194.050 Social Network Analysis

Topic 04: Equivalence-based role mining Final presentation

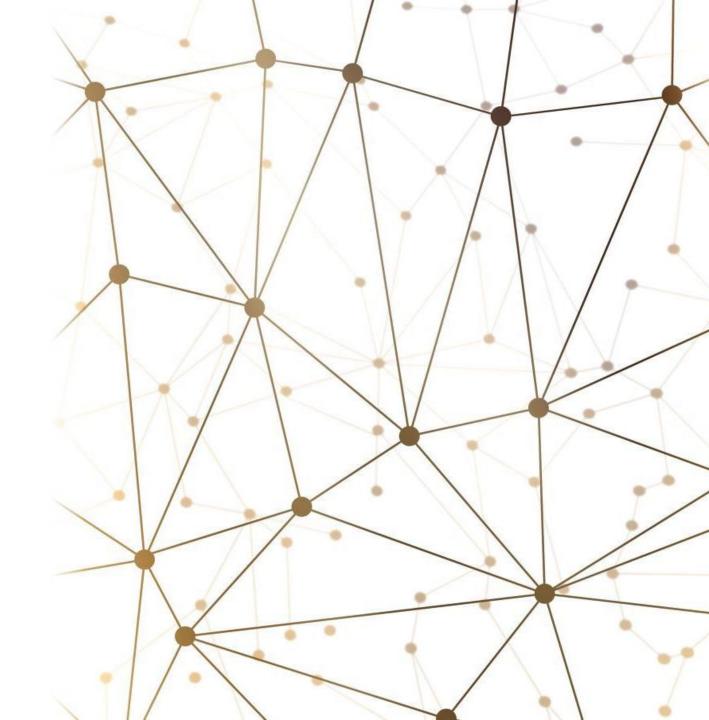
Babiy, Ivan

Chan, Yat Hin

Fan, Wing Yan

Muriasova, Karina

Olsiakova, Terezia



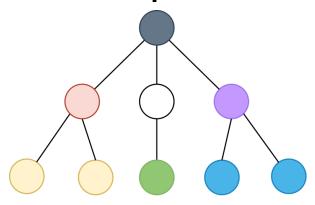
Outline

- Brief review of theoretical background
- Workflow overview
- Methodology
 - Subgraph extraction
 - Algorithms for different equivalence types
 - Clustering analysis
 - o Role interpretation
- Limitations

Brief review of theoretical background

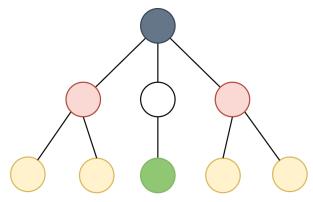
Comparing Positions: how nodes are embedded in its ego-network

Structural equivalence



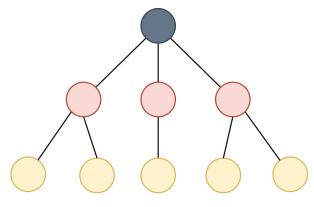
Identical composition of neighbourhood

Automorphic equivalence



Identically-shaped neighbourhood with same degree

Regular equivalence

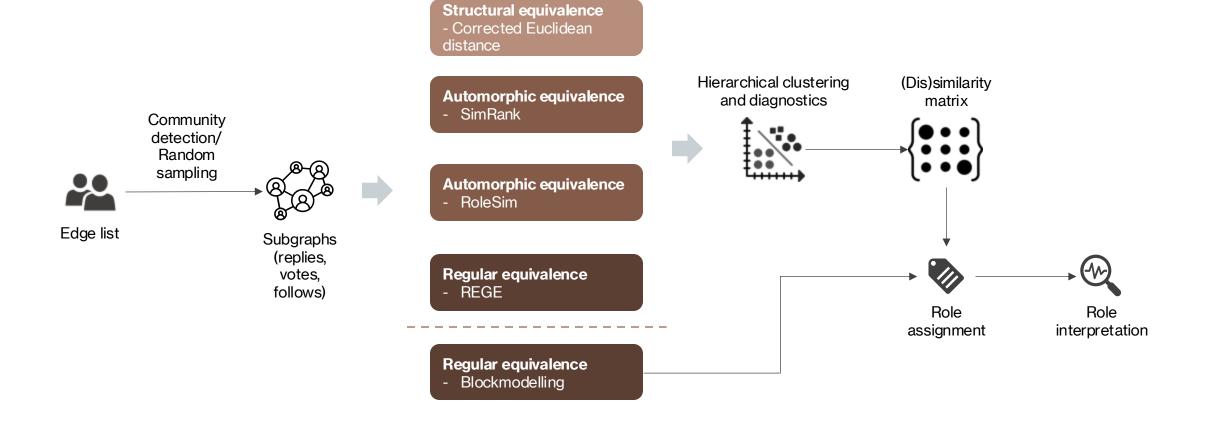


Existence of connections with neighbour(s) of the same class

Most restrictive

Least restrictive

Workflow overview



Methology overview

Equivalence type	Algorithm	Subgraph	Output	Roles found
Structural	Corrected Euclidean distance	Small community	Similarity	Yes
Automorphic	SimRank	Random sampling	Similarity	No
Automorphic	RoleSim	Small community & random sampling	Similarity	No
Regular	REGE	Small community	Similarity	Yes
Regular	Blockmodelling	Small community	Role assignment	Yes

With reference to commonly predefined social network roles, we aim to find 3-8 groups/roles

Subgraph extraction with community detection



1. After creating directed graphs from the edge lists, we select only users who appear in all 3 graphs (follows, votes, replies)



2. We used modularity maximisation to detect communities in the follows graph, the sparest graph. The graph size on which all our algorithms can run within reasonable time is chosen

type	nodes	edges
replies	253	1905
votes	253	6905
follows	253	436



3. The same users in the community are extracted from all 3 graphs

Finding structural equivalence classes with Corrected Euclidean Distance

Key Concept

• Structural equivalence occurs when two nodes have identical relationships to all others, a rare condition in real-world networks

Corrected Euclidean Distance:

- Self-loops and reciprocal links
- Returns zero for equivalent nodes.

$$d_E(v_i, v_j; p) = \sqrt{\sum_{\substack{k=1\\k\neq i,j}}^n (Q_{ik} - Q_{jk})^2 + (Q_{ki} - Q_{kj})^2 + p((Q_{ii} - Q_{jj})^2 + (Q_{ij} - Q_{ji})^2}$$

Algorithm's possible results

- Dendrograms to visualize hierarchical clustering
- Blockmodels that categorize nodes into equivalence classes
- Role Assignments, offering insight into patterns and node functions

Hierarchical clustering Results

- Was applied but showed no strong equivalence clusters
- This highlights the network's heterogeneity or lack of distinct structural roles

Finding automorphic equivalence classes with SimRank and RoleSim

Automorphism

Automorphic equivalence aims to capture nodes that are the same based on their neighborhoods

Similarity Measures

- SimRank simple procedure, but does not satisfy necessary axioms for detecting automorphism
- RoleSim more robust, guarantees the desired properties

Similarity matrices clustering results

- No meaningful clusters were be detected
- The network does not exhibit any such relationship where nodes participate in the network from similar contexts (neighborhoods)

Finding regular equivalence classes with REGE

- Consider "profile" of node's connection
- Suitable for directed data

Similarity Measures

- Utilise interim similarity scores as equivalence measures between iterations.
- Do not guarantee convergence to a similarity score, hence iterative approach
- Customary choice of iterations is 3

Similarity matrices clustering results

 Clusters of significant quality were detected for each graph type

Algorithm:

- 1. Define \mathbf{M}^t as a matrix where $0 \leq \mathbf{M}^t_{ij} \leq 1$ quantifies the degree to which nodes v_i and v_j are regularly equivalent ($\mathbf{M}^t_{ij} = 1$ if they are perfectly equivalent, $\mathbf{M}_{ij} = 0$ if they are perfectly inequivalent) after the t^{th} iteration of REGE. Initialize $\mathbf{M}^0_{ij} = 1$ for all i and j and t = 0.
- 2. Select a maximum number of iterations x.
- 3. for t = 0, 1, ..., x compute:

$$\mathbf{M}_{ij}^{(t+1)} = \frac{\sum_{k=1}^{g} \max_{m=1}^{g} \mathbf{M}_{km}^{t} (_{ij} \mathcal{M}_{km} +_{ji} \mathcal{M}_{km})}{\sum_{k=1}^{g} \max_{m}^{*} (_{ji} \mathbf{M} \mathbf{a} \mathbf{x}_{km} +_{ji} \mathbf{M} \mathbf{a} \mathbf{x}_{km})}$$

4. return \mathbf{M}^{x} .

Motivation

Difficult to find exact equivalence in real-life scenario

Setting:

Approximation with Similarity matrices from algorithms

Approach: Agglomerative hierarchical clustering

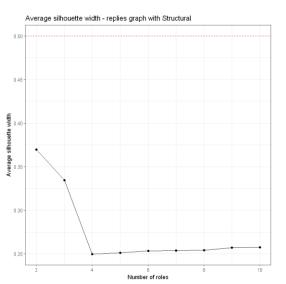
- Similarity definition:
 Complete linkage / Ward's method
- Distance measure:

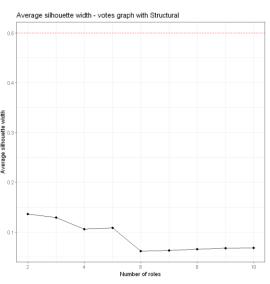
Structural: Compatible variant of Euclidean distance Automorphic, Regular: Euclidean distance

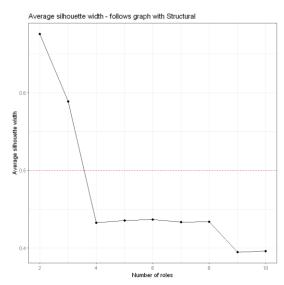
Cluster number decision and cluster quality evaluation:
 Based on average silhouette width supported by dendrogram

Structural equivalence

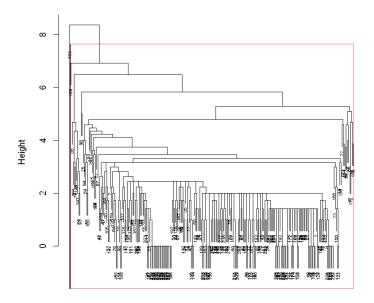
• Clusters identified but with insignificant number of members







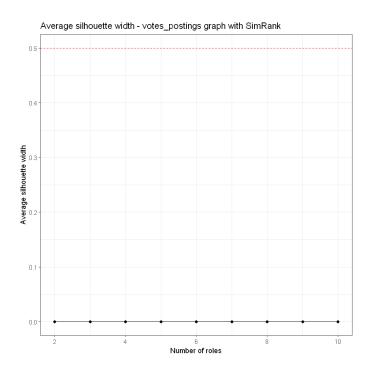
Hierarchical clustering dendrogram - follows graph with Structural

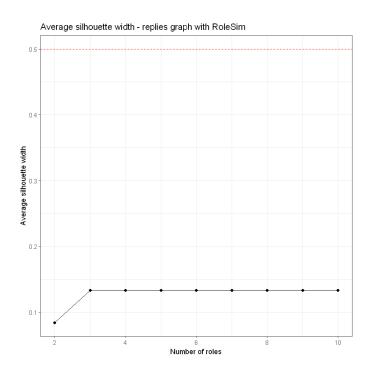


nodes hclust (*, "complete")

Automorphic equivalence – SimRank and RoleSim

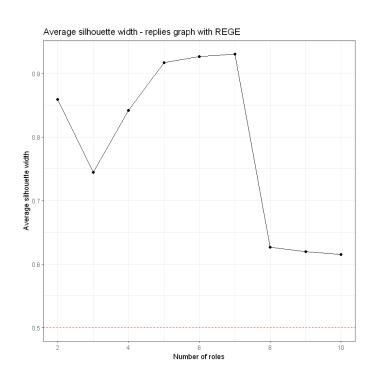
- Either no clusters identified, or clusters identified but with insignificant number of members
- Examples:

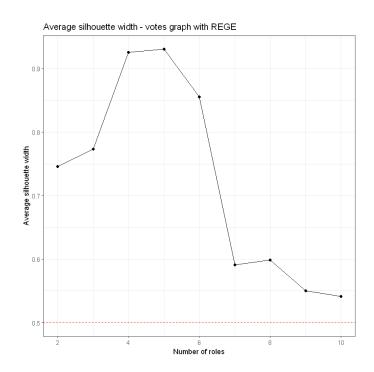


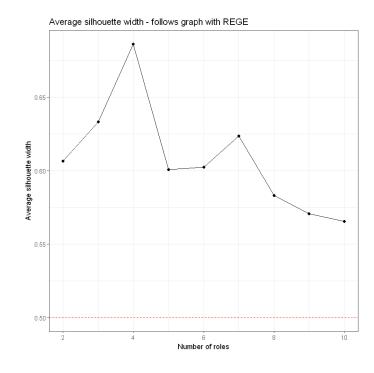


Regular equivalence - REGE

Candidates with similar silhouette width in replies and votes graph for further decision





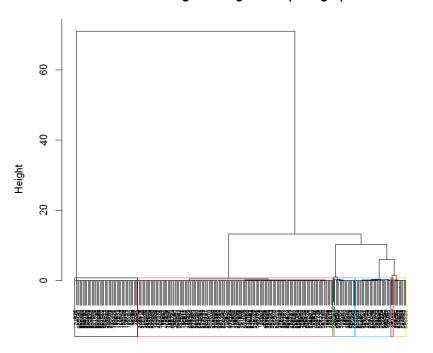


Average silhouette width - replies graph with REGE

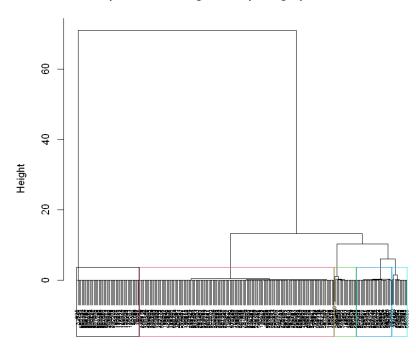
Regular equivalence - REGE

• Balance between role number and role size, e.g. in replies graph

Hierarchical clustering dendrogram - replies graph with REGE



Updated dendrogram - replies graph with REGE



Finding regular equivalence classes with blockmodelling

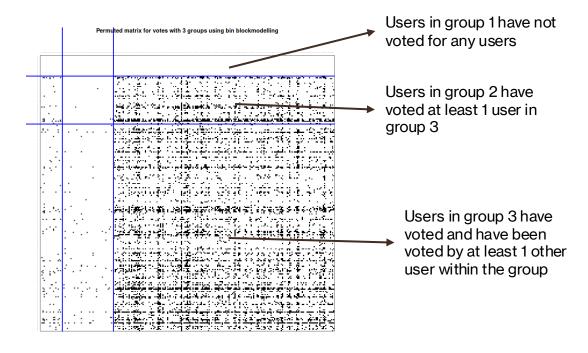
Algorithm:

- 1. Define *k* classes $C = \{C_1, \dots C_k\}$ and randomly assign nodes into the *k* classes.
- 2. Permute A so that nodes in the same class are in consecutive rows.
- 3. For a block describing the relations between the members of classes R_i and R_j define the measure $d(R_i, R_j)$ as:

$$d(R_i, R_j) = \min$$

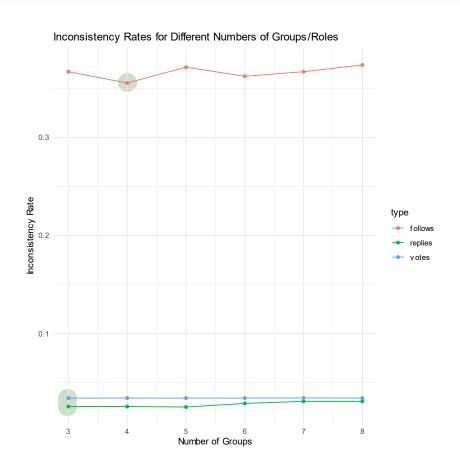
 $\begin{cases} # \text{ of rows and columns of the block with at least one 1} \\ # \text{ of rows and columns of the block that only contain 0} \end{cases}$

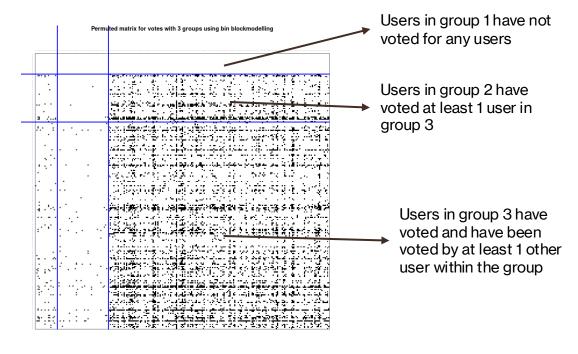
- 4. Initialize $S(\mathcal{C}) = \sum_{i,j} d(R_i, R_j)$.
- 5. For every transformation t(C) that moves a node from one class to another or interchanges a node between two classes, compute S(t(C)).
- 6. Let $S^* = \min_t S(t(\mathcal{C}))$ be the transformation with the smallest value of S.
- 7. If $0 < S^* < S$:
 - a. Set $S = S^*$ and C = t(C).
 - b. Repeat from step 5.
- 8. Otherwise, if S = 0 return C.
- 9. After trying all transformations, return $\mathcal C$ with and label it approximate solution.



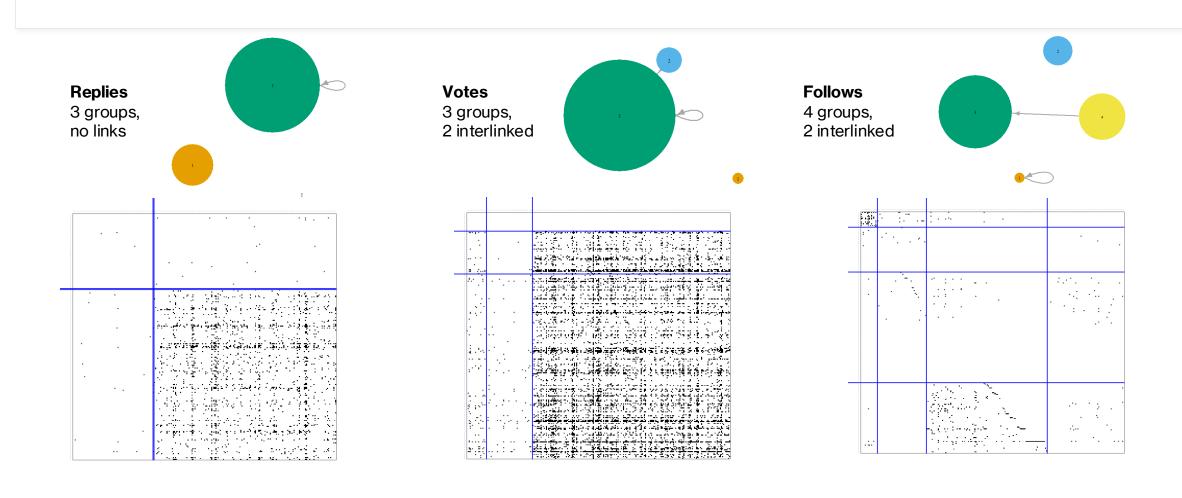
Connections in empty blocks and empty rows/columns in connected blocks are inconsistent with the role definition, to be minimised

Finding regular equivalence classes with blockmodelling



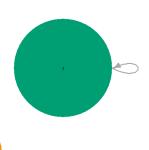


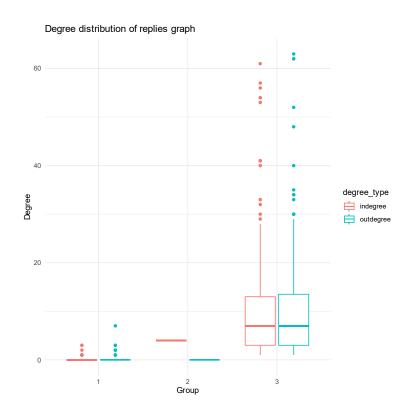
Connections in empty blocks and empty rows/columns in connected blocks are inconsistent with the role definition, to be minimised

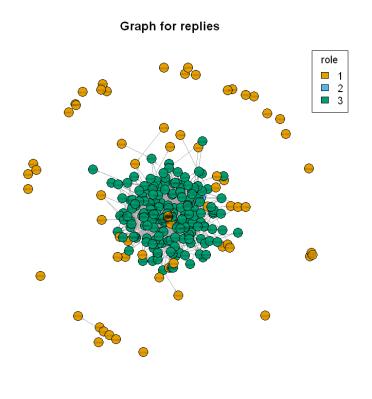


Replies

- Group 1
 - o Some individuals not linked
 - o Interacts with G3 replies
- Group 2
 - o Has only one user
- Group 3
 - Largest
 - o Many interactions, also reflexive
 - o role: conversation starters

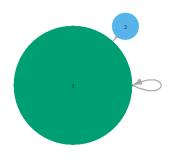


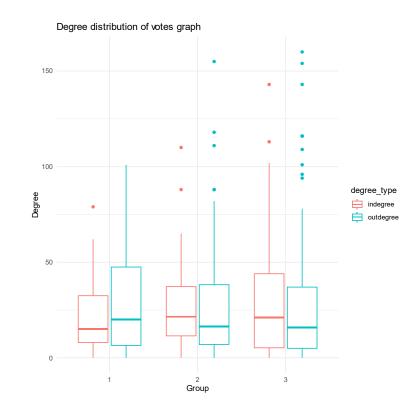


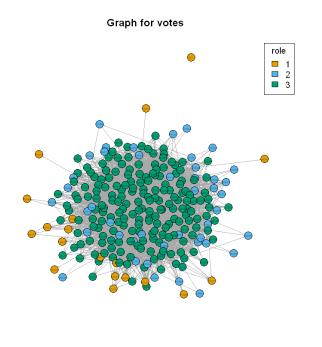


Votes

- Group 1
 - o Smallest, few interactions
- Group 2
 - Interacts with G3, multiple interactions
- Group 3
 - o More received votes than voted
- Weak resemblance to predefined roles

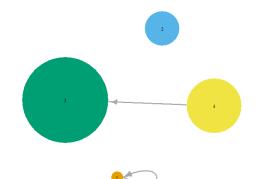


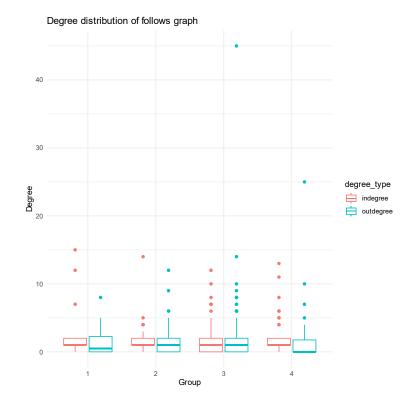


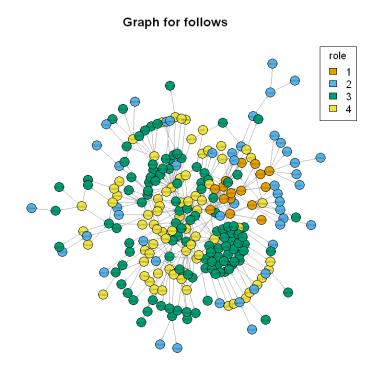


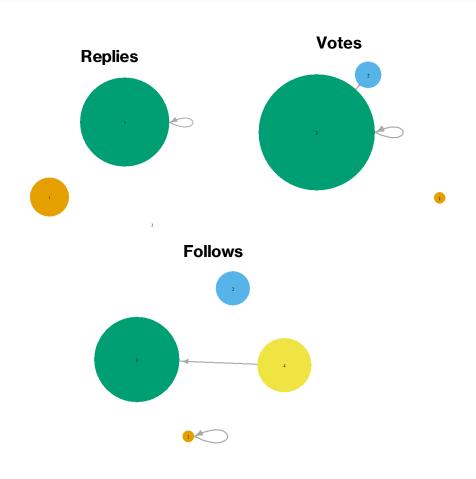
Follows

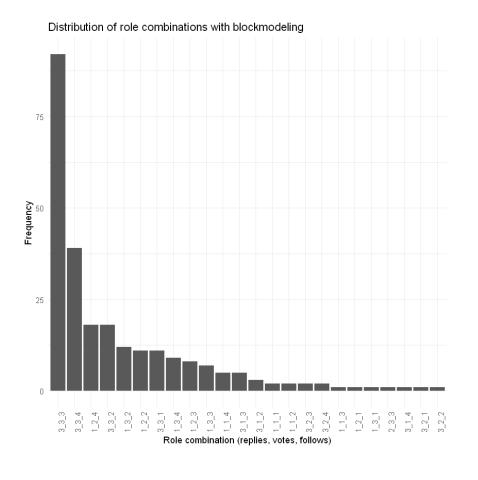
- Group 1
 - o more isolated
- Group 2
 - o on the periphery of the network
- Group 3 & Group 4
 - Interlinked
- · No resemblance to predefined roles











Unlike the blockmodeling approach, the resulting roles are not as direct and interpretable

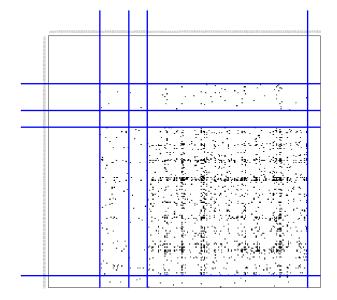
Replies 5 groups

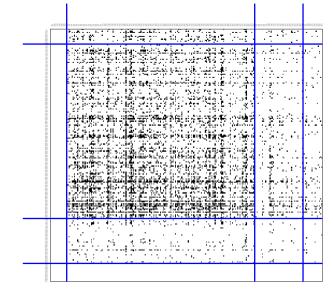
Votes 4 groups Follows
4 groups

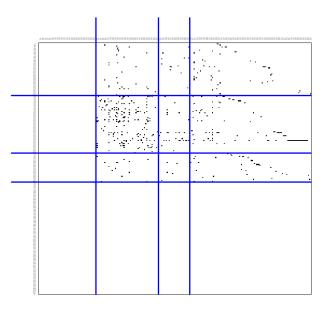
Partitioned matrix for replies graph with REGE

Partitioned matrix for votes graph with REGE

Partitioned matrix for follows graph with REGE





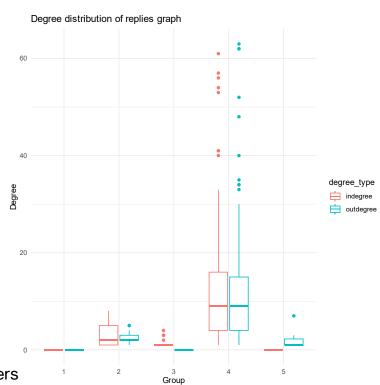


Replies

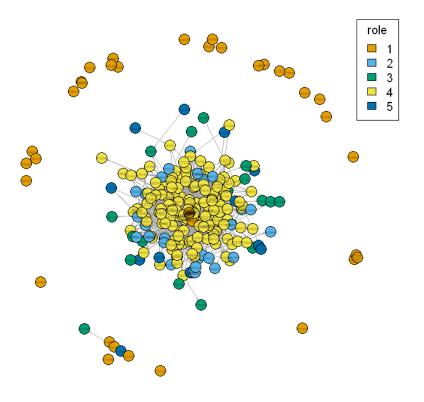
- Group 1
 - Rather isolated, no or little interactions with other nodes
 - o Potential role: inactive browser
- Group 2
 - o Form a community with itself
 - Potential role: a small isolated online community
- Group 3
 - o Do not reply to other nodes
 - o Receive replies from other nodes
- Group 4
 - Largest group
 - o Form a community with itself
 - o Potential role: the majority active users



- Reply to other nodes
- o Do not receive replies from other nodes

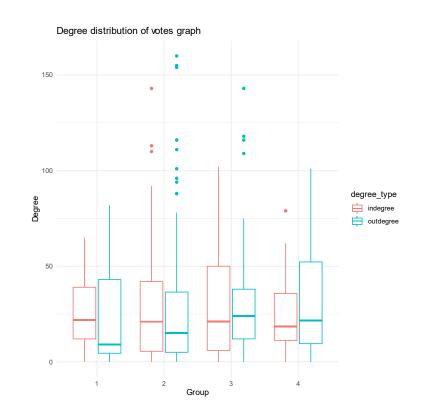


Graph for replies with REGE

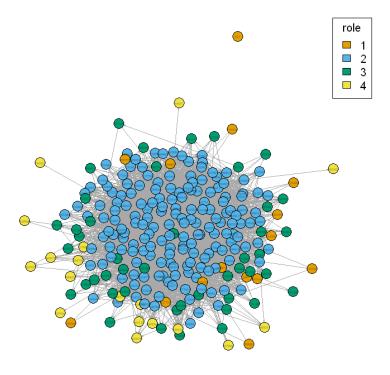


Votes

- Group 1
 - Vote others
- Group 2
 - Largest
 - o interlinked
- Group 3
 - o Form a community
- Group 4
 - o Receive votes



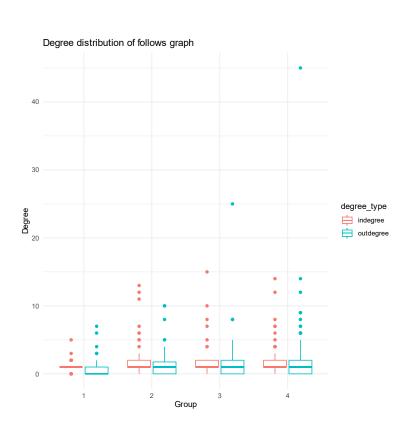
Graph for votes with REGE



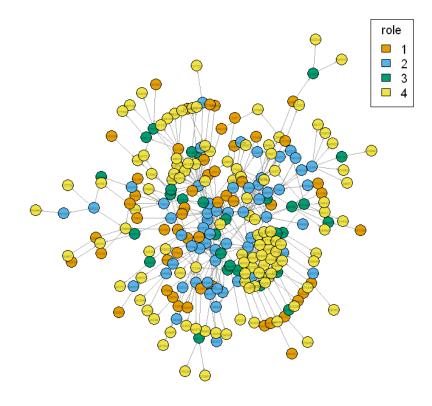
Follows

More comparable group size

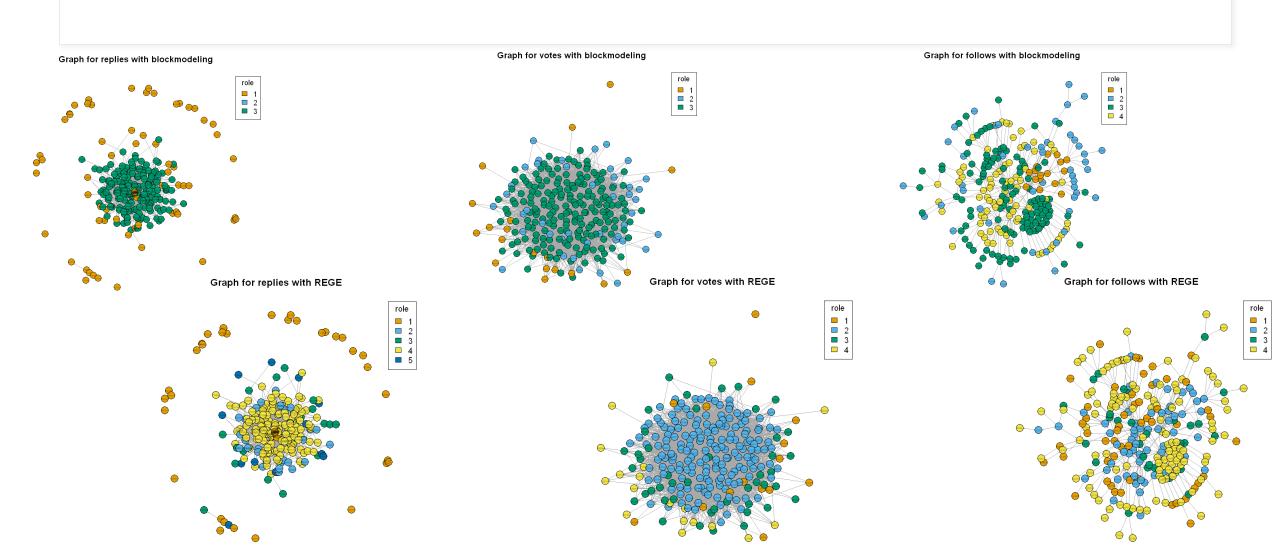
- Group 1
 - o Follow other group
- Group 2
 - Follow other group and within its own group
- Group 3
 - o Follow other group and
 - Being followed by other group
- Group 4
 - Followed by other group
 - Potential role: celebrities or influencers



Graph for follows with REGE



Overview of grouping by blockmodeling and REGE



Limitations

- Small graph
- Subgraphs are based on existing community detected in the follows graph, which might influence results
- Only unvalued graph used
- Interpretation based on metrics might not be intended for roles found by equivalence classes, since roles are based on node relationships, not on metrics

References

- Doran, D. (2017). Network Role Mining and Analysis. Springer International Publishing, Chapter 3.
- Murphy, P. (2019). An Introduction to Linear Regression in R. RPubs. https://rpubs.com/pjmurphy/436072
- Matjašič, Miha & Cugmas, Marjan & Žiberna, Aleš (2021). "blockmodeling: an R package for Generalized Blockmodeling," SocArXiv b8cxp, Center for Open Science.