

UTSA Confounding 2011 1

Statistical Literacy: Confounding

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Augsburg College
Director, W. M. Keck Statistical Literacy Project
Vice President, National Numeracy Network
US Rep., International Statistical Literacy Project

January 13, 2011
University of Texas San Antonio (UTSA)
Slides at www.StatLit.org/pdf/2011-Schield-UTSA-Confounding-Slides.pdf

2011 2

Statistical Literacy

Statistical literacy is **the ability to read and interpret summary statistics in everyday life.**

Statistical Literacy studies

- (1) the relation between statistical associations and causation, and
- (2) the full-range of influences on a statistic or on a statistical association. [Take CARE]

2011 3

Take CARE: Context

The influence of factors **taken into account** by

- data broken out by subgroups in tables and graphs
- averages, ratios and comparisons of averages and ratios
- epidemiological models (cf., deaths attributed to obesity)
- regression models and
- the study design (cf., longitudinal vs. cross-sectional; experiment vs. observational study).

The influence of related factors (confounders) **not taken into account** in the study and **not blocked** by the study design.

2011 4

Controlling for a confounder can DECREASE an association

MN has 3.8 times as much prison expense as ME

State	Total	# Inmates	Per Inmate
MN	\$184M	4,865	\$37,825
ME	\$48M	1,424	\$33,711

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MN has 25% more prison expense *per inmate* than ME

2011 5

Controlling for a confounder can NULLIFY an association

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2011 6

Controlling for a confounder can REVERSE an association

CA has 50% more prison expense than NY

State	Total	# Inmates	Per Inmate
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CA has almost twice as many inmates as NY

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2011 7

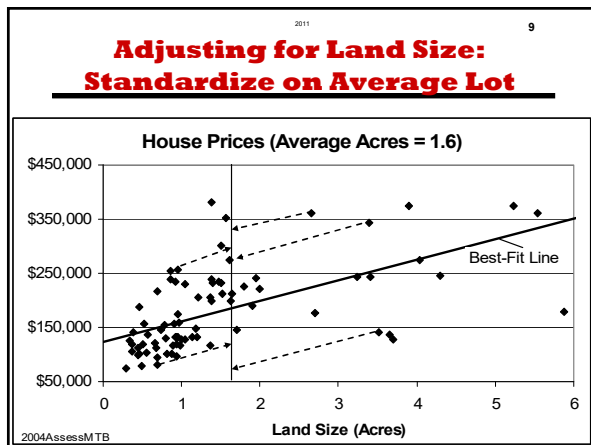
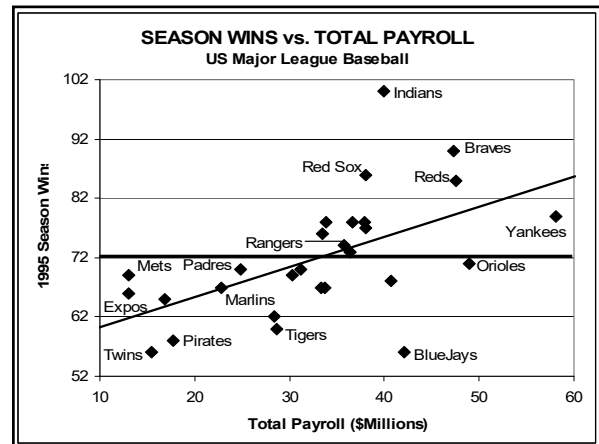
**Controlling for a confounder
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MN has 27% more prison expense than IA

State	Total	# Inmates	Per Inmate
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MN has 18% fewer inmates than IA

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2011 10

SAT VERBAL SCORES: FLAT

GROUP	1981	2002	CHANGE
White	519 (85%)	527 (65%)	8
Black	412 (9%)	431 (11%)	19
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2011 11

**Multivariate Analysis
can be Complex**

To simplify, consider cases with

- a binary outcome,
- a binary predictor and
- a binary confounder.

What are the necessary conditions for nullification or a reversal?

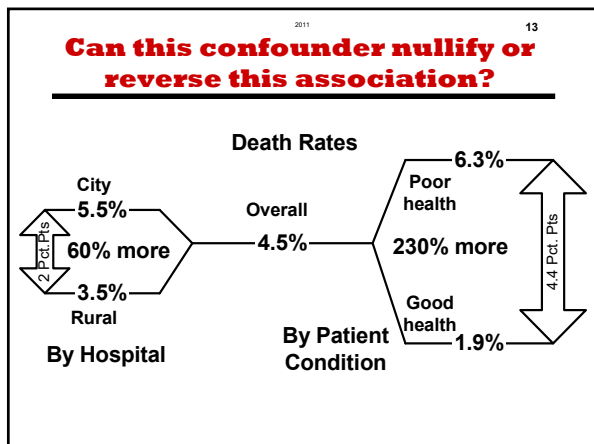
See Schield (1999) and Schield and Burnham (2003)

2011 12

**City Hospital:
Hospital of Death??**

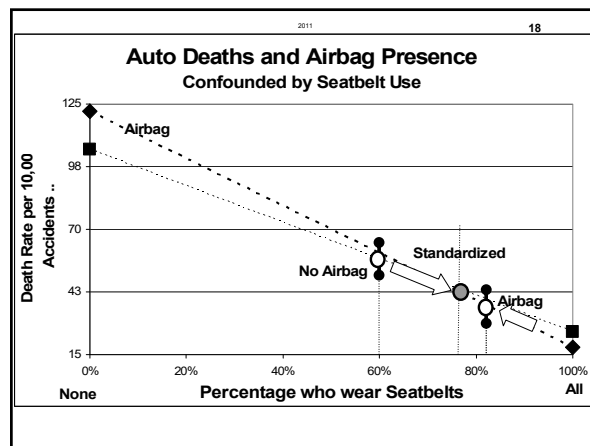
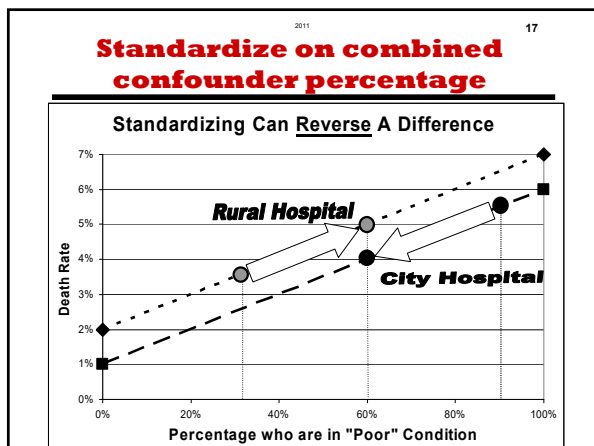
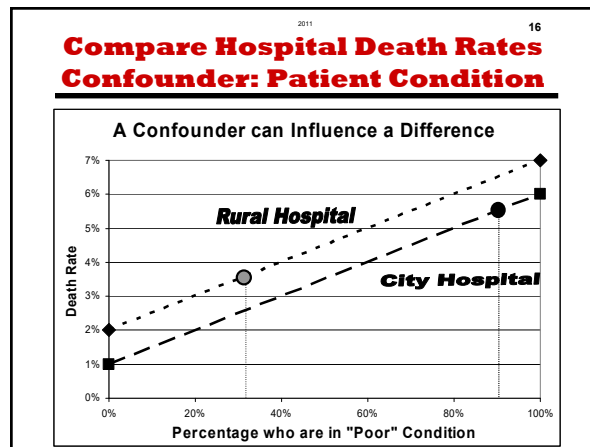
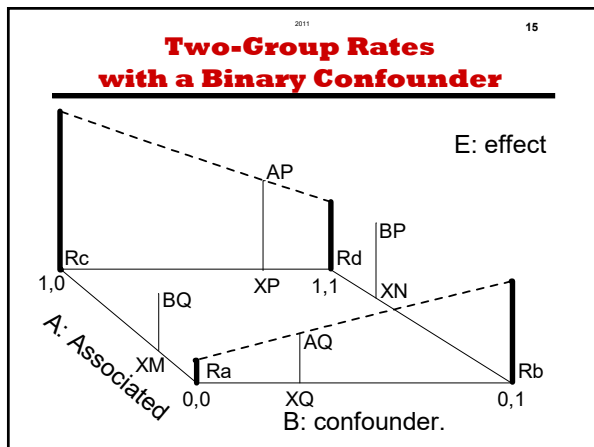
Hospital	Total	Died	Death Rate
City	1,000	55	5.50%
Rural	1,000	35	3.50%
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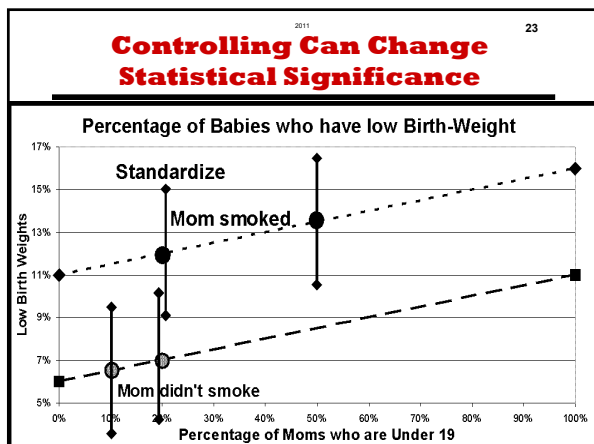
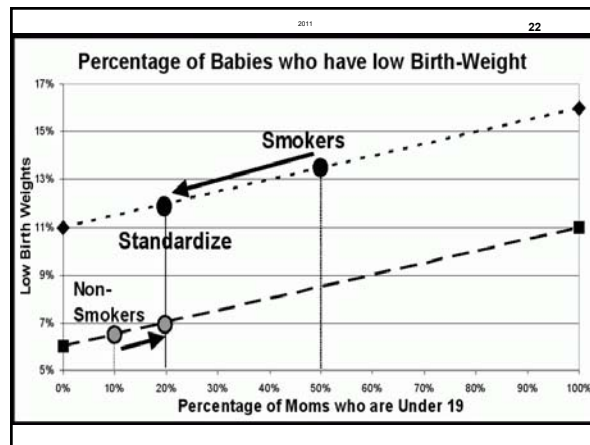
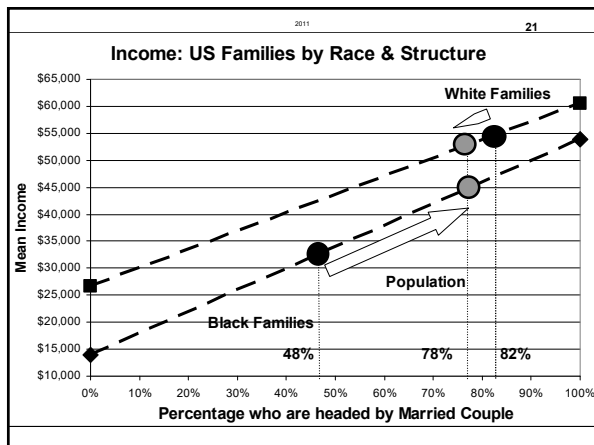
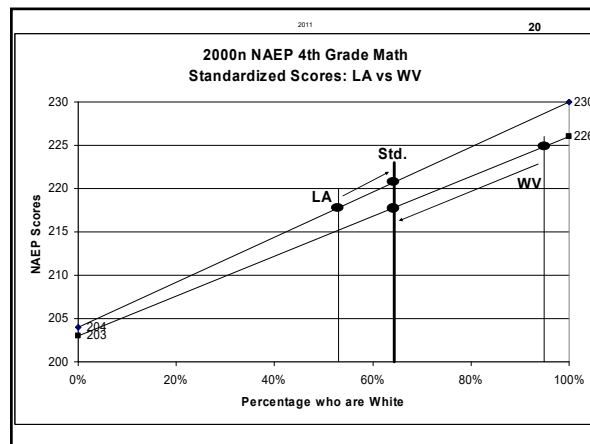
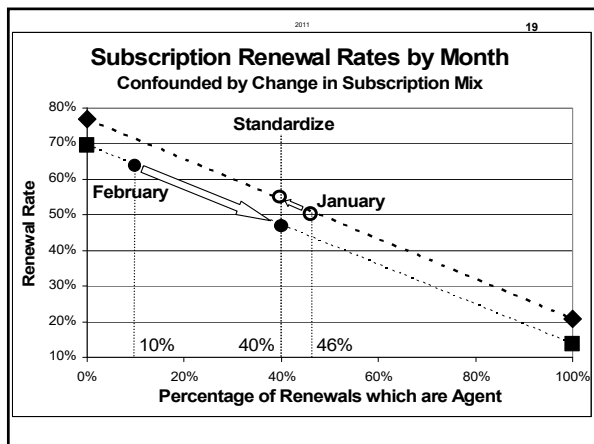
Condition	Total	Died	Death Rate
Good	800	15	1.90%
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Confounder Reverses; City Hospital is Better

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2011 24

Conclusion

Statistical educators must show students how confounders can influence associations and change statistical significance. The failure of educators to do this may be seen as “*statistical negligence*.”

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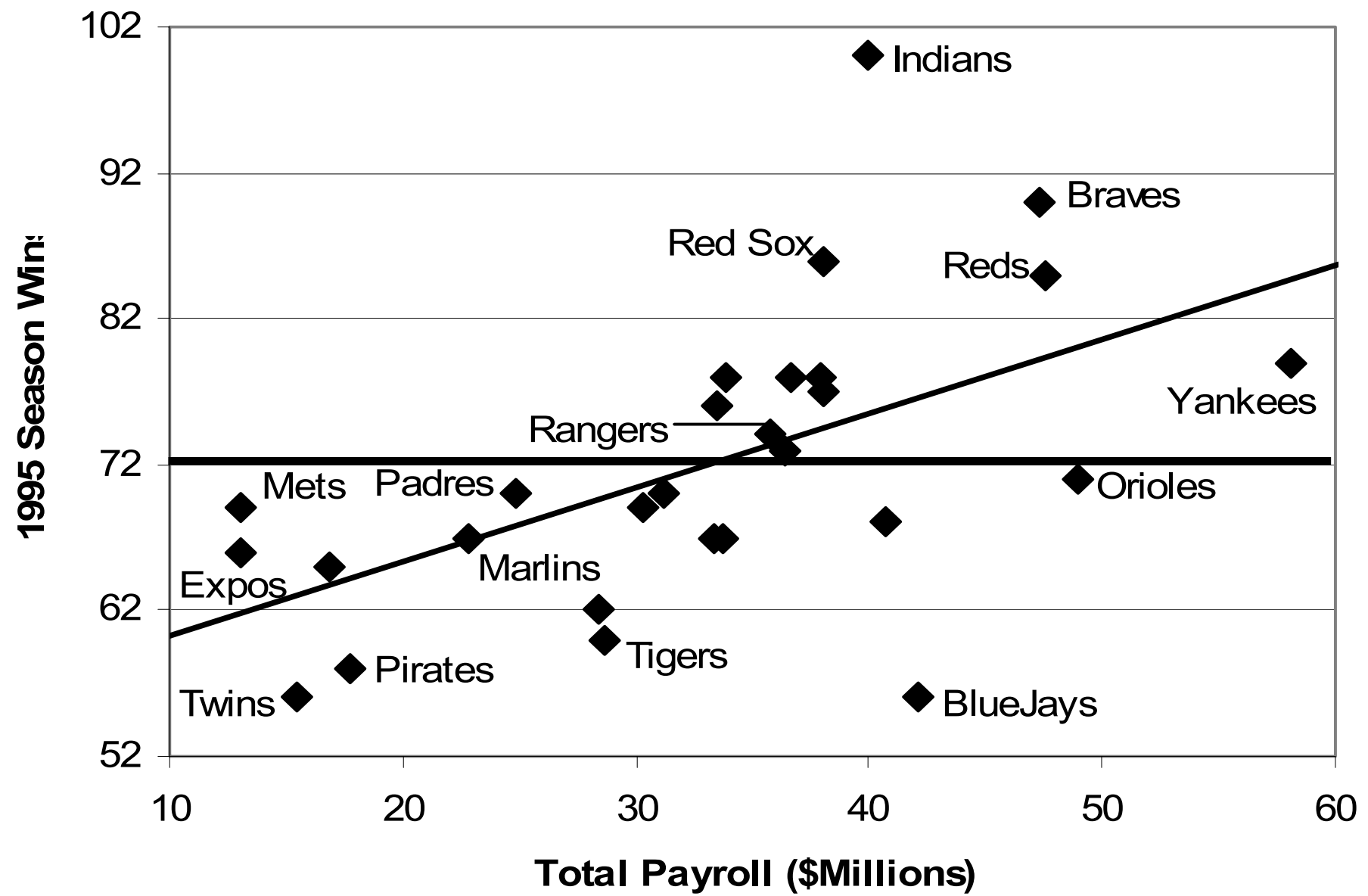
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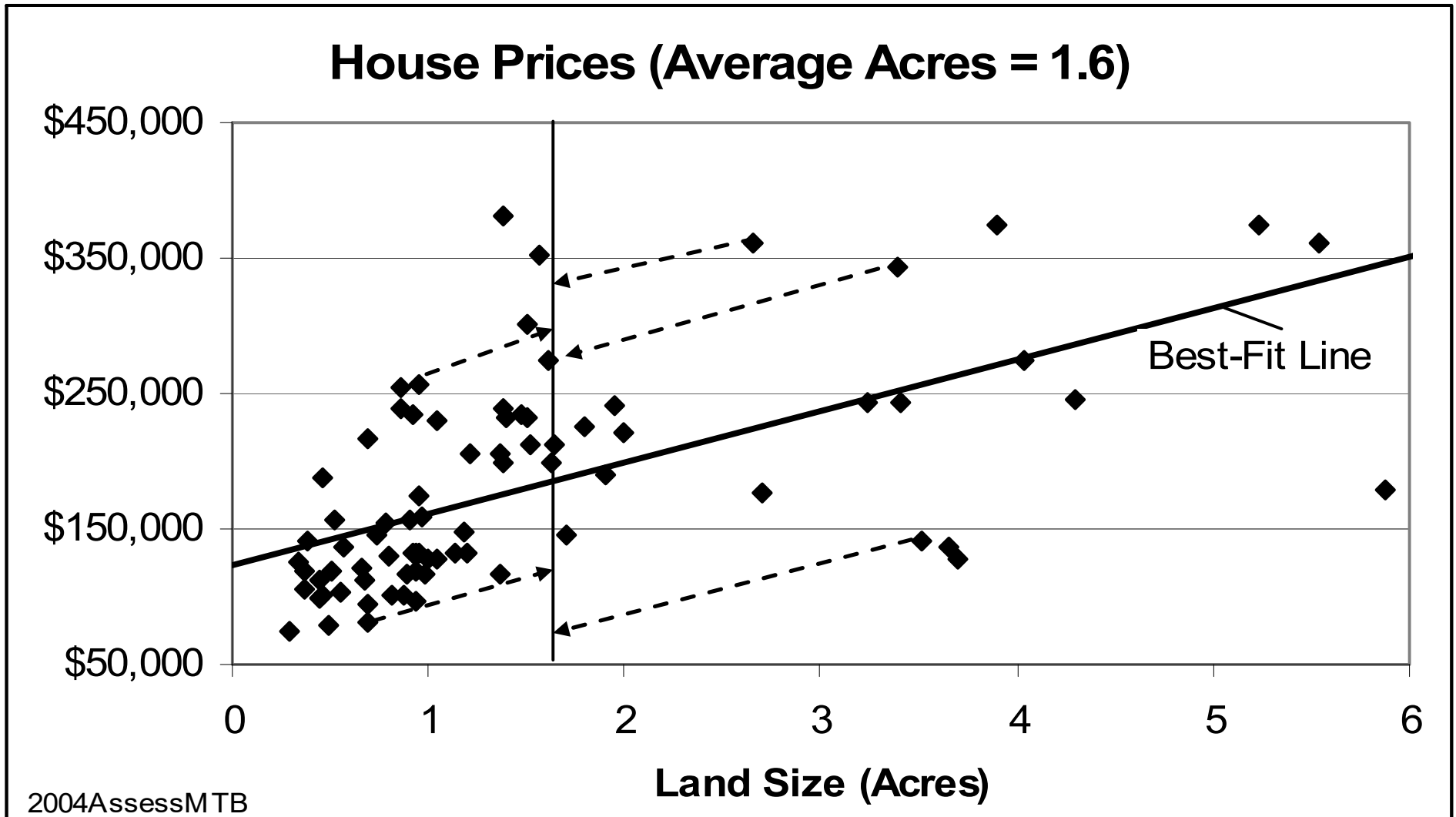
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SEASON WINS vs. TOTAL PAYROLL

US Major League Baseball



Adjusting for Land Size: Standardize on Average Lot



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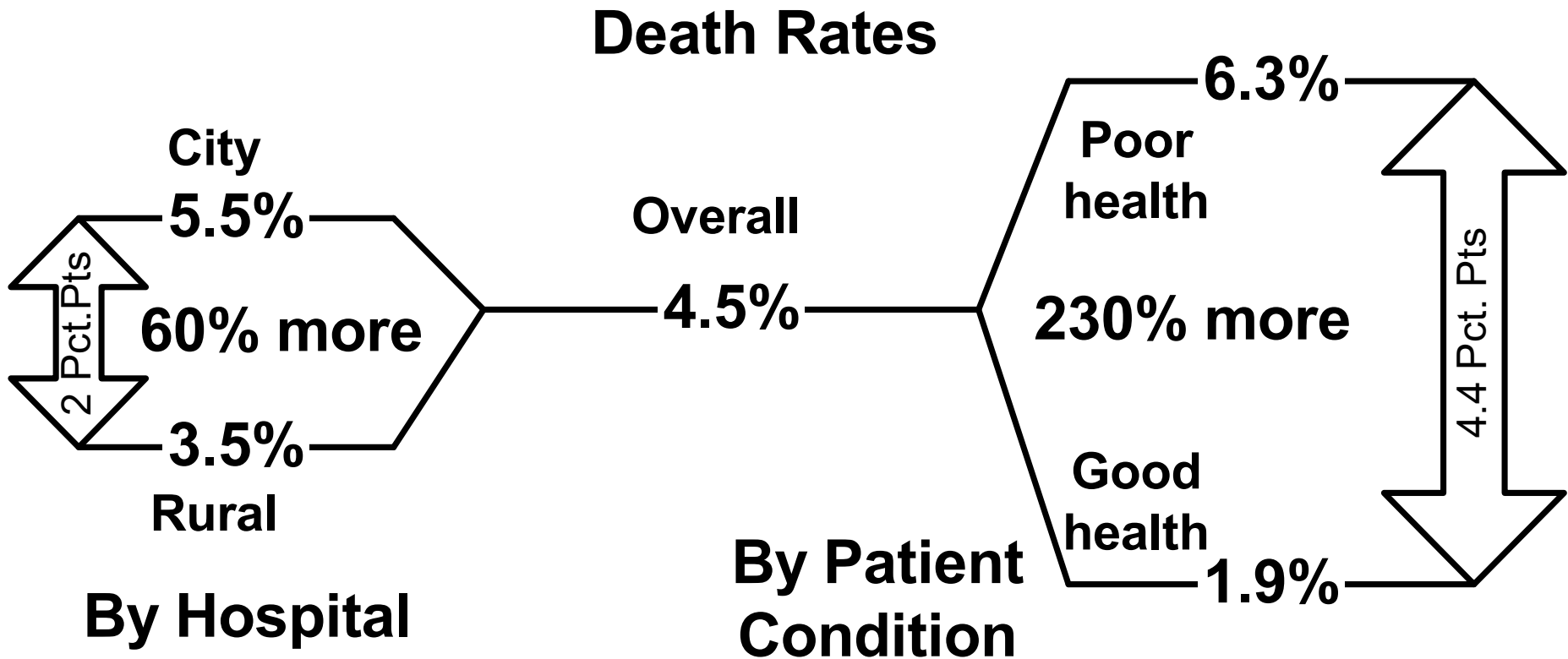
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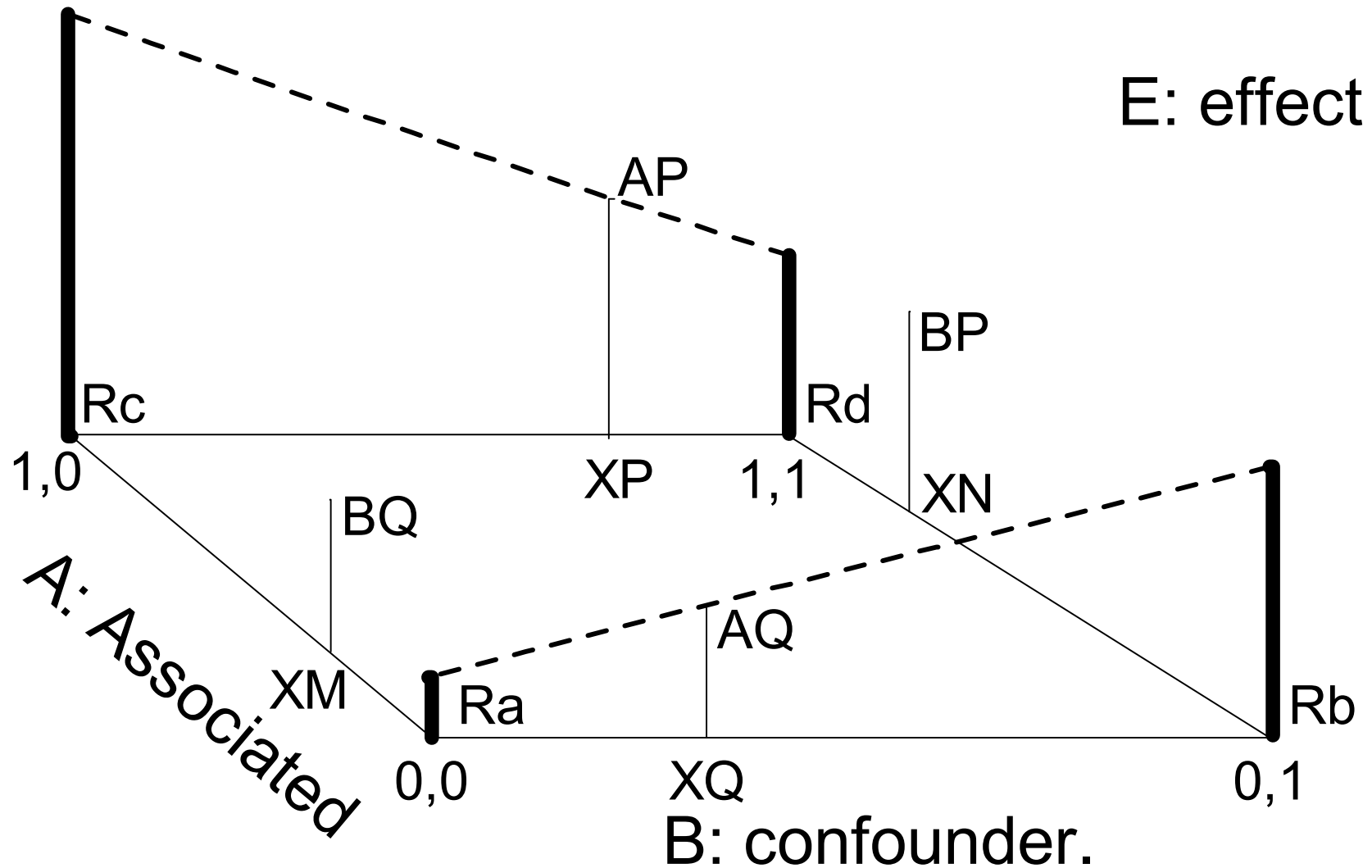
Can this confounder nullify or reverse this association?



Confounder Reverses; City Hospital is Better

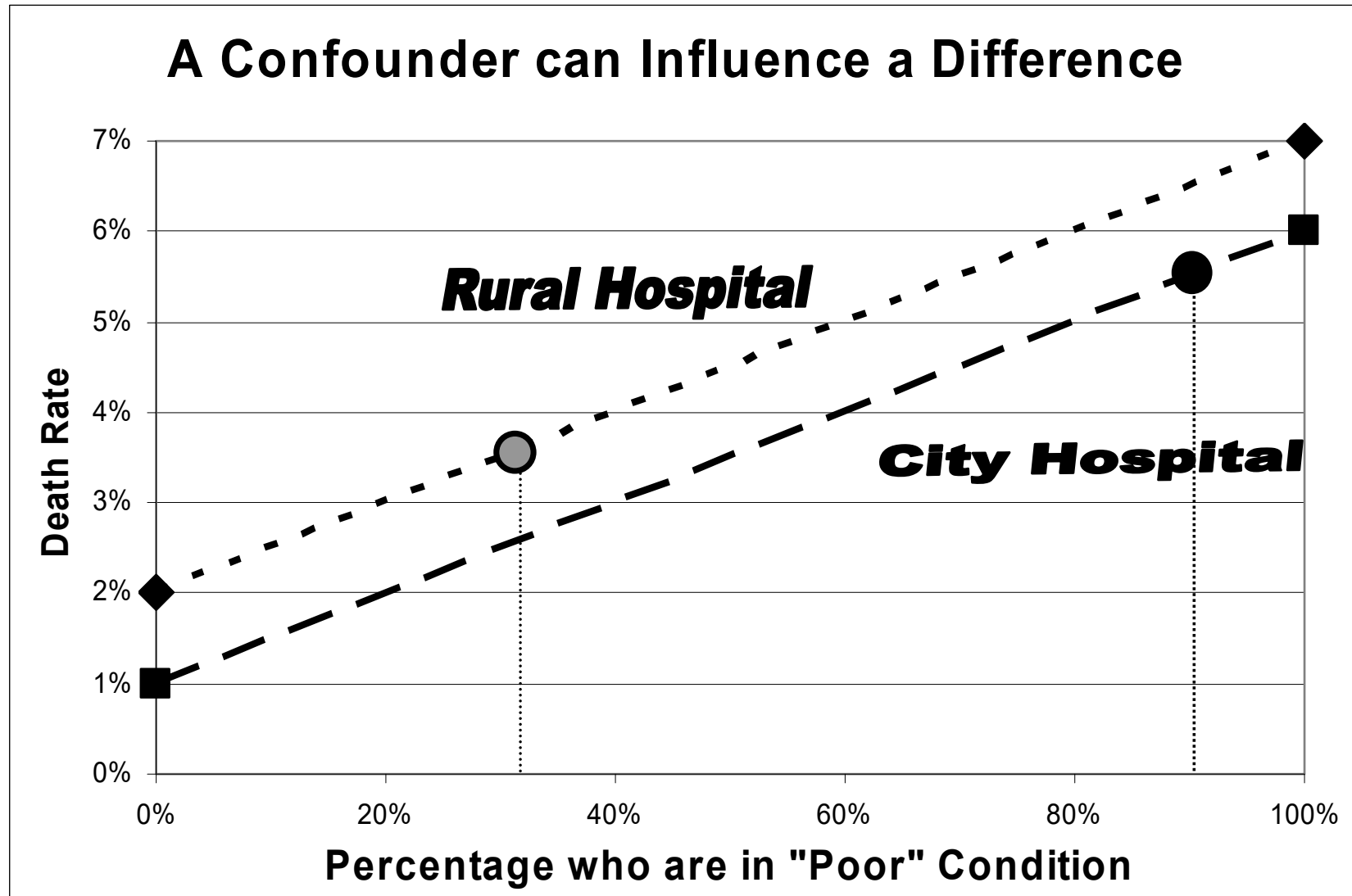
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Two-Group Rates with a Binary Confounder

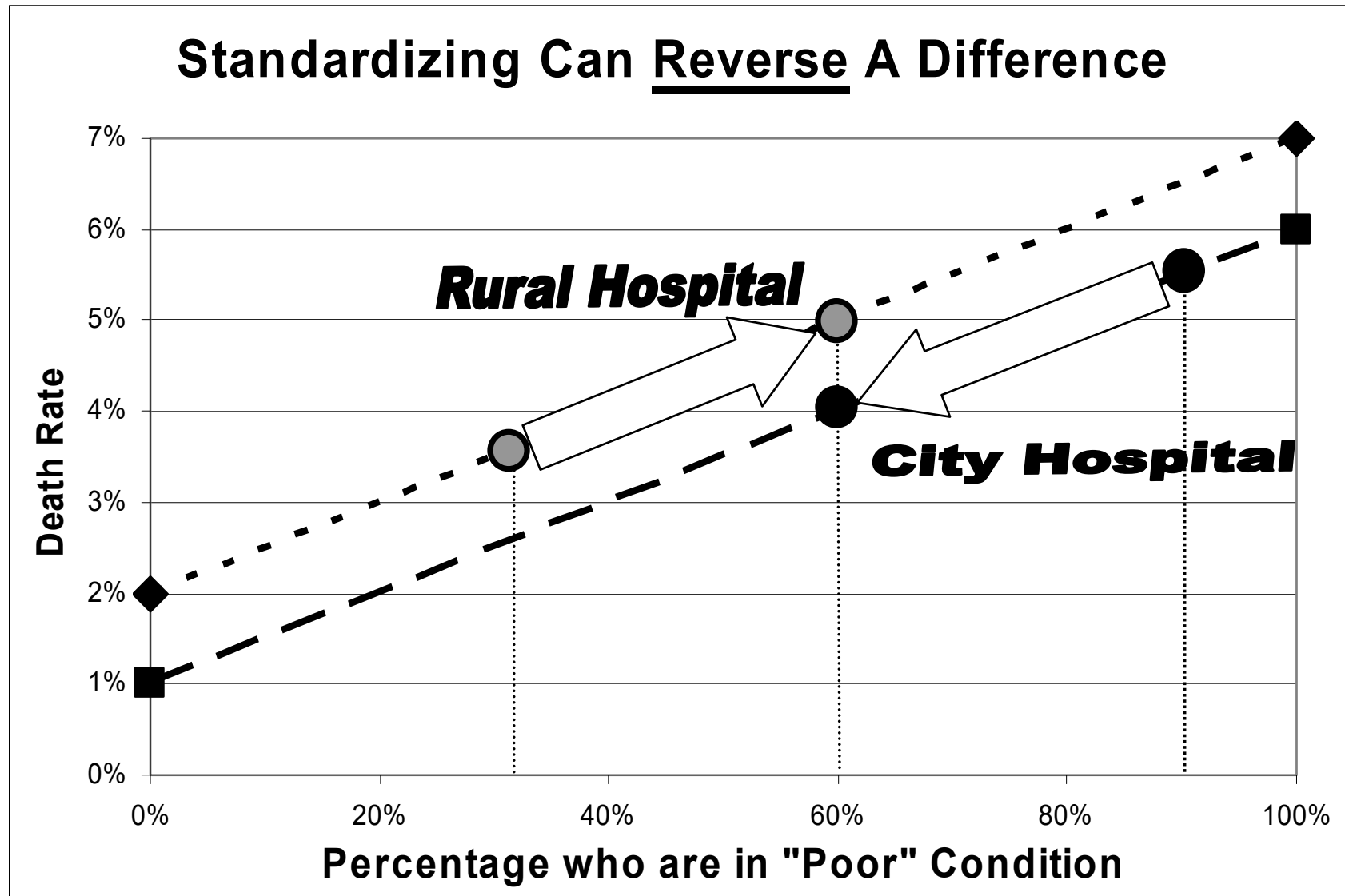


Compare Hospital Death Rates

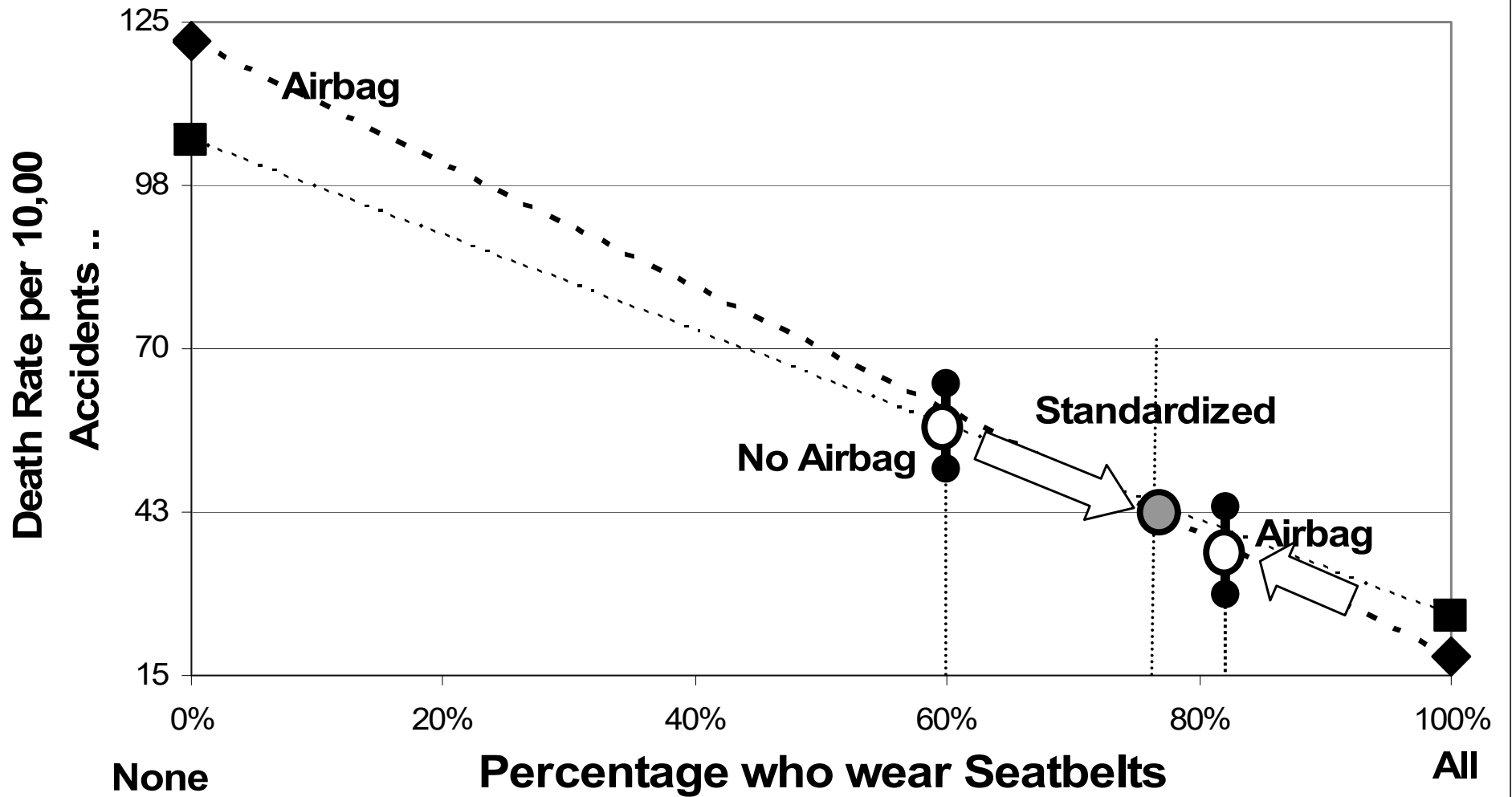
Confounder: Patient Condition



Standardize on combined confounder percentage

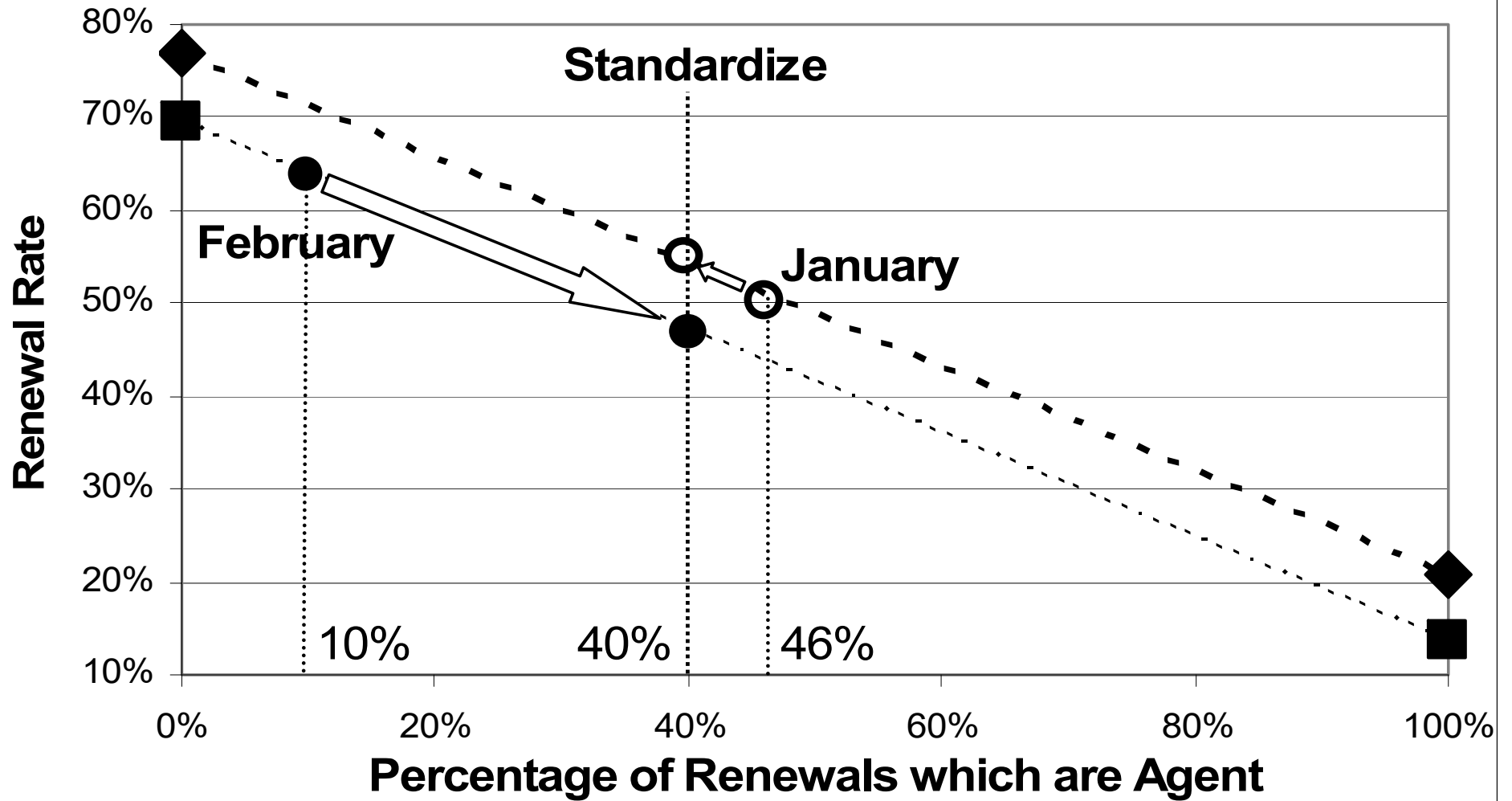


Auto Deaths and Airbag Presence Confounded by Seatbelt Use

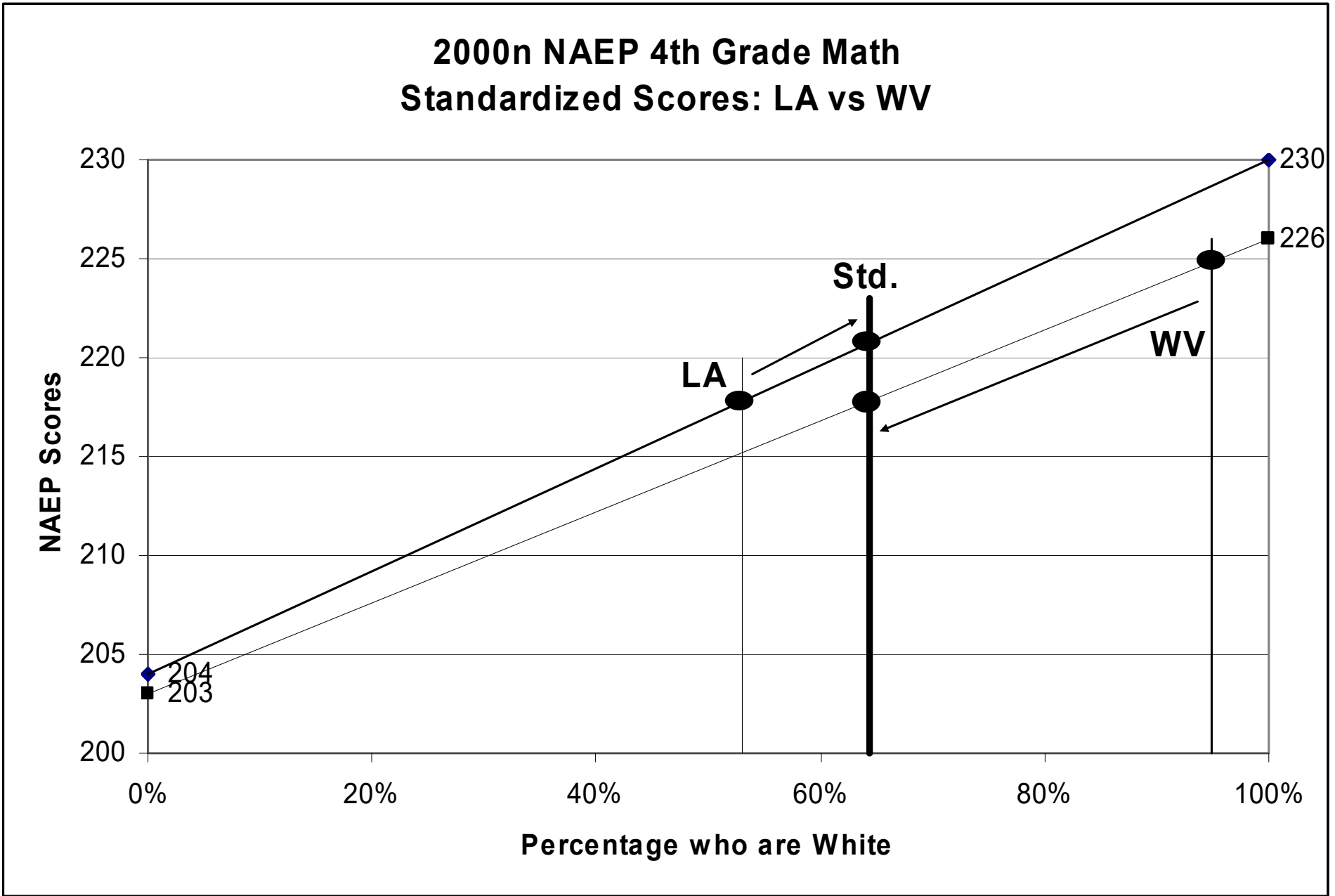


Subscription Renewal Rates by Month

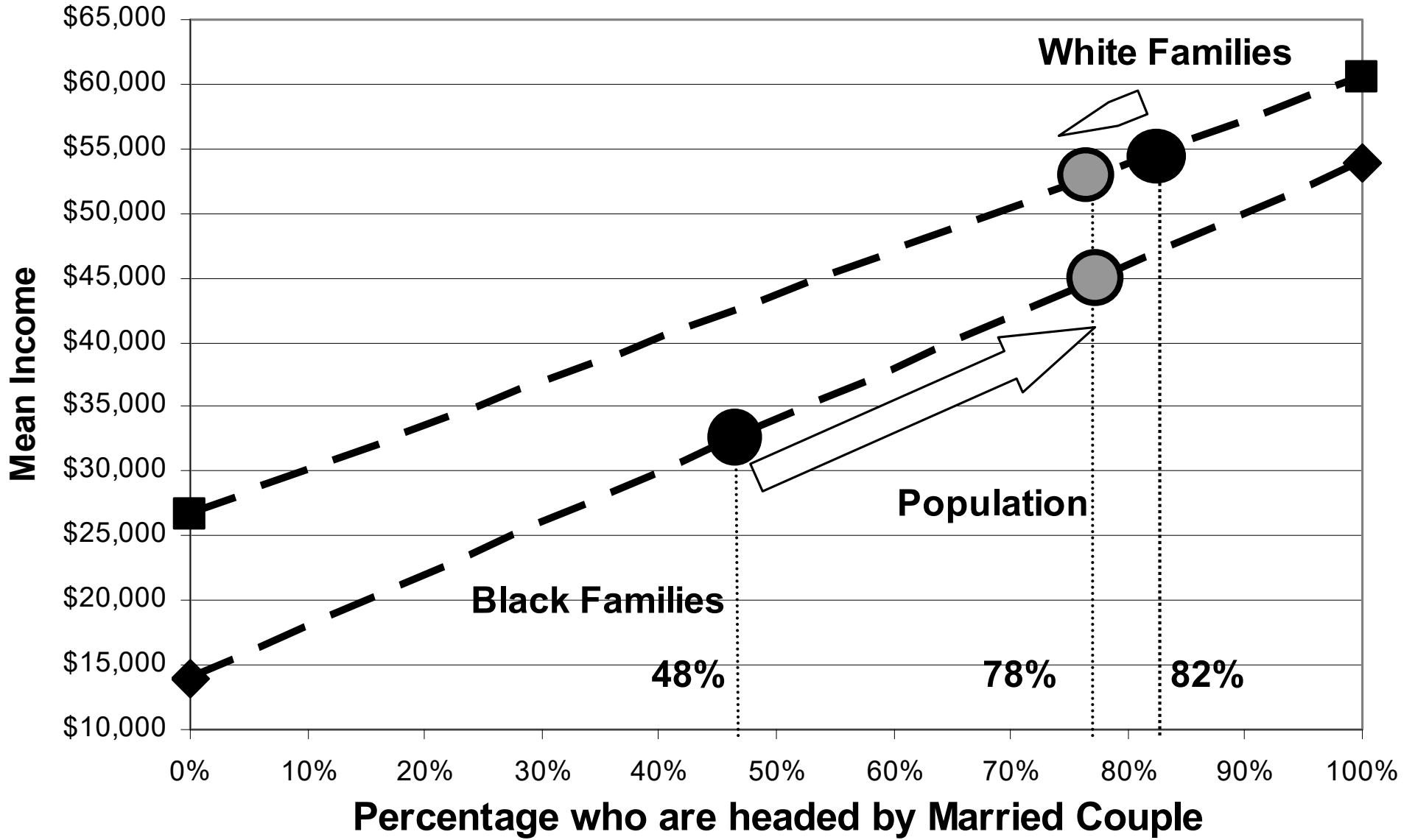
Confounded by Change in Subscription Mix

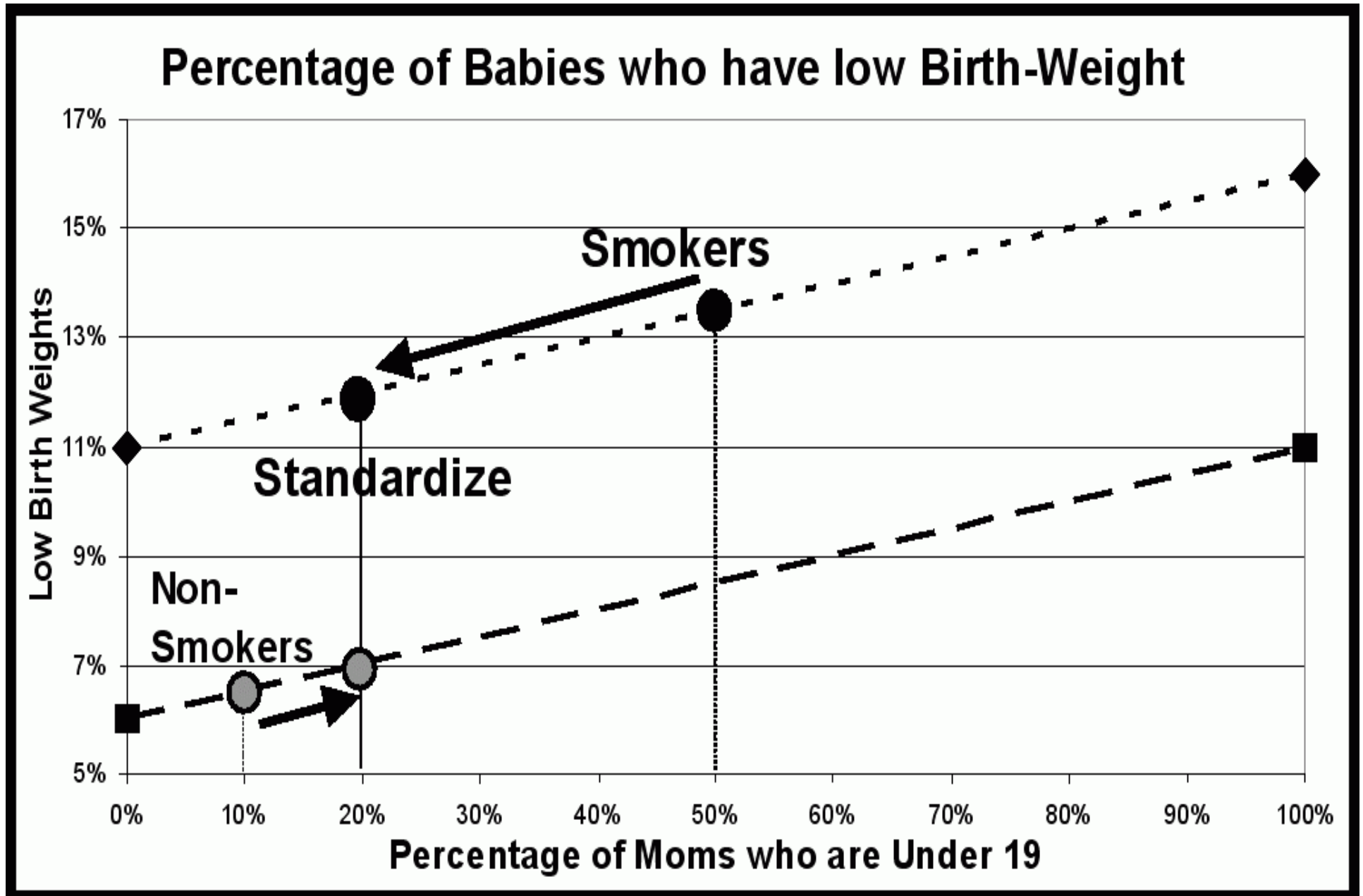


2000n NAEP 4th Grade Math Standardized Scores: LA vs WV

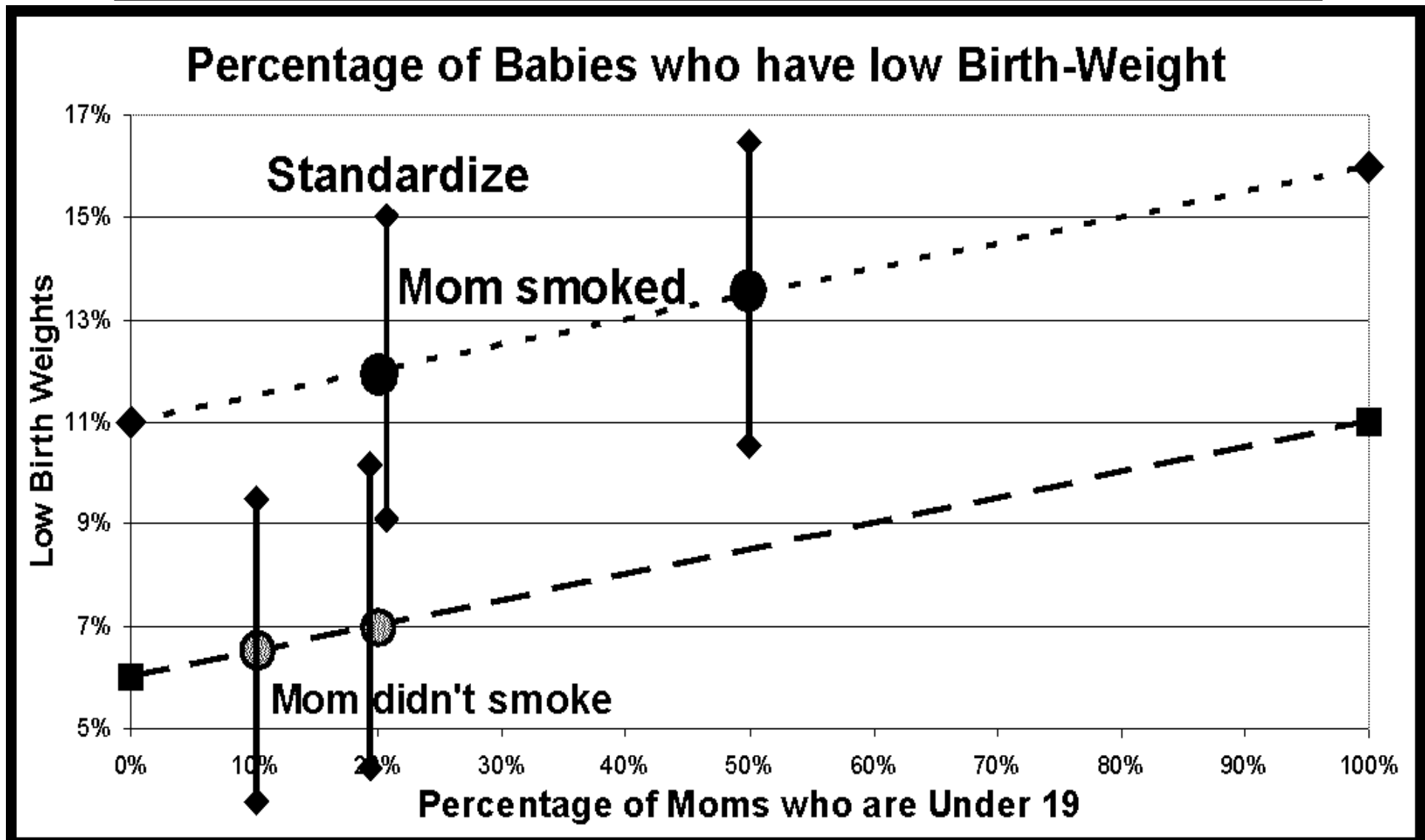


Income: US Families by Race & Structure





Controlling Can Change Statistical Significance



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