Package 'Synth'

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Title Synthetic Control Group Method for Comparative Case Studies
Author
Jens Hainmueller < jhain@stanford.edu> and Alexis Diamond <adiamond@fas.harvard.edu></adiamond@fas.harvard.edu>
Maintainer Jens Hainmueller < jhain@stanford.edu>
Description Implements the synthetic control group method for comparative case studies as described in Abadie and Gardeazabal (2003) and Abadie, Diamond, and Hainmueller (2010, 2011, 2014). The synthetic control method allows for effect estimation in settings where a single unit (a state, country, firm, etc.) is exposed to an event or intervention. It provides a data-driven procedure to construct synthetic control units based on a weighted combination of comparison units that approximates the characteristics of the unit that is exposed to the intervention. A combination of comparison units often provides a better comparison for the unit exposed to the intervention than any comparison unit alone.
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basque Panel Data from Spanish Regions to demonstrate the use of the Synthetic Control Method

Description

The dataset contains information from 1955–1997 on 17 Spanish regions. It was used by Abadie and Gardeazabal (2003), which studied the economic effects of conflict, using the terrorist conflict in the Basque Country as a case study. This paper used a combination of other Spanish regions to construct a synthetic control region resembling many relevant economic characteristics of the Basque Country before the onset of political terrorism in the 1970s. The data contains per-capita GDP (the outcome variable), as well as population density, sectoral production, investment, and human capital (the predictor variables) for the relevant years, and is used here to demonstrate the implementation of the synthetic control method with the synth library.

Usage

data(basque)

Format

A panel dataframe made up of 18 units: 1 treated (no 17; the Basque country) and 16 control regions (no. 2-16,18). Region no. 1 is the average for the whole country of Spain. 1 outcome variable (gdpcap). 13 predictor variables (6 sectoral production shares, 6 highest educational attainment categories, population density, and the investment rate). Region names and numbers are stored in regionno and regionname. 42 time periods (1955 - 1997). All columns have self-explanatory column names. For reference the variables are:

- regionno
 - : Region Number.
- regionname
 - : Region Name.
- year
 - : Year.
- gdpcap
 - : real GDP per capita (in 1986 USD, thousands).
- sec.agriculture
 - : production in agriculture, forestry, and fishing sector as a percentage of total production.
- sec.energy
 - : production in energy and water sector as a percentage of total production.
- sec.industry
 - : production in industrial sector as a percentage of total production.

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- sec.construction
 - : production in construction and engineering sector as a percentage of total production.
- sec.energy
 - : production in marketable services sector as a percentage of total production.
- sec.energy
 - : production in Nonmarketable services sector as a percentage of total production.
- school.illit
 - : number of illiterate persons.
- school.prim
 - : number of persons with primary education or without studies.
- school.med
 - : number of persons with some high school education.
- school.high
 - : number of persons with high school degree.
- school.post.high
 - : number of persons with tertiary education.
- popdens
 - : population density (persons per square kilometer).
- invest
 - : gross total investment as a share of GDP.

Source

Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

Abadie, A., Diamond, A., Hainmueller, J. (2011). Synth: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

collect.optimx

Collect results from optimx optimization methods

Description

An internal function that collects the results from the different optimization methods run by optimx. It stores the parameter and function values and extracts the results for the best performing method (minimum or maximum).

Usage

```
collect.optimx(res, opt = "min")
```

Arguments

res Output from a call to optimx().

opt Either "min" or "max" to extract results for he methods that obtained the mini-

mum or maximum function value across the methods.

Value

out.list Dataframe with results from the different methods.

par Parameter values from method that attained minimum/maximum across the meth-

ods.

value Function value from method that attained minimum/maximum across the meth-

ods.

Author(s)

Jens Hainmueller

See Also

Also see optimx.

dataprep	Constructs a list of matrices from panel dataset to be loaded into
	synth()

Description

The synth function takes a standard panel dataset and produces a list of data objects necessary for running synth and other Synth package functions to construct synthetic control groups according to the methods outlined in Abadie and Gardeazabal (2003) and Abadie, Diamond, Hainmueller (2010, 2011, 2014) (see references and example).

User supplies a dataframe ("foo"), chooses predictors, special predictors (explained below), the operators that act upon these predictors, the dependent variable, identifies the columns associated with unit numbers, time periods (and unit names, when available), as well as the treated unit, the control units, the time-period over which to select the predictors, the time-period over which to optimize, and the time-period over which outcome data should be plotted.

The output of dataprep contains a list of matrices. This list object can be directly loaded into synth.

Usage

Arguments

foo The dataframe with the panel data.

predictors A vector of column numbers or column-name character strings that identifies

the predictors' columns. All predictors have to be numeric.

predictors.op A character string identifying the method (operator) to be used on the predictors.

Default is "mean". rm.na = T is hardwired into the code. See *Details*.

special.predictors

A list object identifying additional numeric predictors and their associated pretreatment years and operators (analogous to "predictors.op" above). See *De-

aument years and operators (analogous to predictor

tails*.

dependent A scalar identifying the column number or column-name character string that

corresponds to the numeric dependent (outcome) variable.

unit.variable A scalar identifying the column number or column-name character string asso-

ciated unit numbers. The unit varibale has to be numeric.

time.variable A scalar identifying column number or column-name character string associated

with period (time) data. The time variable has to be numeric.

treatment.identifier

A scalar identifying the "unit.variable" number or a character string giving the "unit.name "of the treated unit. If a character is supplied, a unit.names.variable

also has to be supplied to identify the treated unit.

controls.identifier

A scalar identifying the "unit.variable" numbers or a vector of character strings giving the "unit.name"s of control units. If a character is supplied, a unit.names.variable

also has to be supplied to identify the control units unit.

time.predictors.prior

A numeric vector identifying the pretreatment periods over which the values for the outcome predictors should be averaged.

time.optimize.ssr

A numeric vector identifying the periods of the dependent variable over which the loss function should be minimized (i.e. the periods over which mean squared prediction error (MSPE), that is the sum of squared residuals between treated

and the synthetic control unit, are minimized.

time.plot A vector identifying the periods over which results are to be plotted with gaps.plot

and path.plot.

unit.names.variable

A scalar or column-name character string identifying the column with the names of the units. This variable has to be of mode character.

Details

The predictors op argument is a character string that provides a function (eg., "mean", "median", etc.) identifying the name of the operator to be applied to the predictors over the given time period.

The special.predictors argument is a list object that contains one or more lists of length = 3. The required components of each of these lists are:

(a) scalar column number associated with that predictor (b) vector of time-period number(s) desired (eg., 1998:2003) (c) character-string identifying the name of the operation to be applied (ie., "mean", "median", etc.)

eg., special.predictors <- list(listc(x1, 1990:2000, "mean"), listc(x2, 1980:1983, "median"), listc(x3, 1980, "mean"))

indicates that predictor x1, should be used with its values averaged over periods 1990:2000; predicator x2 should be used with its median values over periods 1980:1983; x3 should be used with the values from period 1980 only.

Value

X1	matrix of treated predictor data. nrows = number of predictors and (possibly) special predictors. ncols = one.
X0	matrix of controls' predictor data. nrows = number of predictors and (possibly) special predictors. ncols = number of control units.
Z1	matrix of treated outcome data for the pre-treatment periods over which MSPE is to be minimized. nrows = number of pre-treatment periods. ncols = one.
Z0	matrix of controls' outcome data for the pre-treatment periods over which MSPE is to be minimized. nrows = number of pre-treatment periods. ncols = number of control units.
Y1plot	matrix of outcome data for treated unit to be used for results plotting. nrows = number of periods. ncols = one.
Y0plot	matrix of outcome data for control units to be used for results plotting. nrows = number of periods. ncols = number of control units.
names.and.numbers	
	dataframe with two columns showing all unit numbers and corresponding unit
	names.

Author(s)

tag

Jens Hainmueller and Alexis Diamond

References

Abadie, A., Diamond, A., Hainmueller, J. (2014). Comparative Politics and the Synthetic Control Method. *American Journal of Political Science* Forthcoming 2014.

a list of all arguments in initial function call.

Synthetic: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

Abadie, A., Diamond, A., Hainmueller, J. (2011). Synth: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

Abadie A, Diamond A, Hainmueller J (2010). Synthetic Control Methods for Comparative Case Studies: Estimating the Effect of California's Tobacco Control Program. *Journal of the American Statistical Association* 105 (490) 493–505.

Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

See Also

```
synth, gaps.plot, path.plot, synth.tab
```

Examples

```
## The usual sequence of commands is:
## 1. dataprep() for matrix-extraction
## 2. synth() for the construction of the synthetic control group
## 3. synth.tab(), gaps.plot(), and path.plot() to summarize the results
## Below we provide two examples.
## First Example: Toy panel dataset
# load data
data(synth.data)
# create matrices from panel data that provide inputs for synth()
dataprep.out<-
 dataprep(
  foo = synth.data,
  predictors = c("X1", "X2", "X3"),
  predictors.op = "mean",
  dependent = "Y",
  unit.variable = "unit.num",
  time.variable = "year",
   special.predictors = list(
     list("Y", 1991, "mean"),
     list("Y", 1985, "mean"),
     list("Y", 1980, "mean")
   treatment.identifier = 7,
  controls.identifier = c(29, 2, 13, 17, 32, 38),
   time.predictors.prior = c(1984:1989),
   time.optimize.ssr = c(1984:1990),
  unit.names.variable = "name",
   time.plot = 1984:1996
  )
## run the synth command to identify the weights
## that create the best possible synthetic
## control unit for the treated.
synth.out <- synth(dataprep.out)</pre>
## there are two ways to summarize the results
## we can either access the output from synth.out directly
round(synth.out$solution.w,2)
# contains the unit weights or
synth.out$solution.v
## contains the predictor weights.
## the output from synth opt
## can be flexibly combined with
```

```
## the output from dataprep to
## compute other quantities of interest
## for example, the period by period
## discrepancies between the
## treated unit and its synthetic control unit
## can be computed by typing
gaps<- dataprep.out$Y1plot-(</pre>
        dataprep.out$Y0plot%*%synth.out$solution.w
        ); gaps
## also there are three convenience functions to summarize results.
## to get summary tables for all information
## (V and W weights plus balance btw.
## treated and synthetic control) use the
## synth.tab() command
synth.tables <- synth.tab(</pre>
      dataprep.res = dataprep.out,
      synth.res = synth.out)
print(synth.tables)
## to get summary plots for outcome trajectories
## of the treated and the synthetic control unit use the
## path.plot() and the gaps.plot() commands
## plot in levels (treated and synthetic)
path.plot(dataprep.res = dataprep.out,synth.res = synth.out)
## plot the gaps (treated - synthetic)
gaps.plot(dataprep.res = dataprep.out,synth.res = synth.out)
## Second example: The economic impact of terrorism in the
## Basque country using data from Abadie and Gardeazabal (2003)
## see JSS paper in the references details
data(basque)
# dataprep: prepare data for synth
dataprep.out <-
 dataprep(
 foo = basque
  ,predictors= c("school.illit",
                 "school.prim",
                 "school.med",
                 "school.high",
                 "school.post.high"
                 ,"invest"
                 )
   ,predictors.op = c("mean")
   , dependent = c("gdpcap")
   ,unit.variable = c("regionno")
   ,time.variable = c("year")
```

```
,special.predictors = list(
   list("gdpcap",1960:1969,c("mean")),
   list("sec.agriculture", seq(1961,1969,2),c("mean")),
   list("sec.energy", seq(1961,1969,2),c("mean")),
   list("sec.industry", seq(1961, 1969, 2), c("mean")),
    list("sec.construction", seq(1961,1969,2),c("mean")),
    list("sec.services.venta", seq(1961, 1969, 2), c("mean")),
    list("sec.services.nonventa", seq(1961,1969,2),c("mean")),
    list("popdens",1969,c("mean")))
    ,treatment.identifier = 17
    ,controls.identifier = c(2:16,18)
    ,time.predictors.prior = c(1964:1969)
    time.optimize.ssr = c(1960:1969)
    ,unit.names.variable = c("regionname")
    ,time.plot
                          = c(1955:1997)
# 1. combine highest and second highest
# schooling category and eliminate highest category
dataprep.out$X1["school.high",] <-</pre>
dataprep.out$X1["school.high",] +
dataprep.out$X1["school.post.high",]
dataprep.out$X1
as.matrix(dataprep.out$X1[
 -which(rownames(dataprep.out$X1)=="school.post.high"),])
dataprep.out$X0["school.high",] <-</pre>
dataprep.out$X0["school.high",] +
dataprep.out$X0["school.post.high",]
dataprep.out$X0
dataprep.out$X0[
-which(rownames(dataprep.out$X0)=="school.post.high"),]
# 2. make total and compute shares for the schooling catgeories
lowest <- which(rownames(dataprep.out$X0)=="school.illit")</pre>
highest <- which(rownames(dataprep.out$X0)=="school.high")</pre>
dataprep.out$X1[lowest:highest,] <-</pre>
(100 * dataprep.out$X1[lowest:highest,]) /
sum(dataprep.out$X1[lowest:highest,])
dataprep.out$X0[lowest:highest,] <-</pre>
100 * scale(dataprep.out$X0[lowest:highest,],
             center=FALSE,
             scale=colSums(dataprep.out$X0[lowest:highest,])
# run synth
synth.out <- synth(data.prep.obj = dataprep.out)</pre>
# Get result tables
synth.tables <- synth.tab(</pre>
                           dataprep.res = dataprep.out,
                           synth.res = synth.out
```

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```
# results tables:
print(synth.tables)
# plot results:
# path
path.plot(synth.res = synth.out,
          dataprep.res = dataprep.out,
          Ylab = c("real per-capita GDP (1986 USD, thousand)"),
          Xlab = c("year"),
          Ylim = c(0,13),
          Legend = c("Basque country", "synthetic Basque country"),
## gaps
gaps.plot(synth.res = synth.out,
          dataprep.res = dataprep.out,
          Ylab = c("gap in real per-capita GDP (1986 USD, thousand)"),
          Xlab = c("year"),
          Ylim = c(-1.5, 1.5),
## To create the placebo studies simply reassign
## the intervention to other units or times (see references for details)
```

fn.V

Loss Function for nested optimization of W and V weights

Description

Loss function for the nested optimization of W and V weights used for constructing synthetic control groups according to the methods outlined in Abadie and Gardeazabal (2003) and Abadie, Diamond, Hainmueller (2010, 2011, 2014) (see references). This function is called by synth internally, and should not be called manually by a normal user.

Usage

```
fn.V(variables.v = stop("variables.v missing"),
X0.scaled = stop("X0.scaled missing"),
X1.scaled = stop("X1.scaled missing"),
Z0 = stop("Z0 missing"),
Z1 = stop("Z1 missing"),
margin.ipop = 5e-04,
sigf.ipop = 5,
bound.ipop = 10,
quadopt = "ipop")
```

fn. V

Arguments

variables.v	1 by k a vector of v weights.
X0.scaled	matrix of controls' predictor data. nrows = number of predictors and (possibly) special predictors. ncols = number of control units.
X1.scaled	matrix of treated predictor data. nrows = number of predictors and (possibly) special predictors. ncols = one.
Z0	matrix of controls' outcome data for the pre-treatment periods over which MSPE is to be minimized. nrows = number of pre-treatment periods. ncols = number of control units.
Z1	matrix of treated outcome data for the pre-treatment periods over which MSPE is to be minimized. nrows = number of pre-treatment periods. ncols = one.
margin.ipop	setting for ipop optimization routine: how close we get to the constrains (see ipop for details)
sigf.ipop	setting for ipop optimization routine: Precision (default: 7 significant figures (see ipop for details)
bound.ipop	setting for ipop optimization routine: Clipping bound for the variables (see ipop for details)
quadopt	string vector that specifies the routine for quadratic optimization over w weights. possible values are "ipop" (see ipop for details)

Value

A scalar that contains the function value.

Author(s)

Jens Hainmueller and Alexis Diamond

References

Abadie, A., Diamond, A., Hainmueller, J. (2014). Comparative Politics and the Synthetic Control Method. *American Journal of Political Science* Forthcoming 2014.

Synthetic: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

Abadie, A., Diamond, A., Hainmueller, J. (2011). Synth: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

Abadie A, Diamond A, Hainmueller J (2010). Synthetic Control Methods for Comparative Case Studies: Estimating the Effect of California's Tobacco Control Program. *Journal of the American Statistical Association* 105 (490) 493–505.

Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

See Also

```
synth, dataprep, gaps.plot, path.plot, synth.tab
```

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gaps.plot	Plots Gap in Outcome Trajectories between the Treated its Synthetic Control Unit

Description

This function plots the gaps in the trajectories of the outcome variable for the treated unit and the synthetic control group constructed by synth and dataprep. The user can specify whether the whole time period or only the pre-treatment period should be plotted.

Usage

Arguments

synth.res	Output list created by synth.
dataprep.res	Output list created by dataprep.
tr.intake	Optional scalar to indicate the time of treatment intake with a vertical line.
Ylab	Optional label for Y axis.
Xlab	Optional label for X axis.
Ylim	Optional Ylim.
Main	Optional main title.
Z.plot	Flag. If true, only pretreatment period is plotted.

Details

The trajectory of the outcome for the synthetic control group is calculated as: dataprep.res\$Y0plot %*% synth.res\$solution.w. You can use this calculation to construct custom made plots.

Value

The plot of trajectories.

Author(s)

Jens Hainmueller and Alexis Diamond

path.plot 13

References

Abadie, A., Diamond, A., Hainmueller, J. (2014). Comparative Politics and the Synthetic Control Method. *American Journal of Political Science* Forthcoming 2014.

Synthetic: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

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Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

See Also

```
dataprep, synth, path.plot, synth.tab
```

path.plot

Plots Outcome Trajectories for Treated Unit and its Synthetic Control Unit

Description

This function plots the trajectories of the outcome variable for the treated unit and the synthetic control group constructed by synth and dataprep. The user can specify whether the whole time period or only the pretreatment period should be plotted.

Usage

```
path.plot(synth.res = NA,
dataprep.res = NA,
tr.intake = NA,
Ylab = c("Y Axis"),
Xlab = c("Time"),
Ylim = NA,
Legend=c("Treated", "Synthetic"),
Legend.position=c("topright"),
Main = NA,
Z.plot = FALSE)
```

Arguments

synth.res Output list created by synth.

dataprep.res Output list created by dataprep.

tr.intake Optional scalar to indicate the time of treatment intake with a vertical line.

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Ylab Optional label for Y axis.
Xlab Optional label for X axis.

Ylim Optional Ylim.

Main Optional main title.

Legend Optional legend text (e.g. c("Treated", "Synthetic")); see ?legend for details.

Legend.position

Optional legend position (e.g. "bottomright"); see ?legend for details.

Z.plot Flag. If true, only pretreatment period is plotted.

Details

The trajectory of the outcome for the synthetic control group is calculated as: dataprep.res\$Y0plot%*% synth.res\$solution.w. You can use this calculation to construct custom made plots.

Value

The plot of trajectories.

Author(s)

Jens Hainmueller and Alexis Diamond

References

Abadie, A., Diamond, A., Hainmueller, J. (2014). Comparative Politics and the Synthetic Control Method. *American Journal of Political Science* Forthcoming 2014.

Synthetic: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

Abadie, A., Diamond, A., Hainmueller, J. (2011). Synth: An R Package for Synthetic Control Methods in Comparative Case Studies. *Journal of Statistical Software* 42 (13) 1–17.

Abadie A, Diamond A, Hainmueller J (2010). Synthetic Control Methods for Comparative Case Studies: Estimating the Effect of California's Tobacco Control Program. *Journal of the American Statistical Association* 105 (490) 493–505.

Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

See Also

dataprep, gaps.plot, synth, synth.tab

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spec.pred.func

Special Predictor Function for Dataprep

Description

This function is called by dataprep to handle special predictors in the process of setting up the dataset to be loaded into synth. It should not be called manually by the normal user.

Usage

Arguments

list.object	NA
tr.numb	NA
co.numb	NA
unit.var	NA
time.var	NA
foo.object	NA
X0.inner	NA
X1.inner	NA

Details

NA

Value

NA

Author(s)

Jens Hainmueller and Alexis Diamond

References

Abadie, A., Diamond, A., Hainmueller, J. (2014). Comparative Politics and the Synthetic Control Method. *American Journal of Political Science* Forthcoming 2014.

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Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

See Also

synth, dataprep, gaps.plot, path.plot, synth.tab

synth

Constructs synthetic control units for comparative case studies

Description

Implements the synthetic control method for causal inference in comparative case studies as developed in Abadie and Gardeazabal (2003) and Abadie, Diamond, Hainmueller (2010, 2011, 2014). synth estimates the effect of an intervention by comparing the evolution of an aggregate outcome for a unit affected by the intervention to the evolution of the same aggregate outcome for a synthetic control group.

synth constructs this synthetic control group by searching for a weighted combination of control units chosen to approximate the unit affected by the intervention in terms of characteristics that are predictive of the outcome. The evolution of the outcome for the resulting synthetic control group is an estimate of the counterfactual of what would have been observed for the affected unit in the absence of the intervention.

synth can also be used to conduct a variety of placebo and permutation tests that produce informative inference regardless of the number of available comparison units and the number of available time-periods. See Abadie and Gardeazabal (2003), Abadie, Diamond, and Hainmueller (2010, 2011, 2014) for details.

synth requires the user to supply four matrices as its main arguments. These matrices are named X0, X1, Z1, and Z0 accordingly. X1 and X0 contain the predictor values for the treated unit and the control units respectively. Z1 and Z0 contain the outcome variable for the pre-intervention period for the treated unit and the control unit respectively. The pre-intervention period refers to the time period prior to the intervention, over which the mean squared prediction error (MSPE) should be minimized. The MSPE refers to the squared deviations between the outcome for the treated unit and the synthetic control unit summed over all pre-intervention periods specified in Z1 and Z0.

Creating the matrices X1, X0, Z1, and Z0 from a (panel) dataset can be tedious. Therefore the Synth library offers a preparatory function called dataprep that allows the user to easily create all inputs required for synth. By first calling dataprep the user creates a single list object called data.prep.obj that contains all essential data elements to run synth.

Accordingly, a usual sequence of commands to implement the synthetic control method is to first call dataprep to prepare the data to be loaded into synth. Then synth is called to construct the synthetic control group. Finally, results are summarized using the functions synth.tab, path.plot, or gaps.plot.

An example of this sequence is provided in the documentation to dataprep. This procedure is strongly recommended. Alternatively, the user may provide his own preprocessed data matrices and load them into synth via the X0, X1, Z1, and Z0 arguments. In this case, no data.prep.obj should be specified.

The output from synth is a list object that contains the weights on predictors (solution.V) and weights on control units (solution.W) that define contributions to the synthetic control unit.

Usage

```
synth(data.prep.obj = NULL,
X1 = NULL, X0 = NULL,
Z0 = NULL, Z1 = NULL,
custom.v = NULL,
optimxmethod = c("Nelder-Mead", "BFGS"),
genoud = FALSE, quadopt = "ipop",
Margin.ipop = 5e-04,
Sigf.ipop = 5,
Bound.ipop = 10,
verbose = FALSE, ...)
```

Arguments

data.prep.obj	the object that comes from running dataprep. This object contains all information about X0, X1, Z1, and Z0. Therefore, if data.prep.obj is supplied, none of X0, X1, Z1, and Z0 should be manually specified!
X1	matrix of treated predictor data, nrows = number of predictors ncols = ones.
X0	matrix of controls' predictor data. nrows = number of predictors. ncols = number of control units (>=2).
Z1	matrix of treated outcome data for the pre-treatment periods over which MSPE is to be minimized. nrows = number of pre-treatment periods. ncols = 1.
Z0	matrix of controls' outcome data for the pre-treatment periods over which MSPE is to be minimized. nrows = number of pre-treatment periods. ncols = number of control units.
custom.v	vector of weights for predictors supplied by the user. uses synth to bypass optimization for solution.V. See details.
optimxmethod	string vector that specifies the optimization algorithms to be used. Permissable values are all optimization algorithms that are currently implemented in the optimx function (see this function for details). This list currently includes

c("Nelder-Mead', 'BFGS', 'CG', 'L-BFGS-B', 'nlm', 'nlminb', 'spg', and 'ucminf"). If multiple algorithms are specified, synth will run the optimization with all chosen algorithms and then return the result for the best performing method. Default is c("Nelder-Mead", "BFGS"). As an additional possibility, the user can also specify 'All' which means that synth will run the results over all algorithms in optimx.

genoud Logical flag. If true, synth embarks on a two step optimization. In the first

step, genoud, an optimization function that combines evolutionary algorithm methods with a derivative-based (quasi-Newton) method to solve difficult optimization problems, is used to obtain a solution. See genoud for details. In the second step, the genoud results are passed to the optimization algorithm(s) chosen in optimxmethod for a local optimization within the neighborhood of the genoud solution. This two step optimization procedure will require much more computing time, but may yield lower loss in cases where the search space

is highly irregular.

quadopt string vector that specifies the routine for quadratic optimization over w weights.

possible values are "ipop" (see ipop for details).

Margin.ipop setting for ipop optimization routine: how close we get to the constrains (see

ipop for details)

Sigf.ipop setting for ipop optimization routine: Precision (default: 7 significant figures

(see ipop for details)

Bound ipop setting for ipop optimization routine: Clipping bound for the variables (see ipop

for details)

verbose Logical flag. If TRUE then intermediate results will be shown.

... Additional arguments to be passed to optimx and or genoud to adjust optimiza-

tion.

Details

As proposed in Abadie and Gardeazabal (2003) and Abadie, Diamond, Hainmueller (2010), the synth function routinely searches for the set of weights that generate the best fitting convex combination of the control units. In other words, the predictor weight matrix V is chosen among all positive definite diagonal matrices such that MSPE is minimized for the pre-intervention period.

Instead of using this data-driven procedures to search for the best fitting synthetic control group, the user may supply his own vector of V weights, based on his subjective assessment of the predictive power of the variables in X1 and X0. In this case, the vector of V weights for each variable should be supplied via the custom. V option in synth and the optimization over the V matrices is bypassed.

Value

solution.v vector of predictor weights.

solution.w vector of weights across the controls.

loss.v MSPE from optimization over v and w weights.

loss.w Loss from optimization over w weights.

custom. v if this argument was specified in the call to synth, this outputs the weight vector

specified.

rgV.optim Results from optimx() minimization. Could be used for diagnostics.

Author(s)

Jens Hainmueller and Alexis Diamond

References

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Abadie, A. and Gardeazabal, J. (2003) Economic Costs of Conflict: A Case Study of the Basque Country *American Economic Review* 93 (1) 113–132.

See Also

```
dataprep, gaps.plot, path.plot, synth.tab
```

Examples

```
## While synth() can be used to construct synthetic control groups
## directly, by providing the X1, X0, Z1, and Z0 matrices, we strongly
## recommend to first run dataprep() to extract these matrices
## and pass them to synth() as a single object
## The usual sequence of commands is:
## 1. dataprep() for matrix-extraction
## 2. synth() for the construction of the synthetic control group
## 3. synth.tab(), gaps.plot(), and path.plot() to summarize the results
## Below we provide two examples
## First Example: Toy panel dataset
# load data
data(synth.data)
# create matrices from panel data that provide inputs for synth()
dataprep.out<-
 dataprep(
  foo = synth.data,
  predictors = c("X1", "X2", "X3"),
  predictors.op = "mean",
  dependent = "Y",
  unit.variable = "unit.num",
   time.variable = "year",
   special.predictors = list(
```

```
list("Y", 1991, "mean"),
      list("Y", 1985, "mean"),
      list("Y", 1980, "mean")
   treatment.identifier = 7,
  controls.identifier = c(29, 2, 13, 17, 32, 38),
   time.predictors.prior = c(1984:1989),
   time.optimize.ssr = c(1984:1990),
  unit.names.variable = "name",
  time.plot = 1984:1996
  )
## run the synth command to identify the weights
## that create the best possible synthetic
## control unit for the treated.
synth.out <- synth(dataprep.out)</pre>
## there are two ways to summarize the results
## we can either access the output from synth.out directly
round(synth.out$solution.w,2)
# contains the unit weights or
synth.out$solution.v
## contains the predictor weights.
## the output from synth opt
## can be flexibly combined with
## the output from dataprep to
## compute other quantities of interest
## for example, the period by period
## discrepancies between the
## treated unit and its synthetic control unit
## can be computed by typing
gaps<- dataprep.out$Y1plot-(</pre>
        dataprep.out$Y0plot%*%synth.out$solution.w
        ); gaps
## also there are three convenience functions to summarize results.
## to get summary tables for all information
## (V and W weights plus balance btw.
## treated and synthetic control) use the
## synth.tab() command
synth.tables <- synth.tab(</pre>
      dataprep.res = dataprep.out,
      synth.res = synth.out)
print(synth.tables)
## to get summary plots for outcome trajectories
## of the treated and the synthetic control unit use the
## path.plot() and the gaps.plot() commands
## plot in levels (treated and synthetic)
path.plot(dataprep.res = dataprep.out,synth.res = synth.out)
```

```
## plot the gaps (treated - synthetic)
gaps.plot(dataprep.res = dataprep.out,synth.res = synth.out)
## Second example: The economic impact of terrorism in the
## Basque country using data from Abadie and Gardeazabal (2003)
## see JSS paper in the references details
data(basque)
# dataprep: prepare data for synth
dataprep.out <-
 dataprep(
 foo = basque
  ,predictors= c("school.illit",
                 "school.prim",
                 "school.med",
                 "school.high",
                 "school.post.high"
                 ,"invest"
                 )
   ,predictors.op = c("mean")
   , dependent = c("gdpcap")
   ,unit.variable = c("regionno")
   ,time.variable = c("year")
   ,special.predictors = list(
   list("gdpcap",1960:1969,c("mean")),
   list("sec.agriculture", seq(1961,1969,2),c("mean")),
   list("sec.energy", seq(1961,1969,2),c("mean")),
   list("sec.industry", seq(1961,1969,2),c("mean")),
   list("sec.construction", seq(1961, 1969, 2), c("mean")),
   list("sec.services.venta", seq(1961, 1969, 2), c("mean")),
    list("sec.services.nonventa", seq(1961,1969,2),c("mean")),
   list("popdens",1969,c("mean")))
    ,treatment.identifier = 17
    ,controls.identifier = c(2:16,18)
    ,time.predictors.prior = c(1964:1969)
    , time.optimize.ssr = c(1960:1969)
    ,unit.names.variable = c("regionname")
    ,time.plot
                          = c(1955:1997)
    )
# 1. combine highest and second highest
# schooling category and eliminate highest category
dataprep.out$X1["school.high",] <-</pre>
dataprep.out$X1["school.high",] +
dataprep.out$X1["school.post.high",]
dataprep.out$X1
                                <-
 as.matrix(dataprep.out$X1[
 -which(rownames(dataprep.out$X1)=="school.post.high"),])
dataprep.out$X0["school.high",] <-</pre>
 dataprep.out$X0["school.high",] +
```

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```
dataprep.out$X0["school.post.high",]
dataprep.out$X0
dataprep.out$X0[
-which(rownames(dataprep.out$X0)=="school.post.high"),]
# 2. make total and compute shares for the schooling catgeories
lowest <- which(rownames(dataprep.out$X0)=="school.illit")</pre>
highest <- which(rownames(dataprep.out$X0)=="school.high")</pre>
dataprep.out$X1[lowest:highest,] <-</pre>
 (100 * dataprep.out$X1[lowest:highest,]) /
 sum(dataprep.out$X1[lowest:highest,])
dataprep.out$X0[lowest:highest,] <-</pre>
 100 * scale(dataprep.out$X0[lowest:highest,],
             center=FALSE,
             scale=colSums(dataprep.out$X0[lowest:highest,])
# run synth
synth.out <- synth(data.prep.obj = dataprep.out)</pre>
# Get result tables
synth.tables <- synth.tab(</pre>
                           dataprep.res = dataprep.out,
                           synth.res = synth.out
# results tables:
print(synth.tables)
# plot results:
# path
path.plot(synth.res = synth.out,
          dataprep.res = dataprep.out,
          Ylab = c("real per-capita GDP (1986 USD, thousand)"),
          Xlab = c("year"),
          Ylim = c(0,13),
          Legend = c("Basque country", "synthetic Basque country"),
## gaps
gaps.plot(synth.res = synth.out,
          dataprep.res = dataprep.out,
          Ylab = c("gap in real per-capita GDP (1986 USD, thousand)"),
          Xlab = c("year"),
          Ylim = c(-1.5, 1.5),
          )
```

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Description

This artificial panel data set is used to demonstrate the use of the Synthetic Control Method.

Usage

```
data(synth.data)
```

Format

A dataframe made up of 8 units: 1 treated (no 7) and 7 control (no. 2,7,13,17,29,32,36,38), 3 predictors (X1, X2, X3), 21 time periods (1980 - 2000), a unit.names.variable column ("names") and an outcome variable column (Y). All columns have column names.

synth.tab Creates Tables that Summarize Results of Synthetic Control Group Method	ир
---	----

Description

This function is called after dataprep and synth in order to create tables summarizing the results of the run of the synthetic control method. The result tables can be latexed directly.

Usage

Arguments

synth.res The list resulting from the call to synth.

dataprep.res The list resulting from the call to dataprep.

round.digit Integer for rounding in tables.

Details

NA

Value

tab.v	The matrix that contains the table of V-weights and respective variable names.
tab.w	The matrix that contains the table of W-weights and respective unit numbers and possibly names.
tab.loss	The matrix that contains the table of W-loss and V-loss

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Author(s)

Jens Hainmueller and Alexis Dimaond

References

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See Also

synth, dataprep, gaps.plot, path.plot

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