

Community Detection and Interpretation

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☰ Education

- B.Sc. Galatasaray University, Computer Engineering, Istanbul, 2005
- M.Sc. Galatasaray University, Computer Engineering, Istanbul, 2010
- Ph.D. INSA de Lyon, Computer Science&Informatics, Lyon, 2014

☰ Experiences

- Database Manager, Is Bankasi, Istanbul, 2005-2006
- Researcher, Tübitak, Gebze, 2006-2009
- Research and Teaching Assistant, Galatasaray University, Computer Engineering, Istanbul, 2009-2016
- Research Consultant, Vodafone, Istanbul, 2014-2015
- Assistant Professor, Galatasaray University, Computer Engineering, Istanbul, 2016-...

Previous Studies



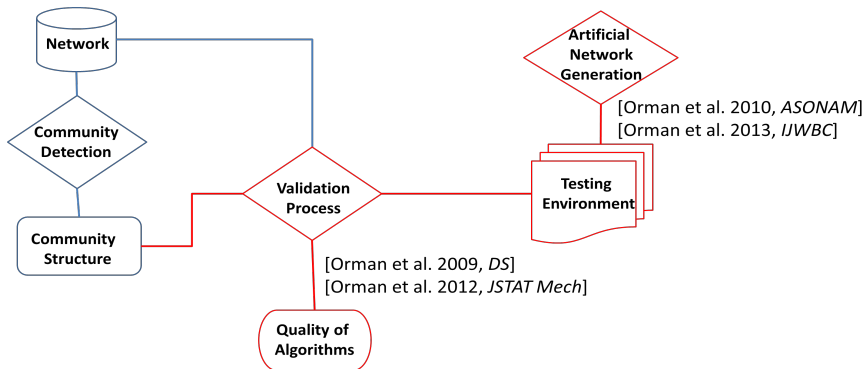
[Fortunato 2010, *Phys. Report*]

[Zhou et al. 2009, *VLDB*][Yang et al. 2013, *ICDM*]

[Aynaud et al. 2011, *Dynamics on and of Complex Networks Vol. 2*]

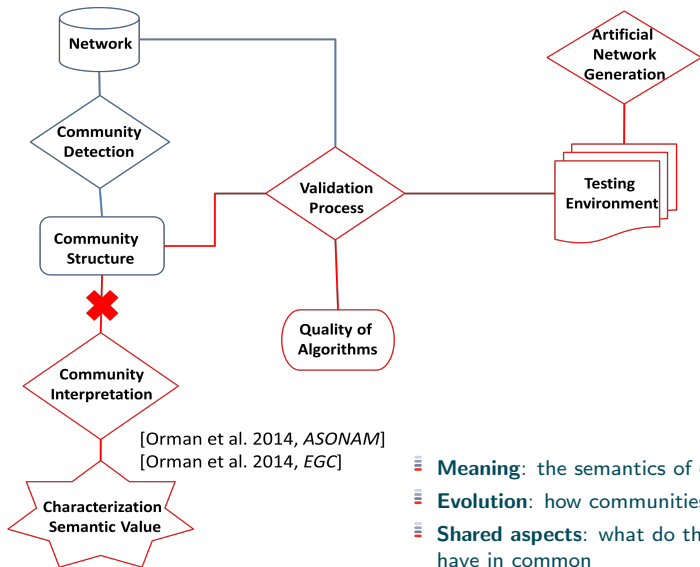
- ≡ Hundreds of methods for plain networks
- ≡ Considering homogeneity for attributed networks
- ≡ Consensual or evolving community structures for dynamic networks

Previous Studies



- ≡ Artificial networks with realistic properties
- ≡ Evaluating algorithms performance by using artificial networks

Previous Studies



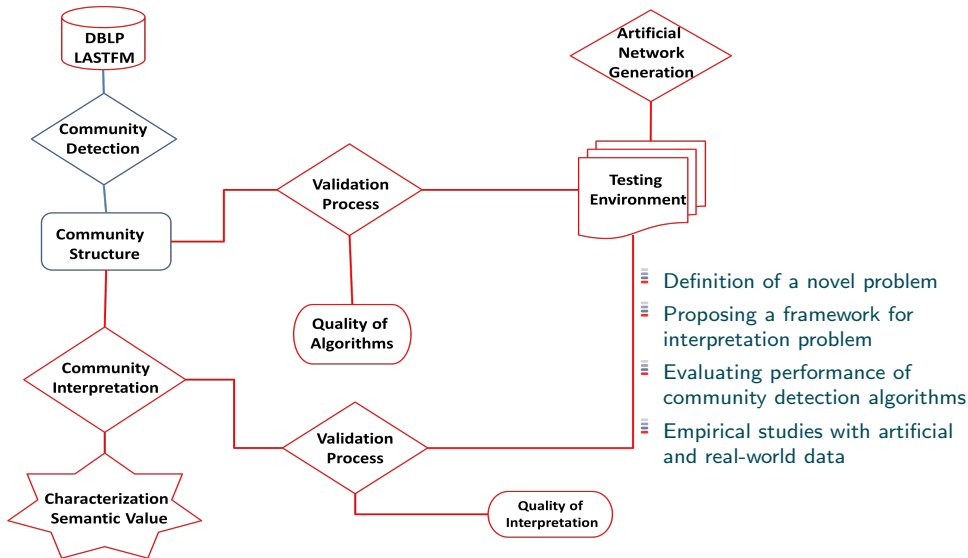
Community in Sociology

Groups of people sharing

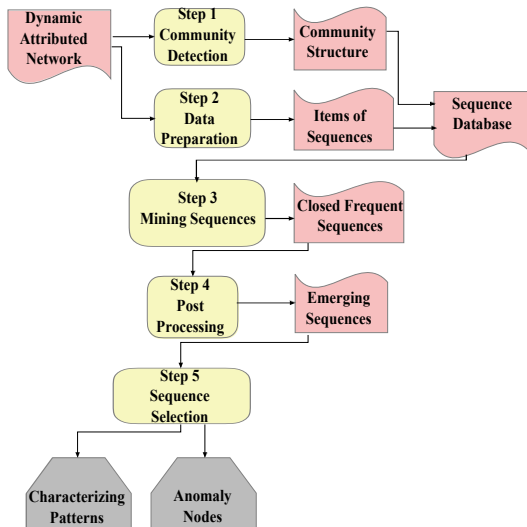
- ≡ A common territory (neighborhood, town, city, etc.)
[Gusfield 1975, *Harper & Row*]
- ≡ Having common relationships (social relationships, family, etc.) or emotions
[McMillan and Chavis 1986, *Journal of Community Psychology*]
- ≡ Common behavior and property
[McMillan and Chavis 1986, *Journal of Community Psychology*]



Community Interpretation Methodology



Community Interpretation Framework

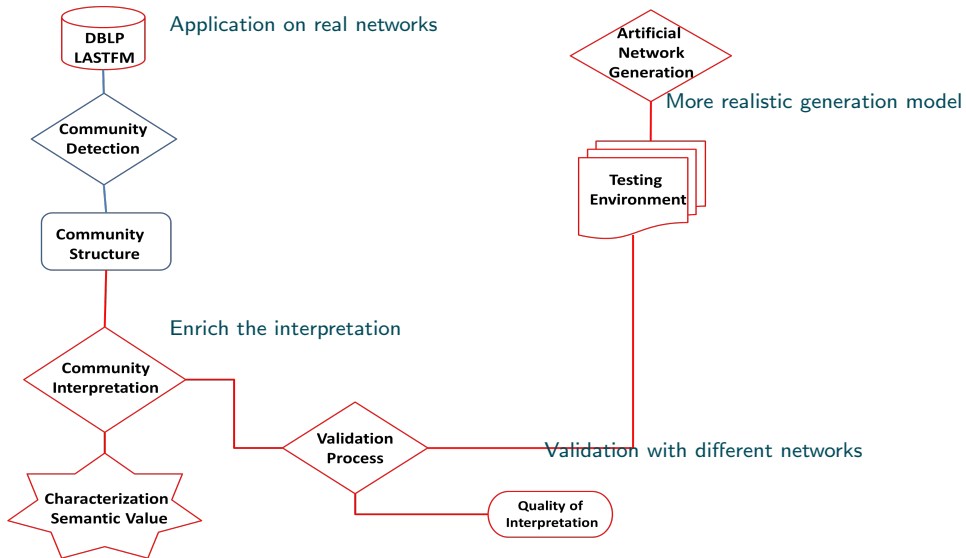


General Results

Lessons from the empirical validation

- ≡ Both topology and attributes support the interpretation process
- ≡ Question of Homogeneity
 - Homogeneity low level
 - Node groups are homogeneous
 - Homogeneous groups tend to be together as time goes by
 - Homogeneity high level
- ≡ Question of Community Evolution
 - Homogeneous node groups tend to merge or expand

Perspectives



Perspectives

Artificial Network Generation

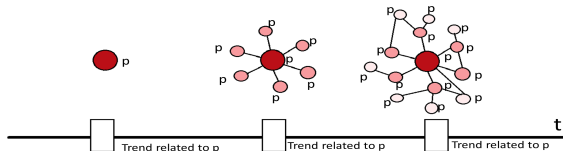
- ≡ Model with realistic attributes
- ≡ Model with combination of community evolution events

Applications

- ≡ Networks from different domains
- ≡ Different types of attributes
 - Textual, image, etc.

Interpretation

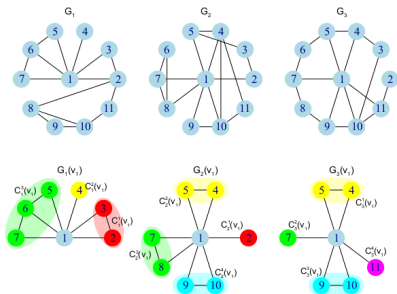
- ≡ Distance between communities
 - Which descriptors specify more the communities?
- ≡ The most representative nodes



Roles of Nodal Changes in Dynamic Networks

- ≡ Assumption that the neighborhood changes reflects the role and position of the node
- ≡ Characterizes each individual node by studying the evolution of its direct neighborhood
- ≡ Define the concept of neighborhood event
 - Six types of such events: birth, death, merge, split, expansion and shrink
- ≡ Study the affects of the evolution of those events

How to find the events

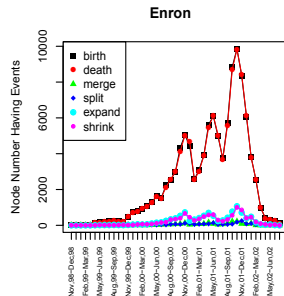
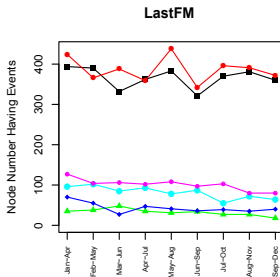
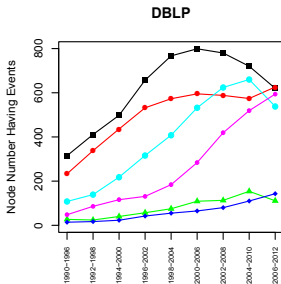


$$M_{1,2}(v_1) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{bmatrix}; M_{2,3}(v_1) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 2 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \end{bmatrix}$$

Experimental Results

Table: Real-world networks used in the experimental evaluation

| Network | Nodes | Active nodes | Time slices | Time span |
|---------|-------|--------------|-------------|--------------|
| DBLP | 2145 | 2046 | 10 | 1990-2012 |
| LastFM | 1701 | 1269 | 10 | Jan-Dec 2013 |
| Enron | 28802 | 28649 | 46 | 1997-2002 |



THANK YOU