SS154 Final - Analyzing South Africa's 2006 Same-Sex Marriage Legalization Impact on Foreign Direct Investments

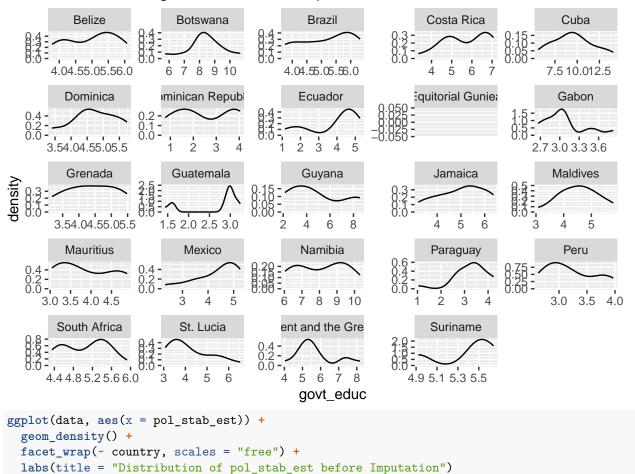
```
# install.packages('vtable')
library(vtable)
## Loading required package: kableExtra
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr
             1.1.4
                       v readr
                                    2.1.5
## v forcats 1.0.0 v stringr 1.5.1
## v ggplot2 3.5.0
                                     3.2.1
                        v tibble
## v lubridate 1.9.3
                     v tidyr
                                     1.3.1
## v purrr
              1.0.2
                                                ----- tidyverse_conflicts() --
## -- Conflicts -----
## x dplyr::filter()
                        masks stats::filter()
## x dplyr::group_rows() masks kableExtra::group_rows()
## x dplyr::lag()
                     masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(dplyr)
library(mice)
##
## Attaching package: 'mice'
## The following object is masked from 'package:stats':
##
##
       filter
## The following objects are masked from 'package:base':
##
##
       cbind, rbind
library(Synth)
## ##
## ## Synth Package: Implements Synthetic Control Methods.
## ## See https://web.stanford.edu/~jhain/synthpage.html for additional information.
# import the data
data <- read.csv("https://docs.google.com/spreadsheets/d/e/2PACX-1vQPw6yEuH2geG6KCkYoD8VcBoMgP1xg9RADG1
colnames(data)
  [1] "Country"
```

```
##
   [2] "Year"
##
   [3] "id"
  [4] "GDP.per.capita..constant.2015.US.."
##
   [5] "Foreign.direct.investment..net.inflows....of.GDP."
   [6] "Foreign.direct.investment..net.outflows....of.GDP."
  [7] "GDP.per.capita..current.US.."
##
  [8] "Trade....of.GDP."
  [9] "Political.Stability.and.Absence.of.Violence.Terrorism..Estimate"
##
## [10] "Government.expenditure.on.education..total....of.GDP."
## [11] "Individuals.using.the.Internet....of.population."
## [12] "Population..total"
## [13] "Taxes.on.international.trade....of.revenue."
## [14] "Taxes.on.international.trade..current.LCU."
## [15] "Real.interest.rate...."
## [16] "Foreign.direct.investment..net.inflows..BoP..current.US.."
## [17] "fdi_bop_per_capita"
head(data)
     Country Year id GDP.per.capita..constant.2015.US...
## 1 Brazil 1990 1
                                                6086.078
## 2 Brazil 1991 1
                                                6043.433
## 3 Brazil 1992 1
                                                5911.687
     Brazil 1993 1
                                                6103.377
## 5 Brazil 1994 1
                                                6358.679
## 6 Brazil 1995 1
                                                6524.519
##
     Foreign.direct.investment..net.inflows....of.GDP.
## 1
                                              0.2140916
## 2
                                              0.2707826
## 3
                                              0.5276628
## 4
                                              0.2947738
## 5
                                              0.5624007
## 6
                                              0.6315860
##
     Foreign.direct.investment..net.outflows....of.GDP.
## 1
                                              0.14395442
## 2
                                              0.24893342
## 3
                                              0.03507511
## 4
                                              0.11202316
## 5
                                              0.18984685
## 6
                                              0.17989608
     GDP.per.capita..current.US.. Trade....of.GDP.
## 1
                         3065.242
                                           15.15560
## 2
                         2656.497
                                           16.59208
## 3
                         2505.378
                                           19.25337
## 4
                         2766.346
                                           19.59932
## 5
                         3393.143
                                           18.67476
## 6
                         4704.960
                                           16.98446
     Political.Stability.and.Absence.of.Violence.Terrorism..Estimate
## 1
## 2
                                                                   NA
## 3
                                                                   NA
## 4
                                                                   NA
## 5
                                                                   NA
## 6
                                                                   NA
     Government.expenditure.on.education..total....of.GDP.
```

```
## 1
                                                          NA
## 2
                                                          NΑ
## 3
                                                          NA
## 4
                                                          NA
## 5
                                                          NA
## 6
                                                     4.56816
     Individuals.using.the.Internet....of.population. Population..total
## 1
                                           0.00000000
                                                                150706446
## 2
                                           0.003288171
                                                                153336445
## 3
                                           0.012946262
                                                                155900790
## 4
                                           0.025498253
                                                                158440875
## 5
                                                                160980472
                                           0.037672709
## 6
                                           0.105138168
                                                                163515328
##
     Taxes.on.international.trade....of.revenue.
## 1
## 2
                                                NA
## 3
                                               NA
## 4
                                               NA
## 5
                                               NA
## 6
##
    Taxes.on.international.trade..current.LCU. Real.interest.rate....
## 1
                                              NA
## 2
                                              NA
                                                                      NA
## 3
                                              NA
                                                                      NA
## 4
                                              NΑ
                                                                      NΑ
## 5
                                              NA
                                                                      NA
## 6
                                              NA
                                                                      NA
     Foreign.direct.investment..net.inflows..BoP..current.US.. fdi_bop_per_capita
## 1
## 2
                                                              NA
                                                                                  NA
## 3
                                                              NA
                                                                                  NA
## 4
                                                              NA
                                                                                  NA
## 5
                                                              NA
                                                                                  NA
## 6
                                                        4.86e+09
                                                                                29.7
# renaming column names
data <-
  data %>%
    rename(
      country = Country,
      year = Year,
      fdi_in = Foreign.direct.investment..net.inflows....of.GDP.,
      fdi_out = Foreign.direct.investment..net.outflows....of.GDP.,
      gdp_cap_2015 = GDP.per.capita..constant.2015.US..,
      gdp_cap = GDP.per.capita..current.US..,
      trade_per_gdp = Trade....of.GDP.,
      real_interest = Real.interest.rate...,
      intnl_trade_tax = Taxes.on.international.trade....of.revenue.,
      intnl_trade_tax_lcu = Taxes.on.international.trade..current.LCU.,
      pol_stab_est = Political.Stability.and.Absence.of.Violence.Terrorism..Estimate,
      govt_educ = Government.expenditure.on.education..total....of.GDP.,
      internet = Individuals.using.the.Internet....of.population.,
      popn = Population..total
```

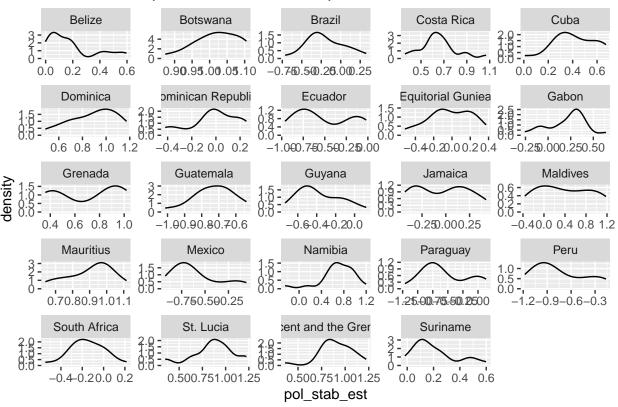
```
head(data)
##
     country year id gdp_cap_2015
                                     fdi_in
                                               fdi_out gdp_cap trade_per_gdp
## 1 Brazil 1990 1
                         6086.078 0.2140916 0.14395442 3065.242
                                                                      15.15560
                         6043.433 0.2707826 0.24893342 2656.497
## 2
     Brazil 1991 1
                                                                      16.59208
## 3 Brazil 1992 1
                         5911.687 0.5276628 0.03507511 2505.378
                                                                      19.25337
## 4 Brazil 1993 1
                         6103.377 0.2947738 0.11202316 2766.346
                                                                      19.59932
## 5 Brazil 1994 1
                         6358.679 0.5624007 0.18984685 3393.143
                                                                      18.67476
                         6524.519 0.6315860 0.17989608 4704.960
## 6 Brazil 1995 1
                                                                      16.98446
    pol_stab_est govt_educ
                                             popn intnl_trade_tax
##
                               internet
## 1
               NA
                         NA 0.00000000 150706446
## 2
               NA
                         NA 0.003288171 153336445
                                                                NA
## 3
               NA
                         NA 0.012946262 155900790
                                                                NA
## 4
               NA
                                                                NA
                         NA 0.025498253 158440875
## 5
               NA
                         NA 0.037672709 160980472
                                                                NA
## 6
               NA
                    4.56816 0.105138168 163515328
                                                                NA
##
     intnl_trade_tax_lcu real_interest
## 1
                      NA
## 2
                      NA
                                    NA
## 3
                      NA
                                    NA
## 4
                      NΑ
                                    NA
## 5
                      NA
                                    NA
## 6
                      NA
                                    NA
    Foreign.direct.investment..net.inflows..BoP..current.US.. fdi_bop_per_capita
##
## 1
                                                                                NA
## 2
                                                             NA
                                                                                NA
## 3
                                                             NΑ
                                                                                NA
## 4
                                                             NA
                                                                                NA
## 5
                                                             NA
                                                                                NA
## 6
                                                       4.86e+09
                                                                              29.7
ggplot(data, aes(x = govt_educ)) +
 geom_density() +
 facet_wrap(~ country, scales = "free") +
 labs(title = "Distribution of govt_educ before Imputation")
## Warning: Removed 304 rows containing non-finite outside the scale range
## (`stat_density()`).
## Warning: Groups with fewer than two data points have been dropped.
## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning
## -Inf
```

Distribution of govt_educ before Imputation



```
## Warning: Removed 216 rows containing non-finite outside the scale range
## (`stat_density()`).
```

Distribution of pol_stab_est before Imputation



```
imputed_data <- data %>%
filter(country != "Equitorial Guniea") %>%
group_by(country) %>% # Group the data by country
nest() %>% # Nest the data for each country
mutate(imputed_govt_educ = map(data, ~ mice(data = .x[, c("year", "govt_educ")], m = 5, method = "pmm mutate(imputed_pol_stab = map(data, ~ mice(data = .x[, c("year", "pol_stab_est")], m = 5, method = "pmm mutate(imputed_data = map2(imputed_govt_educ, imputed_pol_stab, ~bind_cols(complete(.x, action = "long unnest(cols = imputed_data) # Unnest the imputed_data column to bring the imputed values back into the
```

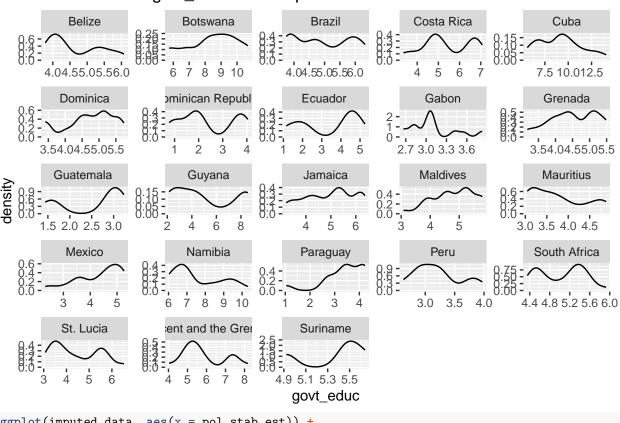
```
## New names:
##
  New names:
  New names:
## New names:
  New names:
## New names:
```

```
## New names:
## New names:
## New names:
## New names:
## * `.imp` -> `.imp...1`
## * `.id` -> `.id...2`
## * `year` -> `year...3`
## * `.imp` -> `.imp...5`
## * `.id` -> `.id...6`
## * `year` -> `year...7`

ggplot(imputed_data, aes(x = govt_educ)) +
    geom_density() +
    facet_wrap(~ country, scales = "free") +
    labs(title = "Distribution of govt_educ after Imputation")
```

Warning: Removed 275 rows containing non-finite outside the scale range
(`stat_density()`).

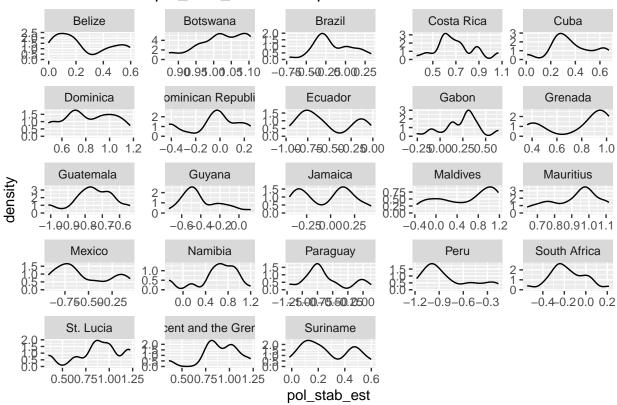
Distribution of govt_educ after Imputation



```
ggplot(imputed_data, aes(x = pol_stab_est)) +
  geom_density() +
  facet_wrap(~ country, scales = "free") +
  labs(title = "Distribution of pol_stab_est after Imputation")
```

Warning: Removed 207 rows containing non-finite outside the scale range
(`stat_density()`).

Distribution of pol_stab_est after Imputation



```
# Perform missing data imputation for govt_educ and pol_stab_est using mice (synthetic control)
imputed_data <- data %>%
  filter(country != "Equitorial Guniea") %>%
  group_by(country) %>%
  nest() %>%
  mutate(imputed_govt_educ = map(data, ~ mice(data = .x[, c("year", "govt_educ")], m = 5, method = "norm
  mutate(imputed_pol_stab = map(data, ~ mice(data = .x[, c("year", "pol_stab_est")], m = 5, method = "n
  mutate(imputed_data = map2(imputed_govt_educ, imputed_pol_stab, ~ left_join(complete(.x, action = "los
  unnest(cols = imputed_data)
## Warning: There were 23 warnings in `mutate()`.
## The first warning was:
## i In argument: `imputed_data = map2(...)`.
## i In group 1: `country = "Belize"`.
## Caused by warning in `left_join()`:
## ! Detected an unexpected many-to-many relationship between `x` and `y`.
## i Row 1 of `x` matches multiple rows in `y`.
## i Row 1 of `y` matches multiple rows in `x`.
## i If a many-to-many relationship is expected, set `relationship =
     "many-to-many" to silence this warning.
## i Run `dplyr::last_dplyr_warnings()` to see the 22 remaining warnings.
# Calculate the average of imputed values for each year within each country
imputed_data_avg <- imputed_data %>%
  group_by(country, year) %>%
  summarise(govt_educ = mean(govt_educ, na.rm = TRUE),
```

pol_stab_est = mean(pol_stab_est, na.rm = TRUE))

Table 1: Summary Statistics

Variable	N	Mean	Std. Dev.	Min	Pctl. 25	Pctl. 75	Max
gdp_cap	720	5037	2985	187	2806	6693	19850
trade_per_gdp	529	76	37	15	48	95	275
pol_stab_est	690	0.2	0.66	$-1.5 \\ 0.31 \\ 0$	-0.28	0.79	1.7
govt_educ	690	4.9	2.1		3.3	5.7	14
internet	713	19	22		0.31	34	81
fdi_bop_per_capita	715	354	1601	-794	0	224	23900

```
## `summarise()` has grouped output by 'country'. You can override using the
## `.groups` argument.
# Merge the averaged imputed data back into the filtered dataset
merged data <- data %>%
  select(-govt_educ, -pol_stab_est) %>% # Remove the original govt_educ and pol_stab_est variables
 left_join(imputed_data_avg, by = c("country", "year"))
# View the updated dataset
head(merged_data)
##
     country year id gdp_cap_2015
                                    fdi_in
                                               fdi_out gdp_cap trade_per_gdp
## 1 Brazil 1990 1
                         6086.078 0.2140916 0.14395442 3065.242
                                                                     15.15560
## 2 Brazil 1991 1
                         6043.433 0.2707826 0.24893342 2656.497
                                                                     16.59208
## 3 Brazil 1992 1
                         5911.687 0.5276628 0.03507511 2505.378
                                                                     19.25337
## 4 Brazil 1993 1
                         6103.377 0.2947738 0.11202316 2766.346
                                                                     19.59932
## 5 Brazil 1994 1
                         6358.679 0.5624007 0.18984685 3393.143
                                                                     18.67476
## 6 Brazil 1995 1
                         6524.519 0.6315860 0.17989608 4704.960
                                                                     16.98446
        internet
                     popn intnl_trade_tax intnl_trade_tax_lcu real_interest
## 1 0.00000000 150706446
                                        NA
                                                            NΑ
                                                                          NΑ
## 2 0.003288171 153336445
                                        NA
                                                            NΑ
## 3 0.012946262 155900790
                                        NA
                                                            NΑ
                                                                          NΑ
## 4 0.025498253 158440875
                                        NA
                                                            NΑ
                                                                          NΑ
                                                            NA
## 5 0.037672709 160980472
                                        NA
                                                                          NΑ
## 6 0.105138168 163515328
                                        NA
                                                            NA
    Foreign.direct.investment..net.inflows..BoP..current.US.. fdi_bop_per_capita
## 1
                                                            NA
                                                                               NA
## 2
                                                                               NA
                                                            NA
## 3
                                                            NA
                                                                               NΑ
## 4
                                                            NA
                                                                               NA
## 5
                                                            NΑ
                                                                               NΑ
## 6
                                                      4.86e+09
                                                                             29.7
##
    govt_educ pol_stab_est
## 1 2.857108 0.344843774
## 2 3.012677 -0.007502549
## 3 3.432674 0.314661194
## 4 3.500593 0.313099208
## 5 3.686193 0.054930021
## 6 4.568160 0.212099695
#descriptive stats table
```

st(merged_data, vars = c("gdp_cap", "trade_per_gdp", "pol_stab_est", "govt_educ", "internet", "fdi_bop_

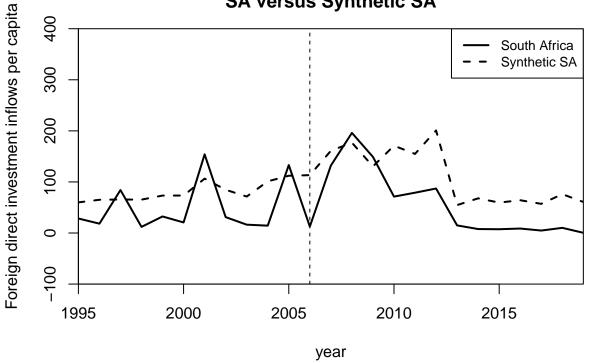
```
# dataprep for synthetic control
dataprep_out <- dataprep(</pre>
  foo = merged_data,
  predictors = c("gdp_cap", "trade_per_gdp", "pol_stab_est", "govt_educ", "internet"),
  # predictors.op = "mean",
  time.predictors.prior = 1995:2005,
  dependent = "fdi_bop_per_capita",
  unit.variable = "id",
  unit.names.variable = "country",
  time.variable = "year",
  treatment.identifier = 24,
  controls.identifier = c(2:5, 7:8, 10, 12:14, 16:20), # c(1:4, 7, 12, 14, 16:20),
  time.optimize.ssr = 1995:2005, # pretreatment for dependent
  time.plot = 1995:2019
)
synth_out <- synth(data.prep.obj = dataprep_out)</pre>
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
## ********
   searching for synthetic control unit
##
##
## *********
## ********
## ********
## MSPE (LOSS V): 2448.899
## solution.v:
## 0.0007954393 0.1208847 0.5294094 7.805e-07 0.3489097
##
## solution.w:
## 0.01641852 0.00595456 0.005184547 0.3304196 0.01750175 0.02019968 0.01049561 0.01719477 0.007194166
synth.tables <- synth.tab(</pre>
  dataprep.res = dataprep_out,
  synth.res = synth_out)
print(synth.tables)
## $tab.pred
                 Treated Synthetic Sample Mean
##
## gdp_cap
                 3815.434 3797.848
                                       3176.999
## trade_per_gdp 45.266
                          45.294
                                       83.102
## pol_stab_est
                  -0.230
                             -0.230
                                         0.018
                   4.715
                             5.244
                                          4.474
## govt_educ
## internet
                   4.682
                             4.682
                                          3.875
##
## $tab.v
##
                 v.weights
```

gdp_cap

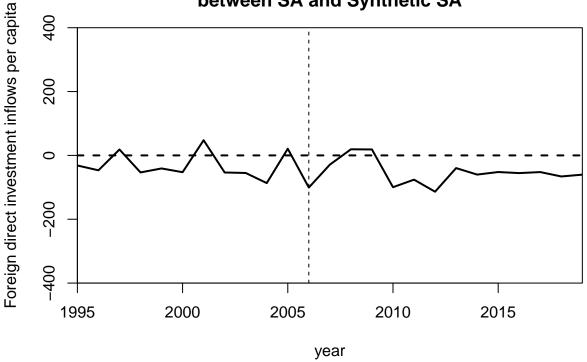
0.001

```
## trade_per_gdp 0.121
## pol_stab_est 0.529
## govt educ
## internet
                 0.349
## $tab.w
      w.weights
                         unit.names unit.numbers
## 2
          0.016
                             Belize
## 3
          0.006
                           Botswana
                                                3
## 4
          0.005
                         Costa Rica
                                                4
## 5
          0.330
                               Cuba
## 7
          0.018 Dominican Republic
                                                7
## 8
                           Ecuador
         0.020
                                                8
## 10
         0.010
                              Gabon
                                               10
## 12
         0.017
                         Guatemala
                                               12
## 13
         0.007
                             Guyana
                                               13
## 14
         0.015
                            Jamaica
                                               14
## 16
         0.013
                          Mauritius
                                               16
## 17
         0.279
                             Mexico
                                               17
## 18
         0.010
                            Namibia
                                               18
## 19
          0.012
                           Paraguay
                                               19
## 20
          0.240
                               Peru
                                               20
##
## $tab.loss
##
              Loss W
                       Loss V
## [1,] 2.257762e-07 2448.899
top_weights <- head(synth.tables$tab.w[order(synth.tables$tab.w$w, decreasing = TRUE), ], 5)
# Extract the top 5 weights
cat("\nRobustness test weights - taking out Cuba from Donor Pool:\n")
##
## Robustness test weights - taking out Cuba from Donor Pool:
print(top_weights)
                         unit.names unit.numbers
##
      w.weights
          0.330
## 5
                               Cuba
          0.279
                                               17
## 17
                             Mexico
## 20
          0.240
                               Peru
                                               20
## 8
          0.020
                           Ecuador
                                                8
## 7
          0.018 Dominican Republic
                                                7
# paths plot
path.plot(synth_out, dataprep_out,
          Ylab = c("Foreign direct investment inflows per capita"),
          Xlab = c("year"),
          Legend = c("South Africa", "Synthetic SA"),
          Main = "Trends in fdi_per_capita:\nSA versus Synthetic SA",
          Ylim = c(-100, 400)
)
abline(v = 2006, lty = 2)
```

Trends in fdi_per_capita: SA versus Synthetic SA



Foreign direct investment inflows per capita Gaps between SA and Synthetic SA



```
# Treatment Effects from Gaps
gaps<- dataprep_out$Y1plot-(
          dataprep_out$Y0plot%*%synth_out$solution.w
          ); gaps</pre>
```

```
##
## 1995
         -31.59382
## 1996
         -46.74574
## 1997
          18.46729
## 1998
         -53.35476
## 1999
         -40.89733
        -52.65242
## 2000
## 2001
          47.54637
         -53.55205
## 2002
## 2003
         -55.05410
## 2004
         -86.75245
## 2005
          20.89254
## 2006 -100.69469
## 2007
         -28.96954
## 2008
          19.12616
## 2009
          18.59559
        -99.64083
## 2010
## 2011
         -75.97929
## 2012 -113.74225
## 2013
        -39.76489
## 2014
         -60.11923
## 2015
         -52.15069
         -55.43398
## 2016
## 2017
        -52.27162
```

```
## 2018 -66.00093
## 2019 -60.33302
```

Robustness Checks

```
# install.packages("remotes")
# remotes::install_github("bcastanho/SCtools")
library(SCtools)
In-space Placebo tests
## Loading required package: future
placebos <- generate.placebos(dataprep_out, synth_out, Sigf.ipop = 5)</pre>
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## ********
   searching for synthetic control unit
##
##
## *********
## *********
## ********
##
## MSPE (LOSS V): 21083.08
##
## solution.v:
## 0.1379261 0.001349831 0.5456375 0.02317939 0.2919072
##
## solution.w:
## 0.01272003 0.1662284 8.5926e-06 0.007432914 0.008613743 0.3021805 0.0103668 0.01458496 0.01756509 0
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## *********
##
   searching for synthetic control unit
##
##
## ********
## ********
## ********
##
## MSPE (LOSS V): 8857.767
##
## solution.v:
## 0.01687796 0.1131923 0.01202475 0.004202498 0.8537025
##
## 0.0003118098 0.0002432953 0.00188681 1.85385e-05 8.2381e-06 0.2062467 4.0027e-06 0.0001106666 4.104
##
```

```
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## *********
## searching for synthetic control unit
##
## ********
## ********
## ********
## MSPE (LOSS V): 8021.158
##
## solution.v:
## 4e-10 0.5818519 5.7e-09 0.2960028 0.1221452
##
## solution.w:
## 2.1e-09 3.8206e-06 0.0001403703 0 1e-10 5e-10 1e-10 1.3e-09 6.1e-09 0.494132 0.5057238 1.35e-08 6e-
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
## ********
## searching for synthetic control unit
##
## *********
## *********
## *********
##
## MSPE (LOSS V): 2930.691
## solution.v:
## 2.47e-08 0.1436044 0.01106524 7.3e-09 0.8453303
##
## solution.w:
## 5.86e-08 1.36e-07 1.248e-07 3.84e-08 0.8837426 5.5e-09 1.14e-08 1.8e-09 2.01e-08 2.68e-08 1.71825e-
##
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
## ********
## searching for synthetic control unit
##
##
## *********
## ********
## ********
## MSPE (LOSS V): 1556.187
##
```

```
## solution.v:
## 0.1512742 0.467467 0.04033938 2.89e-08 0.3409194
##
## solution.w:
## 0.03362408 0.04643646 0.09451097 0.0814824 0.06855493 0.03166498 0.1072659 0.06613143 0.04153417 0.
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## ********
## searching for synthetic control unit
##
##
## ********
## *********
## ********
##
## MSPE (LOSS V): 605.2585
## solution.v:
## 0.5760672 0.3973799 0.02616863 0.0003428146 4.14283e-05
##
## solution.w:
## 0.00210084 0.003842388 6.0489e-06 0.06378552 0.1479542 0.003900305 0.2945273 0.003400642 0.00391026
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## ********
   searching for synthetic control unit
##
##
## *********
## ********
## *********
##
## MSPE (LOSS V): 22394.31
##
## solution.v:
## 0.1546586 0.09924032 0.08586995 0.1937076 0.4665236
## solution.w:
## 0.1937159 0.3138974 2.2e-09 1.5e-09 3.87e-08 0.4923748 1.59e-08 1.09e-08 9.1e-09 1.4e-08 1.33e-08 1
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## ********
##
   searching for synthetic control unit
##
```

##

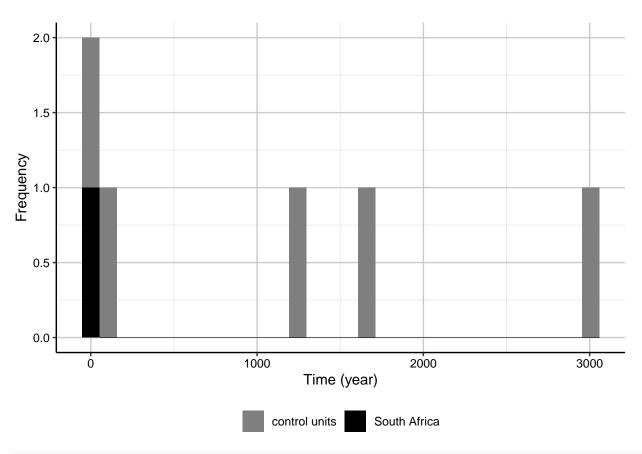
```
## *********
## ********
## ********
##
## MSPE (LOSS V): 3370.051
##
## solution.v:
## 0.5013578 0.0003128123 0.4772426 0.01962149 0.00146533
## solution.w:
## 2e-08 1.64e-08 3.39e-08 1.43e-08 3.508e-07 2.7046e-06 2.33e-08 5.1e-09 2.35e-08 3.46e-08 4.8e-09 4.
##
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## ********
   searching for synthetic control unit
##
##
## ********
## ********
## ********
## MSPE (LOSS V): 644.1798
## solution.v:
## 0.255291 0.1602648 0.1645518 0.184098 0.2357943
##
## solution.w:
## 6.3e-09 0.04929471 1.32e-08 5e-10 3.2e-09 6e-10 2.4e-09 0 1.89e-08 9.43e-08 1.3e-09 0.4490357 0.501
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
   searching for synthetic control unit
##
## ********
## ********
## ********
##
## MSPE (LOSS V): 5648.466
##
## solution.v:
## 0.002244409 0.379159 0.008291287 0.5931836 0.01712172
##
## solution.w:
## 0.05061363 0.02919531 0.1343992 0.03071587 0.06397899 0.05139883 0.04416258 0.05307418 0.1689668 0.
##
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
```

```
##
##
## ********
  searching for synthetic control unit
##
##
##
## ********
## ********
##
## MSPE (LOSS V): 4568.05
##
## solution.v:
## 4.68e-08 0.5502713 0.005704177 0.4440245 6.3e-09
##
## solution.w:
## 5.26662e-05 1.66609e-05 4.39346e-05 1.09794e-05 0.472018 2.28506e-05 0.183043 2.05776e-05 0.3446516
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## *********
   searching for synthetic control unit
##
##
## *********
## ********
## *********
## MSPE (LOSS V): 5554.548
##
## 0.0007936399 1.0981e-06 0.001907755 0.7128054 0.2844921
## solution.w:
## 0.4343238 0.004243963 0.280997 0.02446824 0.0008936169 0.0001786536 0.0004287662 0.0002177069 0.001
##
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
## *********
   searching for synthetic control unit
##
## *********
## ********
## ********
## MSPE (LOSS V): 2719.347
##
## solution.v:
## 0.6197193 0.002311019 0.08643633 0.2909228 0.0006106084
```

```
##
## solution.w:
## 3.78534e-05 0.2172639 0.263903 2.149e-07 0.01946617 0.0004785324 5.69621e-05 0.005150644 0.42467 4.
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
## *********
## searching for synthetic control unit
##
##
## ********
## ********
## ********
## MSPE (LOSS V): 1285.134
##
## solution.v:
## 0.000197648 0.04933212 0.2689966 0.66689 0.01458362
##
## 7.522e-07 2.4008e-06 5.239e-07 1.74637e-05 2.894e-07 2.0935e-06 1.0004e-06 0.5226855 0.2580118 1.42
##
## X1, X0, Z1, Z0 all come directly from dataprep object.
##
##
## ********
## searching for synthetic control unit
##
##
## *********
## ********
## *********
##
## MSPE (LOSS V): 1422.7
##
## solution.v:
## 0.05927164 0.5043216 0.198307 0.1779857 0.06011404
## solution.w:
## 1.5422e-06 1.1871e-06 4.8475e-06 0.02615989 1.7978e-06 0.7918963 6.702e-07 0.03226622 6.512e-07 3.3
## New names:
## * `w.weight` -> `w.weight...1`
## * `w.weight` -> `w.weight...2`
## * `w.weight` -> `w.weight...3`
## * `w.weight` -> `w.weight...4`
## * `w.weight` -> `w.weight...5`
## * `w.weight` -> `w.weight...6`
## * `w.weight` -> `w.weight...7`
## * `w.weight` -> `w.weight...8`
## * `w.weight` -> `w.weight...9`
```

```
## * `w.weight` -> `w.weight...10`
## * `w.weight` -> `w.weight...11`
## * `w.weight` -> `w.weight...12`
## * `w.weight` -> `w.weight...13`
## * `w.weight` -> `w.weight...14`
## * `w.weight` -> `w.weight...15`
plot_placebos(placebos, xlab = "Time (year)", ylab = "Foreign direct investment inflows per capita")
Foreign direct investment inflows per capita
    5000
   -5000
                            2000
                                                                                                 2020
           1995
                                              2005
                                                               2010
                                                                                2015
                                                Time (year)
                                       — Control units — South Africa
```

mspe.plot(placebos, discard.extreme = TRUE, mspe.limit = 1, plot.hist = TRUE, xlab = "Time (year)", yla
`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

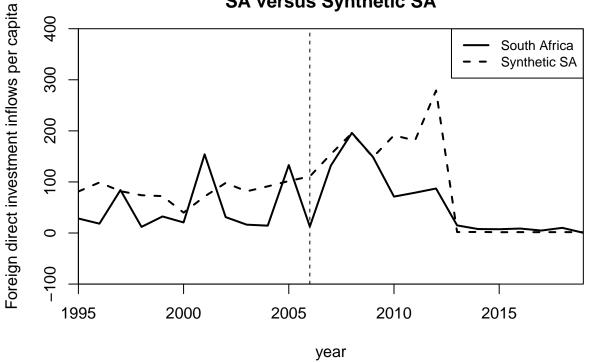


```
# ROBUSTNESS TEST -> taking out highest weighted country (Cuba 5) from synthetic control donor pool
dataprep_out2 <- dataprep(</pre>
  foo = merged data,
  predictors = c("gdp_cap", "trade_per_gdp", "pol_stab_est", "govt_educ", "internet"),
  # predictors.op = "mean",
  time.predictors.prior = 1995:2005,
  dependent = "fdi_bop_per_capita",
  unit.variable = "id",
  unit.names.variable = "country",
  time.variable = "year",
  treatment.identifier = 24,
  controls.identifier = c(2:4, 7:8, 10, 12:14, 16:20), # c(1:4, 7, 12, 14, 16:20),
  time.optimize.ssr = 1995:2005, # pretreatment for dependent
  time.plot = 1995:2019
)
synth_out2 <- synth(data.prep.obj = dataprep_out2)</pre>
```

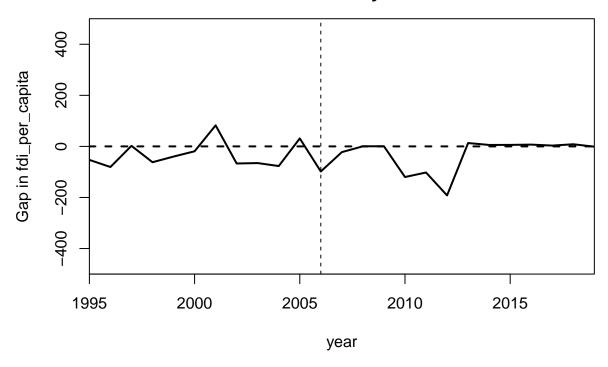
Remove highest weight country

```
searching for synthetic control unit
##
##
## ********
## *********
## ********
## MSPE (LOSS V): 3412.635
##
## solution.v:
## 6.23756e-05 0.7048776 0.0005266627 0.0005036989 0.2940297
##
## solution.w:
## 0.0001023298 0.06968575 6.07595e-05 0.000155608 0.297274 0.0001313628 0.0003541634 2.3947e-05 0.000
synth.tables <- synth.tab(</pre>
 dataprep.res = dataprep_out2,
  synth.res
               = synth_out2)
top_weights <- head(synth.tables$tab.w[order(synth.tables$tab.w$w, decreasing = TRUE), ], 5)
# Extract the top 5 weights
cat("\nRobustness test weights - taking out Cuba from Donor Pool:\n")
##
## Robustness test weights - taking out Cuba from Donor Pool:
print(top_weights)
     w.weights unit.names unit.numbers
         0.512
## 20
                     Peru
## 8
         0.297
                Ecuador
## 17
         0.119
                                    17
                   Mexico
## 3
         0.070
                 Botswana
                                     3
## 2
         0.000
                   Belize
                                     2
# path plot
path.plot(synth_out2, dataprep_out2,
         Ylab = c("Foreign direct investment inflows per capita"),
         Xlab = c("year"),
         Legend = c("South Africa", "Synthetic SA"),
         Main = "Robustness check: Trends in fdi_per_capita:\nSA versus Synthetic SA",
         Ylim = c(-100, 400)
)
abline(v = 2006, lty = 2)
```

Robustness check: Trends in fdi_per_capita: SA versus Synthetic SA



Foreign direct investment inflows per capita Gaps between SA and Synthetic SA



```
# IN-TIME PLACEBO TEST: Choosing a fake treatment year (pre-treatment) 2001

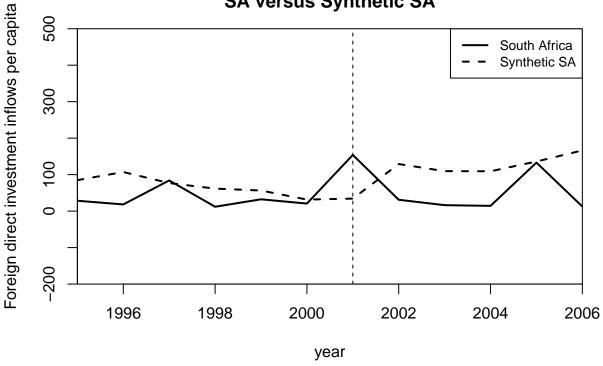
dataprep_out3 <- dataprep(
  foo = merged_data,
  predictors = c("gdp_cap", "trade_per_gdp", "pol_stab_est", "govt_educ", "internet"),
  # predictors.op = "mean",
  time.predictors.prior = 1995:2000,
  dependent = "fdi_bop_per_capita",
  unit.variable = "id",
  unit.names.variable = "country",
  time.variable = "year",
  treatment.identifier = 24,
  controls.identifier = c(2:4, 7:8, 10, 12:14, 16:20),
  time.optimize.ssr = 1995:2000, # pretreatment for dependent
  time.plot = 1995:2006
)

# Run Synth
synth_out3 <- synth(data.prep.obj = dataprep_out3)</pre>
```

In-time Placebo tests

```
searching for synthetic control unit
##
##
## ********
## *********
## ********
## MSPE (LOSS V): 2384.897
##
## solution.v:
## 3.56882e-05 0.02845532 0.2093683 0.7621325 8.1462e-06
##
## solution.w:
## 0.0001658042 0.3368163 0.000354362 0.0001141835 0.0002188454 0.0001000569 0.000136092 0 2.69441e-05
synth.tables <- synth.tab(</pre>
 dataprep.res = dataprep_out3,
  synth.res
               = synth_out3)
top_weights <- head(synth.tables$tab.w[order(synth.tables$tab.w$w, decreasing = TRUE), ], 5)
# Extract the top 5 weights
cat("\nRobustness test weights - taking out Cuba from Donor Pool:\n")
##
## Robustness test weights - taking out Cuba from Donor Pool:
print(top_weights)
      w.weights unit.names unit.numbers
         0.660
## 20
                     Peru
## 3
         0.337
                 Botswana
## 17
         0.002
                                    17
                   Mexico
## 2
         0.000
                   Belize
                                     2
## 4
         0.000 Costa Rica
# path plot
path.plot(synth_out3, dataprep_out3,
         Ylab = c("Foreign direct investment inflows per capita"),
         Xlab = c("year"),
         Legend = c("South Africa", "Synthetic SA"),
         Main = "Robustness check: Trends in fdi_per_capita:\nSA versus Synthetic SA",
         Ylim = c(-200, 500)
)
abline(v = 2001, lty = 2)
```

Robustness check: Trends in fdi_per_capita: SA versus Synthetic SA



fdi_per_capita Gap between SA and Synthetic SA

