

Identifying Relationship Patterns Inside Communities

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Some studies on networks require to isolate groups of elements, known as Communities. Some examples include (i) detection of protein complexes [10]; (ii) mapping metabolic networks [9]; (iii) identification of criminal organizations [5].

There are many algorithms to find communities and a commonly used evaluation approach is the maximization of the Modularity Clustering. This technique was proposed by Newman and Girvan [7][11] and is based on the number of edges that are into a community minus the expected number of edges from *null model*. The *null model* is a graph $G' = (V, E')$, obtained from original network $G = (V, E)$, with $|E'| = |E|$ and each edge is distributed over uniform probability, so each node in G' has the same degree on equivalent node in G . The objective function is to maximize the Equation $\frac{1}{2} \sum_{u,v \in E} (\frac{A_{u,v} - (d_u d_v)}{2m}) \sigma(u, v)$, where the d_w is degree of $w \in V$, $|E| = m$ and function σ returns 1, if u and v are in the same community and 0 otherwise [4] [3]. Some criticism and review over Newman and Girvan model can be seen in [6], [2] and [8].

Community detection is an important problem for Computer and other sciences. Following Agarwal and Kempe [1] one of the most important reasons to make clustering over a network is to identify the function/role of each element in a community.

If the communities have hundreds or thousands of elements, it is important to understand the functions of internal elements, but that will require an automatic process. In this context, we propose to develop a model, capable to identify elements with common features in different communities, based on the connection between elements and communities, agreeing with Newman and Girvan model features.

To understand the patterns within communities and carry out the research, the steps below are necessary. The schedule is as follows:

1. Developing a survey about network clustering and algorithms to solve the problem (aug. 2013 – jun. 2014);
2. Understanding the possible dynamics to elements function classification (jan. 2014 – dec. 2014);
3. Developing a model for dynamics function classification (jan. 2015 – dec. 2015);

4. To propose algorithms and to identify the model computational complexity (mar. 2015 – apr. 2016);
5. Verifying the model with networks instances and understand the adherence (mar. 2015 – jun. 2016).

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