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Confounders as Mathematical Objects

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Paper [2006SchieldBurnhamMAA.pdf](#)
Slides [2006SchieldBurnhamMAA6up.pdf](#)
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Confounding: Can Cause Variation

Observational studies: associations often *confounded*

E is Effect of interest
A is Associated predictor
C is confounder

Confounding not chance-related.
As sample size increases,
* margin of error (influence of chance) decreases,
* influence of confounder is unchanged.

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Confounder Resistance: Rules of Thumb

Sir Richard Doll: No single study is persuasive unless the lower limit of its 95% confidence level falls above a **threefold** increased risk.

“As a general rule of thumb,” says Angell of the New England Journal, “we are looking for a relative risk of **three or more**” before accepting a paper.

Robert Temple, FDA Director of Drug Evaluation, puts it bluntly: “My basic rule is if the relative risk isn’t at least **three or four**, forget it.”

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Problem of Confounding in Epidemiology

John Bailar, epidemiologist: “*There is no reliable way of identifying the dividing line.*”

Epidemiologists need an abstract description of confounding that can generate confounder significance and confounder intervals for Relative Risk. $RP(E:C) = P(E|C) / RP(E|\sim C)$

This description must handle binary data, be meaningful, useful and easy to understand.

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Modeling Spuriousity: A has “no effect” on E

Model relation between 3 binary variables using an ordinary least-squares regression model:

E is Effect of interest
A is Associated predictor
C is confounder

Non-Interactive model: $E(A,C) = b_0 + b_1*A + b_2*C$
Association between A and E is **spurious** if $b_1 = 0$.

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Simplifying: Defining an S Confounder

A, C and E are binary. Top three ratios are observed. Confounder is defined by the bottom three ratios.

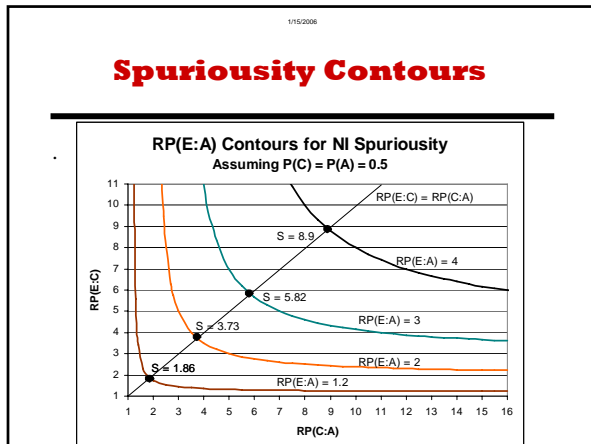
$$RP(E:A) = \frac{P(E|A)}{P(E|\sim A)}$$

$$RP(E:C) = \frac{P(E|C)}{P(E|\sim C)}$$

$$RP(A:C) = \frac{P(A|C)}{P(A|\sim C)}$$

Definition: An “**S confounder**” is a binary confounder with $P(C:A) = RP(E:C) = S$. $P(C) = P(A)$.

An association is “confounder resistant” to a size S confounder if it withstands nullification.

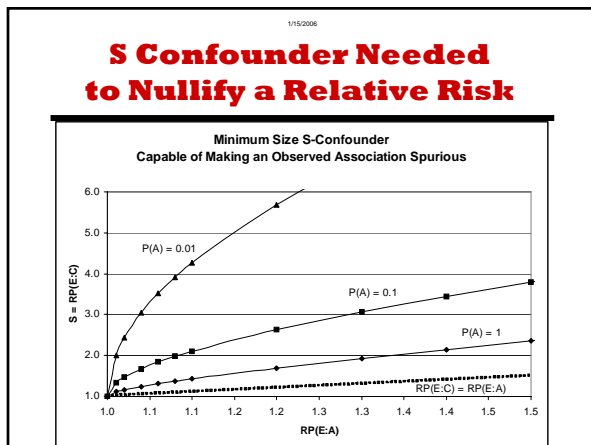


S Confounder Needed to Nullify Relative Risk

Given an observed association, $RP(E:A)$, and a predictor prevalence, $P(A)$, determine the size of the S confounder needed for nullification.

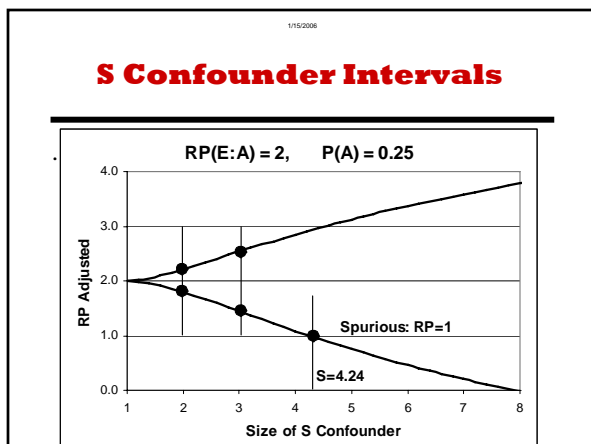
$$\frac{S - 1}{[RP(E : A) - 1]} = 1 + \sqrt{1 + \frac{1}{P(A)[RP(E : A) - 1]}}$$

If $P(A) = 0.5$ and $S = 5$, then $RP(E:A) = 2.6$.
 If $P(A) = 0.4$ and $S = 6$, then $RP(E:A) = 3.1$



Confounder Intervals: Influence of S Confounder

Lower limit: determined by removing the influence of a confounder where $RP(C:A) = RP(E:C) = S$.
 Upper limit: determined by removing the influence of a confounder where $RP(C:A) = 1/RP(E:C) = 1/S$.
 Given the size of an S confounder, the confounder influence can be determined algebraically and illustrated using a standardization diagram.
 For a S confounder of size 2, the confounder interval for $RP(E:A) = 2$ with $P(A) = 0.5$ is [1.67, 3.0].



Recommendations

Review/critique Schield & Burnham (2006) MAA paper: *Confounders as Mathematical Objects*. Copy at www.StatLit.org/pdf/2006SchieldBurnhamMAA.pdf

This 26 page paper is dense: 150 equations with new concepts and new ratio-comparison notation.

In-depth reviews will be acknowledged in the final paper.

If published, this paper will help make confounding a mathematical object. This could open up an entirely new area for statistics and statistical education.