

Project 7: Difference-in-Differences and Synthetic Control

```
# Install and load packages
if (!require("pacman")) install.packages("pacman")

## Loading required package: pacman
devtools::install_github("ebenmichael/augsynth")

## Skipping install of 'augsynth' from a github remote, the SHA1 (0f4f1bcc) has not changed since last :
## Use `force = TRUE` to force installation

pacman::p_load(# Tidyverse packages including dplyr and ggplot2
               tidyverse,
               ggthemes,
               augsynth,
               gsynth)

# set seed
set.seed(44)

# load data
medicaid_expansion <- read_csv('./data/medicaid_expansion.csv')

## Rows: 663 Columns: 5

## -- Column specification -----
## Delimiter: ","
## chr  (1): State
## dbl  (3): year, uninsured_rate, population
## date (1): Date_Adopted
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

Introduction

For this project, you will explore the question of whether the Affordable Care Act increased health insurance coverage (or conversely, decreased the number of people who are uninsured). The ACA was passed in March 2010, but several of its provisions were phased in over a few years. The ACA instituted the “individual mandate” which required that all Americans must carry health insurance, or else suffer a tax penalty. There are four mechanisms for how the ACA aims to reduce the uninsured population:

- Require companies with more than 50 employees to provide health insurance.
- Build state-run healthcare markets (“exchanges”) for individuals to purchase health insurance.
- Provide subsidies to middle income individuals and families who do not qualify for employer based coverage.
- Expand Medicaid to require that states grant eligibility to all citizens and legal residents earning up to 138% of the federal poverty line. The federal government would initially pay 100% of the costs of this expansion, and over a period of 5 years the burden would shift so the federal government would pay 90% and the states would pay 10%.

In 2012, the Supreme Court heard the landmark case *NFIB v. Sebelius*, which principally challenged the constitutionality of the law under the theory that Congress could not institute an individual mandate. The Supreme Court ultimately upheld the individual mandate under Congress's taxation power, but struck down the requirement that states must expand Medicaid as impermissible subordination of the states to the federal government. Subsequently, several states refused to expand Medicaid when the program began on January 1, 2014. This refusal created the "Medicaid coverage gap" where there are individuals who earn too much to qualify for Medicaid under the old standards, but too little to qualify for the ACA subsidies targeted at middle-income individuals.

States that refused to expand Medicaid principally cited the cost as the primary factor. Critics pointed out however, that the decision not to expand primarily broke down along partisan lines. In the years since the initial expansion, several states have opted into the program, either because of a change in the governing party, or because voters directly approved expansion via a ballot initiative.

You will explore the question of whether Medicaid expansion reduced the uninsured population in the U.S. in the 7 years since it went into effect. To address this question, you will use difference-in-differences estimation, and synthetic control.

Data

The dataset you will work with has been assembled from a few different sources about Medicaid. The key variables are:

- **State:** Full name of state
- **Medicaid Expansion Adoption:** Date that the state adopted the Medicaid expansion, if it did so.
- **Year:** Year of observation.
- **Uninsured rate:** State uninsured rate in that year.

Exploratory Data Analysis

Create plots and provide 1-2 sentence analyses to answer the following questions:

- Which states had the highest uninsured rates prior to 2014? The lowest?
- Which states were home to most uninsured Americans prior to 2014? How about in the last year in the data set? **Note:** 2010 state population is provided as a variable to answer this question. In an actual study you would likely use population estimates over time, but to simplify you can assume these numbers stay about the same.

```
medicaid_expansion
```

```
## # A tibble: 663 x 5
##   State      Date_Adopted  year uninsured_rate population
##   <chr>      <date>      <dbl>         <dbl>      <dbl>
## 1 Alabama      NA          2008          0.140    4849377
## 2 Alaska  2015-09-01    2008          0.208     737732
## 3 Arizona  2014-01-01    2008          0.187    6731484
## 4 Arkansas  2014-01-01    2008          0.179    2994079
## 5 California 2014-01-01    2008          0.178   38802500
## 6 Colorado  2014-01-01    2008          0.170    5355856
## 7 Connecticut 2014-01-01    2008         0.0891    3596677
## 8 Delaware  2014-01-01    2008          0.108     935614
## 9 District of Columbia 2014-01-01    2008         0.0805         NA
## 10 Florida    NA          2008          0.209   19893297
## # i 653 more rows
```

```

# highest and lowest uninsured rates
pre2014 <- medicaid_expansion %>%
  filter(year < 2014)

# Filter for the highest and lowest 5% using dplyr
lowest_5pct <- pre2014 %>%
  filter(uninsured_rate <= quantile(uninsured_rate, 0.05))

highest_5pct <- pre2014 %>%
  filter(uninsured_rate >= quantile(uninsured_rate, 0.95))

lowest_uninsured_states <- pull(distinct(lowest_5pct, State))
highest_uninsured_states <- pull(distinct(highest_5pct, State))

print(lowest_uninsured_states)

## [1] "Hawaii" "Massachusetts" "District of Columbia"
## [4] "Vermont"

print(highest_uninsured_states)

## [1] "Florida" "Nevada" "New Mexico" "Utah" "Texas"

```

- Which states were home to most uninsured Americans prior to 2014? How about in the last year in the data set? **Note:** 2010 state population is provided as a variable to answer this question. In an actual study you would likely use population estimates over time, but to simplify you can assume these numbers stay about the same.

```

# most uninsured Americans

pre2014 <-
  pre2014 %>%

  mutate(uninsured_pop = (pre2014$uninsured_rate * pre2014$population))

state_uninsured <-
  pre2014 %>%

  group_by(State) %>%

  summarize(AvgUninsured = mean(uninsured_pop, na.rm = TRUE))

sorted_most <- state_uninsured %>%
  arrange(desc(AvgUninsured), State)
sorted_least <- state_uninsured %>%
  arrange(AvgUninsured, State)

sorted_most

## # A tibble: 51 x 2
##   State      AvgUninsured
##   <chr>      <dbl>
## 1 California 6970785.
## 2 Texas     5372081.
## 3 Florida   4115094.

```

```
## 4 New York          2262138.
## 5 Georgia           1935272.
## 6 Illinois          1693876.
## 7 North Carolina   1622044.
## 8 Ohio              1378521.
## 9 Pennsylvania     1260889.
## 10 Arizona          1180133.
## # i 41 more rows
```

```
sorted_least
```

```
## # A tibble: 51 x 2
##   State      AvgUninsured
##   <chr>      <dbl>
## 1 Vermont      60715.
## 2 North Dakota  74965.
## 3 Wyoming      75130.
## 4 Delaware     91424.
## 5 Hawaii       99643.
## 6 Rhode Island 112053.
## 7 South Dakota 114558.
## 8 Maine        142888.
## 9 New Hampshire 143760.
## 10 Alaska      145397.
## # i 41 more rows
```

What about in the last year of the dataset?

```
medicaid2020 <-
  filter(medicaid_expansion, year == 2020)

# Same analysis as above but with 2020 data

medicaid2020 <-
  medicaid2020 %>%
  mutate(uninsured_pop = (medicaid2020$uninsured_rate * medicaid2020$population))

state_uninsured2020 <-
  medicaid2020 %>%

  group_by(State) %>%

  summarize(AvgUninsured = mean(uninsured_pop, na.rm = TRUE))

sorted_most2020 <- state_uninsured2020 %>%
  arrange(desc(AvgUninsured), State)
sorted_least2020 <- state_uninsured2020 %>%
  arrange(AvgUninsured, State)

sorted_most2020
```

```
## # A tibble: 51 x 2
##   State      AvgUninsured
##   <chr>      <dbl>
## 1 Texas      4960080.
```

```
## 2 California      2987792.
## 3 Florida         2625915.
## 4 Georgia         1353044.
## 5 North Carolina  1123668.
## 6 New York        1026804.
## 7 Illinois        953163.
## 8 Ohio            765215.
## 9 Arizona         760658.
## 10 Pennsylvania   741658.
## # i 41 more rows
```

```
sorted_least2020
```

```
## # A tibble: 51 x 2
##   State      AvgUninsured
##   <chr>      <dbl>
## 1 Vermont    28170.
## 2 Rhode Island 43262.
## 3 North Dakota 51024.
## 4 Hawaii     59622.
## 5 Delaware    61751.
## 6 Wyoming     71851.
## 7 New Hampshire 83589.
## 8 Montana     84957.
## 9 South Dakota 87024.
## 10 Alaska     90003.
## # i 41 more rows
```

The rate of uninsured certainly appears to be more valuable given how much the raw values are contingent upon state population. That being said, seeing Texas overtake California with nearly double the number of uninsured individuals really drives home the premise that Texas might benefit from enrolling in Medicaid Expansion. I would also be interested to see this compared to the state budget or GDP to make an assessment of the premise that refusal to enroll was a “cost issue” rather than a partisan divide. As far as low rates, I’m not surprised to see Massachusetts given their own universal healthcare bill (RomneyCare lol) that likely distinguishes them from many if not all other states.

Difference-in-Differences Estimation

Estimate Model

Do the following:

- Choose a state that adopted the Medicaid expansion on January 1, 2014 and a state that did not. **Hint:** Do not pick Massachusetts as it passed a universal healthcare law in 2006, and also avoid picking a state that adopted the Medicaid expansion between 2014 and 2015.
- Assess the parallel trends assumption for your choices using a plot. If you are not satisfied that the assumption has been met, pick another state and try again (but detail the states you tried).

```
head(medicaid_expansion, n = 60)
```

```
## # A tibble: 60 x 5
##   State      Date_Adopted year uninsured_rate population
##   <chr>      <date>      <dbl>      <dbl>      <dbl>
## 1 Alabama    NA           2008         0.140    4849377
## 2 Alaska     2015-09-01  2008         0.208    737732
## 3 Arizona     2014-01-01  2008         0.187    6731484
```

```
## 4 Arkansas      2014-01-01  2008      0.179    2994079
## 5 California    2014-01-01  2008      0.178    38802500
## 6 Colorado      2014-01-01  2008      0.170    5355856
## 7 Connecticut   2014-01-01  2008      0.0891   3596677
## 8 Delaware      2014-01-01  2008      0.108     935614
## 9 District of Columbia 2014-01-01  2008      0.0805      NA
## 10 Florida      NA          2008      0.209    19893297
## # i 50 more rows
```

```
# Parallel Trends plotting
```

```
medicaid_expansion %>%
```

```
# process
```

```
# -----
```

```
filter(State %in% c("Hawaii", "Pennsylvania")) %>%
```

```
# north dakota, kentucky, nevada, arizona, Date_Adopted %in% c("2016-01-01") |
```

```
# plot
```

```
# -----
```

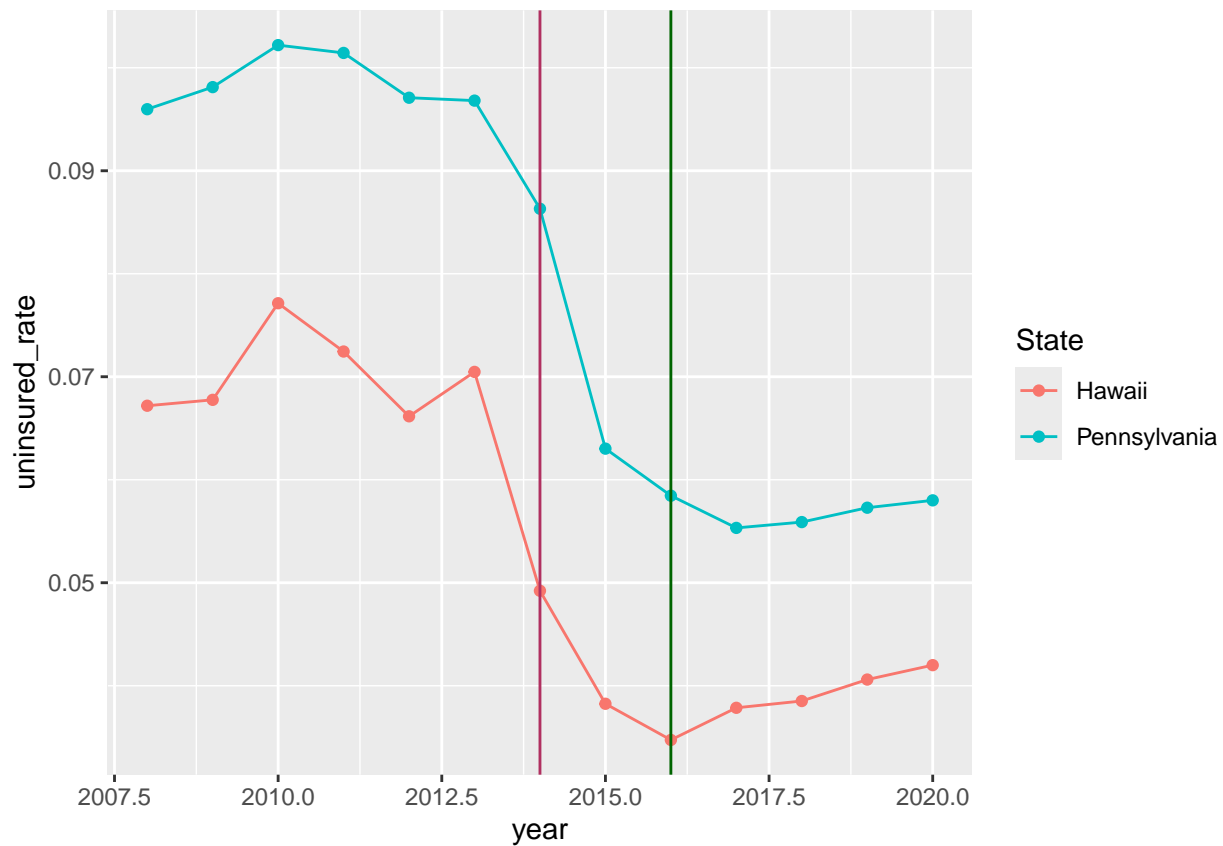
```
ggplot() +
```

```
geom_point(aes(x = year,
               y = uninsured_rate,
               color = State)) +
```

```
geom_line(aes(x = year,
              y = uninsured_rate,
              color = State)) +
```

```
geom_vline(xintercept = 2014, color = "maroon") +
```

```
geom_vline(xintercept = 2016, color = "darkgreen")
```



```

# themes
theme_fivethirtyeight() +
theme(axis.title = element_text()) +

# labels
ggtitle('States Enrolled on Jan 1, 2014 and Jan 1, 2016, \nbefore/after Enrollment') +
xlab('Year') +
ylab('Uninsured Rate')

```

```

## List of 138
## $ line                                     :List of 6
## ..$ colour                               : chr "black"
## ..$ linewidth                             : num 0.545
## ..$ linetype                             : num 1
## ..$ lineend                               : chr "butt"
## ..$ arrow                                : logi FALSE
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_line" "element"
## $ rect                                     :List of 5
## ..$ fill                                  : Named chr "#F0F0F0"
## .. ..- attr(*, "names")= chr "Light Gray"
## ..$ colour                               : logi NA
## ..$ linewidth                             : num 0.545
## ..$ linetype                             : num 0
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_rect" "element"
## $ text                                    :List of 11
## ..$ family                               : chr "sans"
## ..$ face                                  : chr "plain"
## ..$ colour                               : Named chr "#3C3C3C"
## .. ..- attr(*, "names")= chr "Dark Gray"
## ..$ size                                 : num 12
## ..$ hjust                                : num 0.5
## ..$ vjust                                : num 0.5
## ..$ angle                                : num 0
## ..$ lineheight                           : num 0.9
## ..$ margin                               : 'margin' num [1:4] 0points 0points 0points 0points
## .. ..- attr(*, "unit")= int 8
## ..$ debug                                : logi FALSE
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ title                                  : chr "States Enrolled on Jan 1, 2014 and Jan 1, 2016, \nbefore/a
## $ aspect.ratio                           : NULL
## $ axis.title                             :List of 11
## ..$ family                               : NULL
## ..$ face                                  : NULL
## ..$ colour                               : NULL
## ..$ size                                 : NULL
## ..$ hjust                                : NULL
## ..$ vjust                                : NULL
## ..$ angle                                : NULL
## ..$ lineheight                           : NULL
## ..$ margin                               : NULL
## ..$ debug                                : NULL

```

```

## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.x :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : NULL
## ..$ vjust : num 1
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 3points 0points 0points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.x.top :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : NULL
## ..$ vjust : num 0
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 0points 0points 3points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.x.bottom : NULL
## $ axis.title.y :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : NULL
## ..$ vjust : num 1
## ..$ angle : num 90
## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 0points 3points 0points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.y.left : NULL
## $ axis.title.y.right :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : NULL
## ..$ vjust : num 1
## ..$ angle : num -90

```



```

## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 0points 0points 0points 3points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : 'rel' num 0.8
## ..$ hjust : NULL
## ..$ vjust : NULL
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : NULL
## ..$ debug : NULL
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.x :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : NULL
## ..$ vjust : num 1
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 2.4points 0points 0points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.x.top :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : NULL
## ..$ vjust : num 0
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 0points 0points 2.4points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.x.bottom : NULL
## $ axis.text.y :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : num 1

```

```

## ..$ vjust          : NULL
## ..$ angle          : NULL
## ..$ lineheight     : NULL
## ..$ margin         : 'margin' num [1:4] 0points 2.4points 0points 0points
## .. ..- attr(*, "unit")= int 8
## ..$ debug          : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.y.left  : NULL
## $ axis.text.y.right :List of 11
## ..$ family         : NULL
## ..$ face           : NULL
## ..$ colour         : NULL
## ..$ size           : NULL
## ..$ hjust          : num 0
## ..$ vjust          : NULL
## ..$ angle          : NULL
## ..$ lineheight     : NULL
## ..$ margin         : 'margin' num [1:4] 0points 0points 0points 2.4points
## .. ..- attr(*, "unit")= int 8
## ..$ debug          : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.theta   : NULL
## $ axis.text.r       :List of 11
## ..$ family         : NULL
## ..$ face           : NULL
## ..$ colour         : NULL
## ..$ size           : NULL
## ..$ hjust          : num 0.5
## ..$ vjust          : NULL
## ..$ angle          : NULL
## ..$ lineheight     : NULL
## ..$ margin         : 'margin' num [1:4] 0points 2.4points 0points 2.4points
## .. ..- attr(*, "unit")= int 8
## ..$ debug          : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.ticks        : list()
## ..- attr(*, "class")= chr [1:2] "element_blank" "element"
## $ axis.ticks.x      : NULL
## $ axis.ticks.x.top   : NULL
## $ axis.ticks.x.bottom : NULL
## $ axis.ticks.y       : NULL
## $ axis.ticks.y.left  : NULL
## $ axis.ticks.y.right : NULL
## $ axis.ticks.theta   : NULL
## $ axis.ticks.r       : NULL
## $ axis.minor.ticks.x.top : NULL
## $ axis.minor.ticks.x.bottom : NULL
## $ axis.minor.ticks.y.left : NULL
## $ axis.minor.ticks.y.right : NULL
## $ axis.minor.ticks.theta : NULL
## $ axis.minor.ticks.r : NULL

```

```

## $ axis.ticks.length           : 'simpleUnit' num 3points
##   .- attr(*, "unit")= int 8
## $ axis.ticks.length.x         : NULL
## $ axis.ticks.length.x.top     : NULL
## $ axis.ticks.length.x.bottom  : NULL
## $ axis.ticks.length.y         : NULL
## $ axis.ticks.length.y.left    : NULL
## $ axis.ticks.length.y.right   : NULL
## $ axis.ticks.length.theta     : NULL
## $ axis.ticks.length.r         : NULL
## $ axis.minor.ticks.length     : 'rel' num 0.75
## $ axis.minor.ticks.length.x   : NULL
## $ axis.minor.ticks.length.x.top : NULL
## $ axis.minor.ticks.length.x.bottom: NULL
## $ axis.minor.ticks.length.y   : NULL
## $ axis.minor.ticks.length.y.left : NULL
## $ axis.minor.ticks.length.y.right : NULL
## $ axis.minor.ticks.length.theta : NULL
## $ axis.minor.ticks.length.r   : NULL
## $ axis.line                   : list()
##   .- attr(*, "class")= chr [1:2] "element_blank" "element"
## $ axis.line.x                 : NULL
## $ axis.line.x.top             : NULL
## $ axis.line.x.bottom          : NULL
## $ axis.line.y                 : NULL
## $ axis.line.y.left            : NULL
## $ axis.line.y.right           : NULL
## $ axis.line.theta             : NULL
## $ axis.line.r                 : NULL
## $ legend.background           :List of 5
##   ..$ fill                    : NULL
##   ..$ colour                  : logi NA
##   ..$ linewidth               : NULL
##   ..$ linetype                : NULL
##   ..$ inherit.blank: logi FALSE
##   .- attr(*, "class")= chr [1:2] "element_rect" "element"
## $ legend.margin               : 'margin' num [1:4] 6points 6points 6points 6points
##   .- attr(*, "unit")= int 8
## $ legend.spacing              : 'simpleUnit' num 12points
##   .- attr(*, "unit")= int 8
## $ legend.spacing.x            : NULL
## $ legend.spacing.y            : NULL
## $ legend.key                  : NULL
## $ legend.key.size             : 'simpleUnit' num 1.2lines
##   .- attr(*, "unit")= int 3
## $ legend.key.height           : NULL
## $ legend.key.width            : NULL
## $ legend.key.spacing          : 'simpleUnit' num 6points
##   .- attr(*, "unit")= int 8
## $ legend.key.spacing.x        : NULL
## $ legend.key.spacing.y        : NULL
## $ legend.frame                : NULL
## $ legend.ticks                : NULL
## $ legend.ticks.length         : 'rel' num 0.2

```

```

## $ legend.axis.line           : NULL
## $ legend.text                :List of 11
##   ..$ family                : NULL
##   ..$ face                   : NULL
##   ..$ colour                 : NULL
##   ..$ size                   : 'rel' num 0.8
##   ..$ hjust                  : NULL
##   ..$ vjust                  : NULL
##   ..$ angle                  : NULL
##   ..$ lineheight             : NULL
##   ..$ margin                 : NULL
##   ..$ debug                  : NULL
##   ..$ inherit.blank: logi TRUE
##   ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ legend.text.position       : NULL
## $ legend.title               :List of 11
##   ..$ family                : NULL
##   ..$ face                   : NULL
##   ..$ colour                 : NULL
##   ..$ size                   : NULL
##   ..$ hjust                  : num 0
##   ..$ vjust                  : NULL
##   ..$ angle                  : NULL
##   ..$ lineheight             : NULL
##   ..$ margin                 : NULL
##   ..$ debug                  : NULL
##   ..$ inherit.blank: logi TRUE
##   ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ legend.title.position      : NULL
## $ legend.position            : chr "bottom"
## $ legend.position.inside     : NULL
## $ legend.direction           : chr "horizontal"
## $ legend.byrow               : NULL
## $ legend.justification       : chr "center"
## $ legend.justification.top   : NULL
## $ legend.justification.bottom : NULL
## $ legend.justification.left  : NULL
## $ legend.justification.right  : NULL
## $ legend.justification.inside : NULL
## $ legend.location            : NULL
## $ legend.box                 : chr "vertical"
## $ legend.box.just            : NULL
## $ legend.box.margin          : 'margin' num [1:4] 0cm 0cm 0cm 0cm
##   ..- attr(*, "unit")= int 1
## $ legend.box.background      : list()
##   ..- attr(*, "class")= chr [1:2] "element_blank" "element"
## $ legend.box.spacing         : 'simpleUnit' num 12points
##   ..- attr(*, "unit")= int 8
## [list output truncated]
## - attr(*, "class")= chr [1:2] "theme" "gg"
## - attr(*, "complete")= logi TRUE
## - attr(*, "validate")= logi TRUE

```

Assessment: Very difficult to find two states that are parallel but opted in to Medicaid at different times

more than a year apart. I found that Hawaii (adopted Jan 1, 2014) and Pennsylvania (adopted Jan 1, 2015) are *somewhat* similar but definitely still have differences. Concerningly, in 2013, Hawaii's rates spike while Pennsylvania's remain the same.

- Estimates a difference-in-differences estimate of the effect of the Medicaid expansion on the uninsured share of the population. You may follow the lab example where we estimate the differences in one pre-treatment and one post-treatment period, or take an average of the pre-treatment and post-treatment outcomes

```
# Difference-in-Differences estimation

# create a dataset for kansas and colorado
HI_PA <-
  medicaid_expansion %>%
  filter(State %in% c("Pennsylvania","Hawaii")) %>%
  filter(year >= 2014 & year <= 2015)

# pre-treatment difference
# -----
pre_diff <-
  HI_PA %>%
  # filter out only the year we want
  filter(year == 2014) %>%
  # subset to select only vars we want
  select(State,
         uninsured_rate) %>%
  # make the data wide
  pivot_wider(names_from = State,
              values_from = uninsured_rate) %>%
  # subtract to make calculation
  summarise(Pennsylvania - Hawaii)

# post-treatment difference
# -----
post_diff <-
  HI_PA %>%
  # filter out only the quarter we want
  filter(year == 2015) %>%
  # subset to select only vars we want
  select(State,
         uninsured_rate) %>%
  # make the data wide
  pivot_wider(names_from = State,
              values_from = uninsured_rate) %>%
  # subtract to make calculation
  summarise(Pennsylvania - Hawaii)

# diff-in-diffs
# -----
diff_in_diffs <- post_diff - pre_diff
diff_in_diffs

##   Pennsylvania - Hawaii
## 1                -0.01232
```

Discussion Questions

- Card/Krueger’s original piece utilized the fact that towns on either side of the Delaware river are likely to be quite similar to one another in terms of demographics, economics, etc. Why is that intuition harder to replicate with this data?
- **Answer:**
 - On one hand, due to the geographic similarities of the two cities the shared economies are likely to be much stronger. People can and do physically move between the two and intertwine across multiple dimensions. Should that study have been done at the statewide level, differences across the states likely would have resulted in divergent trends and an inappropriate diff-in-diff to assess a counterfactual. Further, how states responded to A) the passing of the ACA, and B) the judicial decision differ across the nation and are not inherently geographically bound.
- What are the strengths and weaknesses of using the parallel trends assumption in difference-in-differences estimates?
- **Answer:**
 - Identifying and assuming parallel trends allows for estimating the average treatment effect on the treated given that we can “reasonably” say that the control groups share similar properties that allows us to isolate the effect of treatment. It can be useful when comparing treatment between two distinct yet similar groups in natural experiments. However, the assumption may not hold robustly. There may be other confounders at the time of treatment that violate the assumption that the control group is similar to/the same as the treatment group through the full time of analysis. And of course a parallel trend may not exist or not be perfectly parallel.

Synthetic Control

Estimate Synthetic Control

Although several states did not expand Medicaid on January 1, 2014, many did later on. In some cases, a Democratic governor was elected and pushed for a state budget that included the Medicaid expansion, whereas in others voters approved expansion via a ballot initiative. The 2018 election was a watershed moment where several Republican-leaning states elected Democratic governors and approved Medicaid expansion. In cases with a ballot initiative, the state legislature and governor still must implement the results via legislation. For instance, Idaho voters approved a Medicaid expansion in the 2018 election, but it was not implemented in the state budget until late 2019, with enrollment beginning in 2020.

Do the following:

- Choose a state that adopted the Medicaid expansion after January 1, 2014. Construct a non-augmented synthetic control and plot the results (both pre-treatment fit and post-treatment differences). Also report the average ATT and L2 imbalance.

First, pick a state that adopted Medicaid after Jan 1, 2014. From above, we can see that Virginia adopted on January 1, 2019.

```
colnames(medicaid_expansion)
```

```
## [1] "State"          "Date_Adopted"   "year"           "uninsured_rate"
## [5] "population"
```

```
# add treatment effect
```

```
medicaid_expansion <-
```

```
  medicaid_expansion %>%
```

```
  # create new treatment flag just to see
```

```
  mutate(treatment = case_when(State == "Virginia" & year >= 2019 ~ 1, # note this adds treatment in 20
                                TRUE ~ 0))
```

```

# view head
head(medicaid_expansion)

## # A tibble: 6 x 6
##   State      Date_Adopted  year uninsured_rate population treatment
##   <chr>      <date>        <dbl>         <dbl>         <dbl>         <dbl>
## 1 Alabama    NA              2008           0.140       4849377         0
## 2 Alaska    2015-09-01      2008           0.208       737732         0
## 3 Arizona    2014-01-01      2008           0.187       6731484         0
## 4 Arkansas   2014-01-01      2008           0.179       2994079         0
## 5 California 2014-01-01      2008           0.178      38802500         0
## 6 Colorado   2014-01-01      2008           0.170       5355856         0

# non-augmented synthetic control

syn <-                                # save object
  augsynth(uninsured_rate ~ treatment, # treatment - use instead of treated bc latter codes 2012.25 as
    State,                             # unit
    year,                              # time
    medicaid_expansion,               # data
    progfunc = "None",                 # plain syn control
    scm = T)                           # synthetic control

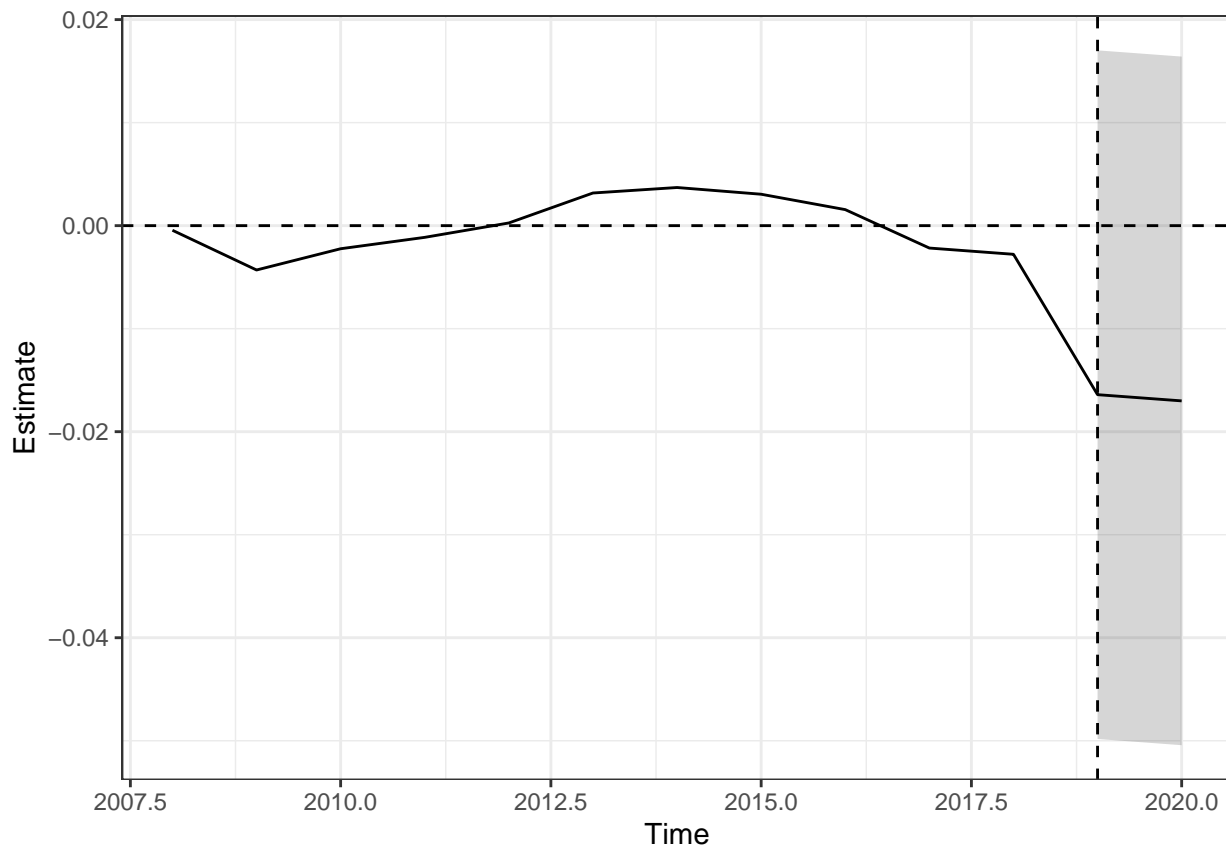
## One outcome and one treatment time found. Running single_augsynth.

# summary to show average ATT and L2 imbalance
summary(syn)

##
## Call:
## single_augsynth(form = form, unit = !!enquo(unit), time = !!enquo(time),
##   t_int = t_int, data = data, progfunc = "None", scm = ..2)
##
## Average ATT Estimate (p Value for Joint Null): -0.0167 ( 0.095 )
## L2 Imbalance: 0.009
## Percent improvement from uniform weights: 88.9%
##
## Avg Estimated Bias: NA
##
## Inference type: Conformal inference
##
## Time Estimate 95% CI Lower Bound 95% CI Upper Bound p Value
## 2019 -0.016 -0.05 0.017 0.079
## 2020 -0.017 -0.05 0.016 0.081

# Plot differences
plot(syn)

```



```
# virginia_synvirginia %>% filter((State == 'Virginia' ) & (year >= 2019))
```

```
syn_sum <- summary(syn)
```

```
# create synthetic Virginia
```

```
# -----
```

```
virginia_synvirginia <-
```

```
  # data
```

```
  medicaid_expansion %>%
```

```
  # filter just Kansas
```

```
  filter(State == "Virginia") %>%
```

```
  # bind columns
```

```
  bind_cols(difference = syn_sum$att$Estimate) %>% # add in estimate
```

```
  # calculate synthetic Virginia
```

```
  mutate(synthetic_virginia = uninsured_rate + difference) # adds the estimate to the observed Kansas t
```

```
# plot
```

```
# -----
```

```
virginia_synvirginia %>%
```

```
  ggplot() +
```

```
  # -----
```

```
  geom_line(aes(x = year,
                 y = uninsured_rate,
                 color = 'Virginia')) +
```

```
  # synthetic Virginia
```

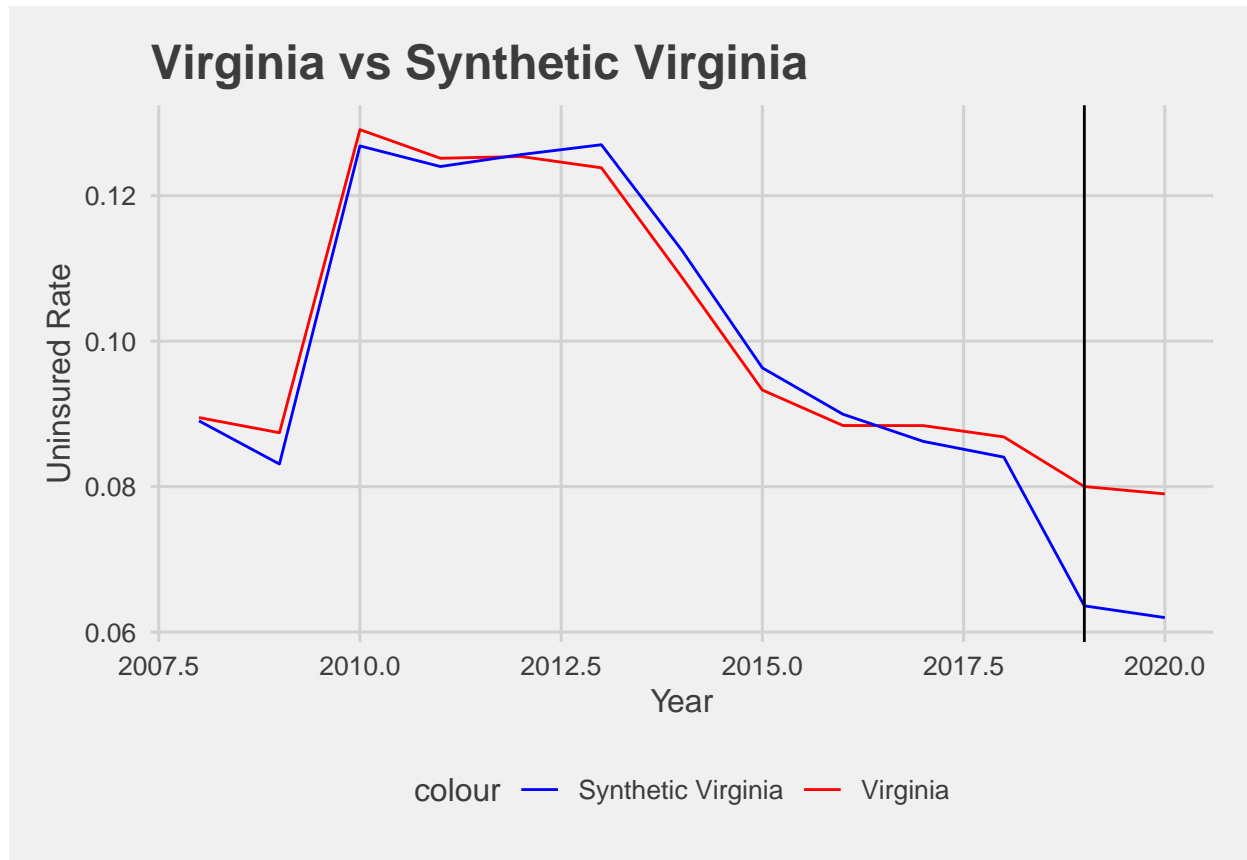
```
  # -----
```



```

geom_line(aes(x = year,
              y = synthetic_virginia,
              color = 'Synthetic Virginia')) +
scale_color_manual(values = c('Virginia' = 'red', 'Synthetic Virginia' = 'blue')) +
geom_vline(aes(xintercept = 2019)) +
theme_fivethirtyeight() +
theme(axis.title = element_text()) +
ggtitle('Virginia vs Synthetic Virginia') +
xlab('Year') +
ylab('Uninsured Rate')

```



- Re-run the same analysis but this time use an augmentation (default choices are Ridge, Matrix Completion, and GSynth). Create the same plot and report the average ATT and L2 imbalance.

```

# augmented synthetic control
ridge_syn <-
  augsynth(uninsured_rate ~ treatment,
           State,
           year,
           medicaid_expansion,
           progfunc = "ridge", # specify
           scm = T)

## One outcome and one treatment time found. Running single_augsynth.
summary(ridge_syn)

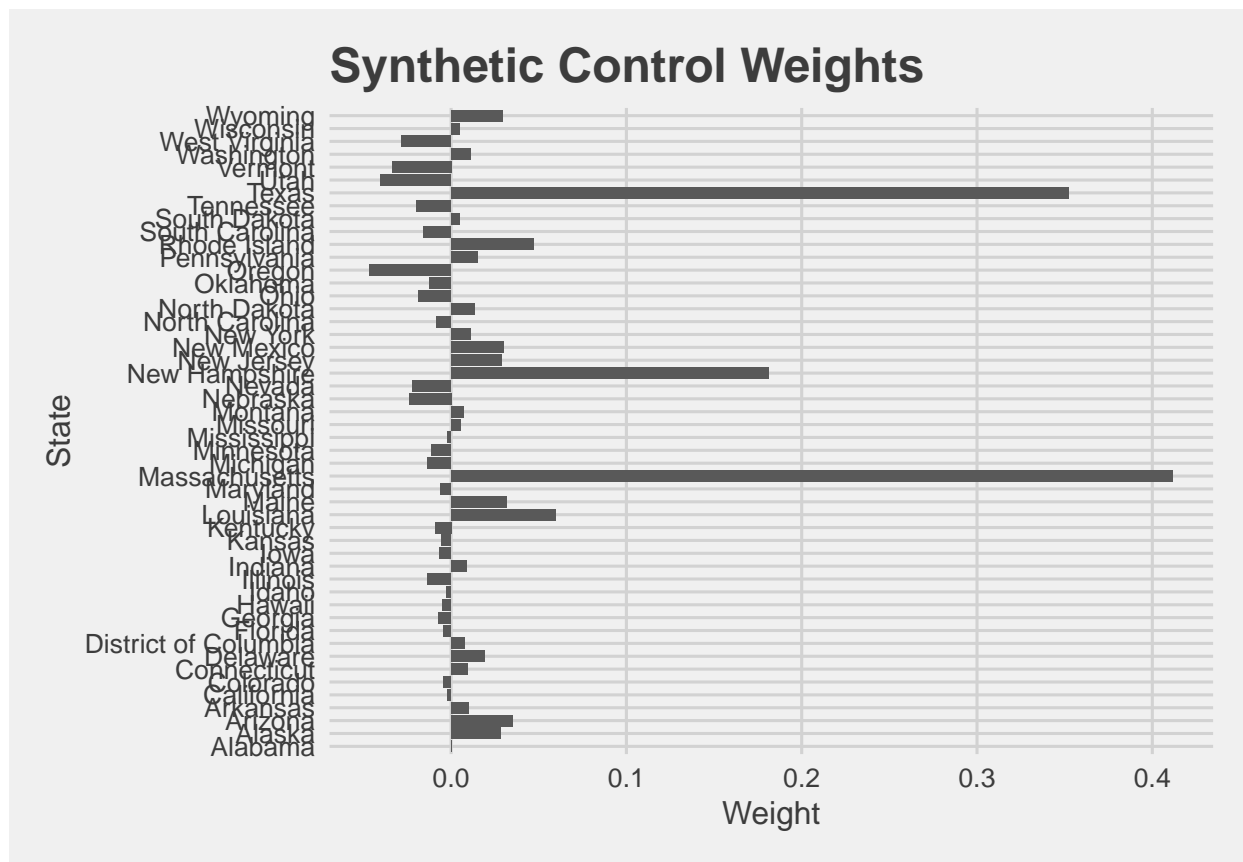
##

```

```
## Call:
## single_augsynth(form = form, unit = !!enquo(unit), time = !!enquo(time),
##   t_int = t_int, data = data, progfunc = "ridge", scm = ..2)
##
## Average ATT Estimate (p Value for Joint Null): -0.0143 ( 0.061 )
## L2 Imbalance: 0.001
## Percent improvement from uniform weights: 99.2%
##
## Avg Estimated Bias: -0.002
##
## Inference type: Conformal inference
##
## Time Estimate 95% CI Lower Bound 95% CI Upper Bound p Value
## 2019 -0.014 -0.043 0.015 0.070
## 2020 -0.015 -0.043 0.014 0.079
```

- Plot barplots to visualize the weights of the donors.

```
# barplots of weights
# -----
data.frame(ridge_syn$weights) %>%
  tibble::rownames_to_column('State') %>%
  ggplot() +
  geom_bar(aes(x = State, y = ridge_syn.weights),
    stat = 'identity') +
  coord_flip() + # coord flip
  theme_fivethirtyeight() +
  theme(axis.title = element_text()) +
  ggtitle('Synthetic Control Weights') +
  xlab('State') +
  ylab('Weight')
```



HINT: Is there any preprocessing you need to do before you allow the program to automatically find weights for donor states?

- We may want to remove states that adopted a form of universal healthcare prior to or separate from the medicaid expansion. Massachusetts is the obvious example here by adopting their own healthcare system in 2006.

Discussion Questions

- What are the advantages and disadvantages of synthetic control compared to difference-in-differences estimators?
- **Answer:**
 - Synthetic control allows for much closer matching as it takes data from all other controls in the dataset, rather than trying to make an educated guess. However, diff-in-diff may allow for drawing conclusions beyond the specific unit analysis. In other words, doing a diff-in-diff between two states may give a broader insight into how medicaid adoption affected uninsured rates elsewhere, whereas synthetic control will tell us more about the effects of medicaid adoption in the state we synthesize but not elsewhere.
- One of the benefits of synthetic control is that the weights are bounded between $[0,1]$ and the weights must sum to 1. Augmentation might relax this assumption by allowing for negative weights. Does this create an interpretation problem, and how should we balance this consideration against the improvements augmentation offers in terms of imbalance in the pre-treatment period?
- **Answer:**
 - Allowing negative weights can complicate the interpretation that each donor contributes “X%” to

the synthetic set. It suggests that some control units might effectively “counteract” others, which can be harder to justify or explain in a causal inference context.

Staggered Adoption Synthetic Control

Estimate Multisynth

Do the following:

- Estimate a multisynth model that treats each state individually. Choose a fraction of states that you can fit on a plot and examine their treatment effects.

```
# multisynth model states
# Removing Massachusetts & add new treatment variable
medicaid_expansion_clean <- medicaid_expansion %>%
  filter(!State == "Massachusetts") %>%

  # create "treatment" - year collective bargaining was adopted
  mutate(Date_Adopted = ifelse(is.na(Date_Adopted),
                                Inf, Date_Adopted),

          adopted = 1 * (year >= as.numeric(format(as.Date(Date_Adopted, format = "%Y-%m-%d"), "%Y"))))

medicaid_expansion_clean
```

```
## # A tibble: 650 x 7
##   State      Date_Adopted  year uninsured_rate population treatment adopted
##   <chr>          <dbl> <dbl>         <dbl>      <dbl>      <dbl> <dbl>
## 1 Alabama          Inf  2008          0.140    4849377         0         0
## 2 Alaska      16679  2008          0.208     737732         0         0
## 3 Arizona      16071  2008          0.187    6731484         0         0
## 4 Arkansas      16071  2008          0.179    2994079         0         0
## 5 California      16071  2008          0.178   38802500         0         0
## 6 Colorado      16071  2008          0.170    5355856         0         0
## 7 Connecticut      16071  2008         0.0891    3596677         0         0
## 8 Delaware      16071  2008          0.108     935614         0         0
## 9 District of C~  16071  2008         0.0805         NA         0         0
## 10 Florida          Inf  2008          0.209   19893297         0         0
## # i 640 more rows
```

```
# Multisynth model states
# -----
# Setting nu to 0 to treat each state individually
no_pool_syn <- multisynth(uninsured_rate ~ adopted,
                          State,          # unit
                          year,           # time
                          nu = 0,         # no pooling
                          medicaid_expansion_clean, # data
                          n_leads = 2)
```

```
no_pool_synsum <- summary(no_pool_syn)
```

```
no_pool_synsum$att
```

```
##   Time      Level      Estimate  Std.Error  lower_bound
## 1   -12      Average  1.025101e-02 0.037957396 -0.0648191846
```

## 2	-11	Average	1.160478e-02	0.026734532	-0.0425560692
## 3	-10	Average	3.432105e-03	0.011716643	-0.0190546194
## 4	-9	Average	-5.185198e-04	0.007580385	-0.0158867256
## 5	-8	Average	-1.848474e-03	0.008369585	-0.0156006257
## 6	-7	Average	-1.163056e-03	0.007227458	-0.0124020708
## 7	-6	Average	-1.136448e-04	0.007167483	-0.0126611594
## 8	-5	Average	-7.142314e-04	0.006064815	-0.0120605201
## 9	-4	Average	2.720578e-04	0.003258371	-0.0062644474
## 10	-3	Average	-3.903377e-04	0.003605776	-0.0077116302
## 11	-2	Average	-8.841279e-04	0.003892036	-0.0087190759
## 12	-1	Average	-9.166283e-05	0.003466892	-0.0074552654
## 13	0	Average	-1.173916e-02	0.004748010	-0.0212930770
## 14	1	Average	-1.817259e-02	0.006550081	-0.0309284125
## 15	NA	Average	-1.440935e-02	0.005159329	-0.0246273620
## 16	-12	Alaska	NA	NaN	NA
## 17	-11	Alaska	NA	NaN	NA
## 18	-10	Alaska	NA	NaN	NA
## 19	-9	Alaska	NA	NaN	NA
## 20	-8	Alaska	NA	NaN	NA
## 21	-7	Alaska	1.027723e-03	0.023082311	-0.0402930653
## 22	-6	Alaska	7.243849e-04	0.017800827	-0.0313787810
## 23	-5	Alaska	-1.195273e-02	0.024951418	-0.0461080769
## 24	-4	Alaska	4.610996e-03	0.007427316	-0.0104625804
## 25	-3	Alaska	6.778125e-03	0.008352612	-0.0130776303
## 26	-2	Alaska	-2.317406e-03	0.010999430	-0.0183712245
## 27	-1	Alaska	1.128911e-03	0.004954784	-0.0091367641
## 28	0	Alaska	-1.065096e-02	0.012512980	-0.0278209506
## 29	1	Alaska	-4.996361e-03	0.005882979	-0.0159787189
## 30	NA	Alaska	-7.823663e-03	0.008904028	-0.0217715106
## 31	-12	Arizona	NA	NaN	NA
## 32	-11	Arizona	NA	NaN	NA
## 33	-10	Arizona	NA	NaN	NA
## 34	-9	Arizona	NA	NaN	NA
## 35	-8	Arizona	NA	NaN	NA
## 36	-7	Arizona	NA	NaN	NA
## 37	-6	Arizona	2.625765e-03	0.024122081	-0.0414435148
## 38	-5	Arizona	-2.946809e-03	0.013306461	-0.0219916454
## 39	-4	Arizona	-3.901417e-03	0.015042918	-0.0302540513
## 40	-3	Arizona	-2.336919e-03	0.011427891	-0.0227669765
## 41	-2	Arizona	3.943136e-03	0.007575458	-0.0113743321
## 42	-1	Arizona	2.616243e-03	0.007384575	-0.0118825809
## 43	0	Arizona	-2.067609e-02	0.016629048	-0.0453036983
## 44	1	Arizona	-3.170106e-02	0.025456749	-0.0658238817
## 45	NA	Arizona	-2.618858e-02	0.020990629	-0.0555618953
## 46	-12	Arkansas	NA	NaN	NA
## 47	-11	Arkansas	NA	NaN	NA
## 48	-10	Arkansas	NA	NaN	NA
## 49	-9	Arkansas	NA	NaN	NA
## 50	-8	Arkansas	NA	NaN	NA
## 51	-7	Arkansas	NA	NaN	NA
## 52	-6	Arkansas	2.488362e-03	0.026397683	-0.0530609294
## 53	-5	Arkansas	-2.706397e-03	0.016905364	-0.0321668602
## 54	-4	Arkansas	9.865953e-04	0.010700928	-0.0184076346
## 55	-3	Arkansas	1.429651e-03	0.011251309	-0.0212821890

## 56	-2	Arkansas	-1.852202e-03	0.010558665	-0.0222650232
## 57	-1	Arkansas	-3.460092e-04	0.010432431	-0.0222419245
## 58	0	Arkansas	-2.147545e-02	0.028360010	-0.0571013614
## 59	1	Arkansas	-2.998838e-02	0.034564757	-0.0692828948
## 60	NA	Arkansas	-2.573191e-02	0.031410668	-0.0627754024
## 61	-12	California	NA	NaN	NA
## 62	-11	California	NA	NaN	NA
## 63	-10	California	NA	NaN	NA
## 64	-9	California	NA	NaN	NA
## 65	-8	California	NA	NaN	NA
## 66	-7	California	NA	NaN	NA
## 67	-6	California	-6.332163e-11	0.014405867	-0.0224134074
## 68	-5	California	7.062909e-11	0.014506824	-0.0243288263
## 69	-4	California	1.736669e-11	0.005972630	-0.0119335181
## 70	-3	California	-1.434475e-10	0.006797815	-0.0138460683
## 71	-2	California	9.116377e-11	0.007865408	-0.0156675535
## 72	-1	California	2.760953e-11	0.009310717	-0.0171304255
## 73	0	California	-2.939468e-02	0.031418116	-0.0651908786
## 74	1	California	-5.028501e-02	0.053606829	-0.1086660456
## 75	NA	California	-3.983985e-02	0.042474614	-0.0862674978
## 76	-12	Colorado	NA	NaN	NA
## 77	-11	Colorado	NA	NaN	NA
## 78	-10	Colorado	NA	NaN	NA
## 79	-9	Colorado	NA	NaN	NA
## 80	-8	Colorado	NA	NaN	NA
## 81	-7	Colorado	NA	NaN	NA
## 82	-6	Colorado	3.411354e-03	0.024422341	-0.0435058608
## 83	-5	Colorado	-3.552840e-03	0.011742331	-0.0254621896
## 84	-4	Colorado	1.924324e-03	0.004496013	-0.0068610794
## 85	-3	Colorado	1.918107e-03	0.007762177	-0.0132755616
## 86	-2	Colorado	-2.129560e-03	0.010151124	-0.0186619499
## 87	-1	Colorado	-1.571385e-03	0.013825769	-0.0254236111
## 88	0	Colorado	-1.172853e-02	0.025550407	-0.0482160634
## 89	1	Colorado	-2.533451e-02	0.031699363	-0.0616749252
## 90	NA	Colorado	-1.853152e-02	0.028404387	-0.0546806722
## 91	-12	Connecticut	NA	NaN	NA
## 92	-11	Connecticut	NA	NaN	NA
## 93	-10	Connecticut	NA	NaN	NA
## 94	-9	Connecticut	NA	NaN	NA
## 95	-8	Connecticut	NA	NaN	NA
## 96	-7	Connecticut	NA	NaN	NA
## 97	-6	Connecticut	-1.263251e-04	0.012137355	-0.0236993332
## 98	-5	Connecticut	6.919339e-05	0.012035450	-0.0241210088
## 99	-4	Connecticut	-1.432031e-03	0.009699305	-0.0212071322
## 100	-3	Connecticut	-5.046973e-03	0.010594551	-0.0216127063
## 101	-2	Connecticut	3.190982e-03	0.007938451	-0.0140109987
## 102	-1	Connecticut	3.345153e-03	0.007493652	-0.0109025191
## 103	0	Connecticut	-2.165019e-03	0.008322045	-0.0176710793
## 104	1	Connecticut	-1.438232e-03	0.013844942	-0.0272391587
## 105	NA	Connecticut	-1.801626e-03	0.010338623	-0.0198092130
## 106	-12	Delaware	NA	NaN	NA
## 107	-11	Delaware	NA	NaN	NA
## 108	-10	Delaware	NA	NaN	NA
## 109	-9	Delaware	NA	NaN	NA

## 110	-8	Delaware	NA	NaN	NA
## 111	-7	Delaware	NA	NaN	NA
## 112	-6	Delaware	4.698354e-04	0.029996603	-0.0587071395
## 113	-5	Delaware	-1.178458e-03	0.026209084	-0.0499331342
## 114	-4	Delaware	6.165954e-04	0.015084691	-0.0282647479
## 115	-3	Delaware	-3.182807e-03	0.019093283	-0.0398788617
## 116	-2	Delaware	-1.628232e-03	0.020983492	-0.0406741094
## 117	-1	Delaware	4.903066e-03	0.008243953	-0.0128257460
## 118	0	Delaware	4.390070e-03	0.007071538	-0.0099835986
## 119	1	Delaware	-1.210016e-02	0.011174822	-0.0271201112
## 120	NA	Delaware	-3.855047e-03	0.004314982	-0.0115922257
## 121	-12	District of Columbia	NA	NaN	NA
## 122	-11	District of Columbia	NA	NaN	NA
## 123	-10	District of Columbia	NA	NaN	NA
## 124	-9	District of Columbia	NA	NaN	NA
## 125	-8	District of Columbia	NA	NaN	NA
## 126	-7	District of Columbia	NA	NaN	NA
## 127	-6	District of Columbia	4.078235e-03	0.016014620	-0.0269985117
## 128	-5	District of Columbia	-4.334165e-03	0.007672711	-0.0191184809
## 129	-4	District of Columbia	1.799168e-03	0.002103907	-0.0023576950
## 130	-3	District of Columbia	5.574322e-03	0.002789504	-0.0004376945
## 131	-2	District of Columbia	-6.438455e-03	0.016267403	-0.0270339205
## 132	-1	District of Columbia	-6.791057e-04	0.005990714	-0.0104579049
## 133	0	District of Columbia	1.517938e-02	0.010410561	-0.0088493345
## 134	1	District of Columbia	2.024933e-02	0.013720273	-0.0102343129
## 135	NA	District of Columbia	1.771435e-02	0.011828290	-0.0095418663
## 136	-12	Hawaii	NA	NaN	NA
## 137	-11	Hawaii	NA	NaN	NA
## 138	-10	Hawaii	NA	NaN	NA
## 139	-9	Hawaii	NA	NaN	NA
## 140	-8	Hawaii	NA	NaN	NA
## 141	-7	Hawaii	NA	NaN	NA
## 142	-6	Hawaii	8.876372e-12	0.013876650	-0.0243373512
## 143	-5	Hawaii	-9.900775e-12	0.012693241	-0.0222781599
## 144	-4	Hawaii	-2.434455e-12	0.007543233	-0.0137734088
## 145	-3	Hawaii	2.010841e-11	0.007491924	-0.0148432835
## 146	-2	Hawaii	-1.277931e-11	0.009222890	-0.0178335297
## 147	-1	Hawaii	-3.870293e-12	0.004824954	-0.0098095764
## 148	0	Hawaii	2.364820e-03	0.005044671	-0.0086941364
## 149	1	Hawaii	2.551249e-03	0.010249761	-0.0180143050
## 150	NA	Hawaii	2.458034e-03	0.006528840	-0.0107350098
## 151	-12	Idaho	2.304296e-03	0.025621375	-0.0452442623
## 152	-11	Idaho	-2.183838e-03	0.018519826	-0.0358393418
## 153	-10	Idaho	7.250498e-03	0.013489748	-0.0183518116
## 154	-9	Idaho	-8.595266e-03	0.009659612	-0.0230998701
## 155	-8	Idaho	1.218516e-03	0.004714972	-0.0075943815
## 156	-7	Idaho	1.150191e-03	0.006882804	-0.0110816597
## 157	-6	Idaho	-8.992386e-04	0.002423652	-0.0055600867
## 158	-5	Idaho	9.360247e-04	0.010353644	-0.0174682181
## 159	-4	Idaho	-9.584729e-05	0.011271906	-0.0187820549
## 160	-3	Idaho	1.291580e-03	0.011597018	-0.0185581780
## 161	-2	Idaho	1.776667e-03	0.008875529	-0.0148169600
## 162	-1	Idaho	-4.153583e-03	0.018226299	-0.0312936817
## 163	0	Idaho	-1.308111e-03	0.015204043	-0.0246980155

## 164	1	Idaho	NA	NaN	NA
## 165	NA	Idaho	-1.308111e-03	0.015204043	-0.0246980155
## 166	-12	Illinois	NA	NaN	NA
## 167	-11	Illinois	NA	NaN	NA
## 168	-10	Illinois	NA	NaN	NA
## 169	-9	Illinois	NA	NaN	NA
## 170	-8	Illinois	NA	NaN	NA
## 171	-7	Illinois	NA	NaN	NA
## 172	-6	Illinois	-5.280943e-04	0.009117789	-0.0177887747
## 173	-5	Illinois	6.320909e-04	0.008224401	-0.0155319688
## 174	-4	Illinois	1.955541e-04	0.002596658	-0.0050591150
## 175	-3	Illinois	-9.334521e-04	0.009839738	-0.0187340634
## 176	-2	Illinois	8.370655e-04	0.004407576	-0.0083806640
## 177	-1	Illinois	-2.031641e-04	0.003264849	-0.0069192964
## 178	0	Illinois	-6.831126e-03	0.010437168	-0.0223328338
## 179	1	Illinois	-1.585249e-02	0.021484688	-0.0434655010
## 180	NA	Illinois	-1.134181e-02	0.015918376	-0.0327273008
## 181	-12	Indiana	NA	NaN	NA
## 182	-11	Indiana	NA	NaN	NA
## 183	-10	Indiana	NA	NaN	NA
## 184	-9	Indiana	NA	NaN	NA
## 185	-8	Indiana	NA	NaN	NA
## 186	-7	Indiana	-4.746103e-06	0.005371382	-0.0100454270
## 187	-6	Indiana	4.391358e-06	0.004690898	-0.0089796921
## 188	-5	Indiana	-1.361134e-05	0.003216849	-0.0068858666
## 189	-4	Indiana	-2.374178e-05	0.003626045	-0.0068430475
## 190	-3	Indiana	2.861172e-05	0.002780322	-0.0059971685
## 191	-2	Indiana	2.471800e-05	0.001394227	-0.0025072834
## 192	-1	Indiana	-1.562186e-05	0.002722789	-0.0052480355
## 193	0	Indiana	-4.896250e-03	0.005356414	-0.0136661302
## 194	1	Indiana	-1.351991e-02	0.014351836	-0.0293953708
## 195	NA	Indiana	-9.208082e-03	0.009637716	-0.0215334003
## 196	-12	Iowa	NA	NaN	NA
## 197	-11	Iowa	NA	NaN	NA
## 198	-10	Iowa	NA	NaN	NA
## 199	-9	Iowa	NA	NaN	NA
## 200	-8	Iowa	NA	NaN	NA
## 201	-7	Iowa	NA	NaN	NA
## 202	-6	Iowa	-8.297807e-12	0.012056002	-0.0222593421
## 203	-5	Iowa	9.255346e-12	0.011047411	-0.0199070372
## 204	-4	Iowa	2.275749e-12	0.006536528	-0.0127020241
## 205	-3	Iowa	-1.879763e-11	0.006654507	-0.0131486914
## 206	-2	Iowa	1.194625e-11	0.008355326	-0.0162566145
## 207	-1	Iowa	3.617995e-12	0.003588018	-0.0071473272
## 208	0	Iowa	-8.208280e-03	0.008224082	-0.0204803918
## 209	1	Iowa	-6.110681e-03	0.005512971	-0.0151989572
## 210	NA	Iowa	-7.159481e-03	0.004582164	-0.0142071568
## 211	-12	Kentucky	NA	NaN	NA
## 212	-11	Kentucky	NA	NaN	NA
## 213	-10	Kentucky	NA	NaN	NA
## 214	-9	Kentucky	NA	NaN	NA
## 215	-8	Kentucky	NA	NaN	NA
## 216	-7	Kentucky	NA	NaN	NA
## 217	-6	Kentucky	-7.023438e-04	0.016341901	-0.0312420578

## 218	-5	Kentucky	6.916848e-04	0.015588950	-0.0304324043
## 219	-4	Kentucky	1.579115e-03	0.008902969	-0.0154682476
## 220	-3	Kentucky	5.145222e-06	0.009319635	-0.0206768182
## 221	-2	Kentucky	-2.042224e-03	0.012248156	-0.0243631717
## 222	-1	Kentucky	4.686232e-04	0.005662458	-0.0099847390
## 223	0	Kentucky	-3.270426e-02	0.033109319	-0.0688873318
## 224	1	Kentucky	-4.939311e-02	0.042455409	-0.0955079142
## 225	NA	Kentucky	-4.104869e-02	0.037552129	-0.0800108723
## 226	-12	Louisiana	NA	NaN	NA
## 227	-11	Louisiana	NA	NaN	NA
## 228	-10	Louisiana	NA	NaN	NA
## 229	-9	Louisiana	NA	NaN	NA
## 230	-8	Louisiana	1.401961e-03	0.016968885	-0.0331798173
## 231	-7	Louisiana	-1.546136e-03	0.010906510	-0.0227186906
## 232	-6	Louisiana	3.544046e-04	0.004863799	-0.0093989099
## 233	-5	Louisiana	1.164666e-03	0.004041927	-0.0067123430
## 234	-4	Louisiana	-9.895302e-04	0.005984501	-0.0112737923
## 235	-3	Louisiana	-7.468527e-04	0.003193174	-0.0072158816
## 236	-2	Louisiana	8.391783e-04	0.006171535	-0.0121464554
## 237	-1	Louisiana	-4.776909e-04	0.015762755	-0.0298702249
## 238	0	Louisiana	-1.205470e-02	0.023758324	-0.0486219719
## 239	1	Louisiana	-3.358185e-02	0.044601062	-0.0904330820
## 240	NA	Louisiana	-2.281828e-02	0.033908527	-0.0691393266
## 241	-12	Maryland	NA	NaN	NA
## 242	-11	Maryland	NA	NaN	NA
## 243	-10	Maryland	NA	NaN	NA
## 244	-9	Maryland	NA	NaN	NA
## 245	-8	Maryland	NA	NaN	NA
## 246	-7	Maryland	NA	NaN	NA
## 247	-6	Maryland	-4.145015e-07	0.012601479	-0.0251743438
## 248	-5	Maryland	4.968849e-07	0.009936815	-0.0194324752
## 249	-4	Maryland	5.656665e-08	0.002984425	-0.0056677373
## 250	-3	Maryland	-9.219880e-07	0.011667553	-0.0198305372
## 251	-2	Maryland	7.423690e-07	0.006926076	-0.0125681301
## 252	-1	Maryland	4.066901e-08	0.003627199	-0.0060971590
## 253	0	Maryland	-4.179886e-04	0.006582224	-0.0118813943
## 254	1	Maryland	3.877541e-03	0.006785955	-0.0090555028
## 255	NA	Maryland	1.729776e-03	0.006062285	-0.0097009208
## 256	-12	Michigan	NA	NaN	NA
## 257	-11	Michigan	NA	NaN	NA
## 258	-10	Michigan	NA	NaN	NA
## 259	-9	Michigan	NA	NaN	NA
## 260	-8	Michigan	NA	NaN	NA
## 261	-7	Michigan	NA	NaN	NA
## 262	-6	Michigan	-4.896385e-04	0.013882882	-0.0258723169
## 263	-5	Michigan	5.485453e-04	0.014999760	-0.0292649700
## 264	-4	Michigan	4.250109e-04	0.004649377	-0.0089648506
## 265	-3	Michigan	-6.605248e-04	0.008113365	-0.0160016349
## 266	-2	Michigan	5.226093e-04	0.007264238	-0.0144985720
## 267	-1	Michigan	-3.460021e-04	0.008130494	-0.0153335829
## 268	0	Michigan	-8.025371e-03	0.011718151	-0.0258322918
## 269	1	Michigan	-1.369117e-02	0.018357711	-0.0386203790
## 270	NA	Michigan	-1.085827e-02	0.014980936	-0.0321255508
## 271	-12	Minnesota	NA	NaN	NA

## 272	-11	Minnesota	NA	NaN	NA
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## 274	-9	Minnesota	NA	NaN	NA
## 275	-8	Minnesota	NA	NaN	NA
## 276	-7	Minnesota	NA	NaN	NA
## 277	-6	Minnesota	-4.920693e-11	0.009536900	-0.0193760101
## 278	-5	Minnesota	5.488555e-11	0.010424074	-0.0201529325
## 279	-4	Minnesota	1.349559e-11	0.005812529	-0.0102571421
## 280	-3	Minnesota	-1.114724e-10	0.005828003	-0.0106502504
## 281	-2	Minnesota	7.084297e-11	0.005898739	-0.0111188630
## 282	-1	Minnesota	2.145525e-11	0.002374266	-0.0045822493
## 283	0	Minnesota	-3.542872e-03	0.004392798	-0.0110345730
## 284	1	Minnesota	-5.321175e-03	0.003841413	-0.0120440162
## 285	NA	Minnesota	-4.432024e-03	0.003321594	-0.0101020421
## 286	-12	Montana	NA	NaN	NA
## 287	-11	Montana	NA	NaN	NA
## 288	-10	Montana	NA	NaN	NA
## 289	-9	Montana	NA	NaN	NA
## 290	-8	Montana	4.453454e-04	0.029038431	-0.0601861821
## 291	-7	Montana	1.703790e-03	0.028551405	-0.0579397392
## 292	-6	Montana	-1.267496e-02	0.014696409	-0.0342975599
## 293	-5	Montana	7.908725e-03	0.009947414	-0.0157753465
## 294	-4	Montana	9.668764e-03	0.011652496	-0.0171816507
## 295	-3	Montana	1.325011e-03	0.010194842	-0.0173875686
## 296	-2	Montana	-6.278959e-03	0.016733460	-0.0360458672
## 297	-1	Montana	-2.097720e-03	0.024327259	-0.0475064507
## 298	0	Montana	-2.522622e-02	0.045286249	-0.0945912018
## 299	1	Montana	-2.396532e-02	0.042868213	-0.0910784056
## 300	NA	Montana	-2.459577e-02	0.044058709	-0.0921605116
## 301	-12	Nebraska	-3.667427e-03	0.015938296	-0.0354921988
## 302	-11	Nebraska	5.059002e-03	0.013386758	-0.0220773866
## 303	-10	Nebraska	2.174985e-03	0.010248508	-0.0161709495
## 304	-9	Nebraska	8.734972e-04	0.006247963	-0.0119717650
## 305	-8	Nebraska	-4.222909e-03	0.010975336	-0.0228426888
## 306	-7	Nebraska	-3.552310e-03	0.009902128	-0.0181326515
## 307	-6	Nebraska	-4.880591e-03	0.003813757	-0.0112468922
## 308	-5	Nebraska	-5.260393e-03	0.004126885	-0.0111957917
## 309	-4	Nebraska	1.049622e-02	0.018844819	-0.0260195092
## 310	-3	Nebraska	4.305912e-03	0.012950707	-0.0213678018
## 311	-2	Nebraska	4.564940e-03	0.009400277	-0.0139345947
## 312	-1	Nebraska	-5.890924e-03	0.005357478	-0.0133802159
## 313	0	Nebraska	-2.017920e-03	0.006286162	-0.0135549649
## 314	1	Nebraska	NA	NaN	NA
## 315	NA	Nebraska	-2.017920e-03	0.006286162	-0.0135549649
## 316	-12	Nevada	NA	NaN	NA
## 317	-11	Nevada	NA	NaN	NA
## 318	-10	Nevada	NA	NaN	NA
## 319	-9	Nevada	NA	NaN	NA
## 320	-8	Nevada	NA	NaN	NA
## 321	-7	Nevada	NA	NaN	NA
## 322	-6	Nevada	-7.968625e-04	0.023971909	-0.0403928429
## 323	-5	Nevada	8.873256e-04	0.023334149	-0.0355196315
## 324	-4	Nevada	7.379966e-04	0.009615545	-0.0178017221
## 325	-3	Nevada	-1.008670e-03	0.012152249	-0.0273241545

##	326	-2	Nevada	7.938569e-04	0.014875611	-0.0264499159
##	327	-1	Nevada	-6.136469e-04	0.015595864	-0.0284044216
##	328	0	Nevada	-3.877113e-02	0.042771102	-0.0904363417
##	329	1	Nevada	-4.845291e-02	0.056069919	-0.1101445300
##	330	NA	Nevada	-4.361202e-02	0.049379270	-0.1004788579
##	331	-12	New Hampshire	NA	NaN	NA
##	332	-11	New Hampshire	NA	NaN	NA
##	333	-10	New Hampshire	NA	NaN	NA
##	334	-9	New Hampshire	NA	NaN	NA
##	335	-8	New Hampshire	NA	NaN	NA
##	336	-7	New Hampshire	NA	NaN	NA
##	337	-6	New Hampshire	6.607513e-04	0.011684446	-0.0252807959
##	338	-5	New Hampshire	-7.057918e-04	0.007588740	-0.0167499950
##	339	-4	New Hampshire	-1.860400e-03	0.005309921	-0.0110031569
##	340	-3	New Hampshire	-5.562685e-03	0.013791658	-0.0267204098
##	341	-2	New Hampshire	3.964452e-03	0.005734285	-0.0069371988
##	342	-1	New Hampshire	3.503674e-03	0.005095872	-0.0062446672
##	343	0	New Hampshire	7.535342e-03	0.010960706	-0.0153440217
##	344	1	New Hampshire	1.736218e-03	0.007794391	-0.0119136790
##	345	NA	New Hampshire	4.635780e-03	0.008637148	-0.0134602504
##	346	-12	New Jersey	NA	NaN	NA
##	347	-11	New Jersey	NA	NaN	NA
##	348	-10	New Jersey	NA	NaN	NA
##	349	-9	New Jersey	NA	NaN	NA
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##	351	-7	New Jersey	NA	NaN	NA
##	352	-6	New Jersey	-5.790568e-04	0.017211443	-0.0287722502
##	353	-5	New Jersey	4.180303e-04	0.017300978	-0.0291163987
##	354	-4	New Jersey	-9.176344e-04	0.009260094	-0.0181476291
##	355	-3	New Jersey	-1.113657e-03	0.008987226	-0.0183712079
##	356	-2	New Jersey	4.185996e-04	0.011133042	-0.0194704844
##	357	-1	New Jersey	1.773718e-03	0.010563371	-0.0179053643
##	358	0	New Jersey	-3.417452e-03	0.012382814	-0.0233459233
##	359	1	New Jersey	-1.862684e-02	0.012790361	-0.0348138348
##	360	NA	New Jersey	-1.102215e-02	0.011459970	-0.0243348428
##	361	-12	New Mexico	NA	NaN	NA
##	362	-11	New Mexico	NA	NaN	NA
##	363	-10	New Mexico	NA	NaN	NA
##	364	-9	New Mexico	NA	NaN	NA
##	365	-8	New Mexico	NA	NaN	NA
##	366	-7	New Mexico	NA	NaN	NA
##	367	-6	New Mexico	3.267756e-03	0.024513702	-0.0417373459
##	368	-5	New Mexico	-3.495510e-03	0.011722245	-0.0234306109
##	369	-4	New Mexico	1.055547e-03	0.009200518	-0.0156298335
##	370	-3	New Mexico	2.933805e-03	0.007413261	-0.0103411305
##	371	-2	New Mexico	-3.047410e-03	0.013313685	-0.0214979628
##	372	-1	New Mexico	-7.141884e-04	0.007112820	-0.0132599062
##	373	0	New Mexico	-2.495741e-02	0.028713792	-0.0576269526
##	374	1	New Mexico	-4.559683e-02	0.051110576	-0.1014023976
##	375	NA	New Mexico	-3.527712e-02	0.039892787	-0.0798167339
##	376	-12	New York	NA	NaN	NA
##	377	-11	New York	NA	NaN	NA
##	378	-10	New York	NA	NaN	NA
##	379	-9	New York	NA	NaN	NA

## 380	-8	New York	NA	NaN	NA
## 381	-7	New York	NA	NaN	NA
## 382	-6	New York	1.598151e-05	0.013493518	-0.0294342557
## 383	-5	New York	-1.726374e-05	0.009917406	-0.0221723111
## 384	-4	New York	4.526955e-06	0.005245284	-0.0105340530
## 385	-3	New York	1.177596e-05	0.006016599	-0.0116981625
## 386	-2	New York	-8.287400e-06	0.006920581	-0.0124980117
## 387	-1	New York	-6.733288e-06	0.005735612	-0.0116738175
## 388	0	New York	4.465051e-04	0.006162014	-0.0103689703
## 389	1	New York	3.426964e-03	0.005945348	-0.0080912295
## 390	NA	New York	1.936735e-03	0.005944418	-0.0091903337
## 391	-12	North Dakota	NA	NaN	NA
## 392	-11	North Dakota	NA	NaN	NA
## 393	-10	North Dakota	NA	NaN	NA
## 394	-9	North Dakota	NA	NaN	NA
## 395	-8	North Dakota	NA	NaN	NA
## 396	-7	North Dakota	NA	NaN	NA
## 397	-6	North Dakota	7.365070e-04	0.007246470	-0.0137676375
## 398	-5	North Dakota	-7.978633e-04	0.007549573	-0.0162675441
## 399	-4	North Dakota	-1.998166e-03	0.010348618	-0.0208266728
## 400	-3	North Dakota	-4.367890e-03	0.007376853	-0.0153331026
## 401	-2	North Dakota	3.184796e-03	0.007240797	-0.0133216337
## 402	-1	North Dakota	3.242616e-03	0.005511497	-0.0070590245
## 403	0	North Dakota	-2.870206e-03	0.006812089	-0.0152837077
## 404	1	North Dakota	1.265383e-02	0.022229363	-0.0303284726
## 405	NA	North Dakota	4.891811e-03	0.013736269	-0.0210987196
## 406	-12	Ohio	NA	NaN	NA
## 407	-11	Ohio	NA	NaN	NA
## 408	-10	Ohio	NA	NaN	NA
## 409	-9	Ohio	NA	NaN	NA
## 410	-8	Ohio	NA	NaN	NA
## 411	-7	Ohio	NA	NaN	NA
## 412	-6	Ohio	-3.226225e-12	0.016279129	-0.0252939590
## 413	-5	Ohio	3.598524e-12	0.016866537	-0.0297760388
## 414	-4	Ohio	8.848200e-13	0.006943526	-0.0141409548
## 415	-3	Ohio	-7.308612e-12	0.007178005	-0.0146949327
## 416	-2	Ohio	4.644771e-12	0.008957145	-0.0182001926
## 417	-1	Ohio	1.406694e-12	0.010554963	-0.0191638308
## 418	0	Ohio	-8.337684e-03	0.012993919	-0.0282133400
## 419	1	Ohio	-9.242049e-03	0.015196338	-0.0306901723
## 420	NA	Ohio	-8.789867e-03	0.014071500	-0.0293180063
## 421	-12	Oregon	NA	NaN	NA
## 422	-11	Oregon	NA	NaN	NA
## 423	-10	Oregon	NA	NaN	NA
## 424	-9	Oregon	NA	NaN	NA
## 425	-8	Oregon	NA	NaN	NA
## 426	-7	Oregon	NA	NaN	NA
## 427	-6	Oregon	-3.095536e-03	0.026571843	-0.0497216028
## 428	-5	Oregon	3.177208e-03	0.029730464	-0.0556878514
## 429	-4	Oregon	6.465552e-03	0.011359646	-0.0178610101
## 430	-3	Oregon	-4.548286e-04	0.014603499	-0.0296717140
## 431	-2	Oregon	-5.597237e-03	0.019667765	-0.0396354337
## 432	-1	Oregon	-4.951587e-04	0.016309612	-0.0296994206
## 433	0	Oregon	-2.984363e-02	0.039722088	-0.0776775203

## 434	1	Oregon	-3.869174e-02	0.050587751	-0.0963807442
## 435	NA	Oregon	-3.426769e-02	0.045136355	-0.0870291323
## 436	-12	Pennsylvania	NA	NaN	NA
## 437	-11	Pennsylvania	NA	NaN	NA
## 438	-10	Pennsylvania	NA	NaN	NA
## 439	-9	Pennsylvania	NA	NaN	NA
## 440	-8	Pennsylvania	NA	NaN	NA
## 441	-7	Pennsylvania	-9.303645e-04	0.011159970	-0.0189361032
## 442	-6	Pennsylvania	9.710774e-04	0.011064714	-0.0179873658
## 443	-5	Pennsylvania	3.626142e-04	0.007510738	-0.0155957316
## 444	-4	Pennsylvania	-1.502775e-03	0.004977692	-0.0112624592
## 445	-3	Pennsylvania	1.319086e-04	0.006549635	-0.0151928358
## 446	-2	Pennsylvania	-1.228539e-03	0.005561617	-0.0123705837
## 447	-1	Pennsylvania	2.196078e-03	0.012433473	-0.0235655713
## 448	0	Pennsylvania	-6.346238e-03	0.009687319	-0.0237993944
## 449	1	Pennsylvania	-2.522150e-03	0.012912656	-0.0294041852
## 450	NA	Pennsylvania	-4.434194e-03	0.011208744	-0.0267264954
## 451	-12	Rhode Island	NA	NaN	NA
## 452	-11	Rhode Island	NA	NaN	NA
## 453	-10	Rhode Island	NA	NaN	NA
## 454	-9	Rhode Island	NA	NaN	NA
## 455	-8	Rhode Island	NA	NaN	NA
## 456	-7	Rhode Island	NA	NaN	NA
## 457	-6	Rhode Island	1.873017e-03	0.019369247	-0.0322376132
## 458	-5	Rhode Island	-2.649163e-03	0.025996277	-0.0448520205
## 459	-4	Rhode Island	1.767065e-05	0.010277599	-0.0172814253
## 460	-3	Rhode Island	-4.244557e-03	0.006689052	-0.0130208775
## 461	-2	Rhode Island	-7.050353e-04	0.012378842	-0.0212488324
## 462	-1	Rhode Island	5.708068e-03	0.018799777	-0.0275397825
## 463	0	Rhode Island	-2.607626e-02	0.019984596	-0.0524884136
## 464	1	Rhode Island	-2.878355e-02	0.022421602	-0.0579378707
## 465	NA	Rhode Island	-2.742990e-02	0.021192551	-0.0552131419
## 466	-12	Utah	3.211616e-02	0.100993411	-0.1536692820
## 467	-11	Utah	4.427464e-02	0.092692149	-0.1321134313
## 468	-10	Utah	6.203015e-03	0.018205328	-0.0272444973
## 469	-9	Utah	5.657675e-03	0.019739808	-0.0301444876
## 470	-8	Utah	-5.031201e-03	0.021569954	-0.0388729470
## 471	-7	Utah	-9.093646e-03	0.023855427	-0.0447586818
## 472	-6	Utah	-3.626450e-03	0.011809243	-0.0217460775
## 473	-5	Utah	-3.927796e-03	0.012006250	-0.0222283280
## 474	-4	Utah	-1.989720e-02	0.023573720	-0.0463838164
## 475	-3	Utah	-1.404148e-02	0.024017913	-0.0467222927
## 476	-2	Utah	-1.840027e-02	0.024023333	-0.0473244307
## 477	-1	Utah	-1.423345e-02	0.022398220	-0.0438129164
## 478	0	Utah	-1.293536e-02	0.021724185	-0.0423158386
## 479	1	Utah	NA	NaN	NA
## 480	NA	Utah	-1.293536e-02	0.021724185	-0.0423158386
## 481	-12	Vermont	NA	NaN	NA
## 482	-11	Vermont	NA	NaN	NA
## 483	-10	Vermont	NA	NaN	NA
## 484	-9	Vermont	NA	NaN	NA
## 485	-8	Vermont	NA	NaN	NA
## 486	-7	Vermont	NA	NaN	NA
## 487	-6	Vermont	1.644172e-03	0.068223087	-0.1287258935

## 488	-5	Vermont	-5.973458e-04	0.061428399	-0.1165527025
## 489	-4	Vermont	-8.107928e-05	0.029970926	-0.0486734371
## 490	-3	Vermont	-2.335120e-03	0.032960025	-0.0547381424
## 491	-2	Vermont	-6.393527e-05	0.036297602	-0.0613657562
## 492	-1	Vermont	1.433308e-03	0.031129420	-0.0506322169
## 493	0	Vermont	-6.342753e-03	0.030015054	-0.0528626602
## 494	1	Vermont	-5.839137e-03	0.023023210	-0.0481325395
## 495	NA	Vermont	-6.090945e-03	0.025919488	-0.0484204287
## 496	-12	Virginia	NA	NaN	NA
## 497	-11	Virginia	-7.306868e-04	0.034724219	-0.0591293489
## 498	-10	Virginia	-1.900077e-03	0.036246623	-0.0575939380
## 499	-9	Virginia	-9.985294e-06	0.011581062	-0.0184420762
## 500	-8	Virginia	-4.902555e-03	0.007971018	-0.0182597925
## 501	-7	Virginia	7.779925e-04	0.009491652	-0.0187839299
## 502	-6	Virginia	8.975205e-04	0.006493561	-0.0099748318
## 503	-5	Virginia	2.685063e-03	0.011000633	-0.0182783849
## 504	-4	Virginia	3.326303e-03	0.012867846	-0.0201730783
## 505	-3	Virginia	2.600563e-03	0.016660026	-0.0289633426
## 506	-2	Virginia	-1.284143e-03	0.012263720	-0.0252922079
## 507	-1	Virginia	-1.459996e-03	0.010043907	-0.0206759271
## 508	0	Virginia	-1.401384e-02	0.007154205	-0.0246454268
## 509	1	Virginia	-1.472141e-02	0.007497121	-0.0258703843
## 510	NA	Virginia	-1.436762e-02	0.007314591	-0.0252486658
## 511	-12	Washington	NA	NaN	NA
## 512	-11	Washington	NA	NaN	NA
## 513	-10	Washington	NA	NaN	NA
## 514	-9	Washington	NA	NaN	NA
## 515	-8	Washington	NA	NaN	NA
## 516	-7	Washington	NA	NaN	NA
## 517	-6	Washington	-2.101208e-04	0.014059299	-0.0263769983
## 518	-5	Washington	1.225242e-04	0.015213856	-0.0280590028
## 519	-4	Washington	-1.670990e-04	0.005428772	-0.0085319667
## 520	-3	Washington	-3.650552e-04	0.005410329	-0.0100435511
## 521	-2	Washington	-6.773906e-05	0.007612819	-0.0128304500
## 522	-1	Washington	6.874899e-04	0.012227161	-0.0209373272
## 523	0	Washington	-3.232985e-02	0.021058752	-0.0595313031
## 524	1	Washington	-4.561742e-02	0.030208607	-0.0830792813
## 525	NA	Washington	-3.897364e-02	0.025555946	-0.0709795283
## 526	-12	West Virginia	NA	NaN	NA
## 527	-11	West Virginia	NA	NaN	NA
## 528	-10	West Virginia	NA	NaN	NA
## 529	-9	West Virginia	NA	NaN	NA
## 530	-8	West Virginia	NA	NaN	NA
## 531	-7	West Virginia	NA	NaN	NA
## 532	-6	West Virginia	4.085459e-04	0.009212959	-0.0187415148
## 533	-5	West Virginia	-4.661518e-04	0.008713304	-0.0170232486
## 534	-4	West Virginia	-1.521051e-03	0.004088859	-0.0085984538
## 535	-3	West Virginia	4.406058e-03	0.009969879	-0.0148652084
## 536	-2	West Virginia	-1.916591e-03	0.007539303	-0.0117903657
## 537	-1	West Virginia	-9.108102e-04	0.004739824	-0.0097393180
## 538	0	West Virginia	-3.322117e-02	0.028565260	-0.0649126802
## 539	1	West Virginia	-5.064441e-02	0.045456050	-0.0992161580
## 540	NA	West Virginia	-4.193279e-02	0.036984567	-0.0818675903
##	upper_bound				

## 1	0.078134387
## 2	0.059022215
## 3	0.026859978
## 4	0.012964496
## 5	0.016818169
## 6	0.014407099
## 7	0.015015184
## 8	0.011916145
## 9	0.006332432
## 10	0.006205907
## 11	0.005825194
## 12	0.006073928
## 13	-0.002682828
## 14	-0.005366612
## 15	-0.004522737
## 16	NA
## 17	NA
## 18	NA
## 19	NA
## 20	NA
## 21	0.041346130
## 22	0.032073580
## 23	0.035407328
## 24	0.016185006
## 25	0.021513585
## 26	0.017600165
## 27	0.011281682
## 28	0.013393484
## 29	0.006467355
## 30	0.009827969
## 31	NA
## 32	NA
## 33	NA
## 34	NA
## 35	NA
## 36	NA
## 37	0.044461138
## 38	0.023310869
## 39	0.025768368
## 40	0.017839550
## 41	0.015485512
## 42	0.014924099
## 43	0.014535866
## 44	0.021268720
## 45	0.018219981
## 46	NA
## 47	NA
## 48	NA
## 49	NA
## 50	NA
## 51	NA
## 52	0.055363551
## 53	0.028722839
## 54	0.019453151

## 55	0.023344153
## 56	0.020428427
## 57	0.021989787
## 58	0.034048055
## 59	0.036551788
## 60	0.034806507
## 61	NA
## 62	NA
## 63	NA
## 64	NA
## 65	NA
## 66	NA
## 67	0.029132225
## 68	0.029351267
## 69	0.009965563
## 70	0.011690605
## 71	0.013429691
## 72	0.016027577
## 73	0.030327259
## 74	0.051547429
## 75	0.040363514
## 76	NA
## 77	NA
## 78	NA
## 79	NA
## 80	NA
## 81	NA
## 82	0.044361598
## 83	0.019712538
## 84	0.010212072
## 85	0.016682901
## 86	0.017379918
## 87	0.024766921
## 88	0.036951259
## 89	0.034478288
## 90	0.035714775
## 91	NA
## 92	NA
## 93	NA
## 94	NA
## 95	NA
## 96	NA
## 97	0.022615344
## 98	0.023738098
## 99	0.019458350
## 100	0.016446014
## 101	0.017974565
## 102	0.015614759
## 103	0.015535827
## 104	0.022500010
## 105	0.016340144
## 106	NA
## 107	NA
## 108	NA

## 109	NA
## 110	NA
## 111	NA
## 112	0.059277850
## 113	0.048804708
## 114	0.026623412
## 115	0.036621099
## 116	0.038858355
## 117	0.019720521
## 118	0.015705571
## 119	0.010533047
## 120	0.005191573
## 121	NA
## 122	NA
## 123	NA
## 124	NA
## 125	NA
## 126	NA
## 127	0.025269953
## 128	0.010080263
## 129	0.004433351
## 130	0.010152380
## 131	0.025090702
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## 133	0.030373122
## 134	0.038344361
## 135	0.034572175
## 136	NA
## 137	NA
## 138	NA
## 139	NA
## 140	NA
## 141	NA
## 142	0.026528406
## 143	0.024184110
## 144	0.012869536
## 145	0.014176792
## 146	0.017549034
## 147	0.009334280
## 148	0.011622956
## 149	0.019531143
## 150	0.012422131
## 151	0.044478385
## 152	0.030903072
## 153	0.027635350
## 154	0.014587792
## 155	0.009812186
## 156	0.013794201
## 157	0.004553078
## 158	0.017671366
## 159	0.018431367
## 160	0.020369904
## 161	0.016445118
## 162	0.028457640

## 163	0.025213222
## 164	NA
## 165	0.025213222
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## 169	NA
## 170	NA
## 171	NA
## 172	0.018847866
## 173	0.013407066
## 174	0.005372952
## 175	0.017780085
## 176	0.008944414
## 177	0.007236378
## 178	0.013458823
## 179	0.024504782
## 180	0.018770426
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## 182	NA
## 183	NA
## 184	NA
## 185	NA
## 186	0.010329582
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## 189	0.007002527
## 190	0.005173870
## 191	0.002819800
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## 193	0.006334092
## 194	0.014000117
## 195	0.010303897
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## 197	NA
## 198	NA
## 199	NA
## 200	NA
## 201	NA
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## 203	0.022006591
## 204	0.011782865
## 205	0.012626003
## 206	0.015560167
## 207	0.006908200
## 208	0.009387154
## 209	0.005502900
## 210	0.003287005
## 211	NA
## 212	NA
## 213	NA
## 214	NA
## 215	NA
## 216	NA

```
## 217 0.031818311
## 218 0.029304177
## 219 0.017281736
## 220 0.018775130
## 221 0.023834757
## 222 0.012232355
## 223 0.030608312
## 224 0.031305354
## 225 0.029821972
## 226      NA
## 227      NA
## 228      NA
## 229      NA
## 230 0.035761288
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## 232 0.009746053
## 233 0.008208953
## 234 0.011090323
## 235 0.004698703
## 236 0.011026488
## 237 0.027770672
## 238 0.032340656
## 239 0.052097329
## 240 0.041434817
## 241      NA
## 242      NA
## 243      NA
## 244      NA
## 245      NA
## 246      NA
## 247 0.022689668
## 248 0.015967688
## 249 0.005712041
## 250 0.021535635
## 251 0.013408693
## 252 0.007285044
## 253 0.011770163
## 254 0.012715760
## 255 0.010856996
## 256      NA
## 257      NA
## 258      NA
## 259      NA
## 260      NA
## 261      NA
## 262 0.025311960
## 263 0.027469693
## 264 0.006908501
## 265 0.014425531
## 266 0.015023630
## 267 0.014915291
## 268 0.015988968
## 269 0.022918538
## 270 0.019456250
```

## 271	NA
## 272	NA
## 273	NA
## 274	NA
## 275	NA
## 276	NA
## 277	0.019411754
## 278	0.017718114
## 279	0.011167041
## 280	0.011685826
## 281	0.010766910
## 282	0.004880455
## 283	0.005649917
## 284	0.002751088
## 285	0.002670764
## 286	NA
## 287	NA
## 288	NA
## 289	NA
## 290	0.060699405
## 291	0.059903213
## 292	0.017206487
## 293	0.024864220
## 294	0.030390790
## 295	0.018827764
## 296	0.028809898
## 297	0.045089006
## 298	0.063971465
## 299	0.063460395
## 300	0.063704578
## 301	0.028832779
## 302	0.028086323
## 303	0.021562226
## 304	0.014025620
## 305	0.019861003
## 306	0.015843243
## 307	0.003101641
## 308	0.003434026
## 309	0.035930877
## 310	0.025946031
## 311	0.017204807
## 312	0.005637100
## 313	0.010949751
## 314	NA
## 315	0.010949751
## 316	NA
## 317	NA
## 318	NA
## 319	NA
## 320	NA
## 321	NA
## 322	0.048773383
## 323	0.047808348
## 324	0.017631029

```
## 325 0.020219342
## 326 0.030303265
## 327 0.026801666
## 328 0.044087920
## 329 0.057171245
## 330 0.050297387
## 331      NA
## 332      NA
## 333      NA
## 334      NA
## 335      NA
## 336      NA
## 337 0.024974180
## 338 0.013287017
## 339 0.008981010
## 340 0.021687191
## 341 0.015190771
## 342 0.013272527
## 343 0.025904094
## 344 0.016452813
## 345 0.020944668
## 346      NA
## 347      NA
## 348      NA
## 349      NA
## 350      NA
## 351      NA
## 352 0.029513766
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## 354 0.017023197
## 355 0.015466803
## 356 0.019512161
## 357 0.020173250
## 358 0.020060400
## 359 0.008700856
## 360 0.010423563
## 361      NA
## 362      NA
## 363      NA
## 364      NA
## 365      NA
## 366      NA
## 367 0.039934894
## 368 0.016914412
## 369 0.017567666
## 370 0.015252767
## 371 0.022296467
## 372 0.013415540
## 373 0.030680730
## 374 0.051359055
## 375 0.041139401
## 376      NA
## 377      NA
## 378      NA
```

## 379	NA
## 380	NA
## 381	NA
## 382	0.025919399
## 383	0.020737037
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## 385	0.013221952
## 386	0.014814745
## 387	0.011940612
## 388	0.013017724
## 389	0.015153156
## 390	0.013923056
## 391	NA
## 392	NA
## 393	NA
## 394	NA
## 395	NA
## 396	NA
## 397	0.014430121
## 398	0.015174543
## 399	0.019039339
## 400	0.010862182
## 401	0.017190587
## 402	0.011993532
## 403	0.012347530
## 404	0.039792746
## 405	0.025713117
## 406	NA
## 407	NA
## 408	NA
## 409	NA
## 410	NA
## 411	NA
## 412	0.033305410
## 413	0.032944633
## 414	0.011697036
## 415	0.011840466
## 416	0.015890432
## 417	0.017517170
## 418	0.017118866
## 419	0.020048810
## 420	0.018601262
## 421	NA
## 422	NA
## 423	NA
## 424	NA
## 425	NA
## 426	NA
## 427	0.047350747
## 428	0.057606456
## 429	0.025716653
## 430	0.029087267
## 431	0.033907525
## 432	0.029063150

```
## 433 0.044784373
## 434 0.056749460
## 435 0.050990444
## 436 NA
## 437 NA
## 438 NA
## 439 NA
## 440 NA
## 441 0.020608802
## 442 0.022048040
## 443 0.015599868
## 444 0.008088038
## 445 0.013159929
## 446 0.008535682
## 447 0.025746359
## 448 0.016022580
## 449 0.025664740
## 450 0.020907135
## 451 NA
## 452 NA
## 453 NA
## 454 NA
## 455 NA
## 456 NA
## 457 0.033914460
## 458 0.041295571
## 459 0.017297247
## 460 0.008136164
## 461 0.020617640
## 462 0.034636697
## 463 0.019149175
## 464 0.022571882
## 465 0.020860528
## 466 0.186525375
## 467 0.177409073
## 468 0.033990494
## 469 0.036296763
## 470 0.033725217
## 471 0.035453791
## 472 0.018036996
## 473 0.018209079
## 474 0.026027332
## 475 0.032358508
## 476 0.028499108
## 477 0.029250937
## 478 0.029081826
## 479 NA
## 480 0.029081826
## 481 NA
## 482 NA
## 483 NA
## 484 NA
## 485 NA
## 486 NA
```

```
## 487 0.110723919
## 488 0.101778665
## 489 0.057728247
## 490 0.061435385
## 491 0.069222936
## 492 0.057052234
## 493 0.050489999
## 494 0.041422082
## 495 0.041420582
## 496 NA
## 497 0.060665978
## 498 0.064928561
## 499 0.018389838
## 500 0.010860700
## 501 0.019628311
## 502 0.012534444
## 503 0.018582771
## 504 0.022142179
## 505 0.026226049
## 506 0.015884485
## 507 0.016186055
## 508 0.002078094
## 509 0.002508740
## 510 0.002105746
## 511 NA
## 512 NA
## 513 NA
## 514 NA
## 515 NA
## 516 NA
## 517 0.026111207
## 518 0.027128900
## 519 0.011602160
## 520 0.009522832
## 521 0.015042443
## 522 0.020203797
## 523 0.014192372
## 524 0.017845801
## 525 0.015807247
## 526 NA
## 527 NA
## 528 NA
## 529 NA
## 530 NA
## 531 NA
## 532 0.017261121
## 533 0.015182772
## 534 0.008005999
## 535 0.020101524
## 536 0.012862472
## 537 0.008056480
## 538 0.021470530
## 539 0.035984751
## 540 0.028022682
```

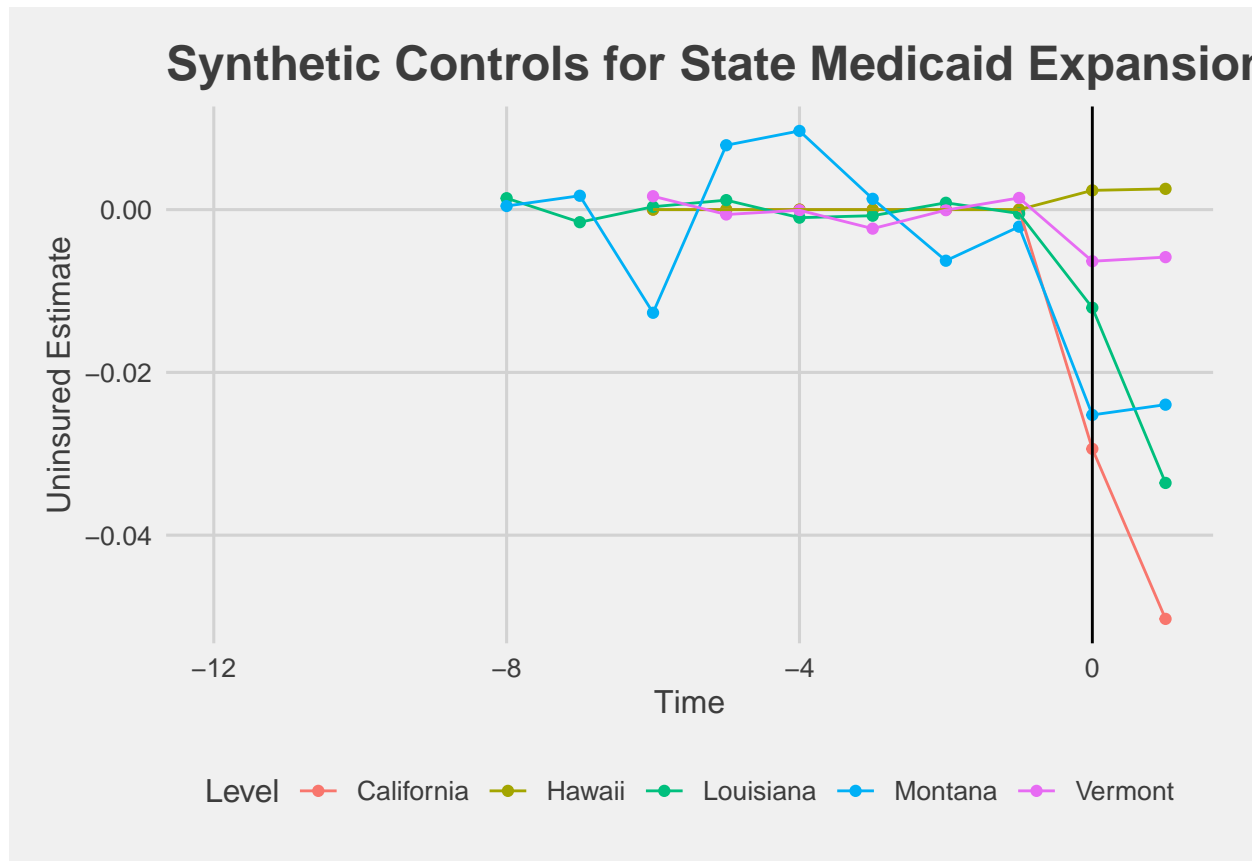


```
# -----
no_pool_synsum_filt_att <- as.data.frame(no_pool_synsum$att) %>% filter(Level %in% c("Hawaii", "California"))

no_pool_synsum_filt_att %>%
  ggplot(aes(x = Time, y = Estimate, color = Level)) +
  geom_point() +
  geom_line() +
  geom_vline(xintercept = 0) +
  theme_fivethirtyeight() +
  theme(axis.title = element_text(),
        legend.position = "bottom") +
  ggtitle('Synthetic Controls for State Medicaid Expansion') +
  xlab('Time') +
  ylab('Uninsured Estimate')
```

```
## Warning: Removed 31 rows containing missing values or values outside the scale range
## (`geom_point()`).
```

```
## Warning: Removed 31 rows containing missing values or values outside the scale range
## (`geom_line()`).
```



- Estimate a multisynth model using time cohorts. For the purpose of this exercise, you can simplify the treatment time so that states that adopted Medicaid expansion within the same year (i.e. all states that adopted expansion in 2016) count for the same cohort. Plot the treatment effects for these time cohorts.

```
# multisynth model time cohorts
```

```

# -----
ppool_syn_time <- multisynth(uninsured_rate ~ adopted,
                             State,
                             year,
                             medicaid_expansion_clean,
                             n_leads = 6,
                             time_cohort = TRUE)           # time cohort set to TRUE

# save summary
ppool_syn_time_summ <- summary(ppool_syn_time)

ppool_syn_time_summ

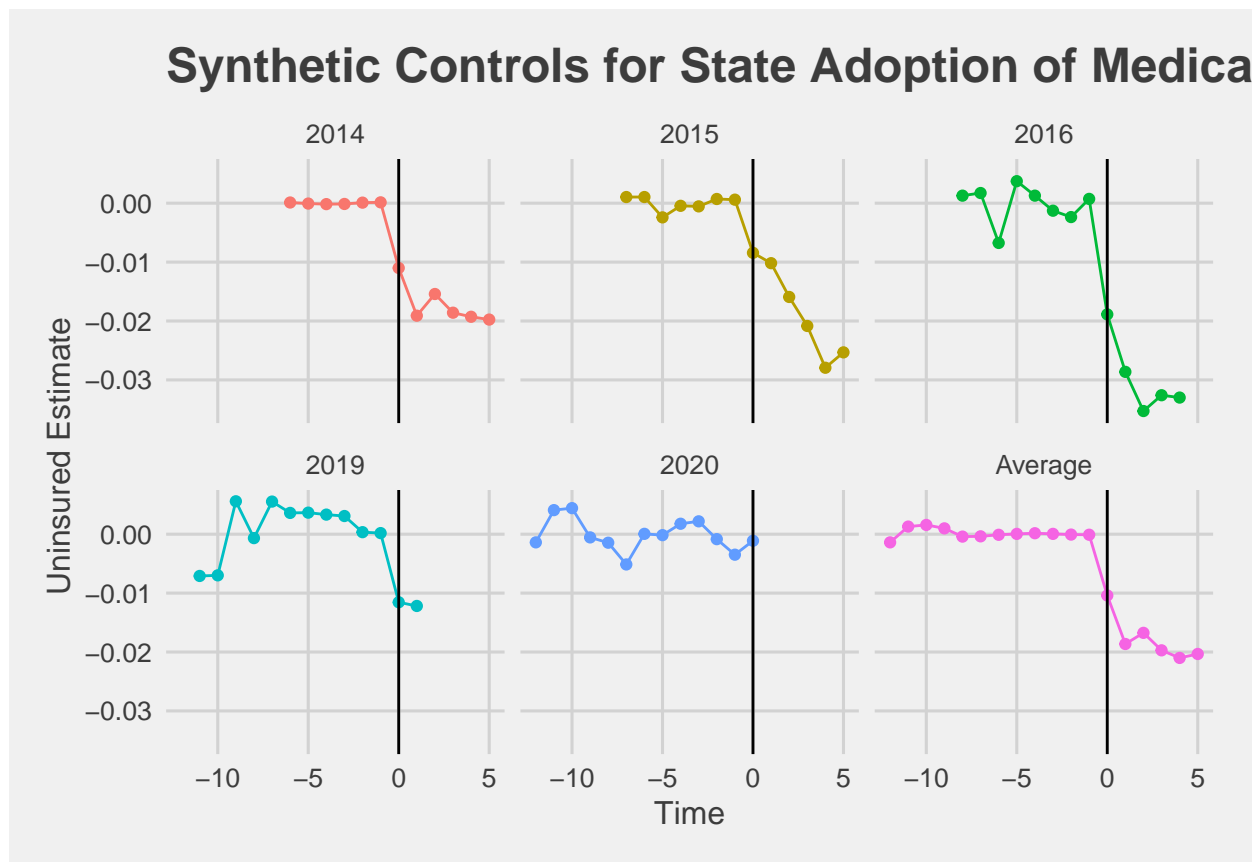
##
## Call:
## multisynth(form = uninsured_rate ~ adopted, unit = State, time = year,
##           data = medicaid_expansion_clean, n_leads = 6, time_cohort = TRUE)
##
## Average ATT Estimate (Std. Error): -0.016 (0.006)
##
## Global L2 Imbalance: 0.001
## Scaled Global L2 Imbalance: 0.008
## Percent improvement from uniform global weights: 99.2
##
## Individual L2 Imbalance: 0.005
## Scaled Individual L2 Imbalance: 0.018
## Percent improvement from uniform individual weights: 98.2
##
## Time Since Treatment   Level   Estimate   Std.Error lower_bound upper_bound
##                      0 Average -0.01039793 0.004731448 -0.02038151 -0.001681147
##                      1 Average -0.01864128 0.005918010 -0.03017164 -0.006974274
##                      2 Average -0.01674715 0.005971125 -0.02819730 -0.005390146
##                      3 Average -0.01971271 0.006150733 -0.03181973 -0.008017841
##                      4 Average -0.02100310 0.006051889 -0.03298715 -0.009754213
##                      5 Average -0.02033303 0.005577805 -0.03088554 -0.009006378

ppool_syn_time_summ$att %>%
  ggplot(aes(x = Time, y = Estimate, color = Level)) +
  geom_point() +
  geom_line() +
  geom_vline(xintercept = 0) +
  theme_fivethirtyeight() +
  theme(axis.title = element_text(),
        legend.position = 'None') +
  ggtitle('Synthetic Controls for State Adoption of Medicare') +
  xlab('Time') +
  ylab('Uninsured Estimate') +
  facet_wrap(~Level)

## Warning: Removed 32 rows containing missing values or values outside the scale range
## (`geom_point()`).

## Warning: Removed 32 rows containing missing values or values outside the scale range
## (`geom_line()`).

```



Discussion Questions

- One feature of Medicaid is that it is jointly administered by the federal government and the states, and states have some flexibility in how they implement Medicaid. For example, during the Trump administration, several states applied for waivers where they could add work requirements to the eligibility standards (i.e. an individual needed to work for 80 hours/month to qualify for Medicaid). Given these differences, do you see evidence for the idea that different states had different treatment effect sizes?
- **Answer:**
 - When analyzing on a state-by-state basis there does seem to be a slight difference between treatment effect, with some states having a slightly positive ATT and others having an ATT of nearly -0.04. This somewhat dissolves when considering time cohorts, where each group falls between 0 and -0.02. Even here, the range of ATT is very small and not as large as I would have expected given the different conditions for treatment.
- Do you see evidence for the idea that early adopters of Medicaid expansion enjoyed a larger decrease in the uninsured population?
- **Answer:**
 - It can be difficult to assess this robustly given that there appears to be a continued decline in the uninsured population for some years after treatment in the early adopters. However, in the year of treatment there does not appear to be a significant difference in the treatment effect.

General Discussion Questions

- Why are DiD and synthetic control estimates well suited to studies of aggregated units like cities, states, countries, etc?
- **Answer:**
 - DiD can effectively control for unmeasured, time-invariant variables by comparing changes over time between treatment and control groups, assuming similar trends absent the treatment. It allows you to draw insights despite unknown confounders that are often present in natural experiments and thus are useful for large-scale policy changes that affect multiple regions. SCM allows for a similar analysis when the DiD assumption of pre-treatment similarities does not hold, making it easier to perform a treatment analysis at regional scales when multiple factors may influence the outcome.
- What role does selection into treatment play in DiD/synthetic control versus regression discontinuity? When would we want to use either method?
- **Answer:**
 - DiD and SCM handle nonrandom selection by assuming parallel trends or matching pretreatment characteristics, respectively, making them suitable for analyses with multiple periods. Regression discontinuity uses a sharp cutoff for treatment assignment, making it suitable for natural experiments where treatment is assigned by a specific threshold. DiD or SCM are more suitable for broader policy impacts across units while regression discontinuity applies when precise cutoff-based treatment assignment allows for quasi-experimental conditions.