# Tipping Point Sensitivity Analyses

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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
- Every subject has a nonzero probability of receiving either exposure

- 1 The exposure-outcome effect
- **The exposure-unmeasured counfounder effect**
- 3 The unmeasured confounder-outcome effect

# What will tip our confidence bound to cross zero?

- 1 The exposure-outcome effect
- **The exposure-unmeasured counfounder effect**
- 3 The unmeasured confounder-outcome effect

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

# **Tipping point**

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

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

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

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- **2** The exposure-unmeasured counfounder effect
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# **Tipping point**

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

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

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

# **Tipping point**

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

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

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

# **Tipping Point**

$$eta_{UO}(LB_{obs},\delta) = rac{LB_{obs}}{\delta}$$

# **Tipping Point**

$$\delta(LB_{obs},eta_{UO})=rac{LB_{obs}}{eta_{UO}}$$

# tipr

#### Main function

tip\_coef()

effect\_observed: observed exposure - outcome effect

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

#### **Main function**

tip\_coef()

exposure\_confounder\_effect: scaled mean difference between the unmeasured confounder in the exposed and unexposed population

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

#### **Main function**

tip\_coef()

confounder\_outcome\_effect: relationship between the unmeasured confounder and outcome

- 1 The exposure-outcome effect
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#### **Main function**

tip\_coef()

effect\_observed

exposure\_confounder\_effect

confounder\_outcome\_effect

#### **Main function**

tip\_coef()

specify one, it will estimate the other

effect

exposure\_confounder\_effect

confounder\_outcome\_effect

#### **Example**

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

```
library(tipr)
tip coef(effect observed = 2.4,
          exposure confounder_effect = 0.3)
## # A tibble: 1 \times 4
##
     effect observed exposure confounder effect
                <fdb>
##
                                             <fdh>>
                  2.4
## 1
                                               0.3
##
     confounder outcome effect n unmeasured confounders
                           <fdb>
##
                                                      <fdb>>
## 1
                               8
```

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 tip\_coef() function to conduct a sensitivity analysis for the estimate from your previous exercises.