Semiparametric Estimation of Long-Term Treatment Effects

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The package longterm implements the semiparametric estimators of long-term average treatment effects developed in Chen and Ritzwoller [2021].

Setup

We observe two data sets. One data set contains measurements of the short-term outcomes of a randomized evaluation of a binary treatment of interest. The other data set contains observational measurements of the joint distribution of short-term and long-term outcomes. We are interested in estimating the average treatment effect of the treatment on a long-term outcome of interest, and are willing to impose the identifying assumptions specified in Chen and Ritzwoller [2021]. Chen and Ritzwoller [2021] develop a semiparametric estimator of this parameter in this context with the "Double/Debiased Machine Learning" construction of Chernozhukov et al. [2018].

Example

We illustrate the use of the package using a data set from Banerjee et al. [2015]. The dataset is included in the package as the list graduation. Run ?graduation for a full description of the dataset. Further details on the contents and construction of this data set is available in the Supplemental Appendix to Chen and Ritzwoller [2021]. See ?longterm for full descriptions of the arguments and returned values of longterm.

```
library(longterm)

#> Loading required package: rlang

#> Loading required package: ggplot2

#> Loading required package: lattice

#> Loading required package: glmnet

#> Loading required package: Matrix

#> Loading required package: Matrix

#> Loading required package: grf

#> Loading required package: grf

#> Loading required package: agboost

#> Loading required package: dplyr

#> Attaching package: 'dplyr'

#> The following object is masked from 'package:xgboost':

#>
```

```
slice
#> The following objects are masked from 'package:stats':
#>
#>
       filter, lag
#> The following objects are masked from 'package:base':
#>
       intersect, setdiff, setequal, union
#> Loading required package: tidyselect
#> Loading required package: haven
## Assign each observation to `experimental` and `observational` groups randomly
graduation$observe <- as.numeric(runif(nrow(graduation)) <= 0.5)</pre>
## Choose Pre-treatment covariates, Short-term Outcomes, and Long-term Outcome
X_vars <- c("ctotal_pcmonth_bsl", "cnonfood_pcmonth_bsl",</pre>
            "cfood_pcmonth_bsl", "cdurable_pcmonth_bsl",
            "asset_index_bsl",
                                 "asset_prod_index_bsl")
S_vars <- c("ctotal_pcmonth_end", "cnonfood_pcmonth_end",</pre>
            "cfood_pcmonth_end", "cdurable_pcmonth_end",
                                "asset_prod_index_end")
            "asset_index_end",
Y_var <- "ctotal_pcmonth_fup"</pre>
## Assume treatment is observed in the observational sample
# Average effect for the experimental population
est 1 1 <- longterm(graduation, S vars = S vars, X vars = X vars, Y var = Y var,
                    obs = TRUE, estimand = TRUE, type = "glmnet")
# Average effect for the observational population
est_1_0 <- longterm(graduation, S_vars = S_vars, X_vars = X_vars, Y_var = Y_var,
                    obs = TRUE, estimand = FALSE, type = "glmnet")
## Assume treatment is not observed in the observational sample
# Average effect for the experimental population
est_0_1 <- longterm(graduation, S_vars = S_vars, X_vars = X_vars, Y_var = Y_var,</pre>
                    obs = FALSE, estimand = TRUE, type = "glmnet")
# Average effect for the observational population
est_0_0 <- longterm(graduation, S_vars = S_vars, X_vars = X_vars, Y_var = Y_var,
                    obs = FALSE, estimand = FALSE, type = "glmnet")
## Display results
kable(data.frame(treatment observed = c("Yes", "Yes", "No", "No"),
                 subpopulation = c("Observational", "Experimental",
                                    "Observational", "Experimental").
                 estimate = c(est_1_1$hat_tau, est_1_0$hat_tau,
                              est_0_1$hat_tau, est_0_0$hat_tau),
                 standard_error = c(est_1_1$se, est_1_0$se,
                                     est_0_1$se, est_0_0$se),
                 lower_ci = c(est_1_1$ci[1], est_1_0$ci[1],
```

treatment_observed	subpopulation	estimate	standard_error	lower_ci	upper_ci
Yes	Observational	6.768485	6.245453	-5.472378	19.009348
Yes	Experimental	8.021267	6.174411	-4.080357	20.122891
No	Observational	3.076402	2.702743	-2.220878	8.373681
No	Experimental	4.553722	2.154539	0.330904	8.776541

References

Abhijit Banerjee, Esther Duflo, Nathanael Goldberg, Dean Karlan, Robert Osei, William Parienté, Jeremy Shapiro, Bram Thuysbaert, and Christopher Udry. A multifaceted program causes lasting progress for the very poor: Evidence from six countries. *Science*, 348(6236), 2015.

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Victor Chernozhukov, Denis Chetverikov, Mert Demirer, Esther Duflo, Christian Hansen, Whitney Newey, and James Robins. Double/debiased machine learning for treatment and structural parameters, 2018.