



# Local approaches for community detection in complex networks:

*Preference networks,  
boundary nodes and  
gossip propagation*

**Mürsel Taşgın**

Department of Computer Engineering

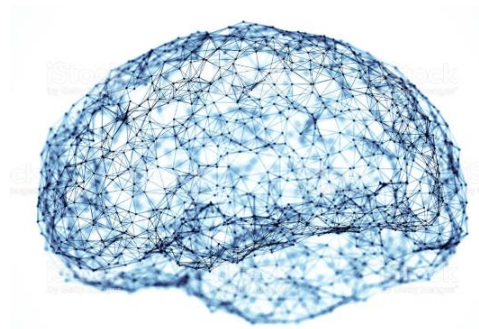
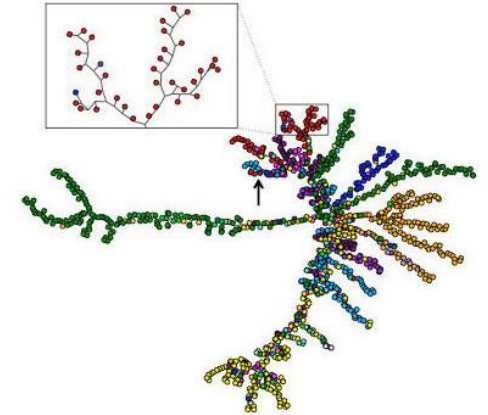
Boğaziçi University



## What is a network?

A collection of entities (nodes) that are interconnected with links (edges).

- people that are friends
- group of interconnected computers
- brain cells that have similar functions
- web pages pointing to each other
- proteins interacting together
- phone calls between people
- authors collaborating on same paper
- airline traffic connections
- underground map of cities



# What makes a network a “complex network”?

- Many agents or individuals interacting with each other
- Each component has its own internal structure and each one perform an specific behavior or function
- The behavior of a small part of the system affects in a non linear way the whole system.

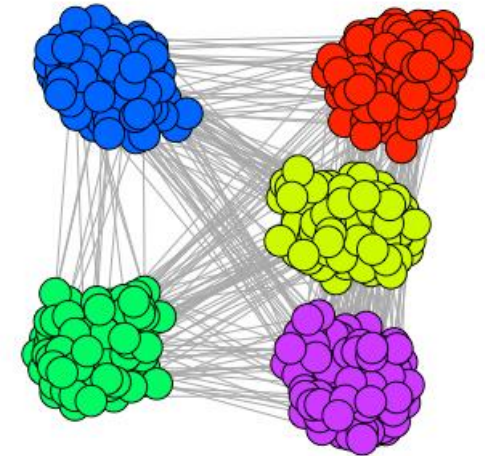
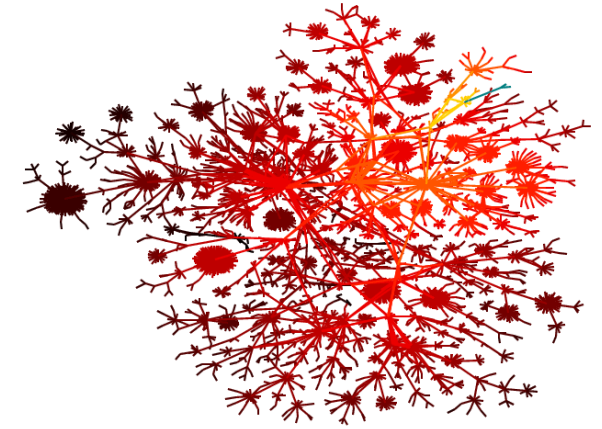
Complex networks present emergent behavior in such a way that any property of the system is not the **simple sum of its parts**.





# Community and community detection

- **Community**  
Group of nodes more connected within community than rest of the network
- **Community detection**  
Task of identifying communities in networks



# Community detection algorithms

Divisive algorithms

GN, centrality-based algorithms

Agglomerative algorithms

Newman's greedy algorithm

Information theoretic approaches

Infomap

Some examples of community detection algorithms

Using AI methods

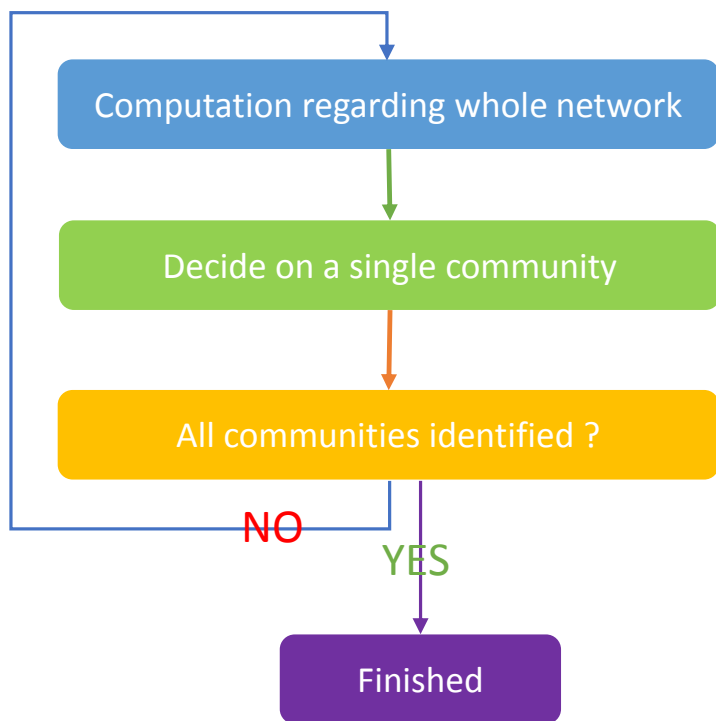
Community detection using genetic algorithm

Label propagation

LPA

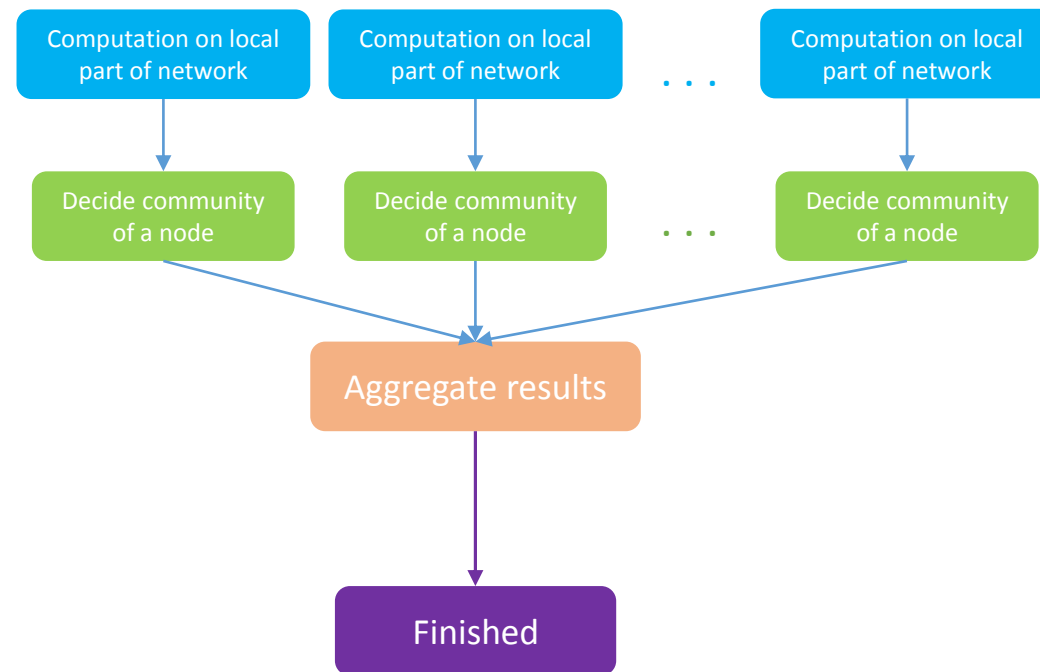
# Global approach vs local approach

## Global approach



- + Decisions with more information
- High time-complexity
- No parallelism, not distributed

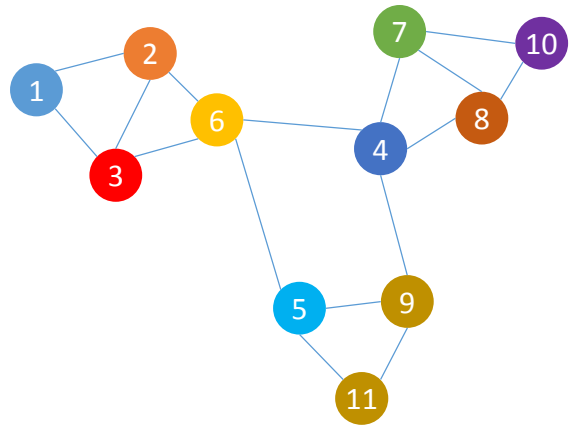
## Local approach



- Decisions with less information
- + Low time-complexity
- + Parallel and distributed execution



# Community detection using preference networks



Original network

1

Ask each node

With whom do you prefer to be in same community?

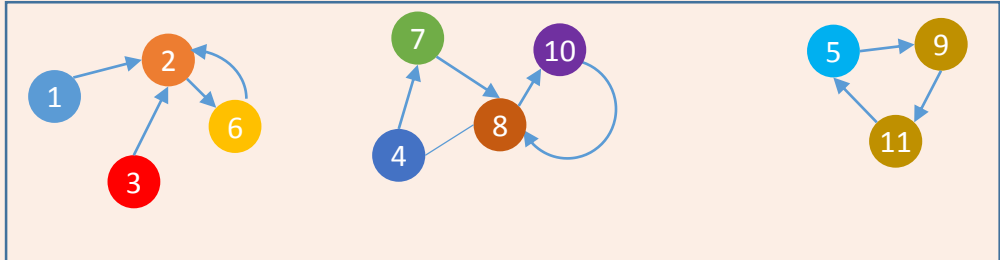
1	2	3	4	5	6	7	8	9	10	11
2	6	2	7	9	2	8	10	11	8	5

Node ID

Preferred node ID

2

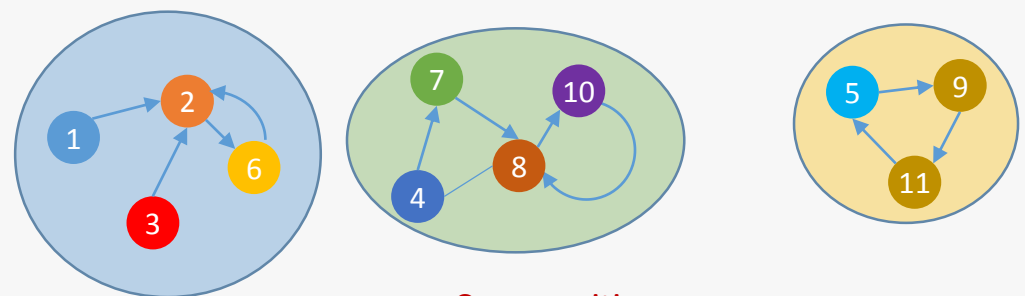
Build a directed «preference network» using answers



Preference network

3

Identify «connected components»; *i.e.* communities



Communities



Tasgin, Mursel, and Haluk O. Bingol. "Community detection using preference networks." *Physica A: Statistical Mechanics and its Applications* 495 (2018): 126-136.

## Deciding the preferred node

Ask each node

With whom do you prefer to be in same community?

### How to decide the preferred node?

Given a node, assign a score to each neighbor and select the neighbor with highest score

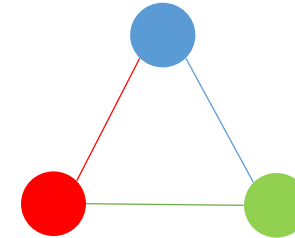
$$p(i) = \arg \max_{j \in A(i)} s_i(j)$$

### Score methods:

- Random number
- Degree of neighbor
- Clustering coefficient of neighbor
- Jaccard similarity
- Number of common neighbors with neighbor
- Gossip spreading capability of neighbor

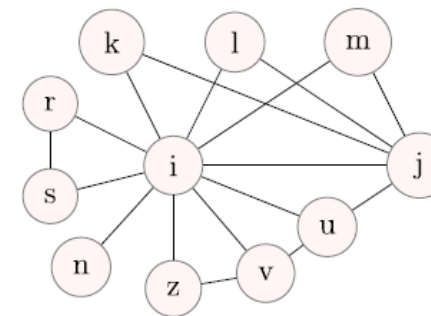


## Triangles and communities



"Strong social tie forms when being part of a triangle in a relation", Simmel 1950

"There exists many triangles within communities while few or no triangle exists between nodes of different communities", Radicchi et al. 2004



### Triangle related metrics

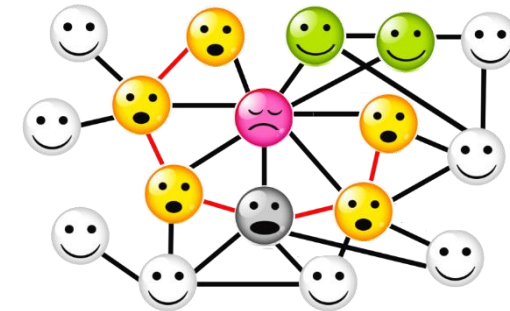
- clustering coefficient
- # common neighbors
- Jaccard similarity
- Gossip spread capability



# Community detection using preference networks

# triplets where $i$ is the center	$\wedge_i$
# triangles around $i$	$\Delta_i$
1-neighborhood of $i$	$\Gamma(i)$
clustering coefficient of $i$	$CC(i) = \frac{\Delta_i}{\wedge_i}$
# common neighbors of $i$ & $j$	$CN(i, j) =  \Gamma(i) \cap \Gamma(j) $
Jaccard similarity of $i$ & $j$	$J(i, j) = \frac{ \Gamma(i) \cap \Gamma(j) }{ \Gamma(i) \cup \Gamma(j) }$
# neighbors of $i$ who hears gossip initiated by $j$	$\Gamma_j(i)$
Gossip spread capability of $j$ around $i$	$\sigma_{i,j} = \frac{ \Gamma_j(i) }{ \Gamma(i) }$

## Gossip spread capability

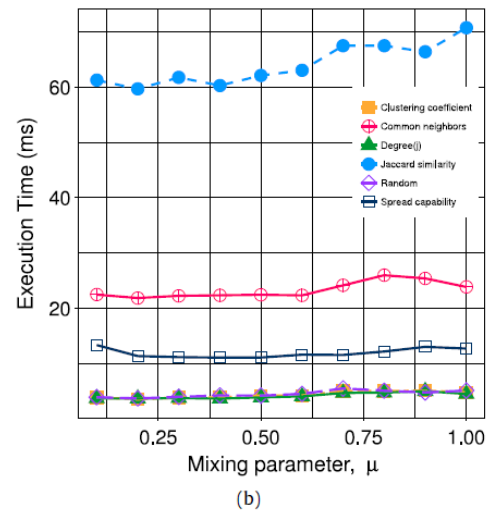
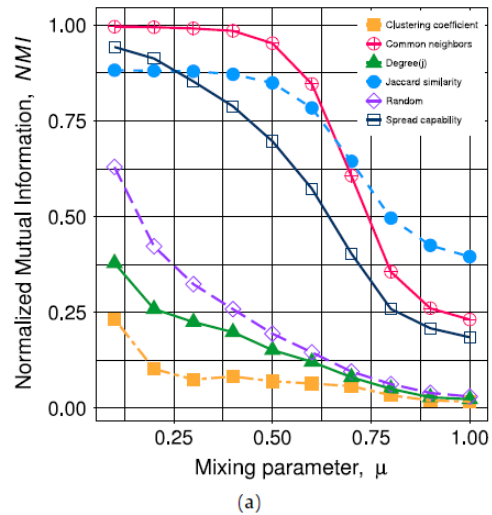


\* Gossip algorithm of Lind et al.

Spread capability of gray node for pink node: **6/8**



## Results on LFR generated networks of 1000 nodes



Different preference metrics

## Findings

- Degree and clustering coefficient score metrics are not good
- Random score is surprisingly better than clustering coefficient
- # common neighbors produces best result
- Jaccard similarity and spread capability also produce good results (Jaccard similarity have high execution time)

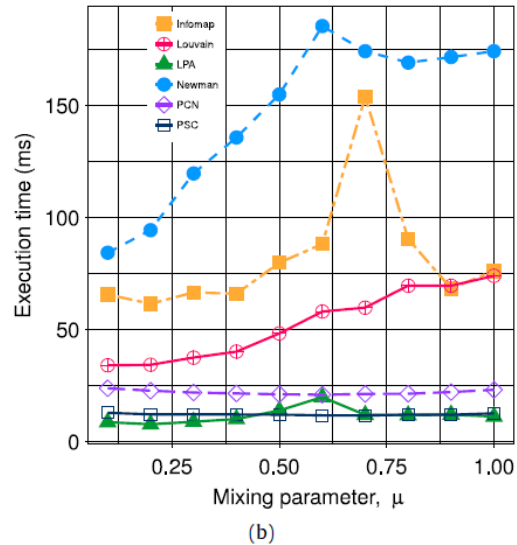
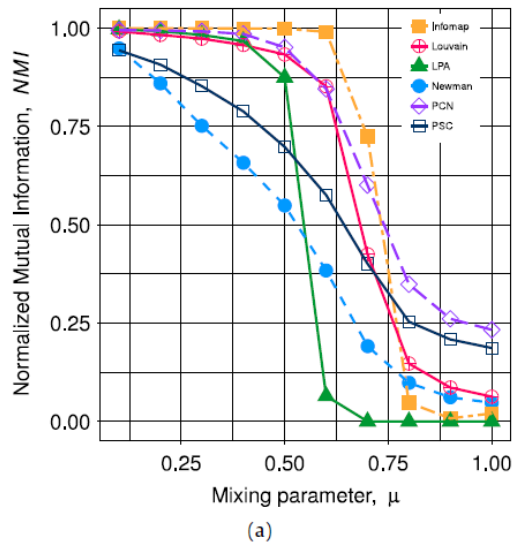
## Conclusion

We select the following 2 score methods for our algorithm

- # common neighbors (PCN)
- Spread capability (PSC)



## Results on LFR generated networks of 1000 nodes



Comparison with some known algorithms

## Results

- Successfully identify dense communities (i.e.  $\mu < 0.5$ )
- Can identify subtle communities (i.e.  $\mu > 0.5$ )
- Have less execution times compared to most of the algorithms



# Community detection using boundary nodes

## A new algorithm based on label propagation algorithm (LPA)

LPA: Each node gets most popular label (*i.e. community identifier*) among the neighbors

## Unnecessary operations of LPA is eliminated

Label update is possible only when interacting nodes have different labels

After a few iterations most of the nodes are surrounded by same labelled nodes

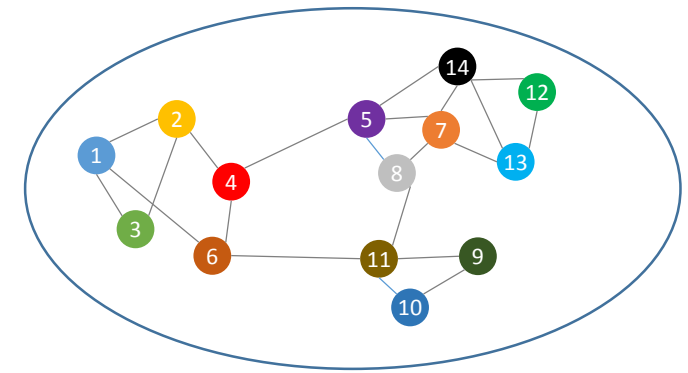
LPA should be applied to only **boundary nodes** (*i.e.* Node having at least 1 neighbor with different label)

Algorithm finds boundary nodes and their correct community label

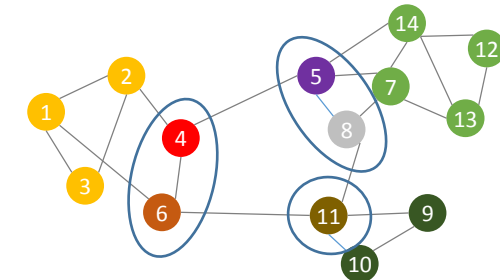
Boundary nodes will divide the network into **communities**

## Label selection rule is enhanced with additional data

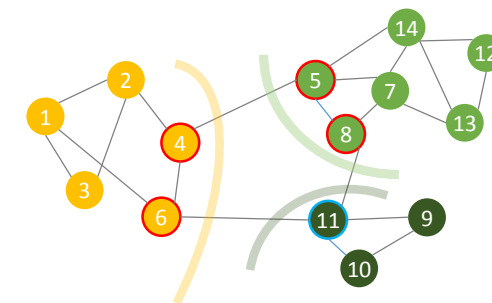
Majority rule + **# common neighbors**



Label propagation on all nodes



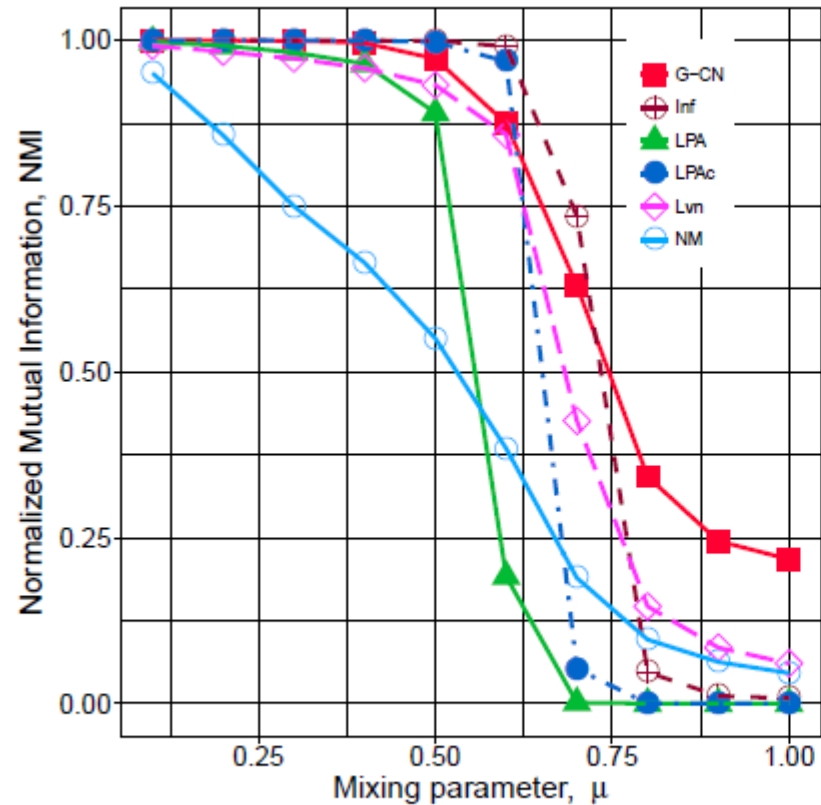
Label propagation only on boundary nodes



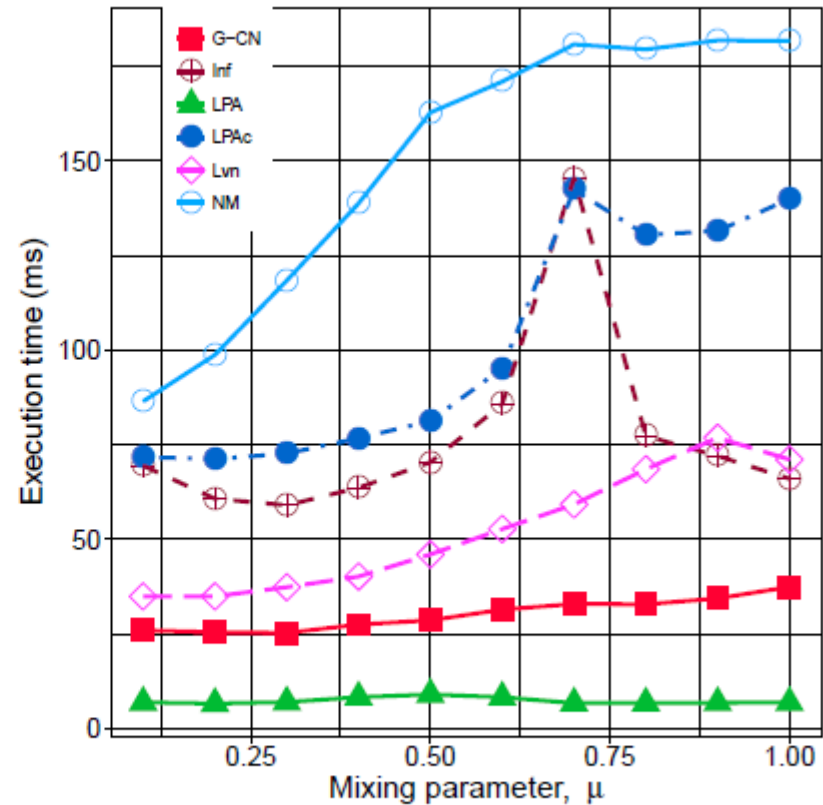
Communities identified



Comparison with known algorithms on LFR generated networks of 1000 nodes



(a)



(b)

