# INFORMATIVE VALUE OF NEGATIVE LINKS FOR GRAPH PARTITIONING

With an application to European Parliament Votes

Modèles et analyse des réseaux : approches mathématiques et informatiques (MARAMI) 14 Octobre 2015

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# SIGNED NETWORKS

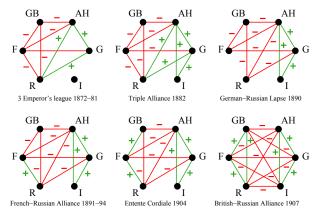
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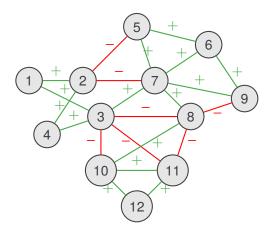
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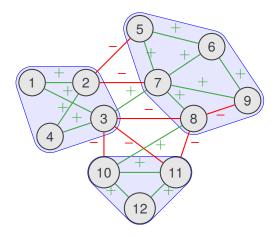
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- Modeling of systems containing antithetical relationships
- Ex.: evolution of pre-WWI alliances between countries [AKR06]



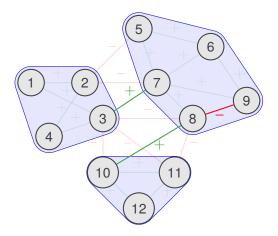
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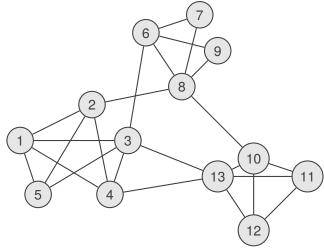
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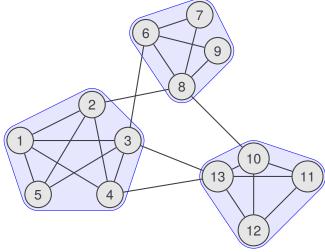
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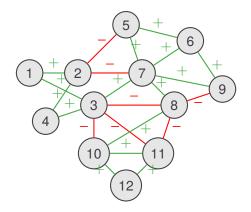
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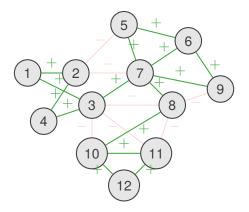
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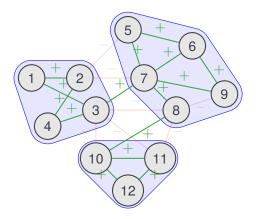
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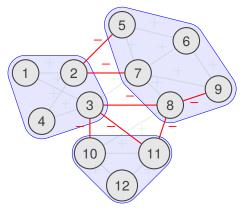
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- O Proposed method:
  - Consider a different dataset, modeling a different type of relationships
  - Compare community detection and correlation clustering algorithms

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#### Raw data:

- Nature: Voting activity at the European Parliament
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  - Ranges from -1 to +1
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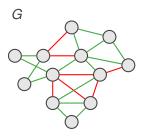
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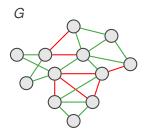
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- Networks:
  - Nodes: Members of the European Parliament (MEPs)
  - Weighted: voting agreement index values
  - Modes: 264 (time × topics)

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  - Parallel Iterated Local Search (pILS) [LDFF15]

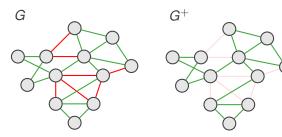
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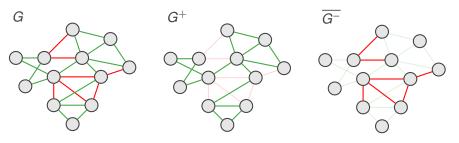
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  - InfoMap [RB08]
  - EdgeBetweenness [NG04]
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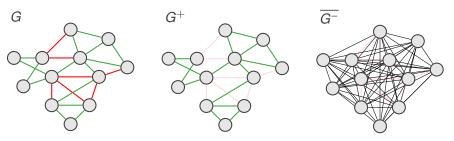
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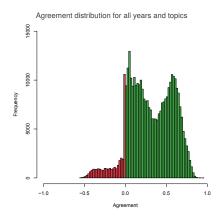


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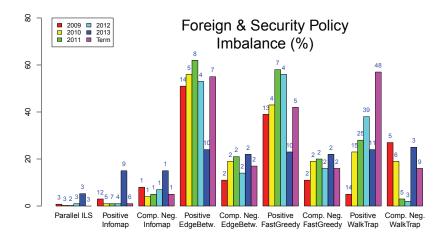


# EXTRACTED NETWORKS

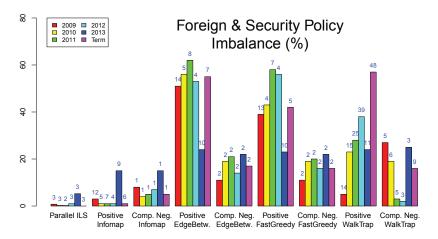
- Same observations for all topics/durations
- Positive side: bimodal distribution
  - Left peak: certain MEPs are frequently absent
  - Right peak: most MEPs often vote similarly
- Negative side: less extreme values
  - Clear majority, in average



#### PARTITION COMPARISON



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Partition comparison: near-zero NMI

# CONCLUSION

- Considering negative links on our dataset leads to:
  - Lower imbalance (at least 3 times better)
  - Only InfoMap outputs CC-like results (imbalance)
  - Different partitions (fewer clusters)
- O Contradiction with Esmailian et al.'s conclusions
  - (but not their results)
- Perspectives:
  - Consider more data, different types of networks (collection)
  - Exhaustive exploration of vote-based extraction methods
  - Political interpretation of the VoteWatch results

○ Data available on Figshare:

o http://dx.doi.org/10.6084/m9.figshare.1545599

R source code available on GitHub:

o https://github.com/CompNet/NetVotes

## REFERENCES

[AKR06]	T. Antal, P. L. Krapivsky, and S. Redner. Social balance on networks: The dynamics of friendship and enmity. <i>Physica D</i> , 224(1-2):130–136, 2006.
[CNM04]	A. Clauset, M. E. J. Newman, and C. Moore. Finding community structure in very large networks. <i>Physical Review E</i> , 70(6):066111, 2004.
[EAJ14]	P. Esmailian, S. E. Abtahi, and M. Jalili. Mesoscopic analysis of online social networks: The role of negative ties. <i>Physical Review E</i> , 90(4):042817, 2014.
[LDFF15]	M. Levorato, L. Drummond, Y. Frota, and R. Figueiredo. An ils algorithm to evaluate structural balance in signed social networks. In ACM Symposium on Applied Computing, pages 1117–1122, 2015.
[NG04]	M. E. J. Newman and M. Girvan. Finding and evaluating community structure in networks. <i>Physical Review E</i> , 69(2):026113, 2004.
[PL05]	P. Pons and M. Latapy. Computing communities in large networks using random walks. Lecture Notes in Computer Science, 3733:284–293, 2005.
[RB08]	M. Rosvall and C. T. Bergstrom. Maps of random walks on complex networks reveal community structure. Proceedings of the National Academy of Sciences, 105(4):1118, 2008.