

Flooding Time in Markov-*Dynamic* Networks

Andrea **Clementi** (*Speaker*)

Claudio **Macci**

Francesco **Pasquale**

University of “**Tor Vergata**”
Rome

Angelo **Monti**

Riccardo **Silvestri**

University of “**La Sapienza**”
Rome

The “dynamic”-world challenge

MAIN RESEARCH GOAL:

Provide a *Mathematical Approach* for

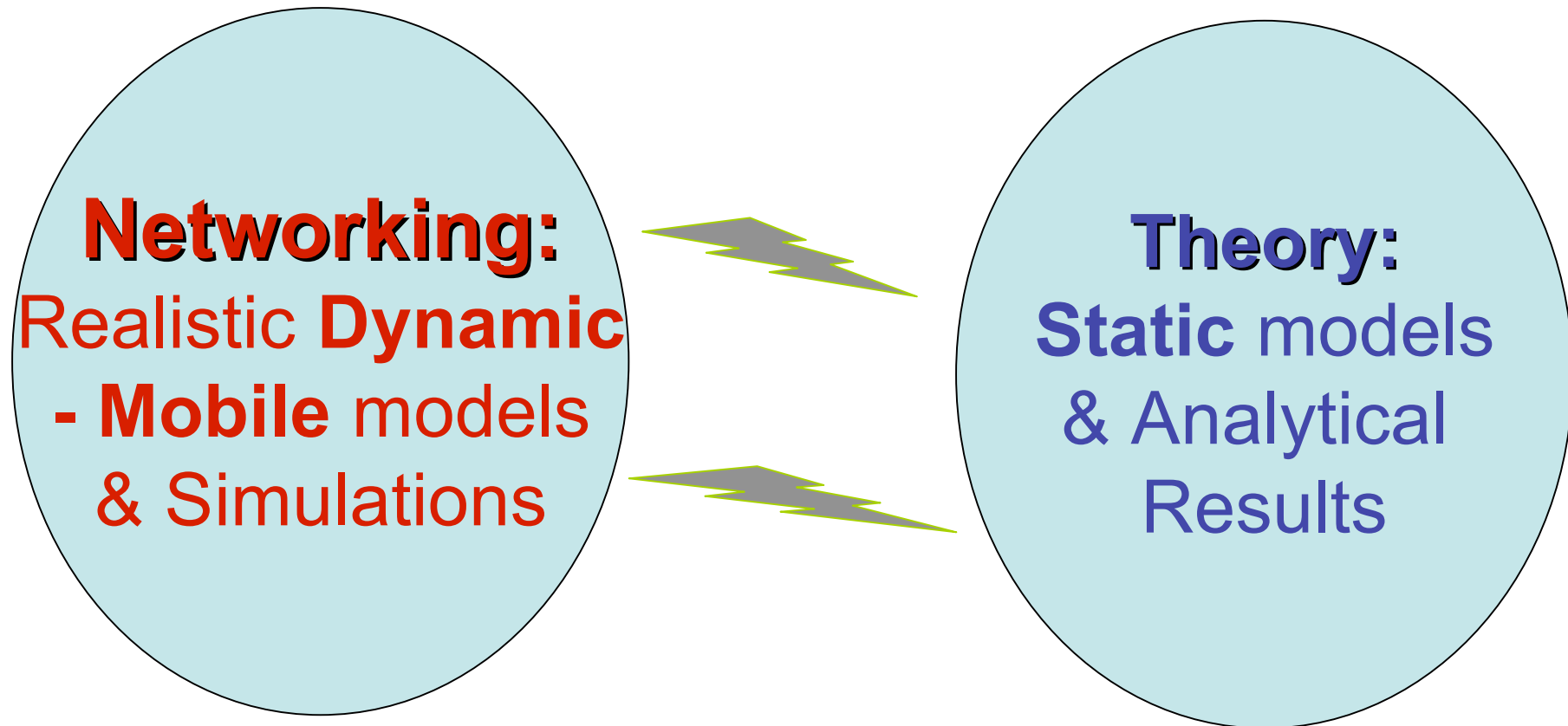
Information Spreading

on

EVOLVING (RADIO) NETWORKS

Practice vs Theory

- Scenario before 2001



Dynamic Networks

A ***dynamic*** network is a sequence of graphs $G(V, E(t))$ ($V = [n]$) $t = 0, 1, \dots$ where

edge sets $E(t)$ may **change**

during the execution of a protocol

$E(0), E(1), \dots, E(t), \dots$

Dynamic Networks

The *Dynamic Network* model is very general: the edge-set sequence can be determined by any **law/adversary**:

- deterministic **worst-case** (Adversarial Networks)
- **random** oblivious changes
- node **mobility** (geometric models)

Theor. Research Advances (2001-...)

2008-..

Markov **Geometric**
RND Changes

This Talk
PODC'08

Markov Time-Depend.
RND Changes

2007

Worst-c.
Changes in
Unkn. topology

Fully-Indep. RND
Changes

2005-6

Worst-c. faults
In Known
topology

2001

Random Dynamic Networks

Previous **analytical** results on communication protocols concern dynamic random models assuming

Fully Spatial and Time Independence

over the **edge set**

Dynamic Networks

- **Goal:**
 - a) Introduce *Stochastic Time-Dependency* on the edge-set sequence
 - b) Provide **analytical** bounds on **Flooding Time**

Markov Dynamic Graphs

We consider the **discrete** version of the

Reciprocity-Graph Model [Wasserman 80]

Evolving Social Networks

Markov-Dynamic Networks

Let $\mathbf{E}(0)$ be any initial edge set and let \mathbf{V}^2 be the set of all possible (undirected) edges.

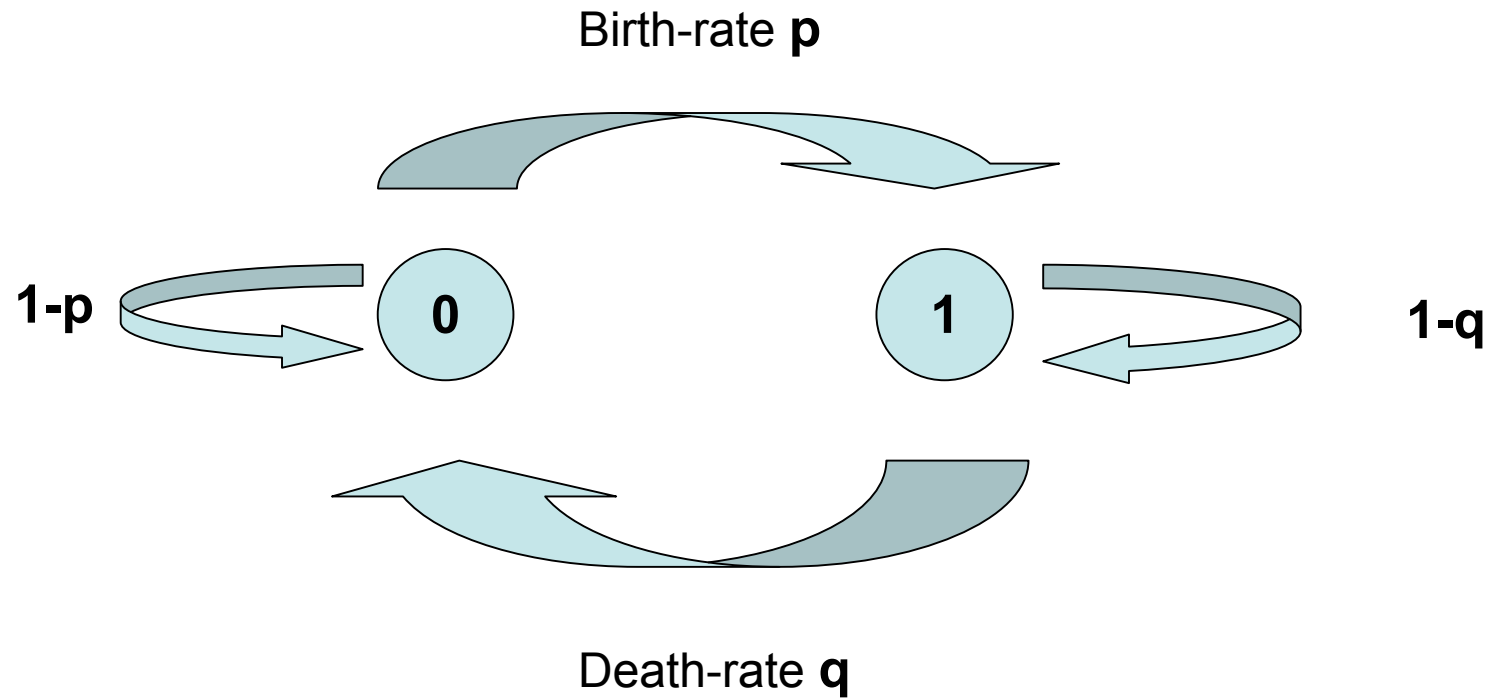
At every time step $t \geq 1$, the existence of any edge $e \in \mathbf{V}^2$ is determined by a *Markov* chain:

- If $e \in \mathbf{E}(t-1)$ then $e \notin \mathbf{E}(t)$ with probability q
- If $e \notin \mathbf{E}(t-1)$ then $e \in \mathbf{E}(t)$ with probability p

$q = q(n)$ is the edge *death*-rate

$p = p(n)$ is the edge *birth*-rate

Markov-Dynamic Graphs



Markov-Dynamic Graphs

Remarks.

- There is *independence* among edges
- There is **Markov dependence** in the time-sequence of every edge.

Markov-Dynamic Graphs

Average node degree in the *stationary* graph distribution

$$d(v) = (n-1) (p / (p+q))$$

- $p = k/n$ where $k = o(n)$ and $q = 1/2$ **then**

$$d(v) \approx k$$

- $p = q = k/n$ where $k = o(n)$ **then**

$$d(v) \approx n$$

So q seems to play a crucial role.....

Special Markov-Dynamic

(Independent) Dynamic-Random Graphs

$$p = 1-q$$

At every time step,

$$G(V, E(t)) \approx G(n, p) \quad (\text{Random Graphs})$$

This is similar to the *Kleinberg et al* dynamic model where, at **every** time step, every node chooses a set of **$k=p(n-1)$** neighbors, independently at random

Special Markov-Dynamic

Densifying networks: $q = 0$

An edge will never die.

Static Networks: $p=q=0$

Flooding

Broadcast Task: A **source** node **s** aims to send a message **m** to all nodes. A node is **informed** if it has received **m**.

Flooding Mechanism: at every time step, every **informed** node sends **m** to all its neighbors.

Flooding

Flooding time is

*the number of time steps required to
inform all nodes.*

Flooding

- In unknown, highly-dynamic networks, **Flooding** is often the “only” simple protocol for broadcasting.
- Flooding time is a natural **lower** bound: it used to **evaluate** other broadcast protocols
- **Flooding** in ***Dynamic*** Networks



Diameter in ***Static*** Networks

Our Results

We determine tight bounds on **Flooding Time**

$$T = T(n, p, q, E(0))$$

on Markov-Dynamic Graphs,

for *all* values (functions) of p and q .

Flooding Time

Upper Bound. For any $0 \leq p, q \leq 1$ and for any $E(0)$,

$$T = O(\log n / \log(1+np))$$

Corollary. Let $p = \Omega(1/n)$ and any q , then Flooding Time is $O(\log n)$ starting from any initial edge set

Previous Related Results

- **fully-independent** random dynamic networks.

Flooding Time is $O(\log n)$ for $k = \Theta(1)$
random neighbors.

[KempeKleinberg02],[Pittel87]

Same result holds in $G(n, p)$ ($p = 1 - q \geq 1/n$)

Flooding Time

Consequence 1:

Under very general assumptions ($p \geq 1/n$,
any q)

Markovian time-dependency over the edge set **does not (asymptotically) decrease** the speed of data propagation in dynamic **random** graphs.

Flooding Time

Important Consequence 2:

(almost dense graphs)

When $p \geq 1/n^\delta$ with fixed $0 < \delta < 1$ and any q ,

Flooding Time is constant and so

it does not depend on the initial topology.

Flooding Time

The Lower Bounds.

- For any $0 \leq p, q \leq 1$ and starting conf. $E(0) = \circ$

$$T = \Omega(\log n / np)$$

- For any $p \geq \log n / n$, $0 \leq q \leq 1$ and $E(0) = \circ$

$$T = \Omega(\log n / \log(1+np))$$

Flooding Time

Note. Flooding time can be *slow*:

$$p \approx 1/n^2 \quad \text{then} \quad T = \Theta(n \log n)$$

Flooding Time

Matching Bounds.

Upper and Lower bounds are **tight** when

$$p \leq 1/n \quad \text{OR} \quad p \geq \log n / n$$

for any $0 \leq q \leq 1$

Death-rate q may play a relevant role only in
range

$$1/n \leq p \leq \log n / n$$

Flooding Time

Range: $1/n \leq p \leq \log n / n$ (by varying q)

- $1/n \leq p \leq \log \log n / n$
 $\log n / \log \log \log n \leq T \leq \log n$ (MAX)
- $\log \log n / n \leq p \leq \log n / n$
 $\log n / \log \log n \leq T \leq \log n / \log \log \log n$
(MIN)

Flooding Time: Densifying Networks

In range $1/n \leq p \leq \log n / n$ we consider case $q=0$ (an edge never dies)

- $T = \Theta(\log n / np)$ when $p \leq \log \log n / n$
- $T = \Theta(\log n / \log \log n)$ when
(plateau!) $\log \log n / n \leq p \leq \log n / n$

Upper Bound: Proof's Overview

Classic approach in *static* random graphs:

(Flooding = Diameter)

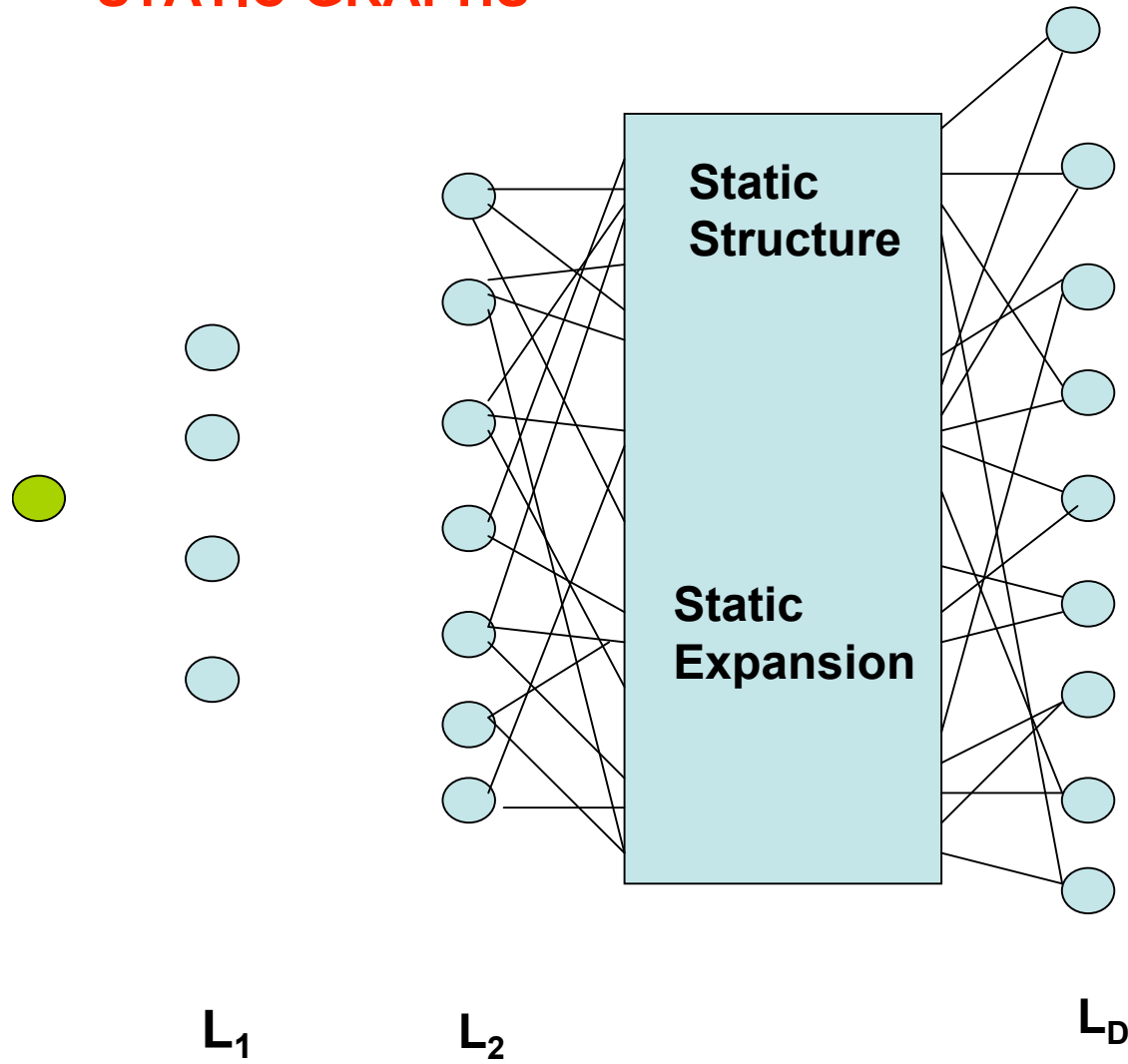
- a) Prove (edge/vertex) *expansion* properties of the graph
- b) Use such properties to evaluate the *height* of BFS Trees

Upper Bound: Proof's Overview

In our **dynamic world** we

- a) Define and prove **time/dynamic expansion** properties of a graph sequence
- b) Use them to evaluate, at every step, the number of **new informed** nodes.

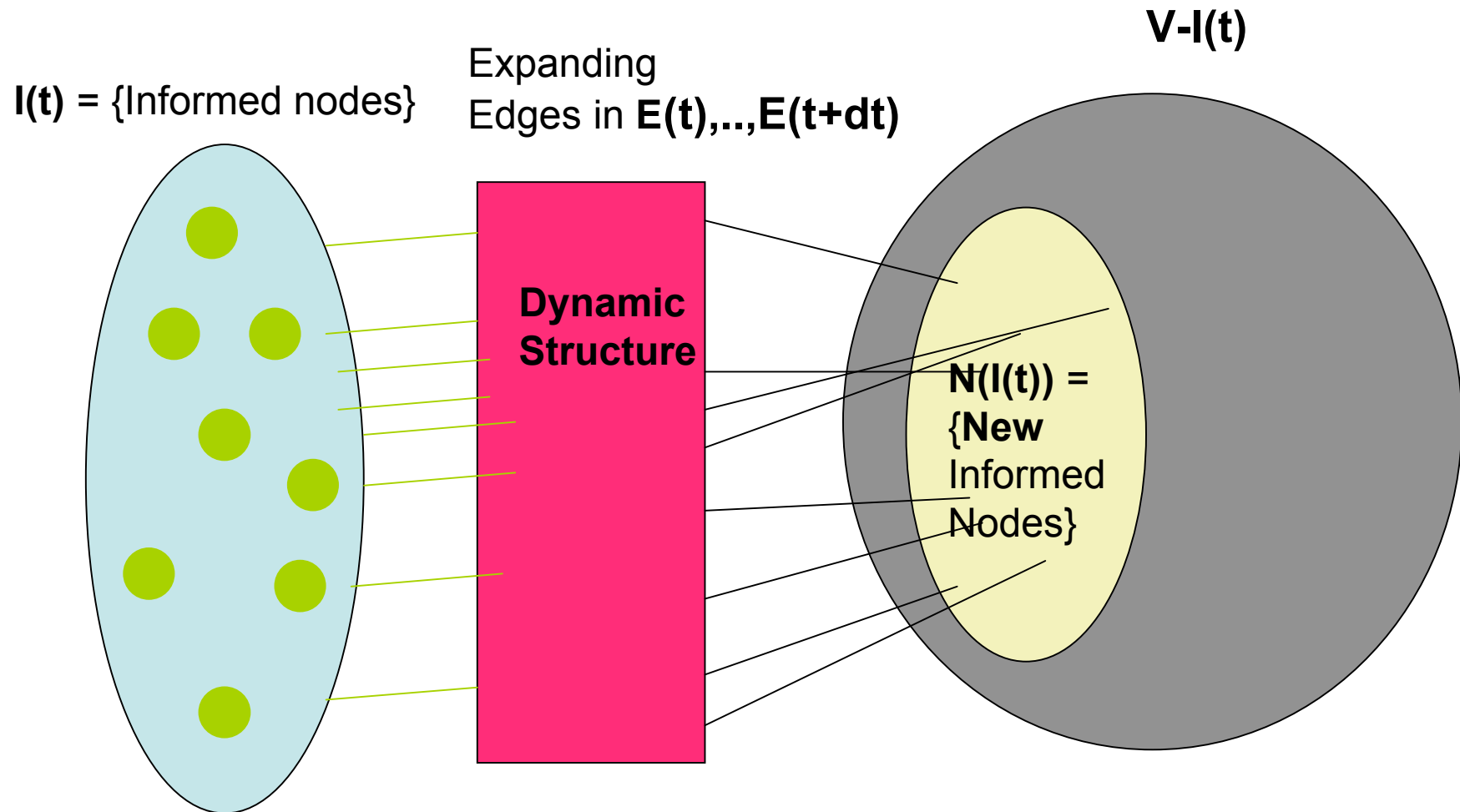
STATIC GRAPHS



Apply **Static**
Expansion between
fixed Levels

Flooding Time
|||
Diameter **D**

Dynamic Graphs: Picture at time t



Conclusions

- We introduce a *Markovian-Dynamic Graph model*
- Derive a *dynamic* version of **expanding** properties of random graphs
- Apply them to analyze the **flooding** process

Future Work

“Immediate” technical questions:

- Tight bound in range $1/n \leq p \leq \log n / n$
- Starting from the *Stationary Distribution*

Current (Challenging) Work

- Introduce *spatial* Markovian-dependency by means of
simple/random node-mobility
- Adopt the same approach:

Dynamic expansion+ Flooding Phases