

#### <u>Title</u>

scest — Estimation for Synthetic Control Methods.

#### Syntax

scest , dfname(string) [p(#) direc(string) Q(#) lb(#) name(string) opt(string)]

## Description

Companion R and  $\underline{\text{Python}}$  packages are described in  $\underline{\text{Cattaneo, Feng, Palomba and Titiunik (2022)}}$ .

Companion commands are:  $\underline{scdata}$  for data preparation,  $\underline{scpi}$  for inference procedures, and  $\underline{scplot}$  for SC plots.

Related Stata, R, and Python packages useful for inference in SC designs are described in the following website:

https://nppackages.github.io/scpi/

For an introduction to synthetic control methods, see <a href="Abadie (2021)"><u>Abadie (2021)</u></a> and references therein.

# Options

dfname(string) specifies the name of the Python object containing the processed
 data created with <u>scdata</u>.

\_\_\_\_ Constraint

These options let the user specify the type of constraint to be imposed to estimate the SC weights. The user controls the lower bound on the weights (option  $\mathbf{lb}$ ), the norm of the weights to be constrained (option  $\mathbf{p}$ ), the direction of the constraint on the norm (option  $\mathbf{dir}$ ), and the size of the constraint on the norm (option  $\mathbf{q}$ ). Alternatively, some popular constraints can be selected through the option  $\mathbf{name}$ . A detailed description of the popular constraints implemented can be found in Cattaneo, Feng, Palomba and Titiunik (2022).

- 1b(#) specifies the lower bound on the weights. The default is 1b(0).
- p(#) sets the type of norm to be constrained. Options are:
  - ${f 0}$  no constraint on the norm of the weights is imposed.  ${f 1}$  a constraint is imposed on the L1 norm of the weights (the default).
  - 2 a constraint is imposed on the L2 norm of the weights.
- direc(string) specifies the direction of the constraint on the norm of the
   weights. Options are:
  - <= the constraint on the norm of the weights is an inequality constraint.
  - == the constraint on the norm of the weights is an equality constraint (the default).
- $\mathbf{Q}$  (#) specifies the size of the constraint on the norm of the weights.
- name(string) specifies the name of the constraint to be used. Options are:
   simplex classic synthetic control estimator where the weights are constrained
   to be non-negative and their L1 norm must be equal to 1.
  - lasso weights are estimated using a Lasso-type penalization
  - ridge weights are estimated using a Ridge-type penalization.
  - ols weights are estimated without constraints using least squares

☐ Others

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\begin{array}{c} \textbf{opt}\,(string) \text{ a string specifying the stopping criteria used by the underling} \\ \text{optimizer } (\underline{nlopt}) \text{ for point estimation. The default is a sequential quadratic} \\ \text{programming } (SQP) \text{ algorithm for nonlinearly constrained gradient-based} \end{array}
         optimization ('SLSQP'). The default value is opt("'maxeval' = 5000, 'xtol_rel' = 1e-8, 'xtol_abs' = 1e-8, 'ftol_rel' = 1e-12, 'ftol_abs' = 1e-12, 'tol_eq' = 1e-8, 'tol_ineq' = 1e-8"). In case a lasso-type constraint is
          implemented, a different optimizer (\underline{\text{cvxpy}}) is used and stopping criteria
          cannot be changed.
Example: Germany Data
     Setup
          . use scpi_germany.dta
     Prepare data
           . scdata gdp, dfname("python_scdata") id(country) outcome(gdp) time(year)
          treatment(status) cointegrated
     Estimate Synthetic Control with a simplex constraint
          . scest, dfname("python_scdata") name(simplex)
Stored results
     scest stores the following in e():
     Scalars
       e (M)
                                       number of features
                                       number of covariates used for adjustment
       e (KM)
                                       number of donors
       e (J)
       e(T1)
                                       number of post-treatment periods
       e (q)
                                       size of the constraint on the norm
     Macros
       e(features)
                                      name of features
       e(outcomevar)
                                       \hbox{\tt name of outcome variable}
       e(constant)
                                       logical indicating the presence of a common constant
                                         across features
       e(cointegrated_data)
                                       logical indicating cointegration
                                       type of norm of the weights used in constrained
       e(p)
                                          estimation
       e(dir)
                                       direction of the constraint on the norm of the
                                         weights
       e (name)
                                       name of constraint used in estimation
     Matrices
       e(T0)
                                       number of pre-treatment periods per feature
       e (A)
                                       pre-treatment features of the treated unit
       e (B)
                                       pre-treatment features of the control units
                                       covariates used for adjustment
       e (C)
                                       predicted values of the features of the treated unit
       e (pred)
       e(res)
                                       residuals e(A) - e(pred)
       e (w)
                                       weights of the controls
                                       coefficients of the covariates used for adjustment
       e(r)
       e (beta)
                                      stacked version of e(w) and e(r)
       e(Y_post)
                                      post-treatment outcome of the treated unit
       e(Y_post_fit)
                                       estimated post-treatment outcome of the treated unit
       e(Y_pre)
                                      pre-treatment outcome of the treated unit
```

estimate pre-treatment outcome of the treated unit

e(Y\_pre\_fit)

- Abadie, A. 2021. <u>Using synthetic controls: Feasibility, data requirements, and methodological aspects.</u> Journal of Economic Literature, 59(2), 391-425.
- Cattaneo, M. D., Feng, Y., and Titiunik, R. 2021. <u>Prediction Intervals for Synthetic Sontrol Methods.</u> Journal of the American Statistical Association, 116(536), 1865-1880.
- Cattaneo, M. D., Feng, Y., Palomba F., and Titiunik, R. 2022. <a href="mailto:scopi: Uncertainty Quantification for Synthetic Control Estimators, arXiv:2202.05984">scpi: Uncertainty Quantification for Synthetic Control Estimators, arXiv:2202.05984</a>.

### <u>Authors</u>

Matias D. Cattaneo, Princeton University, Princeton, NJ. <a href="mailto:cattaneo@princeton.edu">cattaneo@princeton.edu</a>. Yingjie Feng, Tsinghua University, Beijing, China. <a href="mailto:fengyj@sem.tsinghua.edu.cn">fengyj@sem.tsinghua.edu.cn</a>. Filippo Palomba, Princeton University, Princeton, NJ. <a href="mailto:fpalomba@princeton.edu">fpalomba@princeton.edu</a>. Rocio Titiunik, Princeton University, Princeton, NJ. <a href="mailto:titiunik@princeton.edu">titiunik@princeton.edu</a>.