer *i* become good and bad nodes in other layers remain bad.
- OAll bad nodes in layers *j>i* remain in the same node-state.

Full Reversal Algorithm



Full Reversal Algorithm



Full Reversal Algorithm



Full Reversal Algorithm



Full Reversal Algorithm



Full Reversal Algorithm

 Back to our question: how many times do the bad nodes reverse?



Full Reversal Algorithm

- OEvery bad node reverses in each execution exactly one time.
- Each node in layer 1 became good after 1 reversal. Each node in layer 2 needed 2 reversals.

=> Each node in layer *i* needs *i* reversals before it becomes a good node.

Graph has *n* bad nodes
 Layer *i* has *n<sub>i</sub>* nodes

#### Full Reversal Algorithm



 $\Rightarrow \text{Number of reversals: } n_i \cdot 1 + n_2 \cdot 2 + n_3 \cdot 3 + n_4 \cdot 4 + n_5 \cdot 5$  $\Rightarrow \text{Trivial upper bound for } n \text{ bad nodes: } O(n^2)$ 

48

#### Upper Bound Full Reversal Algorithm

 We get an upper bound for the number of reversals in the full reversal algorithm:

For any graph with an initial state with n bad nodes, the full reversal algorithm requires at most  $O(n^2)$  work and time till stabilization.

We will now show that these bounds are tight

## **Lower Bound**

Full Reversal Algorithm

• There is a graph with an initial state containing n bad nodes such that the full reversal algorithm requires  $\Omega(n^2)$  work until stabilization.



○Each node in layer *i* will reverse *i* times ○sum of all reversals is  $1+2+3+...+n = n(n+1)/2 = \Omega(n^2)^{50}$ 

## Lower Bound

Full Reversal Algorithm

• There is a graph with an initial state containing n bad nodes such that the full reversal algorithm requires  $\Omega(n^2)$  time until stabilization.



O[n/2]+1 layers

 First [n/2] layers contain 1 node each last layer contains [n/2] nodes

○ sum of all reversals is 1+2+...+  $\lfloor n/2 \rfloor$ + ( $\lfloor n/2 \rfloor$ +1)· $\lfloor n/2 \rfloor$  = Ω( $n^2$ ) 51

## **Partial Reversal Algorithm**

**Performance Analysis** 

 One might expect that the partial reversal algorithm needs less reversals in the worst case than the full reversal algorithm. Is this true?

Idea: group the bad nodes in levels.

#### **Levels** Partial Reversal Algorithm

 Bad node v is in level i, if the shortest undirected path from v to a good node has length i.



## Some Reversals later

**Partial Reversal Algorithm** 



Upper bound on height

h <sup>max</sup>	+1	+2	+3	+4	+5
					- /

**Partial Reversal Algorithm** 



Upper bound on number of reversals  $h^{\text{max}} - h^{\text{min}} = h^* + 1 + 2 + 3 + 4 + 5$ 

Each reversal increases the height by at least 1.

## Upper Bound

Partial Reversal Algorithm

 A bad node needs in the worst case h\*+n reversals.

• We have n bad nodes: => O( $n \cdot h^* + n^2$ )

## Upper Bound

**Partial Reversal Algorithm** 

- For any initial state with n bad nodes, the partial reversal algorithm requires at most O(n·h\*+n<sup>2</sup>) work and time until the network stabilizes.
  - OProblem: h\* (= h<sup>max</sup> h<sup>min</sup>) may be arbitrarily large

## Lower Bound

**Partial Reversal Algorithm** 

There is a graph with an initial state containing *n* bad nodes, such that the partial reversal algorithm requires Ω(*n*·*h*\*+*n*<sup>2</sup>) work (time) until stabilization.

# Deterministic Reversal Algorithms

Defined by a "height increase" function g.
Heights of different nodes are unique
Node v is sink with height h<sub>v</sub> and adjacent nodes v<sub>n</sub>v<sub>2</sub>,...,v<sub>d</sub> with heights h<sub>n</sub>h<sub>2</sub>,...,h<sub>d</sub>
-> v's height after reversal is g(h<sub>n</sub>h<sub>2</sub>,...,h<sub>v</sub>)
=> Full and partial reversal algorithms are deterministic

#### **Bounds** Deterministic Reversal Algorithms

• There is a graph with an initial state containing n bad nodes such that any deterministic reversal algorithm requires  $\Omega(n^2)$  work (time) until stabilization.

=> Full reversal algorithm is optimal in the worst case, while the partial reversal algorithm is not!

## Overview

Link Reversal Routing Algorithms
OFull Reversal
OPartial Reversal
Equivalence of Executions
Performance Analysis
Results

#### Conclusion

## Results

- Full reversal algorithm requires O(n<sup>2</sup>) work and time (n = nodes which have lost the routes to the destination)
- Partial reversal algorithm requires  $O(n \cdot h^* + n^2)$ work and time ( $h^*$  = nonnegative integer)
- For every deterministic link reversal algorithm, there are initial states which require  $\Omega(n^2)$

## Overview

Link Reversal Routing Algorithms
OFull Reversal
OPartial Reversal
Equivalence of Executions
Performance Analysis
Results

#### Conclusion

## Conclusion

- Full reversal outperforms partial reversal algorithm in the worst case.
- Full reversal is optimal while the partial reversal algorithm is not.
- Number of reversals only depends on the number of bad nodes.
- Is there a variation of the partial reversal algorithm with O(n<sup>2</sup>) in the worst case?
- Partial reversal better in the average case?
- Analysis of non-deterministic algorithms (TORA)
- Algorithms only suited for connected graphs
- What about >1 destinations?



## Thanks for your attention!

Questions