

# Tipping Point Sensitivity Analyses

Lucy D'Agostino McGowan

Wake Forest University

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# Recall: Propensity scores

Rosenbaum and Rubin showed in observational studies, conditioning on **propensity scores** can lead to unbiased estimates of the exposure effect

- 1 There are no unmeasured confounders
- 2 Every subject has a nonzero probability of receiving either exposure

# Quantifying Unmeasured Confounding

What you'll need:

- 1 The exposure-outcome effect
- 2 The unmeasured confounder-exposure effect
- 3 The unmeasured confounder-outcome effect

**What will tip our confidence bound to cross zero?**

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## Tipping point

$$\beta_{UO}(LB_{obs}, \delta)$$

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$\beta_{UO}$ : the **unmeasured confounder-outcome effect**



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## Tipping point

$$\beta_{UO}(LB_{obs}, \delta)$$

$LB_{obs}$ : **limiting bound** - the bound closest to the null

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## Tipping point

$$\beta_{UO}(LB_{obs}, \delta)$$

$\delta$ : **standardized mean difference** of the unmeasured confounder between the exposed and unexposed groups

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# Tipping Point

$$\beta_{UO}(LB_{obs}, \delta) = \frac{LB_{obs}}{\delta}$$

## Tipping Point

$$\delta(LB_{obs}, \beta_{UO}) = \frac{LB_{obs}}{\beta_{UO}}$$







tipr

## Main function

`lm_tip()`

**d: a data frame that includes the observed confidence bounds**

# Quantifying Unmeasured Confounding

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## Main function

`lm_tip()`

**smd: scaled mean difference  
between the unmeasured  
confounder in the exposed and  
unexposed population**

# Quantifying Unmeasured Confounding

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## Main function

lm\_tip()

**outcome\_association: association  
between the unmeasured  
confounder and outcome**

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**Main function**

**lm\_tip()**

d

smd

outcome\_association

# Main function

lm\_tip()

**specify** one, it will **estimate** the other

d

smd

outcome\_association

# Example

**Our causal effect estimate: 3.5 lbs (95% CI 2.4 lbs, 4.4 lbs)**

```
library(tipr)
lm_tip(data.frame(conf.low = 2.4,
                  conf.high = 4.4),
        smd = 0.3)
```

```
## # A tibble: 1 × 5
##   observed_lb observed_ub
##   <dbl>      <dbl>
## 1      2.4      4.4
##   smd outcome_associat... n_unmeasured_conf...
##   <dbl>      <dbl>      <dbl>
## 1    0.3          8          1
```

The observed effect (2.4, 4.4) WOULD be tipped by 1 unmeasured confounder with the following specifications:

**estimated standardized mean difference between the unmeasured confounder in the exposed population and unexposed population: 0.3**

## Your turn

- 1 **Use the `lm_tip()` function to conduct a sensitivity analysis for the estimate from your previous exercises.**

10:00