# Causal Modeling in R: Whole Game

2020-07-29 (updated: 2020-07-28)

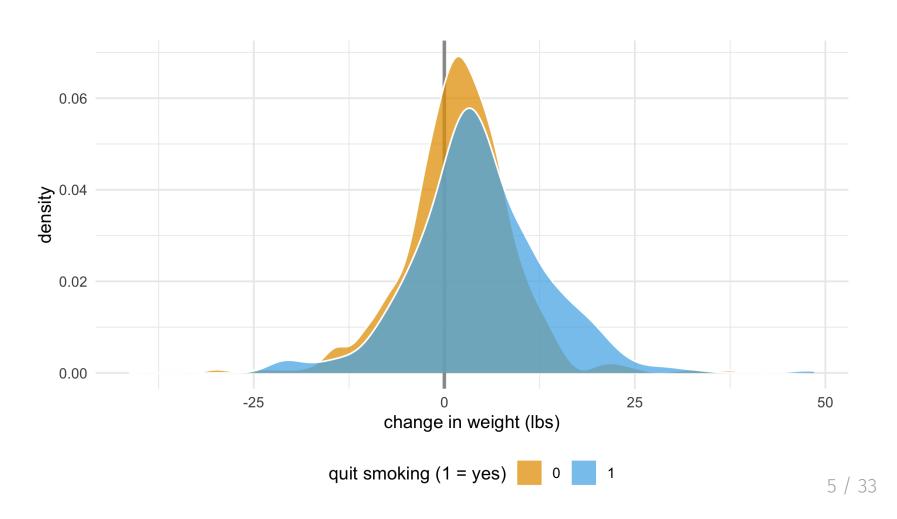
#### **Broad strokes**

- Specify causal question
- Draw assumptions (causal diagram)
- Model assumptions (propensity score)
- 4 Analyze propensities (diagnostics)
- 5 Estimate causal effects (IPW)

# Do people who quit smoking gain weight?

```
library(cidata)
 nhefs_complete_uc <- nhefs_complete %>%
   filter(censored == 0)
 nhefs complete uc
## # A tibble: 1,566 x 67
       segn gsmk death yrdth modth dadth
##
                                                sbp
                                                      dbp sex
##
      <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <fct>
        233
                                                175
###
    1
                 0
                        0
                             NA
                                    NA
                                           NA
                                                        96 0
                                                123
    2
        235
                        0
                                                        80 0
###
                 0
                             NA
                                    NA
                                           NA
    3
        244
                 0
                        0
                                    NA
                                                115
                                                        75 1
###
                             NA
                                           NA
4F4F
        245
                 0
                        1
                             85
                                                148
                                                        78 0
    4
                                           14
        252
                                                118
                                                        77 0
    5
                        0
###
                 0
                             NA
                                    NA
                                           NA
        257
                                                141
                                                        83 1
###
    6
                 0
                        0
                             NA
                                    NA
                                           NA
4F4F
        262
                 0
                        0
                                                132
                                                        69 1
                             NA
                                    NA
                                           NA
                                                        53 1
4F4F
    8
        266
                 0
                        0
                                                100
                             NA
                                    NA
                                           NA
                                                        79 0
    9
        419
                        1
                                                163
###
                 0
                             84
                                    10
                                           13
                        1
                                           17
## 10
        420
                 \odot
                             86
                                    10
                                                184
                                                       106 0
## # ... with 1,556 more rows, and 58 more variables: age <dbl>,
### ##
       race <fct>, income <dbl>, marital <dbl>, school <dbl>,
### ##
       education <fct>, ...
```

# Did those who quit smoking gain weight?



## Did those who quit smoking gain weight?

```
# ~2.5 lbs gained for quit vs. not quit
 nhefs_complete_uc %>%
   group_by(qsmk) %>%
   summarize(
     mean_weight_change = mean(wt82_71),
     sd = sd(wt82_71),
     .groups = "drop"
## # A tibble: 2 x 3
4F4F
      qsmk mean weight change
                                  sd
4F4F
     <dbl>
                         <dbl> <dbl>
## 1
                         1.98 7.45
## 2
                         4.53 8.75
```

### draw your assumptions

### What do I need to control for?

### Multivariable regression: what's the association?

```
lm(
  wt82_71~ qsmk + sex +
    race + age + I(age^2) + education +
    smokeintensity + I(smokeintensity^2) +
    smokeyrs + I(smokeyrs^2) + exercise + active +
    wt71 + I(wt71^2),
    data = nhefs_complete_uc
) %>%
    tidy(conf.int = TRUE) %>%
    filter(term == "qsmk")
```

### Multivariable regression: what's the association?

```
lm(
  wt82 71\sim qsmk + sex +
    race + age + I(age^2) + education +
    smokeintensity + I(smokeintensity^2) +
    smokeyrs + I(smokeyrs^2) + exercise + active +
    wt71 + I(wt71^2),
  data = nhefs_complete uc
) %>%
  tidy(conf.int = TRUE) %>%
  filter(term == "qsmk")
## # A tibble: 1 x 7
## term estimate std.error statistic p.value conf.low
  <chr> <dbl> <dbl>
                                <dbl> <dbl>
                                                 <dbl>
4F4F
## 1 qsmk 3.46 0.438 7.90 5.36e-15
                                                  2.60
## # ... with 1 more variable: conf.high <dbl>
```

### model your assumptions

# counterfactual: what if <u>everyone</u> quit smoking vs. what if <u>no one</u> quit smoking

#### Fit propensity score model

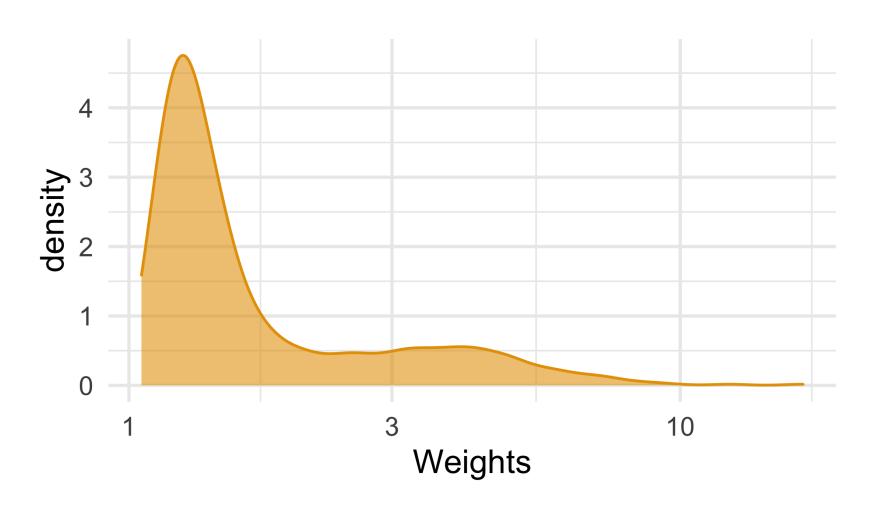
```
propensity_model <- glm(
   qsmk ~ sex +
    race + age + I(age^2) + education +
    smokeintensity + I(smokeintensity^2) +
    smokeyrs + I(smokeyrs^2) + exercise + active +
    wt71 + I(wt71^2),
   family = binomial(),
   data = nhefs_complete_uc
)</pre>
```

### Calculate inverse probability weights

```
nhefs_complete_uc <- propensity_model %>%
  # predict whether quit smoking
  augment(type.predict = "response", data = nhefs_complete_uc) %>%
  # calculate inverse probability
  mutate(wts = 1 / ifelse(qsmk == 0, 1 - .fitted, .fitted))
```

# diagnose your model assumptions

### What's the distribution of weights?



### estimate the causal effects

#### Estimate causal effect with IPW

```
ipw_model <- lm(
  wt82_71 ~ qsmk,
  data = nhefs_complete_uc,
  weights = wts
)

ipw_estimate <- ipw_model %>%
  tidy(conf.int = TRUE) %>%
  filter(term == "qsmk")
```

#### Estimate causal effect with IPW

### Let's fix our confidence intervals with the bootstrap!

```
# fit ipw model for a single bootstrap sample
fit_ipw_not_quite_rightly <- function(split, ...) {
    # get bootstrapped data sample with `rsample::analysis()`
    .df <- analysis(split)

# fit ipw model
lm(wt82_71 ~ qsmk, data = .df, weights = wts) %>%
    tidy()
}
```

```
fit ipw <- function(split, ...) {</pre>
  .df <- analysis(split)</pre>
  # fit propensity score model
  propensity model <- glm(</pre>
    asmk ~ sex +
      race + age + I(age^2) + education +
      smokeintensity + I(smokeintensity^2) +
      smokeyrs + I(smokeyrs^2) + exercise + active +
      wt71 + I(wt71^2),
    family = binomial(),
    data = .df
  # calculate inverse probability weights
  .df <- propensity model %>%
    augment(type.predict = "response", data = .df) %>%
    mutate(wts = 1 / ifelse(qsmk == 0, 1 - .fitted, .fitted))
  # fit correctly bootsrapped ipw model
  lm(wt82\ 71 \sim gsmk, data = .df, weights = wts) %>%
    tidy()
7
```

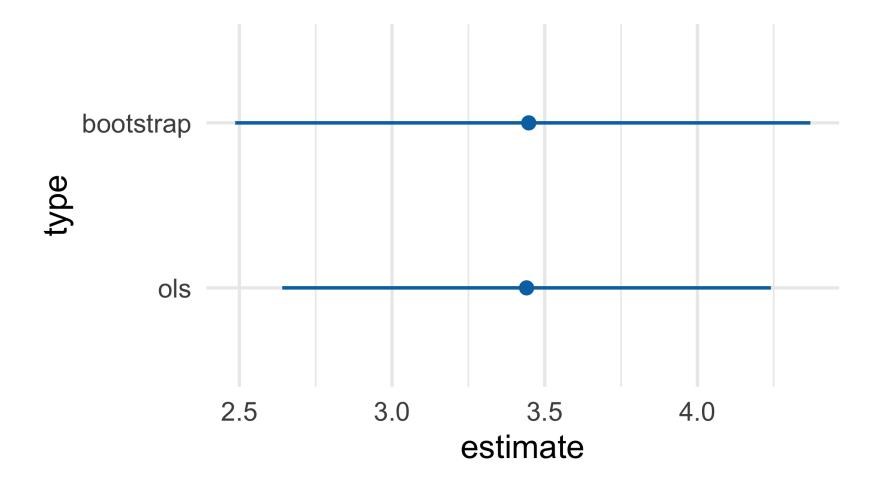
### Using {rsample} to bootstrap our causal effect

```
# fit ipw model to bootstrapped samples
ipw_results <- bootstraps(nhefs_complete, 1000, apparent = TRUE) %>%
    mutate(results = map(splits, fit_ipw))
```

### Using {rsample} to bootstrap our causal effect

```
# get t-statistic-based CIs
boot_estimate <- int_t(ipw_results, results) %>%
  filter(term == "qsmk")
boot_estimate
```

### Using {rsample} to bootstrap our causal effect



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

# Review the R Markdown file... later!

#### Resources

Causal Inference: Comprehensive text on causal inference. Free online.

Causal Inference Notebook: R code to go along with Causal Inference

Bootstrap confidence intervals with {rsample}