

# Optimal Buffer Management Policies for Delay Tolerant Networks

**Chadi BARAKAT**

**INRIA Sophia Antipolis, France**  
**Planète research group**

**Joint work with Amir Krifa and Thrasyvoulos Spyropoulos**

**Email:** `Chadi.Barakat@sophia.inria.fr`

**WEB:** `http://www.inria.fr/planete/chadi`

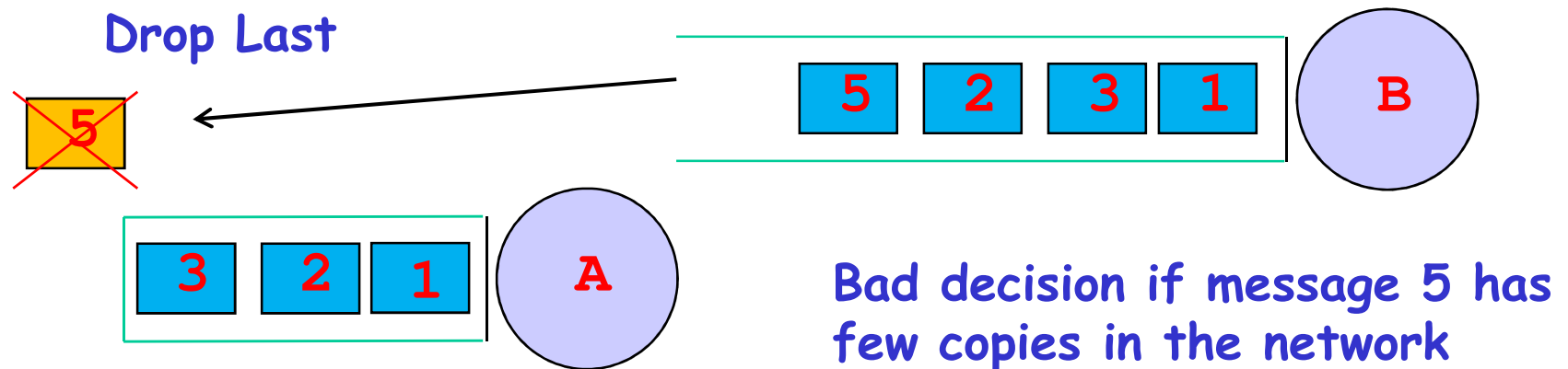
# The DTN principle

- Rely on nodes' mobility to route through disconnected networks
  - Node: human carrying a laptop or PDA, a bus, a car, a satellite, etc
  - Nodes carry messages of each other while moving
  - Nodes replicate messages to increase the chance of delivery
- **DTN routing:** important DTN block that decides whether to replicate a message or not.
  - An important tradeoff:
    - The more the copies the more the chance
    - But the more the load on the network
  - Different propositions: Epidemic, spray and wait, utility-based, etc

# How to Drop Messages in DTNs?

## Another important block

- Routing can limit the number of copies but is not enough ...
  - Routing does not check if resources are available before replication
  - Nodes' buffers can still overflow due to many messages
  - Which message to drop in case of congestion?
    - Last In ? First In ? Oldest ? Youngest ? Other ?
  - An important problem for which there is no clear solution



# Our main contribution

- ❑ An analytical framework to solve the problem in its foundation using the notion of **per-message utility**
  - The utility can be seen as the marginal loss in some global metric (as the delivery ratio or the delivery delay) when dropping a message
- ❑ The optimal drop policy is then simple: **Drop the message with the lowest utility value.**
- ❑ State of the art:
  - Basic comparisons as Drop Front better than Drop Tail (Zhang et al.)
  - RAPID (Levine et al.) is relevant to our work but is for message scheduling and makes stronger assumptions.

# Methodology

- ❑ Consider point-to-point communications
- ❑ Suppose first global knowledge about copies
- ❑ Take a global routing metric as the delay or the delivery rate
- ❑ Estimate its value at the moment of congestion
  - Decisions are taken separately by the nodes
- ❑ Find the best message to drop
  - Associate utilities to messages (marginal loss in global performance)
  - Drop the message with the lowest utility value.
- ❑ Then try to estimate the global knowledge using global information BUT on old messages ...

# Notations / Assumptions

- ❑  $n$  = Number of copies of a message in the network
- ❑  $m$  = Number of nodes that have seen a message
  - Used to calculate the probability of message being delivered
- ❑  $L$  = Number of nodes
- ❑  $\lambda$  = rate at which nodes meet
  - Inter-meeting time exponentially distributed
- ❑ TTL = lifetime of messages

# Case of delivery rate

- Global delivery rate at the time of congestion is then:

$$DR = \sum_{\text{all messages}} \Pr(\text{message delivered})$$

$$\Pr(\text{message delivered}) = \frac{m}{L-1} + \left(1 - \frac{m}{L-1}\right) * \left(1 - \exp(-\lambda n R)\right)$$

- We differentiate DR w.r.t the number of copies to find the best message to drop. It is the one minimizing this value:

$$\left(1 - \frac{m}{L-1}\right) \cdot \lambda \cdot R \cdot \exp(-\lambda \cdot n \cdot R)$$

# Per-message utility relative to the delivery rate

$$\left(1 - \frac{m}{L-1}\right) \cdot \lambda \cdot R \cdot \exp(-\lambda \cdot n \cdot R)$$

This is a function of **global** information on a message. Models its utility (its importance) w.r.t. the global delivery rate.

We call it **per-message utility relative to the delivery rate**.

The resulting policy is called **GBD** (Global knowledge based drop policy)

Global information to be estimated later.



# Case of delivery delay

In the same way, one can find the per-message utility relative to the global delivery delay:

$$\frac{1}{n^2\lambda} * \left(1 - \frac{m}{L-1}\right)$$

Again function of global information on a message.

Not easy to relate our utility-based optimal policies to simple ones as Drop Tail and Drop Front

# Distributed version: How to calculate $n$ and $m$ ?

- ❑  $n$  = number of copies of a message  
 $m$  = number of nodes that have seen the message.
- ❑ One option is to use the value obtained by flooding, but this is not efficient because the info arrives too late.
- ❑ Our solution:
  - Flood global information BUT on old messages.
  - All nodes have the same information after some time.
  - Estimate  $n$  and  $m$  from what has happened to old messages at the same elapsed time.
  - Of course under some stationarity and ergodicity conditions.

# Distributed version (ctd)

- Suppose  $m$  and  $n$  follow two random variables  $M$  and  $N$ .
- We look for estimators that preserve global network performance:
  - Same average delivery delay and same average delivery rate

*Take for example the delivery rate. This translates to:*

real delivery rate = estimated delivery rate

$$E\left[\frac{M(T)}{L-1} + \left(1 - \frac{M(T)}{L-1}\right) * \left(1 - \exp(-\lambda N(T)R_i)\right)\right] = \frac{\hat{m}(T)}{L-1} + \left(1 - \frac{\hat{m}(T)}{L-1}\right) * \left(1 - \exp(-\lambda \hat{n}(T)R_i)\right)$$

Then solve this equation for the estimators of  $m$  and  $n$

- Estimator of  $m$  is taken equal to its average value.

# Distributed version: Message utility expressions

By plugging the estimators in utility expressions, we get:

For the delivery rate:  $\lambda RE \left[ \left( 1 - \frac{M}{L-1} \right) * \exp(-\lambda RN) \right]$

For the delivery delay:  $\frac{E \left[ \frac{L-1-M}{N} \right]^2}{\lambda * (L-1) * (L-1-\bar{m})}$

**History Based Drop (HBD)**

Expectation calculated by summing over old messages.

# Experimental results: Setup

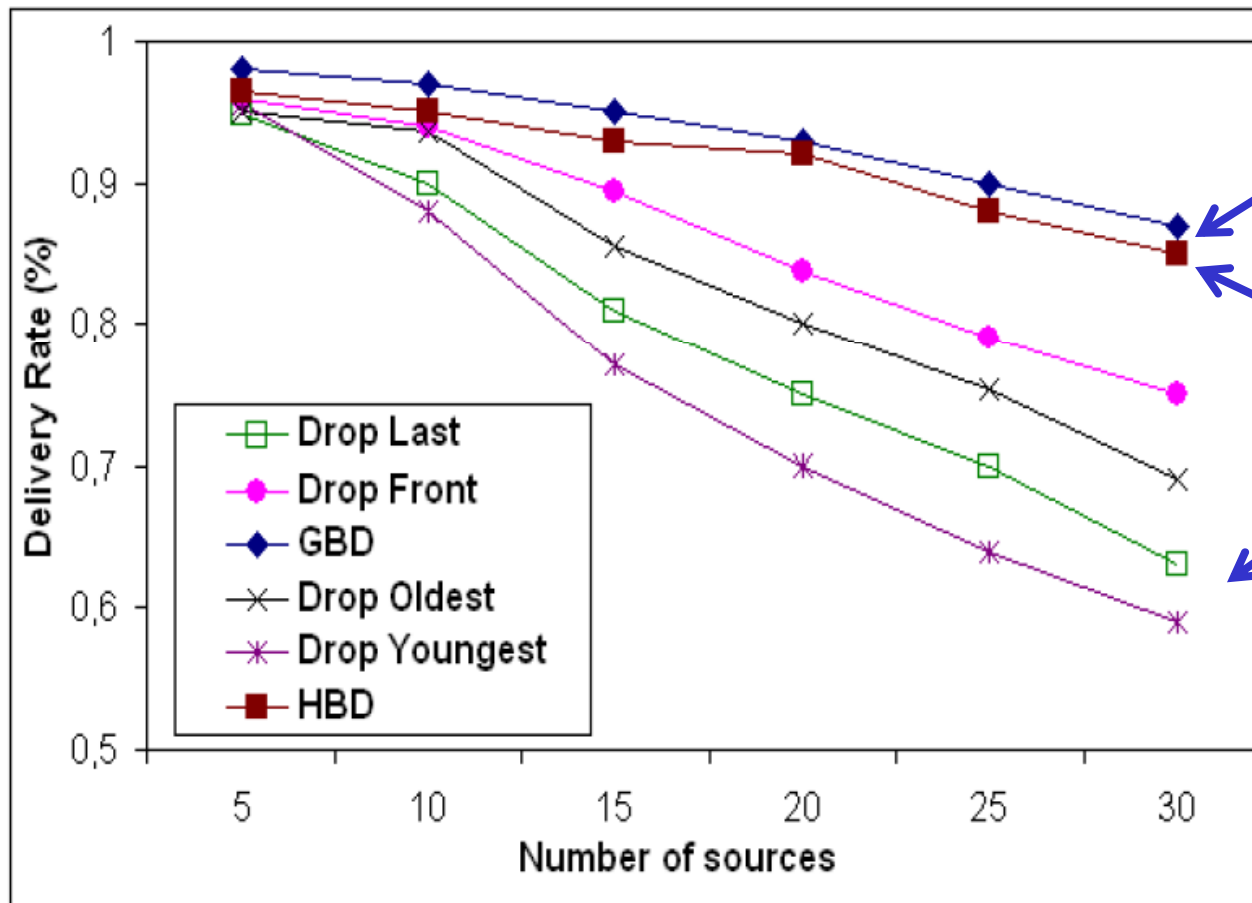
<b>Mobility model</b>	<b>Random Waypoint</b>	<b>Traces du projet ZebraNet</b>	<b>Traces du projet Cabspotting</b>
<b>Simulation duration (s):</b>	<b>5000</b>	<b>5000</b>	<b>36000</b>
<b>Simulated Surface (m<sup>2</sup>):</b>	<b>1000*1000</b>	<b>1500*1500</b>	<b>-</b>
<b>Number of nodes:</b>	<b>30</b>	<b>40</b>	<b>40</b>
<b>Average speed (Km/h) :</b>	<b>6</b>	<b>-</b>	<b>-</b>
<b>TTL (s) :</b>	<b>650</b>	<b>650</b>	<b>7200</b>
<b>Intervalle CBR (s) :</b>	<b>200</b>	<b>200</b>	<b>2100</b>

**DTN architecture added to the ns-2 simulator**

**MAC = 802.11b, range=100m, CBR sources, message size = 1Kbytes  
random sources and destinations**

# Global Delivery Rate

## Random Way Point

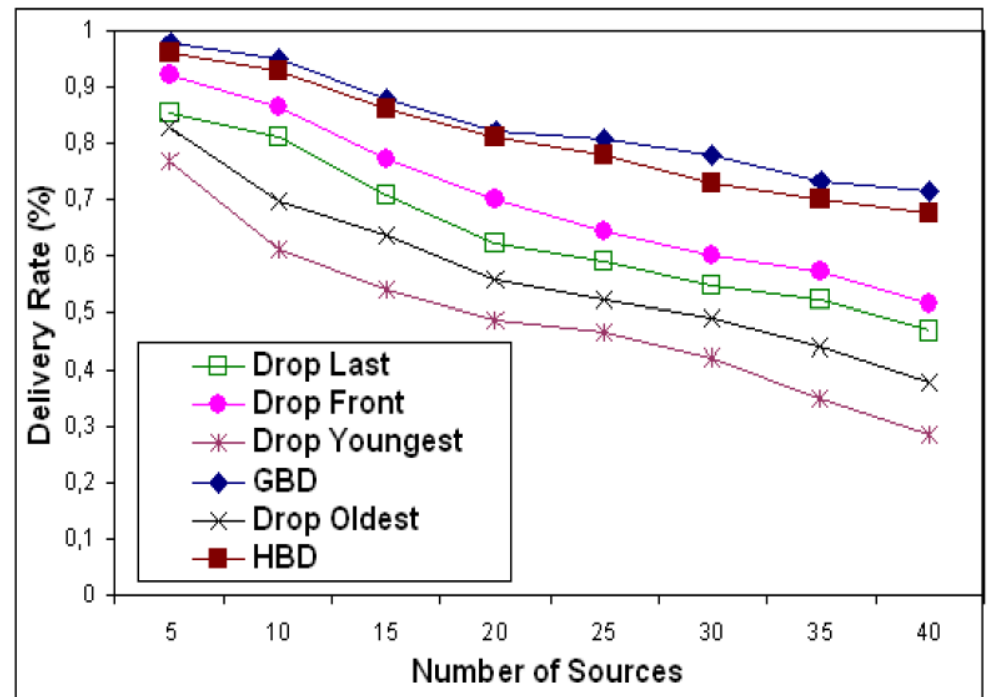
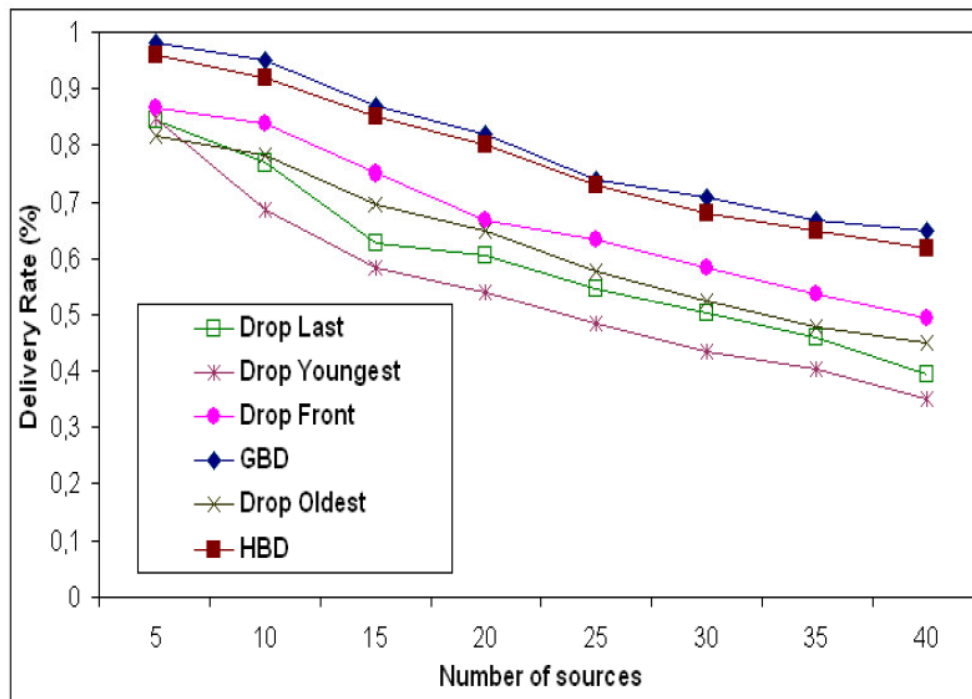


Very close

Almost  
50%  
gain over  
DropTail

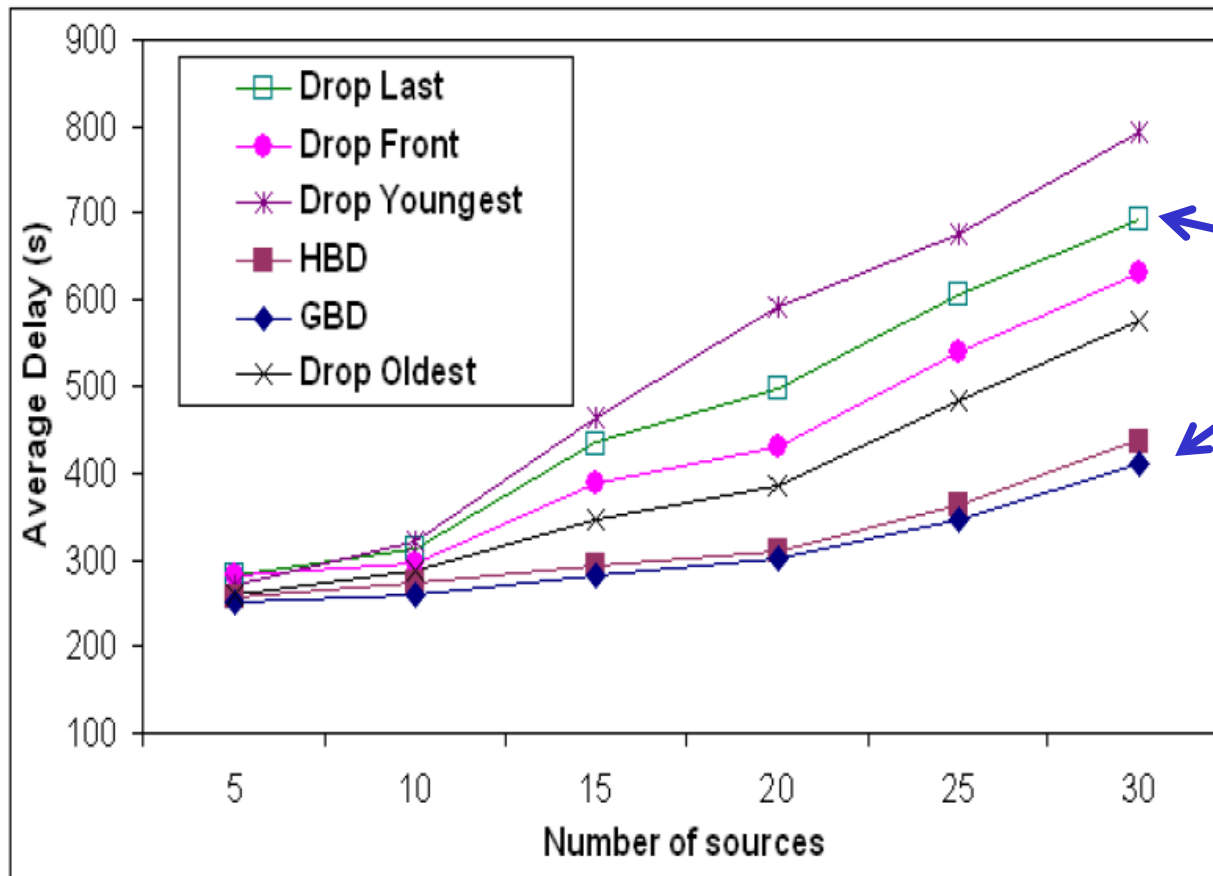
# Message Delivery Rate

## Real Traces



# Global Delivery Delay

## Random Way Point

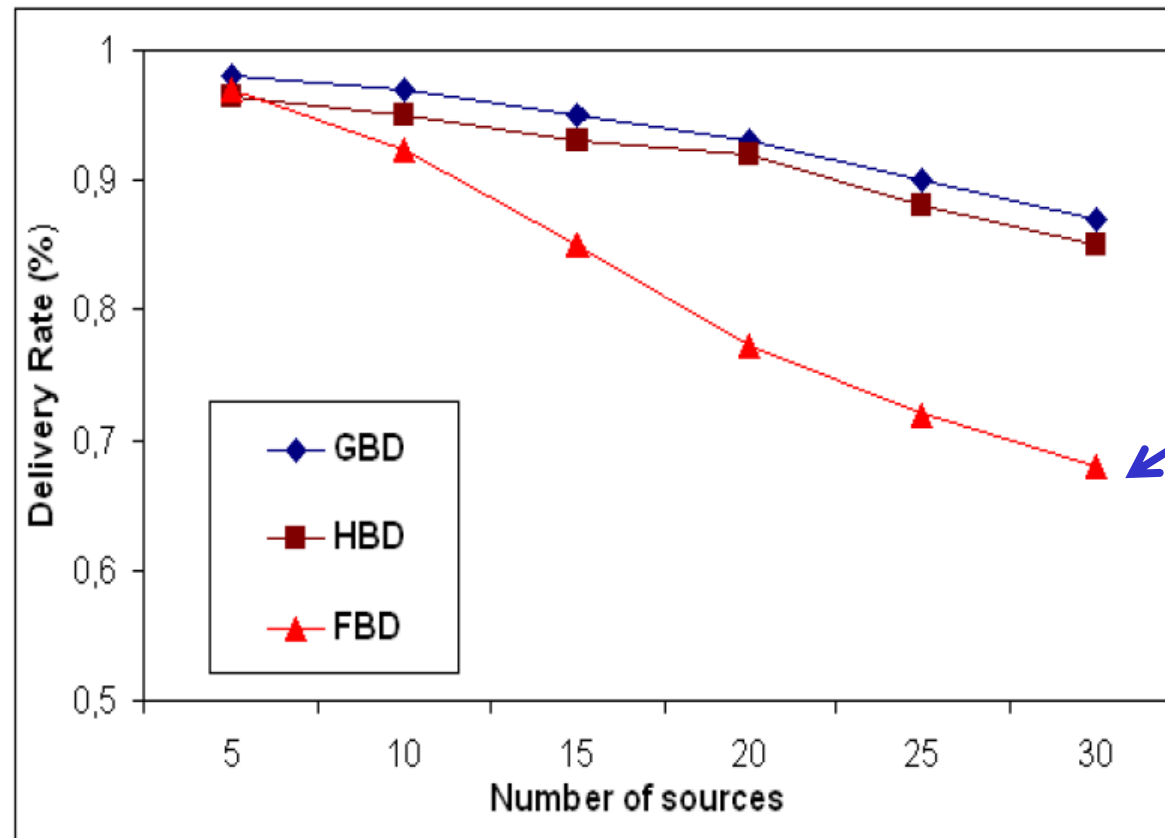


Again,  
almost  
50% gain



# Flooding vs. History

## Random Way Point



RAPID-like

# Conclusions

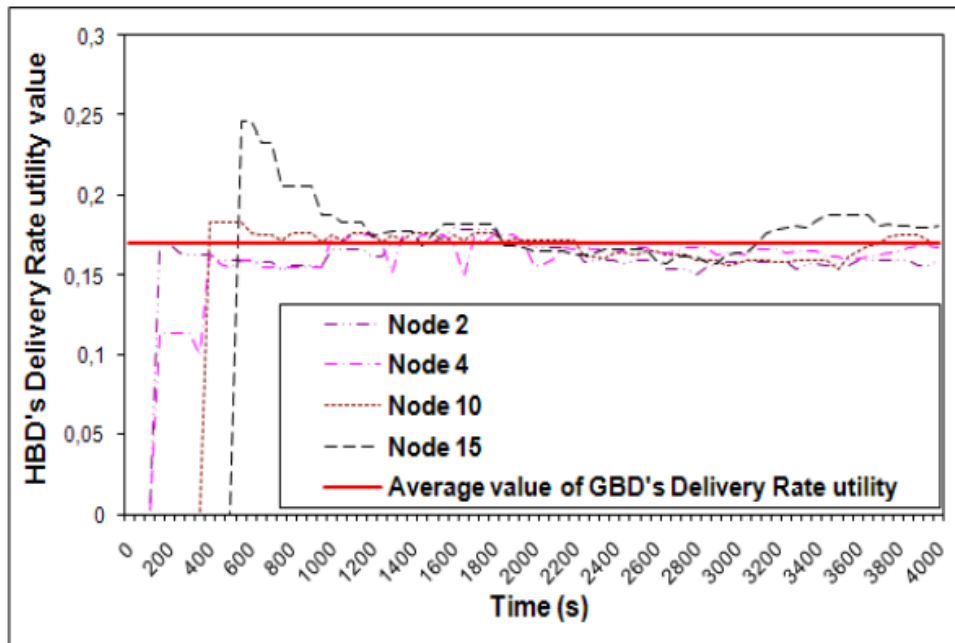
- ❑ Analytical framework for optimal message drop in DTNs
- ❑ Two policies for the cases of delivery rate and delivery delay
- ❑ A distributed version that WORK
- ❑ Validation with a synthetic mobility model and real traces
  
- ❑ Many issues still open e.g., find more global metrics and account for factors as the non stationary of the scenario, the variability of the number of nodes etc.

Thanks for your attendance !

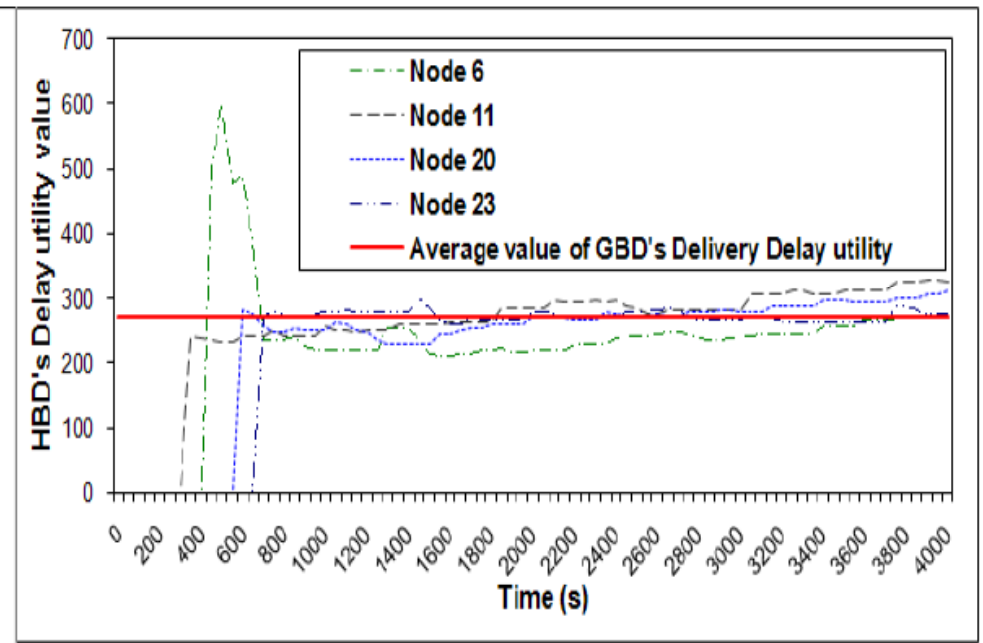
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# Utility values convergence

Small messages of 1Kbytes each (Random Way Point) (for elapsed time  $T = 100s$ )



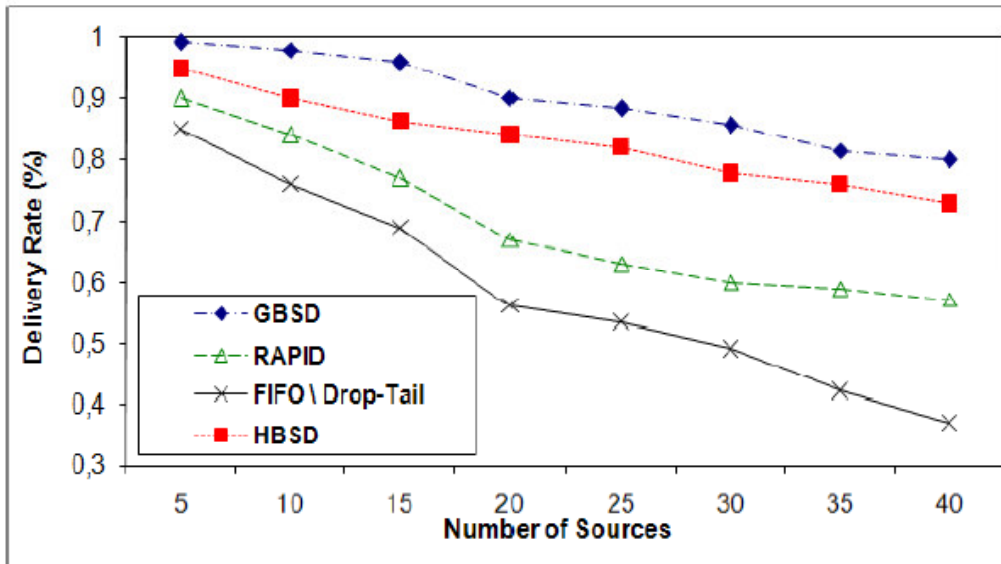
Message Utility for delivery rate



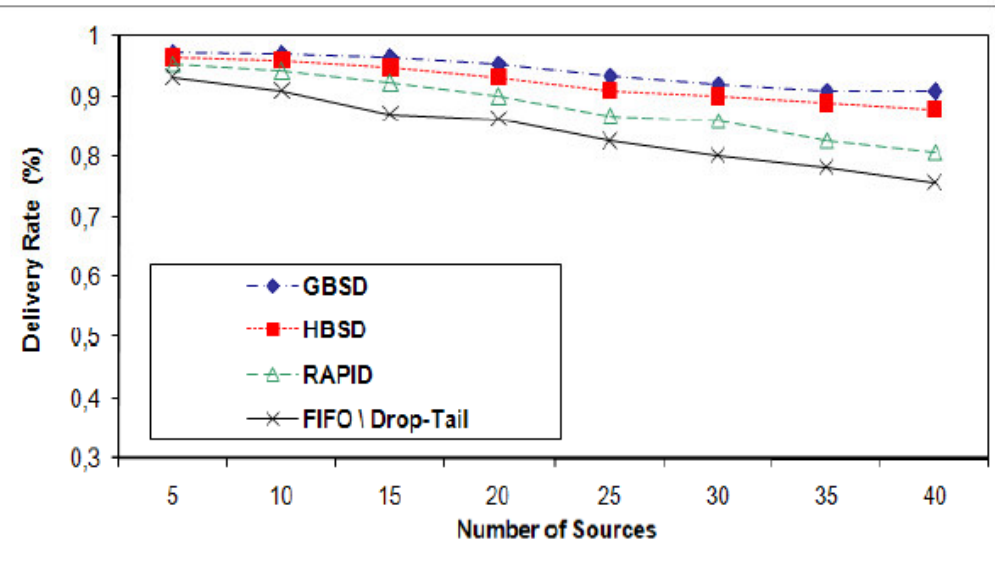
Message Utility for delivery delay

# Bandwidth limitation: Delivery

Large messages of 85 Kbytes each (Taxi trace)



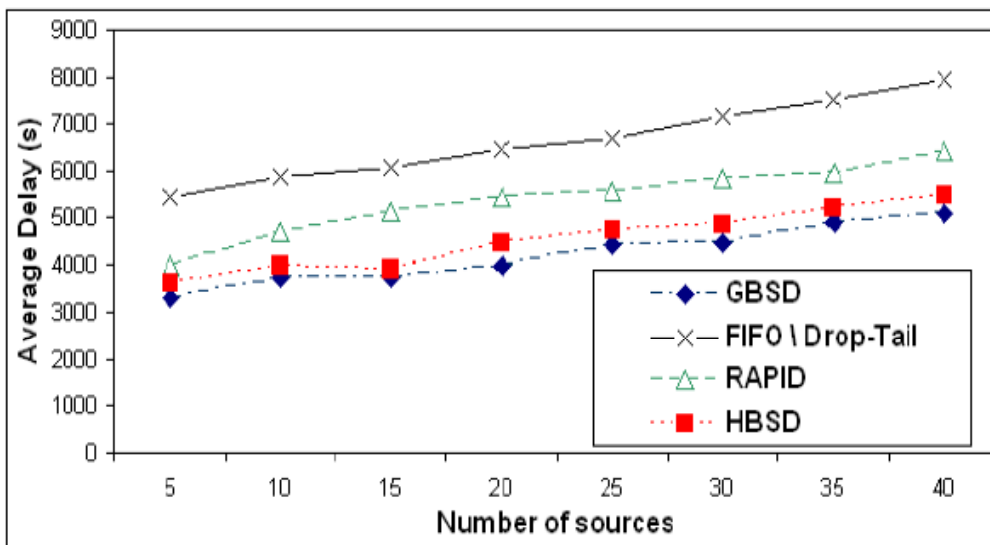
Limited Buffer



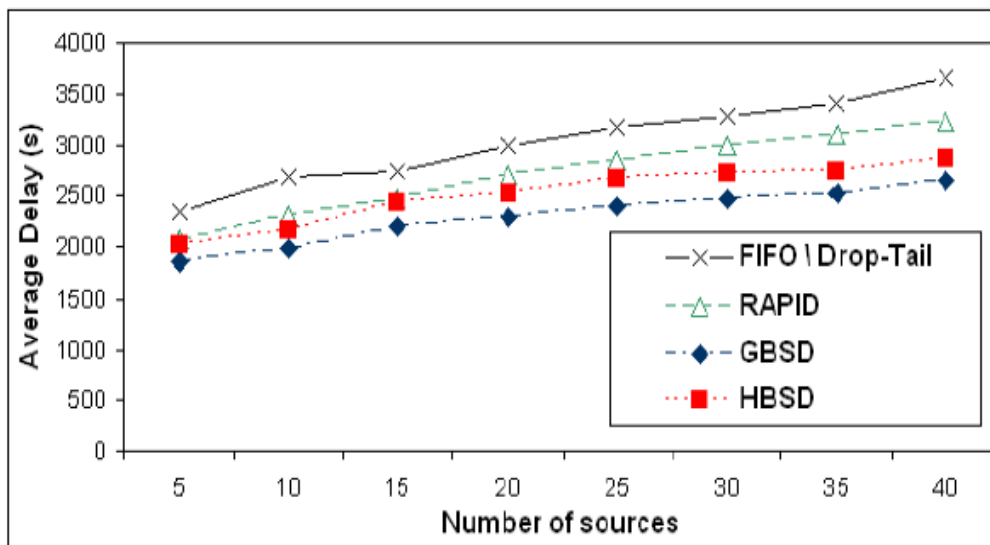
Unlimited Buffer

# Bandwidth limitation: Delay

Large messages of 85 Kbytes each (Taxi trace)



Limited Buffer



Unlimited Buffer

For more details on the bandwidth limited case:

Amir Krifa, Chadi Barakat, Thrasyvoulos Spyropoulos, "An Optimal Joint Scheduling and Drop Policy for Delay Tolerant Networks", to appear in proceedings of the WoWMoM Workshop on Autonomic and Opportunistic Communications, Newport Beach (CA), June 2008.