Whom to Trust in a Signed Network?

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Problem: An observer who learns from link signs



- Nodes of two types: $\theta_i \in \{-1, +1\}$
- Link signs correlate with node types:
 - $\theta_i = \theta_j$: link is positive with probability $r \ge 0.5$
 - $\theta_i \neq \theta_j$: link is negative with probability $r \ge 0.5$
- Source nodes: The observer knows their types
 - Probability that signal is correct: $q \ge 0.5$

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Possible applications of the model

- 1. Two opposing camps: mainstream media and misinformation sources
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- 2. Employee network: Manager attempts to assess hidden qualities of a target employee
- 3. Inter-firm network: Which other firms to trust
- 4. Social networks: E.g., find a suitable roommate
- 5. ...

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Random neighbor heuristic: The outcome



Shaded area: 10th–90th percentile range

Random neighbor heuristic: The outcome

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- As N grows, expected accuracy approaches 0.5 as

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To make sense of a complex world is difficult

See M. Medo, M. S. Mariani, L. Lü, Communications Physics 4, 1, 2021 for more

1. Bayesian solution:

- The probability of a vector of node types, θ , when source node signals are σ and observed link signs, R: $P[\theta|\sigma, R]$
- The probability that given target node *t* is of positive type is

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2. Shortest-path heuristic:

- For all source nodes, $s \in S$, find the shortest path $s \rightarrow t$
- Compute the probability $P(\theta_t = +1|s \rightarrow t)$ for each path
- Aggregate information from all paths as if they were not overlapping

Simulation results



Simulation results



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Theorem (Equivalence)

If path from s to t is unique for all $s \in S$ and the paths from distinct source nodes do not overlap, the Bayesian rule and the shortest path rule are equivalent.



Theorem (Ordering)

For a given network, set of source nodes S and target node t, the expected accuracies of the three rules are be ordered as

 $E[A^{Bayes}] \ge E[A^{ShPath}] \ge E[A^{RNeighbor}].$

Theoretical results

Theorem (Unique Path)

If there is only one source and the path from s to t is unique, then all three rules yield the same expected accuracy

$$E[A^{Bayes}] = E[A^{ShPath}] = E[A^{RNeighbor}]$$



Theorem (Shortest Path Accuracy)

When the source and target node are chosen at random in an Erdös-Rényi network, the expected accuracy of the shortest path decays with the number of nodes, N, as $E[A^{ShPath}] - 0.5 \sim N^{-\gamma_{ShPath}}$ where

$$\gamma_{ShPath} = -\ln(2r-1)/\ln z$$

and z is the mean degree.

Recall: For the random neighbor rule, the scaling exponent is

 $\gamma_{RNeighbor} = 1 - r.$

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- Opinion/trust formation on a signed network
- Different from other opinion formation models (voter model, DeGroot,...)
 - One agent (observer), N subjects
- For simple formation mechanisms, the results are sensitive to noise in the system
- More sophisticated mechanisms yield better results at higher computational costs
- See more here:
 - 1. M. Medo, M. S. Mariani, L. Lü, The fragility of opinion formation in a complex world, Communications Physics 4, 1 (2021)
 - 2. F. Meng, M. Medo, B. Buechel, Whom to Trust in a Signed Network? Optimal Solution and two Heuristic Rules, preprint (2021)

Further questions

- 1. Consider more than two node types
- 2. Correlate link presence with node types (here only link signs depend on types)

- 3. Which rules yield accurate opinions without being excessively demanding?
- 4. Which spurious links distort the results most?
- 5. Combination with social opinion-formation models
- 6. How to empirically study belief formation on signed network data?

7. ...

Thank you for your attention!

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