Testing Independence between Observations from a Single Network

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Question

Testing the existence of correlation in our observations depending on their social relationship.

Motivation

Observations in a study are often collected from a single network within the target population. In this case, it is likely that the observed value of subjects' outcome is dependent on whom they are closely related to. Ignoring this dependence and implementing standard statistical analysis could lead to invalid statistical inference. We propose a test statistic for testing independence between observations from a single network.

Methods

We extend the existing test used for spatial autocorrelation:

Moran's I (continuous observations Y)

Weighted correlation coefficient: close to zero without dependence.

$$I \propto \sum_{i,j=1}^{n} W_{ij}(y_i - \bar{y})(y_j - \bar{y})$$

• Ø (categorical observations Y)

Weighted concordant(+1) and discordant(-1) pairs.

$$\phi \propto \sum_{i,j=1}^{n} W_{ij} \{2I(y_i = y_j) - 1\} / \{P(Y = y_i)P(Y = y_j)\}$$
 reject the null when confidence interval coverage rates droped assumption i.i.d. outcomes should not be used in inference.

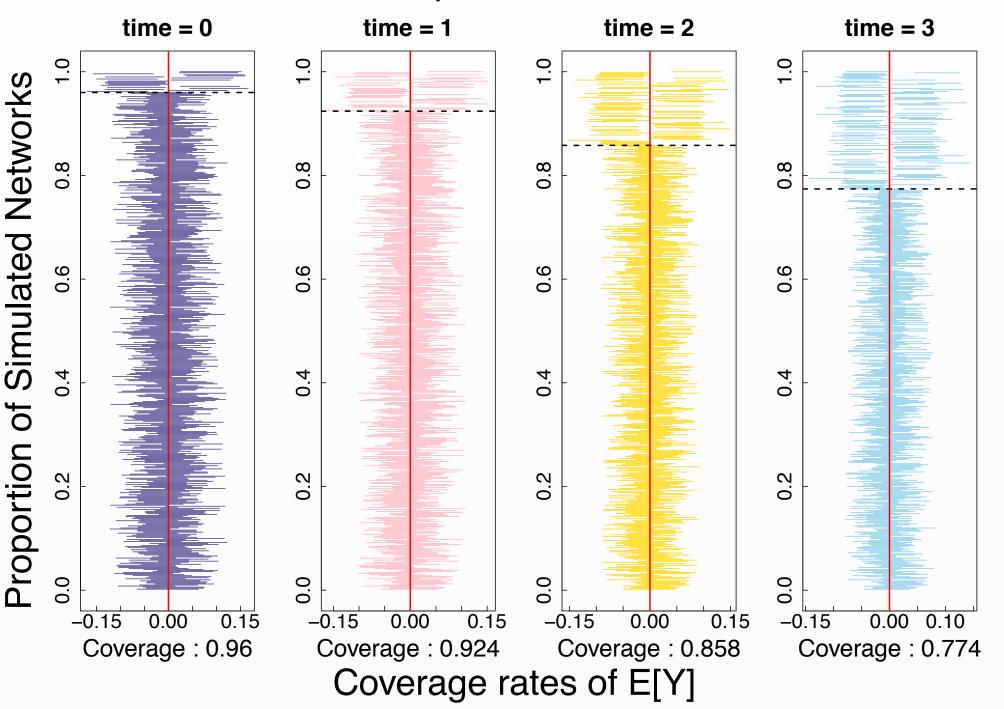
- W_{ij} : closeness measure between i and j.
- Statistics are weighted more when the observations are rare.
- When Y is binary, $I = \emptyset$.
- Asymptotic normality of both I and Ø under some conditions.

Numerical Study

. Peer-dependent continuous outcome of Y

$$Y_i^t = (1-\theta)Y_i^{t-1} + \theta \overline{\tilde{Y}_i^{t-1}},$$

where \tilde{Y}_i^{t-1} is the average of i 's friends at time t-1. As time t increases, the amount of dependence increases.



Standard 95% confidence interval assuming independence on Y leads to decrease in coverage rates. We are more likely to reject the null when confidence interval coverage rates drops;

Testing for Network Dependence				
time	95% CI coverage rate	Moran' I	Power(%)	
t = 0	0.96	0.06	6.40	
t = 1	0.92	0.99	27.80	
t = 2	0.86	2.09	60.60	
t = 3	0.77	3.42	81.80	

Numerical Study

2. Peer-dependent categorical outcome of Y

$$Y_i^t = (1 - \theta)Y_i^{t-1} + \theta Z_i^t$$

where $Y_i^0 \sim Multi(0.1, 0.2, 0.3, 0.25, 0.15)$ and $P(Z_i^t = k) =$ $\sum I(y_i^{t-1}=k)/n$. Assume that we are interested in estimating the simultaneous confidence interval for population proportion for five categories. As time t increases, the amount of correlation in Y also increases, which leads to decreasing coverage rate of confidence interval.

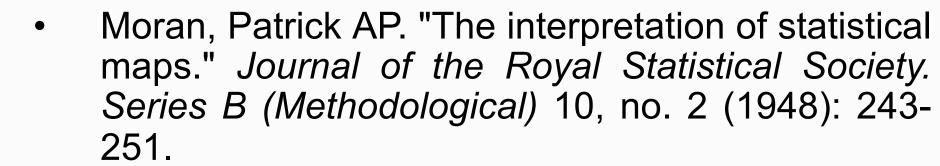
Testing for Network Dependence				
time	95% CI coverage rate	ф	Power(%)	
t = 0	0.94	0.06	7.00	
t = 1	0.85	1.66	48.20	
t = 2	0.75	2.93	82.20	

Example of Collaborative Network

Collaboration Network and Department

Reference

Discussion



If we reject the null hypothesis, it is highly recommended

We used an adjacency matrix as a weight matrix W, and

this is a robust choice for weight which does not depend on

If we have substantive knowledge of the dependence

mechanism, other weight matrices that incorporate this

Even though we use a coverage rate of confidence

intervals as a measure of dependence in our numerical

studies, the existence of correlation between the

observations does not necessarily lead to lower coverage

The test statistic will be finally applied to the observations

from Framingham Heart Study where individuals are

not to make i.i.d. assumption in the observations.

the exact form of dependence.

socially related to each other.

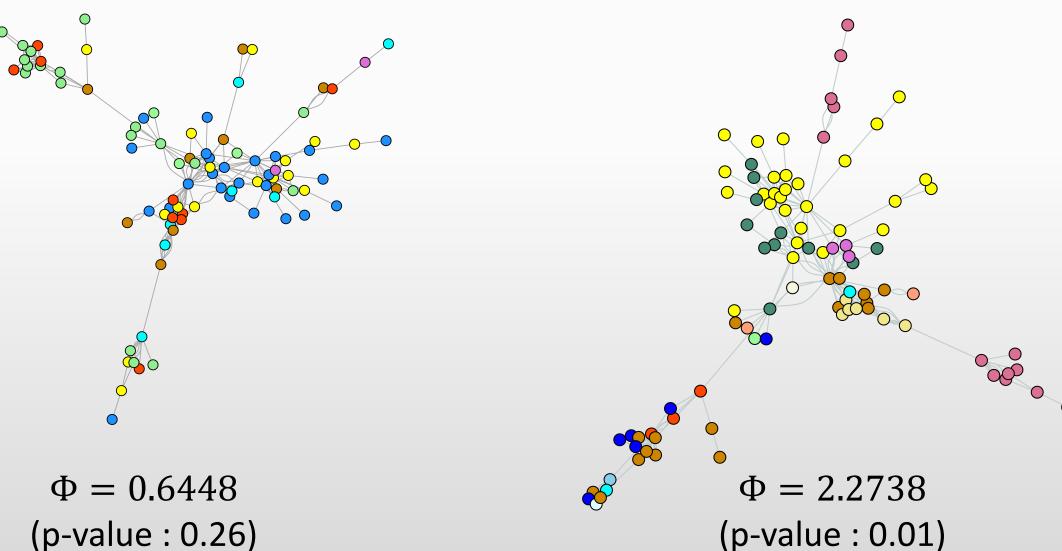
rate than expected.

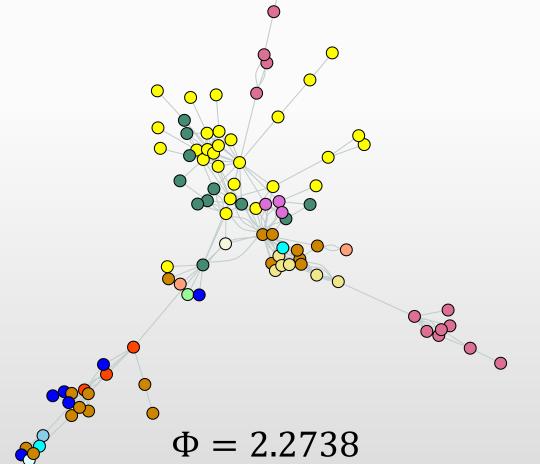
information might be more efficient.

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Christakis, Nicholas A., and James H. Fowler. "The spread of obesity in a large social network over 32 years." New England journal of medicine 357, no. 4 (2007): 370-379.







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