# Stratified Randomized Experiments



The sreg package for R, offers a toolkit for estimating average treatment effects (ATEs) in stratified randomized experiments. 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); and Jiang, Linton, Tang, Zhang (2023).

Dependencies: dplyr, tidyr, extraDistr, rlang Suggests: haven, knitr, rmarkdown, testthat (>= 3.0.0) R version required: >= 2.10 Latest Build (v.1.0.0)

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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

\*\* help

The latest version can be installed using devtools. The official CRAN release will be available soon.

```
library(devtools)
install_github("jutrifonov/sreg")
Downloading GitHub repo jutrifonov/sreg@HEAD
  R CMD build
  checking for file '/private/var/folders/mp/06gjwr8j56zdp5j2vgdkd4z40000gq/T/RtmpZh7j1Y/remotesfbf7659
  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 0.5.8.tar.gz'
* installing *source* package 'sreg' ...
** using staged installation
** R
** data
*** moving datasets to lazyload DB
** byte-compile and prepare package for lazy loading
```

# 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)
```

#### 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, ..., 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. [^\*]: 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.

	Y		S		D		G.id		Ng		<b>x_1</b>	<b>x_2</b>	
		-		- -		-   -		-   -		-   -			-
	-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  $\alpha$ . The example of the printed output is provided below.

```
Saturated Model Estimation Results under CAR with clusters and linear adjustments
Observations: 30000
Clusters: 1000
Number of treatments: 2
Number of strata: 4
Covariates used in linear adjustments: x_1, x_2
Coefficients:
     Tau
          As.se T-stat P-value CI.left(95%) CI.right(95%) Significance
1 0.01614 0.04513 0.35753 0.72069
                                      -0.07232
                                                      0.1046
2 0.78642 0.04642 16.94263 0.00000
                                       0.69545
                                                      0.8774
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.

#### **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</pre>
```

```
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
   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)
Saturated Model Estimation Results under CAR
Observations: 215
Number of treatments: 2
Number of strata: 5
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)
```

Saturated Model Estimation Results under CAR

## 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**

```
sreg.rgen(n, Nmax = 50, n.strata,
          tau.vec = c(0), gamma.vec = c(0.4, 0.2, 1),
          cluster = TRUE, is.cov = TRUE)
```

### 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  $1 \times 3$  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.

#### 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 × 1 vector of treatments indexed by 0, 1, 2, ..., 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

```
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
```

## 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.