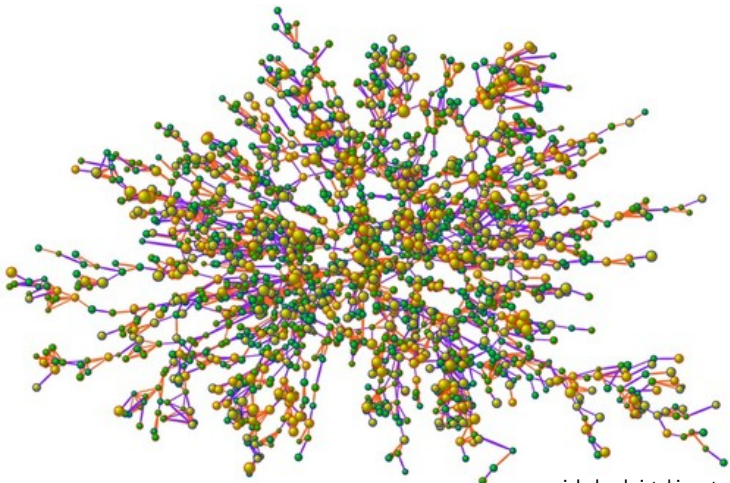


# Social network dependence and the replication crisis

Betsy Ogburn

eogburn@jhsph.edu  
Department of Biostatistics,  
Johns Hopkins University



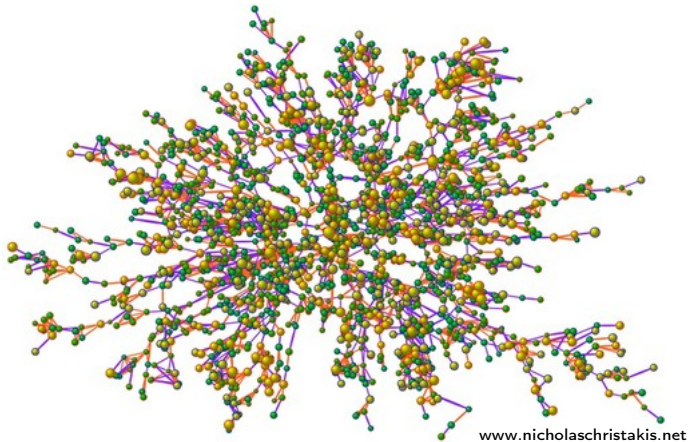
[www.nicholaschristakis.net](http://www.nicholaschristakis.net)

# outline

- ▶ Framingham Heart Study
- ▶ Ignoring network dependence is dangerous
  - ▶ Anticonservative statistical inference
  - ▶ Spurious associations due to dependence
- ▶ Testing for network dependence...
  - ▶ And finding striking evidence of dependence in FHS papers
- ▶ Re-analysis of a FHS peer effects model

# Framingham Heart Study

- ▶ Ongoing cohort study initiated in 1948 to study cardiovascular disease etiology one of the most successful and influential epidemiologic cohort studies in existence
  - ▶ arguably the most important source of data on cardiovascular epidemiology
- ▶ Thousands of papers published using FHS data, all using i.i.d. statistical methods
- ▶  $n \simeq 16,000$ , including multiple members of 1538 extended families, representing a sizable portion of the population of Framingham, MA



- ▶ FHS is a convenience sample that is comprised of members of an interconnected network rather than independent subjects.
- ▶ We expect social network dependence whenever subjects are sampled from one or a small number of schools, communities, hospitals, etc.

# Framingham Heart Study

- ▶ In the early 2000s, Christakis and Fowler discovered information on social ties that allowed them to reconstruct the (partial) social network underlying the cohort.
- ▶ Widely publicized results include significant peer effects for obesity (Christakis and Fowler, 2007), smoking (Christakis and Fowler, 2008), and happiness (Fowler and Christakis, 2008).
- ▶ The FHS has since been used to study peer effects by many other researchers (Pachucki et al., 2011; Rosenquist et al., 2010).
- ▶ The methods used have come under considerable criticism by statisticians, but little attention has been paid to the fact that i.i.d. methods were used for purportedly non-independent data.

# why is (network) dependence a problem?

Lee Y & Ogburn EL (2020)

1. **Anticonservative inference** Failure to adequately account for dependence leads to artificially small p-values, confidence intervals, and standard errors.
2. **Spurious associations** When two variables of exhibit similar types of dependence, association and effect estimates may be biased away from the truth

## anticonservative inference

- ▶ Suppose we're interested in the average height in the Boston suburbs.
- ▶ Let  $Y$  be height, and we will estimate  $E[Y]$  with the sample average from FHS:  $\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$ .
- ▶ If the data are independent, then

$$\text{var}(\bar{Y}) = \frac{1}{n^2} \left\{ \sum_{i=1}^n \sigma^2 \right\} = \frac{\sigma^2}{n}$$

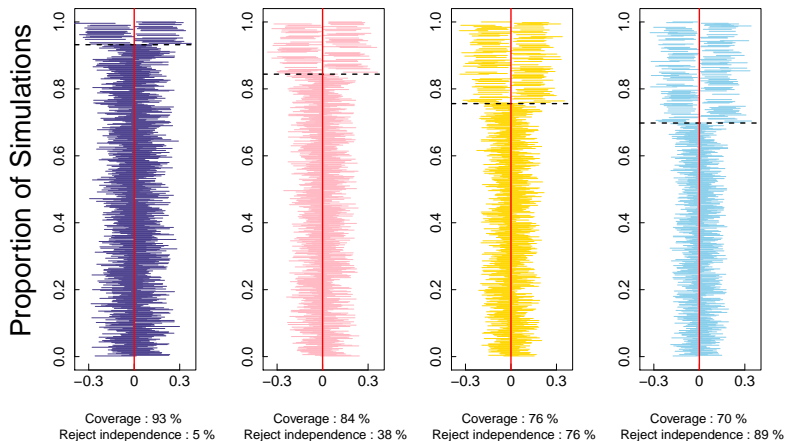
- ▶ But if there is dependence, then

$$\text{var}(\bar{Y}) = \frac{1}{n^2} \left\{ \sum_{i=1}^n \sigma^2 + \sum_{i \neq j} \text{cov}(Y_i, Y_j) \right\}$$



# anticonservative inference

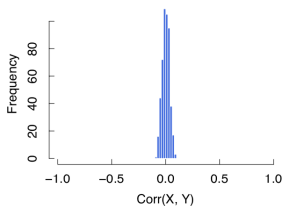
95% confidence intervals for  $\mu$  assuming independence



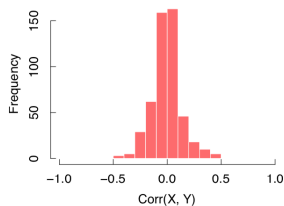
## spurious associations due to dependence

- ▶ When an exposure and an outcome both exhibit dependence across units, e.g. due to space, time, genetics, or social network ties, **estimates of associations—and causal effects—may be concentrated away from the truth.**
- ▶ Even if the exposure and the outcome are **causally and statistically independent** from one another, tests of independence will tend to reject the null.
- ▶ This occurs
  - ▶ in the absence of any confounding
  - ▶ in a representative sample
  - ▶ even if the only interest is in (out-of-sample) prediction
- ▶ Well-known in time series and GWAS; I'm not aware of any acknowledgement of this phenomenon outside of those settings

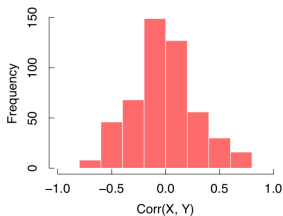
# spurious associations due to network dependence



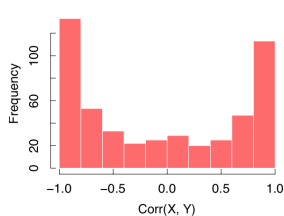
(a) Correlation between iid  $X$  and iid  $Y$



(b) Correlation between  $X$  and  $Y$  generated under direct transmission with large random errors



(c) Correlation between  $X$  and  $Y$  generated under direct transmission with moderate random errors



(d) Correlation between  $X$  and  $Y$  generated under direct transmission with small random errors

## test for network dependence

- ▶ Is it possible that studies based on FHS data report anticonservative s.e.'s (and CIs and p-values) and estimates that are spurious due to network dependence?
- ▶ We adapted Moran's  $I$  to test for network dependence, replacing weighted spatial distances with an adjacency matrix.
- ▶ We tested:
  1. regression residuals: dependence is (circumstantial) evidence of anticonservative inference
  2. outcome of interest and exposure of interest: dependence in both is (circumstantial) evidence of spurious associations

# test for network dependence in FHS papers

## The Impact of Obesity on Left Ventricular Mass and Geometry

The Framingham Heart Study

Michael S. Lauer, MD; Keaven M. Anderson, PhD; William B. Kannel, MD; et al

[> Author Affiliations](#)

JAMA. 1991;266(2):231-236. doi:10.1001/jama.1991.03470020057032

November 28, 1986

## Incidence of Coronary Heart Disease and Lipoprotein Cholesterol Levels The Framingham Study

William P. Castelli, MD; Robert J. Garrison, MS; Peter W. F. Wilson, MD; et al

[> Author Affiliations](#)

JAMA. 1986;256(20):2835-2838. doi:10.1001/jama.1986.03380200073024

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#### Atrial fibrillation as an independent risk factor for stroke: the Framingham Study.

P A Wolf, R D Abbott, W B Kannel



The American Journal of Medicine

Volume 62, Issue 5, May 1977, Pages 707-714



Clinical study

#### High density lipoprotein as a protective factor against coronary heart disease: The Framingham study

Tavia Gordon <sup>1</sup>, William P. Castelli M.D. <sup>1</sup>, Marthana C. Hjortland Ph.D. <sup>1</sup>, William B. Kannel M.D. <sup>1</sup>, Thomas R. Dawber M.D. <sup>1</sup>

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### ARTICLES

#### Reduced heart rate variability and mortality risk in an elderly cohort. The Framingham Heart Study.

H Tsuji, F J Venditti, E S Manders, J C Evans, M G Larson, C L Feldman, D Levy

Is there evidence that obesity is “socially contagious” in FHS?

*The NEW ENGLAND JOURNAL of MEDICINE*

SPECIAL ARTICLE

## The Spread of Obesity in a Large Social Network over 32 Years

Nicholas A. Christakis, M.D., Ph.D., M.P.H., and James H. Fowler, Ph.D.

# Is there evidence that obesity is “socially contagious” in FHS?

- ▶ To assess peer effects of obesity, researchers ran models like this:

$$Y_{ego}^t = \alpha + \beta Y_{alter}^{t-1} + \gamma Y_{alter}^{t-2} + \eta Y_{ego}^{t-1} + \lambda X_{alter,ego} + \varepsilon_{ego}^t$$

- ▶  $Y_{ego}^t$  is the ego's obesity status at time  $t$ ,  $Y_{alter}^{t-1}$  is the alter's obesity status at time  $t - 1$ , and  $\beta$  is interpreted as the effect of interest.

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$$Y_{ego}^t = \alpha + \beta Y_{alter}^{t-1} + \gamma Y_{alter}^{t-2} + \eta Y_{ego}^{t-1} + \lambda X_{alter,ego} + \varepsilon_{ego}^t$$

- ▶ These models were estimated assuming that  $\varepsilon_i$  and  $\varepsilon_j$  are independent for  $i \neq j$  (but  $\varepsilon_i^t$  and  $\varepsilon_j^s$  could be dependent).



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- ▶ We tested for network dependence in the outcome, the predictor of interest, and the regression residuals.
- ▶  $p < 0.01$  for all tests.

# Is there evidence that obesity is “socially contagious” in FHS?

- ▶ Using a new method to account for network dependence ([Ogburn et al. 2020](#)), we re-analyzed the FHS obesity data...
- ▶ ... and found no evidence of peer effects.



# Is there evidence that obesity is “socially contagious” in FHS?

- ▶ We estimated the expected probability of obesity at time  $t$  under a hypothetical intervention to increase the number of each node's obese alters by 1.
- ▶ We estimated a causal risk difference of exactly 0, with 95% confidence interval  $(-0.01, 0.01)$ .

## Is there evidence that obesity is “socially contagious” in FHS?

- ▶ We also estimated the causal effect of an increase (of half a standard deviation) in the average BMI of each subject’s friends.
- ▶ We estimated a causal effect of 0.25, 95% confidence interval  $(-0.47, 0.98)$ .
- ▶ (For context, the empirical mean BMI was 25.51)
- ▶ These analyses are consistent with the hypothesis that the strong results in the original paper are spurious, due to dependence and/or model misspecification rather than true associations or causal effects.

## conclusion

- ▶ Whenever data are dependent, analyses that fail to fully account for dependence can underestimate uncertainty and produce spurious estimates of associations and causal effects.
  - ▶ Spurious associations are a problem for out-of-sample prediction, too!
- ▶ Data may be dependent more often than you might think.
  - ▶ Convenience samples are everywhere in the health and social sciences.
- ▶ Statisticians know how to account for Euclidean dependence; non-Euclidean network dependence is a new frontier and lots more research is needed.

# Thank you

## people

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