

# Group extraction for real-world networks

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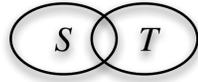
## Background

Complex real-world networks contain characteristic groups of nodes with common linking pattern like densely linked **communities** [1]. These were the focus of most recent work and have diverse applications. However, many real-world networks also contain **other groups of nodes** that can be **overlapping** and other, whereas some parts of the networks reveal **no significant groups**.

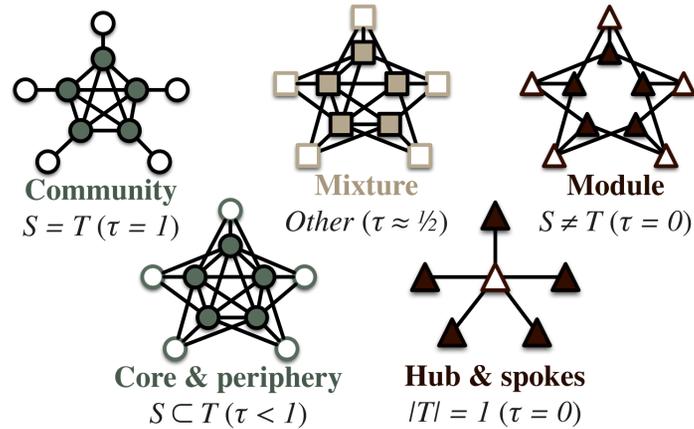
## Group formalism

Let  $S$  be a **group of nodes**,  $T$  the **linking pattern** and  $\tau$  the **group parameter**.

$$\tau(S, T) = \frac{|S \cap T|}{|S \cup T|}$$



## Group examples



## Group criterion

Let  $W$  be the **group criterion**,  $L$  the number of links and  $\mu$  the (harmonic) mean size.

$$W(S, T) = \mu(S, T) (1 - \mu(S, T)) \left( \frac{L(S, T)}{|S||T|} - \frac{L(S, T^c)}{|S||T^c|} \right)$$

$W$  is a **local asymmetric** criterion that **favors** the links between  $S$  and  $T$ , and **penalizes** for the links between  $S$  and  $T^c$ . (Note, however, that  $W$  **disregards** the links with both endpoints in  $S^c$ .) For  $S = T$ ,  $W$  is consistent with a wide class of other models (e.g., *stochastic blockmodel*). [2]

## Group extraction

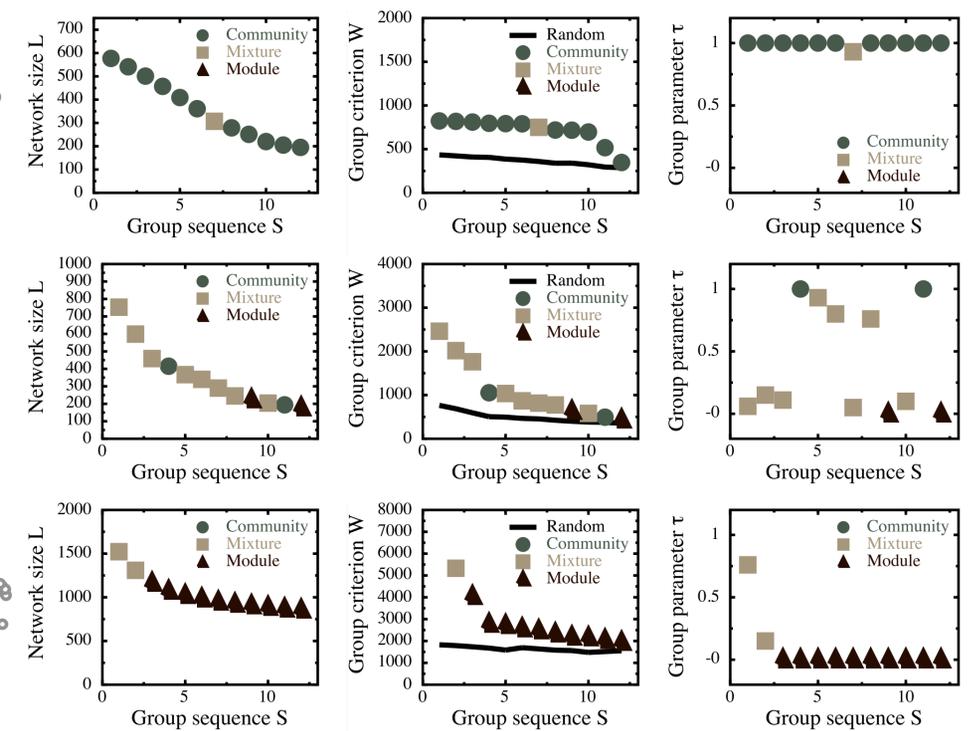
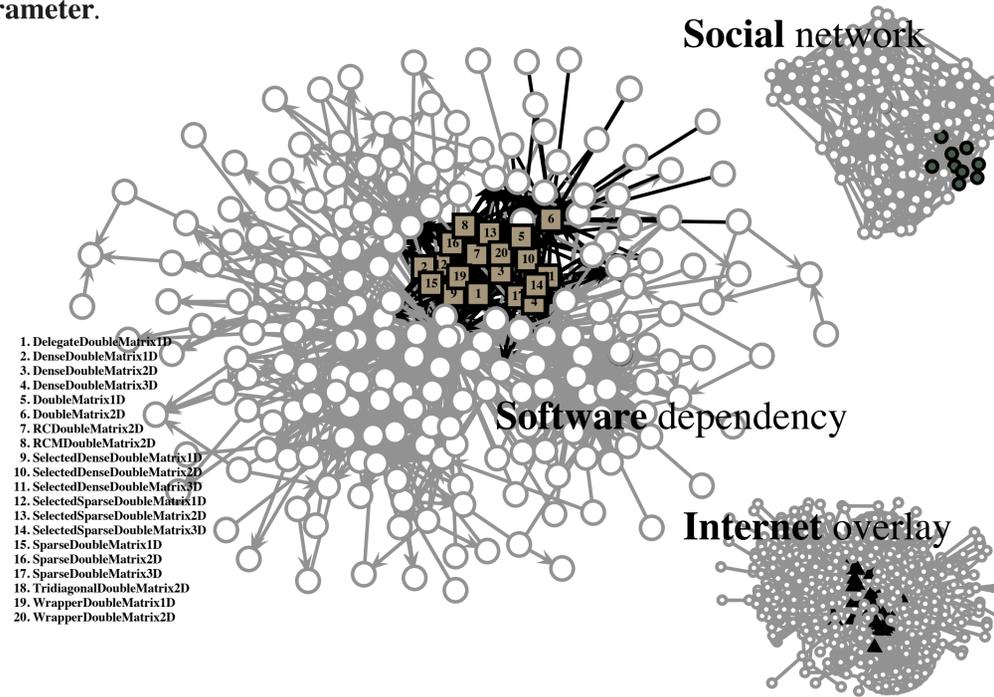
A sequential extraction [2] of groups that can be **overlapping, nested etc.**

- (1) **Find  $S$  and  $T$**  that optimize criterion  $W$  (e.g., *tabu search*).
- (2) **Extract only the explained links** between  $S$  and  $T$  (and isolated nodes).
- (3) **Repeat until  $W$  is larger than expected** in a random graph (by simulation).

## Contributions

1. A simple **formalism and criterion for general groups** of nodes.
2. An **adequate extraction procedure** for statistically significant groups.
3. **Characterization of the group structure** of different real-world networks.

## Groups in real-world networks



Network	Nodes	Links	#	Group  S	$\tau$	Community	Core	Mixture	Module	Background
Author collaborat. [3]	1589	2742	160	5.6	<b>0.94</b>	<b>71% (47%)</b>	0% (0%)	6% (5%)	1% (1%)	<b>22% (47%)</b>
American football [1]	115	613	13	8.6	<b>0.88</b>	<b>59% (83%)</b>	9% (11%)	3% (7%)	0% (0%)	<b>29% (98%)</b>
Lucene search engine	1657	6808	123	12.1	<b>0.55</b>	19% (25%)	1% (2%)	<b>30% (24%)</b>	<b>38% (34%)</b>	<b>11% (49%)</b>
Colt computing [4]	227	963	15	10.3	<b>0.41</b>	7% (11%)	5% (6%)	<b>69% (49%)</b>	4% (6%)	<b>15% (64%)</b>
Word adjacency [3]	112	425	4	11.2	<b>0.28</b>	0% (0%)	0% (0%)	<b>34% (33%)</b>	25% (15%)	<b>41% (99%)</b>
Internet overlay [5]	767	1857	33	10.6	<b>0.08</b>	0% (1%)	12% (4%)	13% (7%)	<b>34% (35%)</b>	<b>41% (80%)</b>
Southern women [6]	32	89	2	4.3	<b>0.00</b>	0% (0%)	0% (0%)	0% (0%)	<b>80% (41%)</b>	<b>20% (47%)</b>

All extracted groups are statistically significant at 1% level.

## References

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- [3] Newman, M. E. J.: Finding community structure in networks using the eigenvectors of matrices. *Phys. Rev. E* 74(3), 036104 (2006).
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