sreg: Stratified Randomized Experiments



The sreg package for R, offers a toolkit for estimating average treatment effects (ATEs) in stratified randomized experiments. It supports a wide range of stratification designs, including matched pairs, k-tuple designs, and larger strata with many units — possibly of unequal size across strata. The package is designed to accommodate scenarios with multiple treatments, and cluster-level treatment assignments, and accommodates optimal linear covariate adjustment based on baseline observable characteristics. The package computes estimators and standard errors based on Bugni, Canay, Shaikh (2018); Bugni, Canay, Shaikh, Tabord-Meehan (2023); Jiang, Linton, Tang, Zhang (2023); Bai, Jiang, Romano, Shaikh, Zhang (2024), Liu (2024); and Cytrynbaum (2024).

```
Dependencies: dplyr, tidyr, extraDistr, rlang
Suggests: haven, knitr, rmarkdown, testthat (>= 3.0.0)
R version required: >= 2.10
Latest Build (v.2.0.0.9001)
```

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Supplementary files

- PDF version of the manual: Download PDF
- Sketch of the derivation of the ATE variance estimator under cluster-level treatment assignment: Download PDF
- Expressions for the multiple treatment case (with and without clusters): Download PDF

Installation

```
*** installing help indices
** building package indices
** installing vignettes
** testing if installed package can be loaded from temporary location
** testing if installed package can be loaded from final location
** testing if installed package keeps a record of temporary installation path
* DONE (srea)
The downloaded source packages are in
        '/private/var/folders/mp/06gjwr8j56zdp5j2vgdkd4z40000gq/T/RtmpVk96vN/downloaded_packages'
library(sreg)
                              Stratified Randomized
#> \___ \| |_) | _|| |
#> ____) | _ <| |__| |
#> |____/ | \_\_\__\ version 1.0.1
#> Type 'citation("sreg")' for citing this R package in publications.
The latest development version can be installed using devtools.
library(devtools)
install_github("jutrifonov/sreg")
Downloading GitHub repo jutrifonov/sreg@HEAD
- R CMD build -
checking for file '/private/var/folders/mp/06gjwr8j56zdp5j2vgdkd4z40000gq/T/RtmpVk96vN/remotes1026130d7f9b4/jut
- preparing 'sreg':
✓ checking DESCRIPTION meta-information ...

    checking for LF line-endings in source and make files and shell scripts

    checking for empty or unneeded directories

    building 'sreg 2.0.0.9000.tar.gz'

Installing package into '/opt/homebrew/lib/R/4.4/site-library'
(as 'lib' is unspecified)
* installing *source* package 'sreg' ...
** using staged installation
** R
** data
*** moving datasets to lazyload DB
** byte-compile and prepare package for lazy loading
*** installing help indices
** building package indices
** installing vignettes
** testing if installed package can be loaded from temporary location
** testing if installed package can be loaded from final location
** testing if installed package keeps a record of temporary installation path
* DONE (sreg)
library(sreg)

  □ Loading sreg

                             Stratified Randomized
```

#> Type 'citation("sreg")' for citing this R package in publications.

The function sreg()

Estimates the ATE(s) and the corresponding standard error(s) for a (collection of) treatment(s) relative to a control.

Syntax

```
sreg(Y, S = NULL, D, G.id = NULL, Ng = NULL, X = NULL, HC1 = TRUE, small.strata = FALSE)
```

Arguments

- Y a numeric vector/matrix/data.frame/tibble of the observed outcomes;
- **S** a numeric vector/matrix/data.frame/tibble of strata indicators 0, 1, 2, ...; if NULL then the estimation is performed assuming no stratification;
- **D** a numeric vector/matrix/data.frame/tibble of treatments indexed by $0, 1, 2, \ldots$, where **D** = **0** denotes the control;
- **G.id** a numeric vector/matrix/data.frame/tibble of cluster indicators; if NULL then estimation is performed assuming treatment is assigned at the individual level;
- Ng a numeric vector/matrix/data.frame/tibble of cluster sizes; if NULL then Ng is assumed to be equal to the number of available observations in every cluster;
- X a matrix/data.frame/tibble with columns representing the covariate values for every observation; if NULL then the estimator without linear adjustments is applied [^*];
- **HC1** a TRUE/FALSE logical argument indicating whether the small sample correction should be applied to the variance estimator;
- small.strata a TRUE/FALSE logical argument indicating whether the estimators for small strata (i.e., strata with few units, such as matched pairs or n-tuples) should be used. [^*]: Note: sreg cannot use individual-level covariates for covariate adjustment in cluster-randomized experiments. Any individual-level covariates will be aggregated to their cluster-level averages.

Data Structure

Here we provide an example of a data frame that can be used with sreg.

| Υ | S | D | G.id | Ng | x_1 | x_2 | |
|-------------|---|---|------|----|-----------|-------------|--|
| | | | | | | | |
| -0.57773576 | 2 | 0 | 1 | 10 | 1.5597899 | 0.03023334 | |
| 1.69495638 | 2 | 0 | 1 | 10 | 1.5597899 | 0.03023334 | |
| 2.02033740 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |
| 1.22020493 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |
| 1.64466086 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |
| -0.32365109 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |
| 2.21008191 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |
| -2.25064316 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |
| 0.37962312 | 4 | 2 | 2 | 30 | 0.8747419 | -0.77090031 | |

Summary

sreg prints a "Stata-style" table containing the ATE estimates, corresponding standard errors, t-statistics, p-values, 95% asymptotic confidence intervals, and significance indicators for different levels α . The example of the printed output is provided below.

```
Saturated Model Estimation Results under CAR
Observations: 2710
Clusters: 100
Number of treatments: 2
Number of strata: 10
Setup: big strata
Standard errors: adjusted (HC1)
Treatment assignment: cluster level
Covariates used in linear adjustments:
_ _ _
Coefficients:
     Tau As.se T-stat P-value CI.left(95%) CI.right(95%) Significance
1 1.13687 0.31181 3.64608 0.00027
                                       0.52574
                                                     1.74799
2 0.66447 0.30263 2.19565 0.02812
                                       0.07133
                                                     1.25761
Signif. codes: 0 `***` 0.001 `**` 0.01 `*` 0.05 `.` 0.1 ` ` 1
```

Return Value

The function returns an object of class sreg that is a list containing the following elements:

- tau.hat a $1 \times |\mathcal{A}|$ vector of ATE estimates, where $|\mathcal{A}|$ represents the number of treatments;
- se.rob a $1 \times |\mathcal{A}|$ vector of standard errors estimates, where $|\mathcal{A}|$ represents the number of treatments;
- **t.stat** a $1 \times |\mathcal{A}|$ vector of t-statistics, where $|\mathcal{A}|$ represents the number of treatments;
- p.value a $1 \times |\mathcal{A}|$ vector of corresponding p-values, where $|\mathcal{A}|$ represents the number of treatments;
- CI.left a $1 \times |\mathcal{A}|$ vector of the left bounds of the 95% as. confidence interval;
- CI.right a $1 \times |\mathcal{A}|$ vector of the right bounds of the 95% as. confidence interval;
- data an original data of the form data.frame(Y, S, D, G.id, Ng, X);
- lin.adj a data.frame representing the covariates that were used in implementing linear adjustments;
- **small.strata** a TRUE/FALSE logical argument indicating whether the estimators for small strata (e.g., matched pairs or n-tuples) were used;
- HC1 a TRUE/FALSE logical argument indicating whether the small sample correction (HC1) was applied to the variance estimator.

Empirical Example

Here, we provide the empirical application example using the data from (Chong et al., 2016), who studied the effect of iron deficiency anemia on school-age children's educational attainment and cognitive ability in Peru. The example replicates the empirical illustration from (Bugni et al., 2019). For replication purposes, the data is included in the package and can be accessed by running data("AEJapp"). This example can be accessed directly in R via help(sreg).

```
library(sreg, dplyr, haven)
```

The description of the dataset can be accessed using help():

```
help(AEJapp)
```

```
We can upload the AEJapp dataset to the R session via data():
data("AEJapp")
data <- AEJapp
It is pretty straightforward to prepare the data to fit the package syntax using dplyr:
Y <- data$gradesq34
D <- data$treatment
S <- data$class_level
data.clean <- data.frame(Y, D, S)</pre>
data.clean <- data.clean %>%
 mutate(D = ifelse(D == 3, 0, D))
Y <- data.clean$Y
D <- data.clean$D
S <- data.clean$S
head(data.clean)
    Y D S
1 11.2 1 1
2 12.4 0 3
3 11.9 0 5
4 13.1 0 1
5 13.4 2 2
6 10.7 0 1
We can take a look at the frequency table of D and S:
table(D = data.clean$D, S = data.clean$S)
  S
D 1 2 3 4 5
 0 15 19 16 12 10
 1 16 19 15 10 10
 2 17 20 15 11 10
Now, it is straightforward to replicate the results from (Bugni et al, 2019) using sreg:
result <- sreg::sreg(Y = Y, S = S, D = D)
print(result)
Saturated Model Estimation Results under CAR
Observations: 215
Number of treatments: 2
Number of strata: 5
Setup: big strata
Standard errors: adjusted (HC1)
Treatment assignment: individual level
Covariates used in linear adjustments:
Coefficients:
       Tau As.se T-stat P-value CI.left(95%) CI.right(95%) Significance
1 -0.05113 0.20645 -0.24766 0.80440
                                         -0.45577
                                                        0.35351
2 0.40903 0.20651 1.98065 0.04763
                                          0.00427
                                                        0.81379
Signif. codes: 0 `***` 0.001 `**` 0.01 `*` 0.05 `.` 0.1 ` ` 1
Besides that, sreg allows adding linear adjustments (covariates) to the estimation procedure:
pills <- data$pills_taken</pre>
age <- data$age months
```

```
data.clean <- data.frame(Y, D, S, pills, age)</pre>
data.clean <- data.clean %>%
 mutate(D = ifelse(D == 3, 0, D))
Y <- data.clean$Y
D <- data.clean$D
S <- data.clean$S
X <- data.frame("pills" = data.clean$pills, "age" = data.clean$age)</pre>
result <- sreg::sreg(Y, S, D, G.id = NULL, X = X)
print(result)
Saturated Model Estimation Results under CAR with linear adjustments
Observations: 215
Number of treatments: 2
Number of strata: 5
Setup: big strata
Standard errors: adjusted (HC1)
Treatment assignment: individual level
Covariates used in linear adjustments: pills, age
Coefficients:
       Tau As.se T-stat P-value CI.left(95%) CI.right(95%) Significance
1 -0.02862 0.17964 -0.15929 0.87344
                                        -0.38071
                                                       0.32348
2 0.34609 0.18362 1.88477 0.05946
                                        -0.01381
                                                       0.70598
Signif, codes: 0 `*** 0.001 `** 0.01 `* 0.05 `. 0.1 ` 1
```

Example (small strata)

Treatment assignment: individual level

Coefficients:

Covariates used **in** linear adjustments: x_1, x_2

Beginning with version 2.0.0+, the sreg package supports experimental designs with small strata (e.g., matched pairs or k-tuples) via the small strata argument in the sreg function. We demonstrate its implementation using simulated data generated by sreg.rgen() under a matched triplets design.

```
> head(data)
                                               Y S D
                                                                                              x_1
1 2.6455170 1 0 5.594675 1.9023835
2 6.6589024 1 2 6.450984 4.2343208
3 4.3412644 1 1 4.787852 3.1895694
4 -0.7592291 2 2 6.240883 0.7458935
5 5.1391241 2 1 6.076305 2.6105942
6 2.3934378 2 0 5.403182 3.4032419
result <- sreg(Y = data\$Y, S = data\$S, D = data\$D, X = data.frame('x_1' = data\$x_1, 'x_2' = data\$x_2), small.stratering | sma
> print(result)
Saturated Model Estimation Results under CAR with linear adjustments
Observations: 300
Number of treatments: 2
Number of strata: 100
Setup: small strata
Strata size (k): 3
Standard errors: adjusted (HC1)
```

data <- sreg.rgen(n = 300, tau.vec = c(1.2, 0.8), cluster = FALSE, small.strata = TRUE, k = 3, treat.sizes = c(1, 0.8)

```
Tau As.se T-stat P-value CI.left(95%) CI.right(95%) Significance
1 1.11577 0.13995 7.97258 0e+00 0.84147 1.39006 ***
2 0.58806 0.13439 4.37574 1e-05 0.32466 0.85147 ***

---
Signif. codes: 0 `***` 0.001 `**` 0.01 `*` 0.05 `.` 0.1 `` 1
```

The function plot.sreg()

Visualizes the estimated average treatment effects (ATEs) and their confidence intervals from an object returned by sreg(). This function defines an S3 method for the generic plot() function for objects of class sreg.

Syntax

```
plot(x,
```

```
treatment labels = NULL,
title = "Estimated ATEs with Confidence Intervals",
bar fill = NULL,
point shape = 23,
point size = 3,
point fill = "white",
point stroke = 1.2,
point color = "black",
label color = "black",
label size = 4,
bg color = NULL,
grid = TRUE,
zero_line = TRUE,
y axis title = NULL,
x_axis_title = NULL,
...)
```

Arguments

- x an object of class sreg, returned by the sreg() function;
- treatment_labels an optional vector of labels to display on the y-axis; if NULL, defaults to "Treatment 1", "Treatment 2", etc.;
- title an optional string specifying the plot title; default is "Estimated ATEs with Confidence Intervals";
- **bar_fill** an optional color specification for the confidence interval bars; can be **NULL** (default viridis scale), a single color, or a vector of two colors for a gradient;
- **point_shape** an **integer** specifying the shape of the point representing the estimated ATE; default is 23 (diamond);
- point size a numeric value specifying the size of the ATE point;
- point_fill a string indicating the fill color of the ATE point shape;
- point_stroke a numeric value for the stroke (border thickness) of the ATE point shape;
- point color a string specifying the outline color of the ATE point;
- label_color a string indicating the color of the text label displaying the estimate and standard error;
- label_size a numeric value for the size of the text label;
- **bg_color** an optional **string** specifying the background color of the plot panel; if **NULL**, the default theme background is used;
- grid a TRUE/FALSE argument indicating whether grid lines should be displayed (TRUE by default);
- **zero_line** a TRUE/FALSE argument indicating whether to include a dashed vertical line at 0 (TRUE by default);
- y axis title an optional string specifying the y-axis title; if NULL, no title is displayed;

- x axis title an optional string specifying the x-axis title; if NULL, no title is displayed;
- ... additional arguments passed to other methods (not used in this method).

Return Value

Invisibly returns the ggplot object used to generate the figure. The function is called primarily for its side effect — rendering the plot. ### Example

```
library("sreg")
library("dplyr")
library("haven")
data <- sreg.rgen(n = 1000, tau.vec = c(-0.3, 0.2), n.strata = 4, cluster = FALSE)
Y <- data$Y
S <- data$S
D <- data$D
X <- data.frame("x_1" = data$x_1, "x_2" = data$x_2)
result <- sreg(Y, S, D, G.id = NULL, Ng = NULL, X)
plot(result)</pre>
```

The function sreg.rgen()

Generates the observed outcomes, treatment assignments, strata indicators, cluster indicators, cluster sizes, and covariates for estimating the treatment effect following the stratified block randomization design under covariate-adaptive randomization (CAR).

Syntax

Arguments

- **n** a total number of observations in a sample;
- Nmax a maximum size of generated clusters (maximum number of observations in a cluster);
- n.strata an integer specifying the number of strata;
- tau.vec a numeric $1 \times |\mathcal{A}|$ vector of treatment effects, where $|\mathcal{A}|$ represents the number of treatments;
- gamma.vec a numeric 1×3 vector of parameters corresponding to covariates;
- **cluster** a TRUE/FALSE argument indicating whether the dgp should use a cluster-level treatment assignment or individual-level;
- is.cov a TRUE/FALSE argument indicating whether the dgp should include covariates or not;
- **small.strata** a TRUE/FALSE argument indicating whether the data-generating process should use a small-strata design (e.g., matched pairs, *n*-tuples);
- k an integer specifying the number of units per stratum when small.strata = TRUE;
- treat.sizes a numeric $1 \times (|\mathcal{A}| + 1)$ vector specifying the number of units assigned to each treatment within a stratum; the first element corresponds to control units (D = 0), the second to the first treatment (D = 1), and so on.

Return Value

- Y a numeric $n \times 1$ vector of the observed outcomes;
- **S** a numeric $n \times 1$ vector of strata indicators:
- **D** a numeric $n \times 1$ vector of treatments indexed by $0, 1, 2, \dots$, where **D** = **0** denotes the control;
- **G.id** a numeric $n \times 1$ vector of cluster indicators;

- Ng a numeric vector/matrix/data.frame of cluster sizes; if NULL then Ng is assumed to be equal to the number of available observations in every cluster;
- X a data.frame with columns representing the covariate values for every observation.

Example

```
library(sreg)
# big stata
data <- sreg.rgen(n = 1000, tau.vec = c(0), n.strata = 4, cluster = TRUE)
> head(data)
         Y S D
                    x_1
                              x 2
1 1.717293 1 0 4.772092 2.4138491
2 2.553695 2 0 5.413440 2.0551019
3 2.237556 3 2 6.611161 0.9300293
4 1.825809 3 1 2.735503 1.7839981
5 5.536280 2 2 2.469239 2.0495611
6 1.628753 2 0 4.887561 2.1327071
# matched pairs (small strata)
data <- sreg.rgen(n = 100, tau.vec = c(1.2), cluster = FALSE, small.strata = TRUE, k = 2, treat.sizes = c(1, 1))
> head(data)
          Y S D
                     x 1
1 2.0393535 1 1 7.904694 1.487941
2 3.3839515 1 0 3.461776 2.832059
3 1.7250989 2 0 3.049906 3.170014
4 3.0991776 2 1 7.437064 1.098371
5 1.7406104 3 1 5.008703 1.750753
6 0.6986514 3 0 3.418835 1.375744
```

References

Bugni, F. A., Canay, I. A., and Shaikh, A. M. (2018). Inference Under Covariate-Adaptive Randomization. *Journal of the American Statistical Association*, 113(524), 1784–1796, doi:10.1080/01621459.2017.1375934.

Bugni, F., Canay, I., Shaikh, A., and Tabord-Meehan, M. (2024+). Inference for Cluster Randomized Experiments with Non-ignorable Cluster Sizes. Forthcoming in the Journal of Political Economy: Microeconomics, doi:10.48550/arXiv.2204.08356.

Jiang, L., Linton, O. B., Tang, H., and Zhang, Y. (2023+). Improving Estimation Efficiency via Regression-Adjustment in Covariate-Adaptive Randomizations with Imperfect Compliance. Forthcoming in Review of Economics and Statistics, doi:10.48550/arXiv.2204.08356.

Bai, Y., Jiang, L., Romano, J. P., Shaikh, A. M., and Zhang, Y. (2024). Covariate adjustment in experiments with matched pairs. *Journal of Econometrics*, 241(1), doi:10.1016/j.jeconom.2024.105740.

Liu, J. (2024). Inference for Two-stage Experiments under Covariate-Adaptive Randomization. doi:10.48550/arXiv.2301.09016.

Cytrynbaum, M. (2024). Covariate Adjustment in Stratified Experiments. *Quantitative Economics*, 15(4), 971–998, doi:10.3982/QE2475