

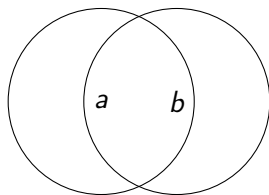
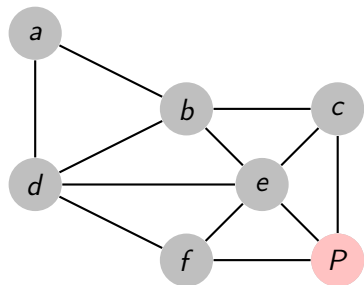
# Routing Algorithms using Random Walks with Tabu Lists

Karine Altisen, Stéphane Devismes, **Pascal Lafourcade** and  
Clément Ponsonnet

Verimag, University of Grenoble

ALGOTEL 2011 : May, Cap Estérel

# Wireless Sensor Network (WSN)

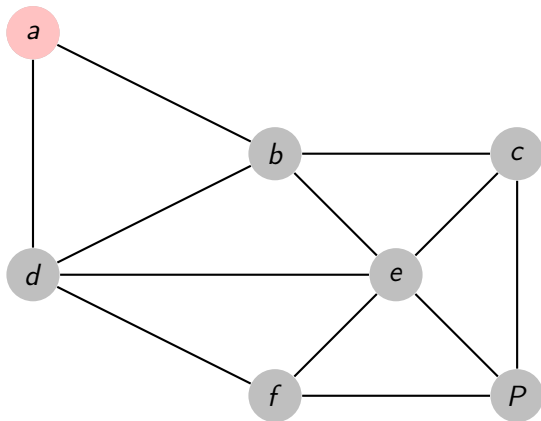


A node has :

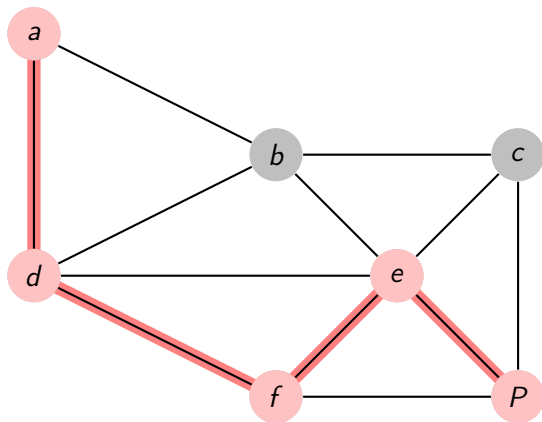
- ▶ low memory,
- ▶ low computation power
- ▶ finite battery.



# Routing



## Routing



# Settings

- ▶ One sink/Multi source
- ▶ Connected
- ▶ Identified
- ▶ Reliable
- ▶ Asynchronous
- ▶ Spontaneous requests

## Hitting Time

Average number of hops to reach the sink

# Plan

Introduction

Random Walks

Random Walks with Tabu Lists

In Messages

In Nodes

Experimental Results

Conclusion

# Plan

Introduction

**Random Walks**

Random Walks with Tabu Lists

In Messages

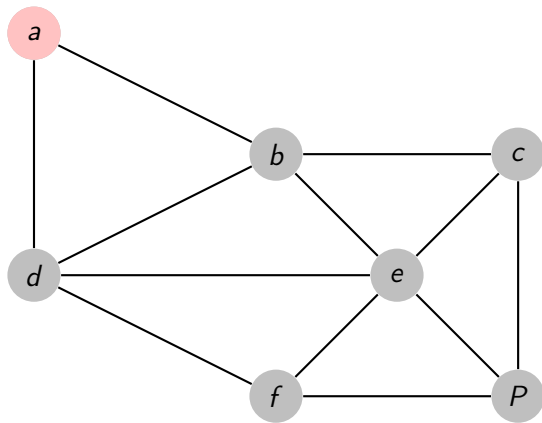
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# Random Walk

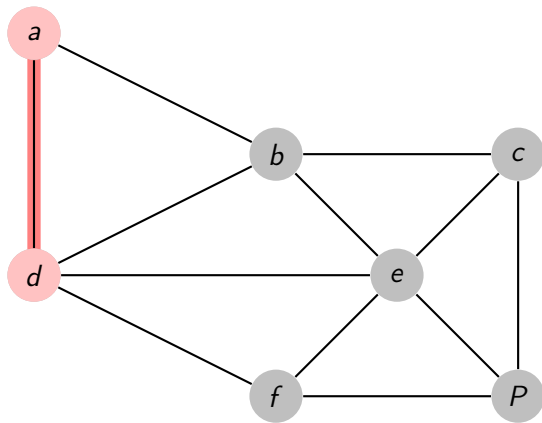
Uniform Law:  $P(u) = \frac{1}{|\mathcal{N}_u|}$





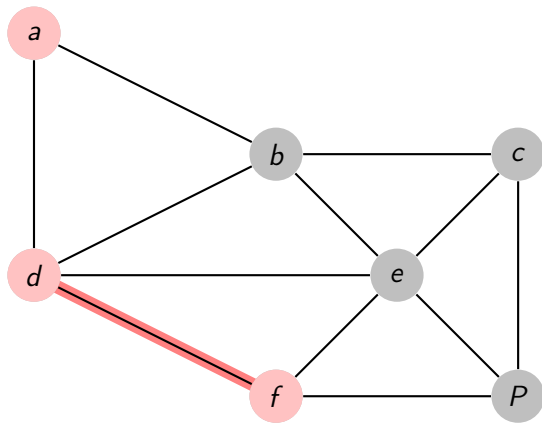
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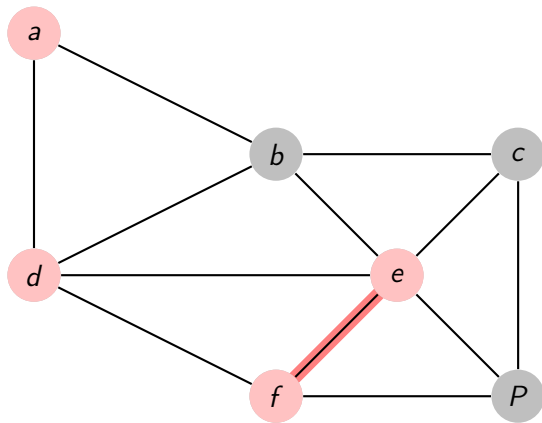
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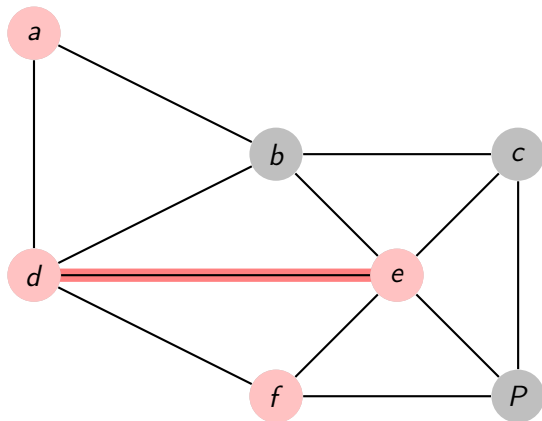
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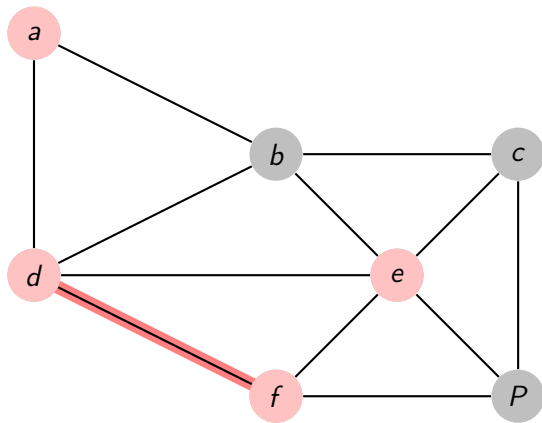
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# Random Walk

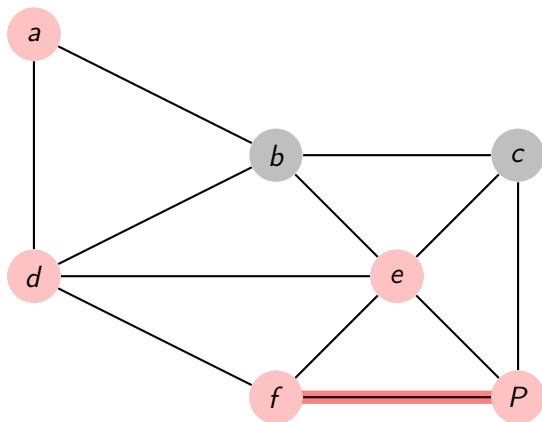
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# Random Walk

Uniform Law:  $P(u) = \frac{1}{|\mathcal{N}_u|}$

Complexity:  $O(N^3)$



## RWLD by Yamashita et al.

$\delta_u$  = number of neighbours of  $u$

$$P(u) = \frac{\delta_u^{-\frac{1}{2}}}{\sum_{w \in \mathcal{N}_v} \delta_w^{-\frac{1}{2}}}$$

Complexity:  $O(N^2)$

Avoiding high degree nodes

## RWLD by Yamashita et al.

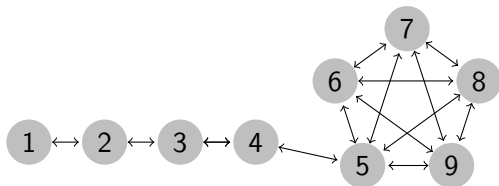
$\delta_u$  = number of neighbours of  $u$

Complexity:  $O(N^2)$

$$P(u) = \frac{\delta_u^{-\frac{1}{2}}}{\sum_{w \in \mathcal{N}_v} \delta_w^{-\frac{1}{2}}}$$

Avoiding high degree nodes

Worst Case:





# Routing by Random Walk

## Pros

- ▶ Message length
- ▶ Tight local computation and memory
- ▶ No overlay
- ▶ No control messages (no load effect)
- ▶ ...

## Cons

- ▶ Hitting time is  $O(N^3)$  (RW) and  $O(N^2)$  (RWLD)

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**Random Walks with Tabu Lists**

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## Random Walk with Tabu Lists

- ▶ Idea : Add memory to help random walks.

Avoiding cycles

- ▶ Store hints about previous choices

Is it a “good” trade-off ?

# Where is the Tabu List?

Two Options

## Where is the Tabu List?

### Two Options

In Messages (Pessimistic)

In Nodes (Optimistic)

Same Goal: Detecting and Avoiding Cycles

## How to deal with Full List?

Small memory and battery  $\Rightarrow$  small finite size of tabu list.

- A node identity appears only once in the Tabu List.

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### Variation of FIFO (First In First out)

- ▶ If  $N \in TL$  then  $N$  becomes first.
- ▶ Otherwise Add  $N$  and Remove the older node.



## How to deal with Full List?

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### Variation of FIFO (First In First out)

- ▶ If  $N \in TL$  then  $N$  becomes first.
- ▶ Otherwise Add  $N$  and Remove the older node.

### RAND

If  $N \notin TL$  then

- ▶ Randomly remove a node in TL.
- ▶ Add  $N$  instead.

Otherwise keep the current TL.

## Parameters of our algorithms

- ▶ 2 Locations of tabu lists: Message vs Nodes
- ▶ 2 Updates : FIFO vs RAND
- ▶  $k$  = Size of tabu list

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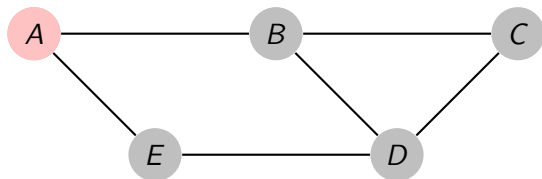
# Idea of TLM

## Avoiding Cycles!

- ▶ Store IDs of visited nodes.
- ▶ Visit new nodes first.

# TLM(2,FIFOupdate)

Size 2 avoids cycle of size 3.



FIFO (First In First Out)

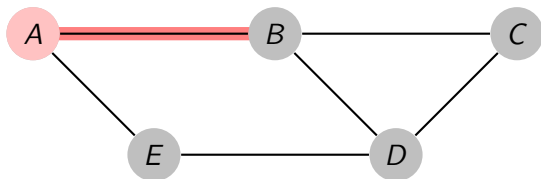
Visited Node : A

Tabu List:



# TLM(2,FIFOupdate)

Size 2 avoids cycle of size 3.



FIFO (First In First Out)

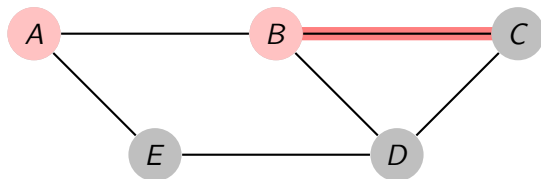
Visited Node : B

Tabu List:

A	
---	--

# TLM(2,FIFOupdate)

Size 2 avoids cycle of size 3.



FIFO (First In First Out)

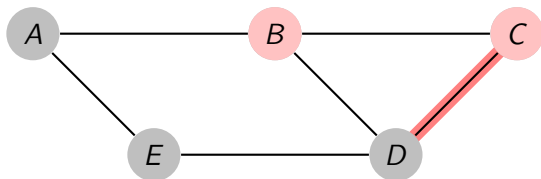
Visited Node : C

Tabu List:



## TLM(2,FIFOupdate)

Size 2 avoids cycle of size 3.



FIFO (First In First Out)

Visited Node : D

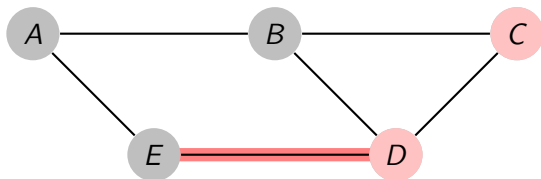
Tabu List:





# TLM(2,FIFOupdate)

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FIFO (First In First Out)

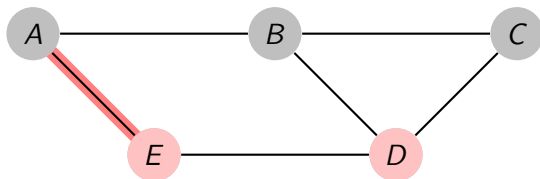
Visited Node : E

Tabu List:



# TLM(2,FIFOupdate)

Size 2 avoids cycle of size 3.



FIFO (First In First Out)

Visited Node : A

Tabu List:



## Summary of TLM

- ▶ Store in each message a list of already visited nodes.
- ▶ Select first nodes not in the Tabu List.
- ▶ Two updates policies : FIFO or RAND

### Dead-end

If all neighbours are in the Tabu List  
then we use RWLD.

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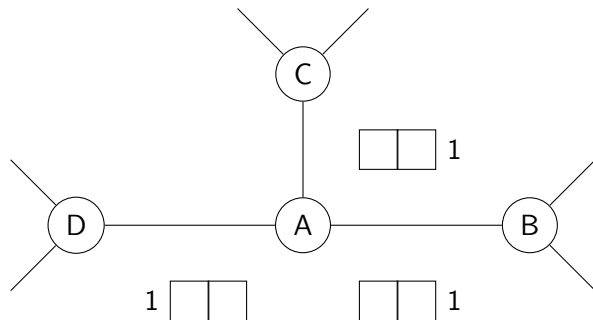
## Idea of TLCN

Detect cycles by learning.  
Avoid to route to cycles.

- ▶ One list per destination : Store message ID.
- ▶ One counter per destination : Detection of cycles.

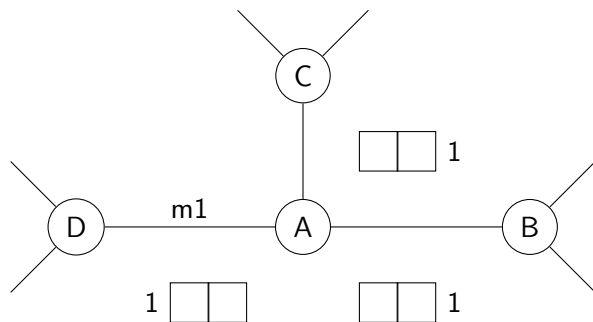
## TLCN(2,FIFOUpdate)

Cycle detection



## TLCN(2,FIFOUpdate)

Cycle detection



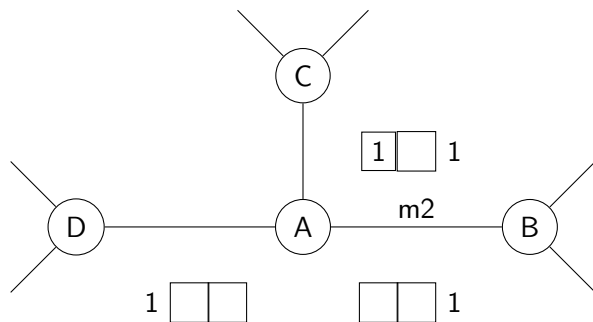
$$P_{\text{TLCN}}^{\text{v}}(\text{Counter})(u) = \frac{\text{Counter}[u]^{-1} \times \delta_u^{-1/2}}{\sum_{w \in \mathcal{N}_v} \text{Counter}[w]^{-1} \times \delta_w^{-1/2}}$$





## TLCN(2,FIFOUpdate)

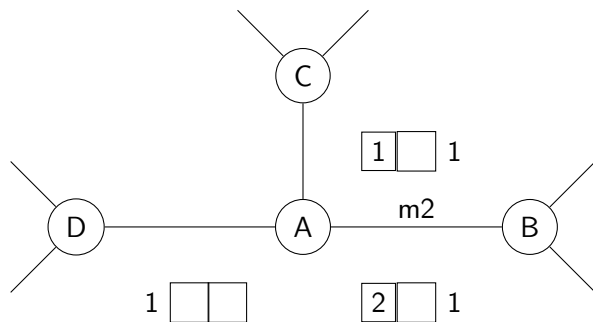
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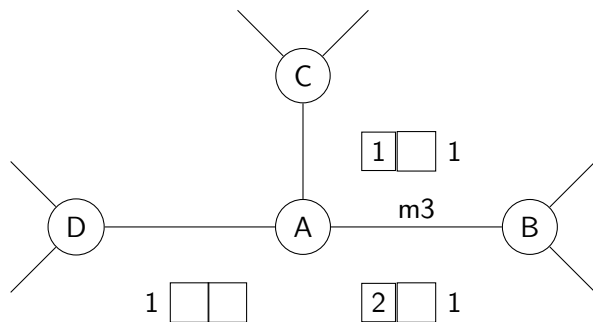
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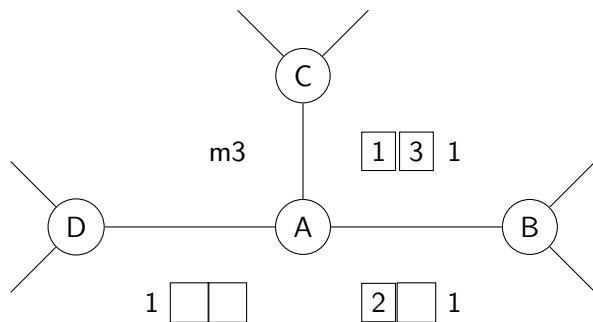
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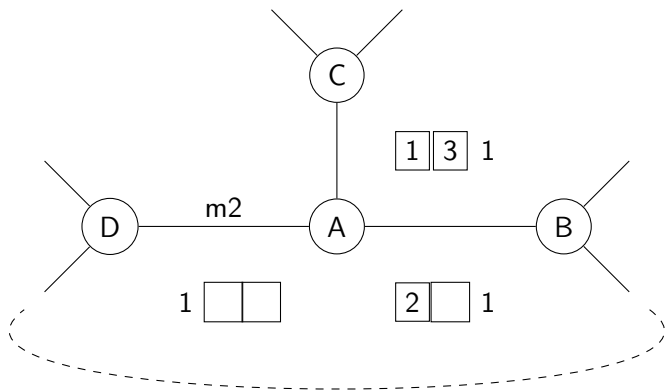
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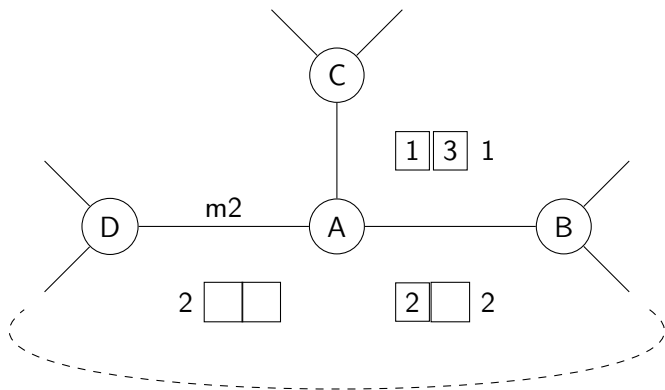
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Cycle detection



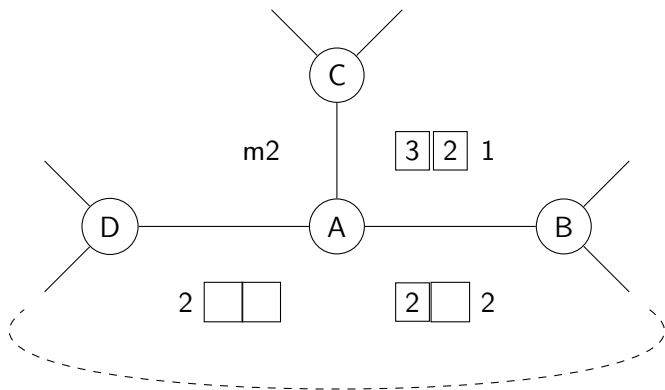
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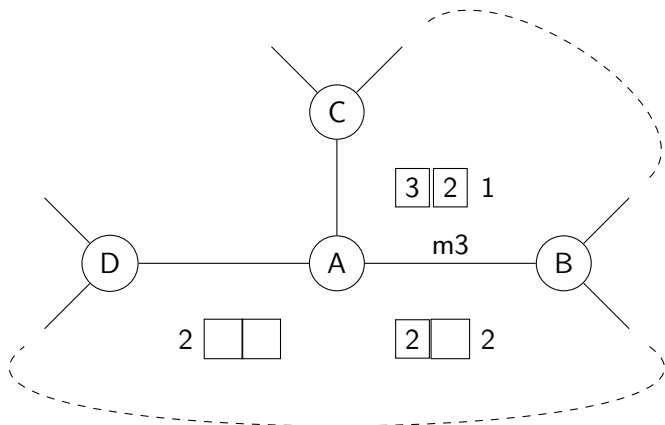
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Cycle detection



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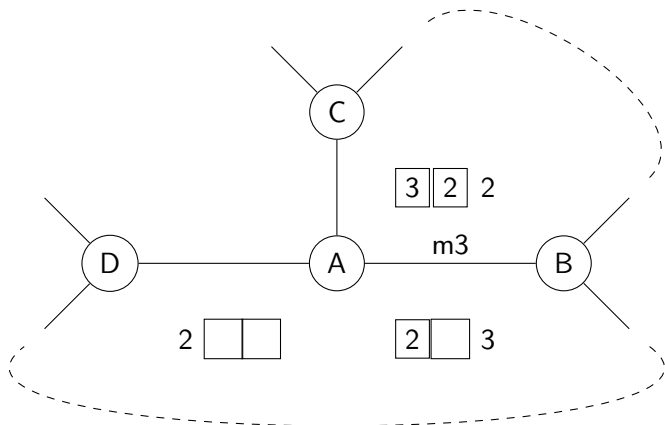
Cycle detection





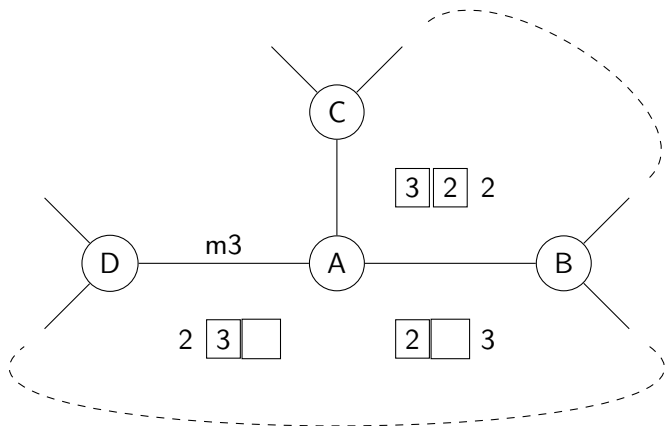
## TLCN(2,FIFOUpdate)

Cycle detection



## TLCN(2,FIFOUpdate)

Cycle detection



## Summary of TLCN

- ▶ In each node for each link :
  - ▶ Tabu List of messages already send.
  - ▶ Counter detecting possible cycle.
- ▶ Prefer to visit link with a low counter.
- ▶ Two updates policies : FIFO or RAND.

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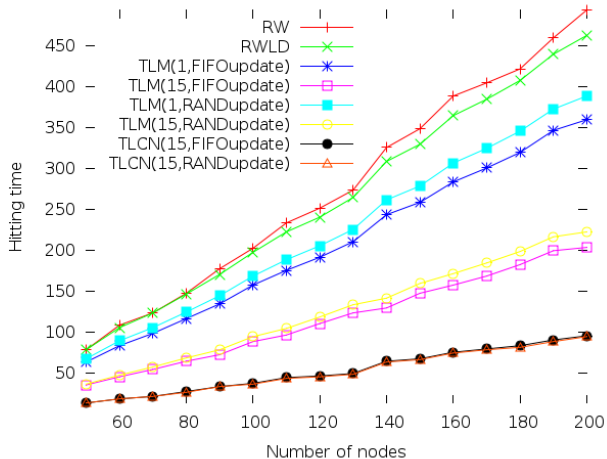
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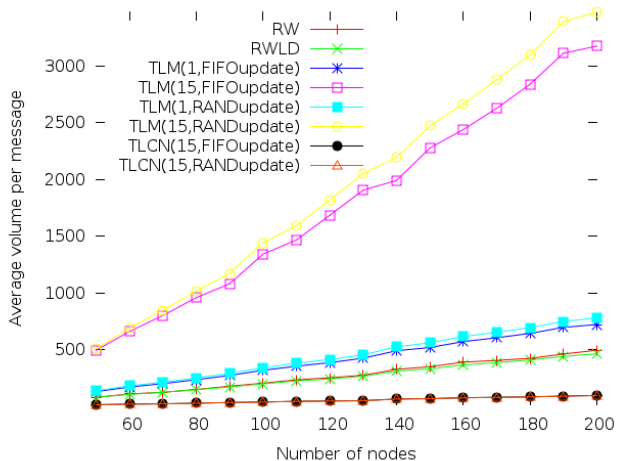
# Experiment Settings

- ▶ Sinalgo(JAVA)
- ▶ Graphs:
  - ▶ UDG.
  - ▶ connected.
  - ▶ one sink/multi-source.
  - ▶ uniform distribution.
  - ▶ Size : 60, 70, . . . , 200 nodes.
- ▶ 100 messages per sources.

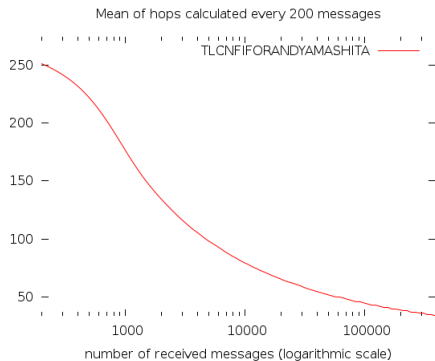
## Hitting Time and Number of Nodes



# Volume and Number of Nodes



# Convergence of TLCN



If  $\text{size}(\text{TL}) = \infty$  then TLCN converges to spanning tree.



# Plan

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## Random Walk with Tabu List

### TLM vs RW

- ▶ Better Hitting time.
- ▶ Introduces an overload.
- ▶ Size of the list has an influence.
- ▶ FIFO is slightly better than RAND.

### TLCN vs RW

- ▶ Improve Random Walk.
- ▶ Size of the list has a small influence.
- ▶ FIFO = RAND

**Conclusion** : TLCN better than TLM better than RW

## Sum Up

	Hitting Time	Volume	Load Sensitivity
TLCN (15,FIFO)	1 <sup>st</sup>	1 <sup>st</sup>	yes
TLCN (15,RAND)	1 <sup>st</sup>	1 <sup>st</sup>	yes
TLM (15,FIFO)	3 <sup>rd</sup>	7 <sup>th</sup>	no
TLM (15,RAND)	4 <sup>th</sup>	8 <sup>th</sup>	no
TLM (1,FIFO)	5 <sup>th</sup>	5 <sup>th</sup>	no
TLM (1,RAND)	6 <sup>th</sup>	6 <sup>th</sup>	no
RWLD	7 <sup>th</sup>	3 <sup>rd</sup>	no
RW	8 <sup>th</sup>	4 <sup>th</sup>	no

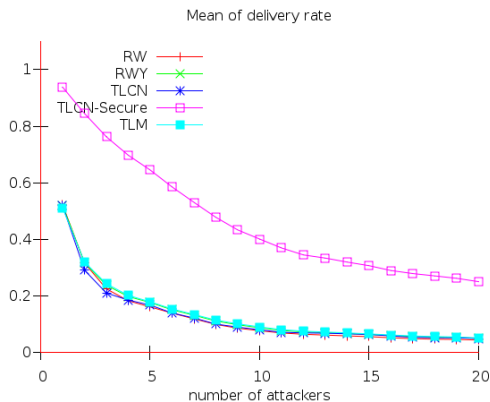
# Next

- ▶ Theoretical study of TLM and TLCN :
  - ▶ for any graph,  $TLM(1, FIFO) > RW$ ,
  - ▶ Find graph such that  $k > 2$   
 $TLM(k, FIFO) <> TLM(k+1, FIFO)$ .
  - ▶ Optimal size for Tabu List in TLCN, TLM.
  - ▶ Convergence of TLCN.
- ▶ Resistance to attackers?
- ▶ How to secure our protocols?

Thanks for your Attention

Questions ?

## Results for Secured Version of TLCN



# Results for Secured Version of TLCN

