Causal Modeling in R: Whole Game

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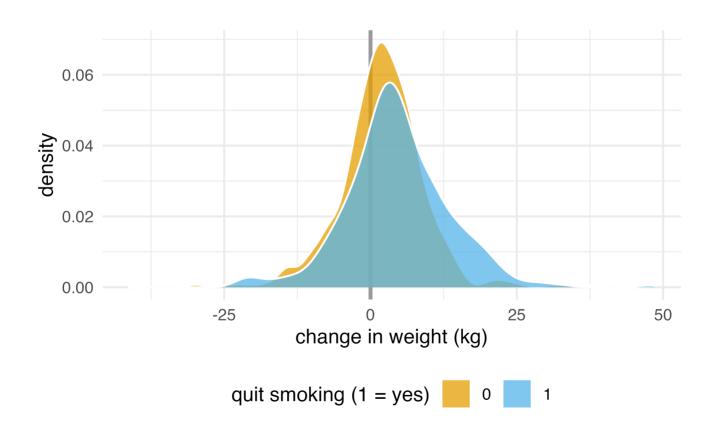
- Specify causal question
- 2 Draw assumptions (causal diagram)
- 3 Model assumptions (e.g. propensity score)
- 4 Analyze propensities (diagnostics)
- 5 Estimate causal effects (e.g. IPW)
- 6 Sensitivity analysis (more later!)

Do people who quit smoking gain weight?

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

education <fct>, ...

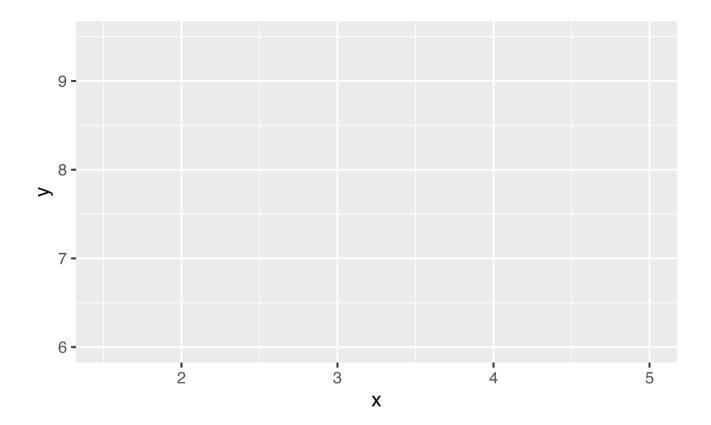
Did those who quit smoking gain weight?

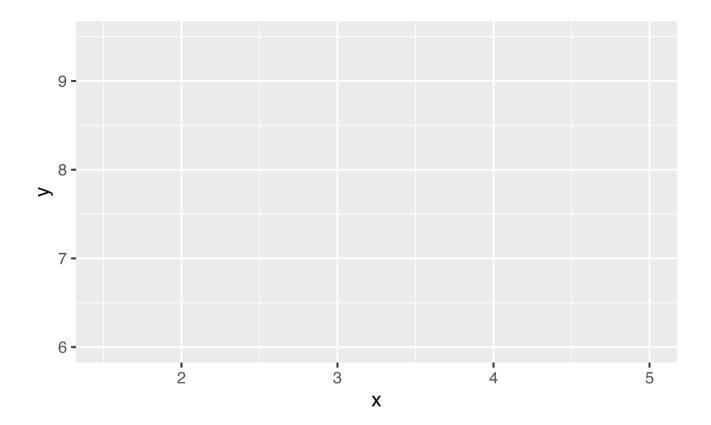


Did those who quit smoking gain weight?

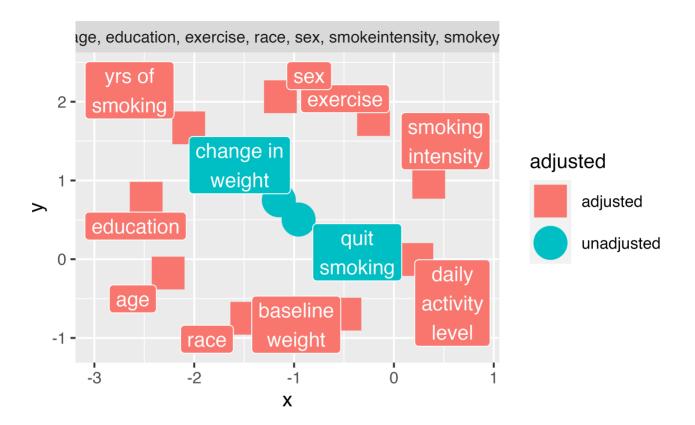
```
# ~2.5 KGs gained for quit vs. not quit
nhefs_complete_uc %>%
  group_by(qsmk) %>%
  summarize(
    mean_weight_change = mean(wt82_71),
    sd = sd(wt82_71),
    .groups = "drop"
)
```

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 × 7
## term estimate std.error statistic p.value conf.low
## <chr> <dbl> <dbl> <dbl> <dbl>
## 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

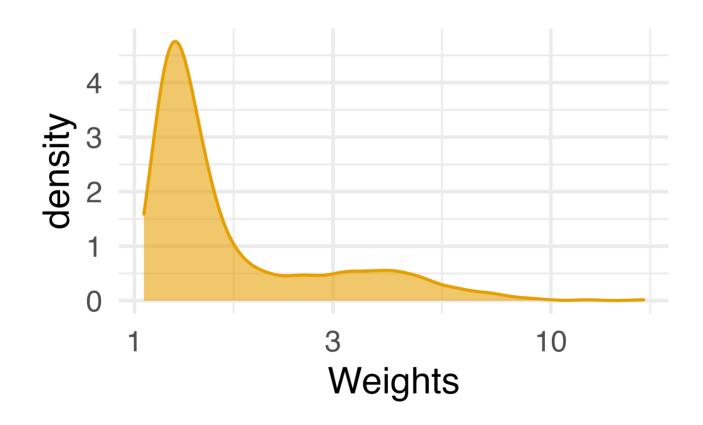
```
propensity_model <- glm(
   qsmk ~ sex +
    race + age + I(age^2) + education +
    smokeintensity + I(smokeintensity^2) +
    smokeyrs + I(smokeyrs^2) + exercise + active +
    wt71 + I(wt71^2),
   family = binomial(),
   data = nhefs_complete_uc
)</pre>
```

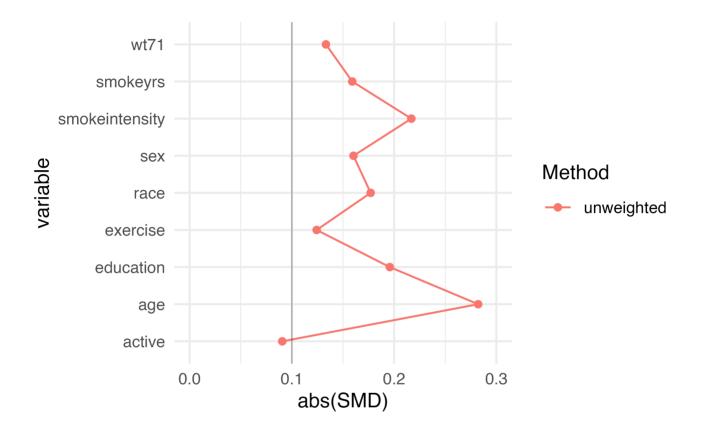
Calculate inverse probability weights

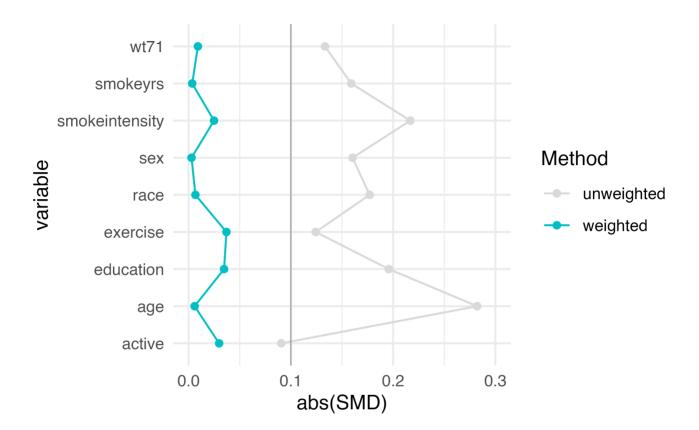
```
nhefs_complete_uc <- propensity_model %>%
  # predict whether quit smoking
  augment(type.predict = "response", data = nhefs_complete_uc) %>%
  # calculate inverse probability
  mutate(wts = 1 / ifelse(qsmk == 0, 1 - .fitted, .fitted))
```

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 robust SEs!

```
# also see robustbase, survey, gee, and others
library(estimatr)
ipw_model_robust <- lm_robust(
    wt82_71 ~ qsmk,
    data = nhefs_complete_uc,
    weights = wts
)

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

Let's fix our confidence intervals with robust SEs!

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>
    qsmk \sim 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(gsmk == 0, 1 - .fitted, .fitted))
  # fit correctly bootsrapped ipw model
  lm(wt82_71 ~ qsmk, data = .df, weights = wts) %>%
    tidy()
}
```

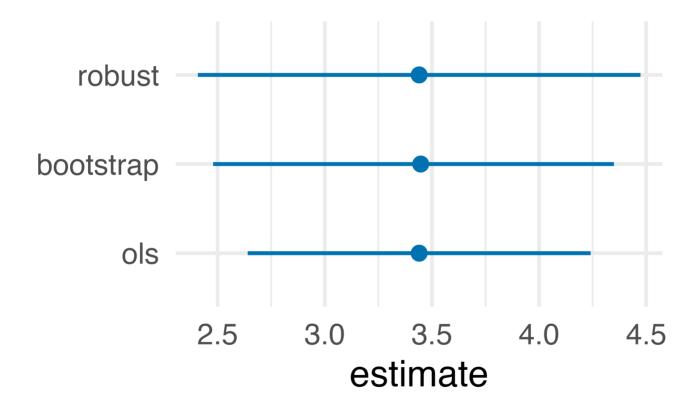
Using {rsample} to bootstrap our causal effect

```
# 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 kg (95% CI 2.4 kg, 4.4 kg)

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}