

# Package: drpop (via r-universe)

September 7, 2024

**Title** Efficient and Doubly Robust Population Size Estimation

**Version** 0.0.3

**Description** Estimation of the total population size from capture-recapture data efficiently and with low bias implementing the methods from Das M, Kennedy EH, and Jewell NP (2021) <[arXiv:2104.14091](https://arxiv.org/abs/2104.14091)>. The estimator is doubly robust against errors in the estimation of the intermediate nuisance parameters. Users can choose from the flexible estimation models provided in the package, or use any other preferred model.

**Depends** stats

**Imports** gam, janitor, reshape2, stringr, tidyr, dplyr, SuperLearner, ggplot2, nnet, nnls, parallel, ranger

**License** GPL-3

**Encoding** UTF-8

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**Repository** <https://mqnjgrid.r-universe.dev>

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informat	<i>A function to check whether a given data table/matrix/data frame is in the appropriate for drpop.</i>
----------	--

---

## Description

A function to check whether a given data table/matrix/data frame is in the appropriate for drpop.

## Usage

```
informat(data, K = 2)
```

## Arguments

data	The data table/matrix/data frame which is to be checked.
K	The number of lists (optional).

## Value

A boolean for whether data is in the appropriate format.

## Examples

```
data = matrix(sample(c(0,1), 2000, replace = TRUE), ncol = 2)
x = matrix(rnorm(nrow(data)*3, 2,1), nrow = nrow(data))
```

```
informat(data = data)
#this returns TRUE
```

```
data = cbind(data, x)
informat(data = data)
#this returns TRUE
```

```
informat(data = data, K = 3)
#this returns FALSE
```

---

plotci	<i>Plot estimated confidence interval of total population size from object of class popsize or popsize_cond.</i>
--------	--

---

### Description

Plot estimated confidence interval of total population size from object of class popsize or popsize\_cond.

### Usage

```
plotci(object, tsize = 12, ...)
```

### Arguments

object	An object of class popsize or popsize_cond.
tsize	The text size for the plots.
...	Any extra arguments passed into the function.

### Value

A ggplot object fig with population size estimates and the 95% confidence intervals.

### References

H. Wickham. ggplot2: Elegant Graphics for Data Analysis. *Springer-Verlag* New York, 2016.

### Examples

```
data = simuldata(n = 10000, l = 1)$data_xstar  
  
p = popsize(data = data, funcname = c("logit", "gam"))  
plotci(p)  
  
data = simuldata(n = 10000, l = 1, categorical = TRUE)$data_xstar  
p = popsize_cond(data = data, condvar = 'catcov')  
plotci(p)
```

---

popsize	<i>Estimate total population size and capture probability using user provided set of models or user provided nuisance estimates.</i>
---------	--

---

### Description

Estimate total population size and capture probability using user provided set of models or user provided nuisance estimates.

### Usage

```
popsize(
  data,
  K = 2,
  j,
  k,
  margin = 0.005,
  filterrows = FALSE,
  nfolds = 5,
  funcname = c("rangerlogit"),
  sl.lib = c("SL.gam", "SL.glm", "SL.glm.interaction", "SL.ranger", "SL.glmnet"),
  getnuis,
  q1mat,
  q2mat,
  q12mat,
  idfold,
  TMLE = TRUE,
  PLUGIN = TRUE,
  Nmin = 100,
  ...
)
```

### Arguments

data	The data frame in capture-recapture format with K lists for which total population is to be estimated. The first K columns are the capture history indicators for the K lists. The remaining columns are covariates in numeric format.
K	The number of lists that are present in the data.
j	The first list to be used for estimation.
k	The second list to be used in the estimation.
margin	The minimum value the estimates can attain to bound them away from zero.
filterrows	A logical value denoting whether to remove all rows with only zeroes.
nfolds	The number of folds to be used for cross fitting.
funcname	The vector of estimation function names to obtain the population size.

sl.lib	Algorithm library for <code>qhat_sl()</code> . See <code>SuperLearner::listWrappers()</code> . Default library includes "gam", "glm", "glmnet", "glm.interaction", "ranger".
getnuis	A list object with the nuisance function estimates and the fold assignment of the rows for cross-fitting or a data.frame with the nuisance estimates.
q1mat	A dataframe with capture probabilities for the first list.
q2mat	A dataframe with capture probabilities for the second list.
q12mat	A dataframe with capture probabilities for both the lists simultaneously.
idfold	The fold assignment of each row during estimation.
TMLE	The logical value to indicate whether TMLE has to be computed.
PLUGIN	The logical value to indicate whether the plug-in estimates are returned.
Nmin	The cutoff for minimum sample size to perform doubly robust estimation. Otherwise, Petersen estimator is returned.
...	Any extra arguments passed into the function. See <code>qhat_rangerlogit()</code> , <code>qhat_sl()</code> , <code>tmle()</code> .

### Value

A list of estimates containing the following components for each list-pair, model and method (PI = plug-in, DR = doubly-robust, TMLE = targeted maximum likelihood estimate):

result	A dataframe of the below estimated quantities. <ul style="list-style-type: none"> <li>• psi The estimated capture probability.</li> <li>• sigma The efficiency bound.</li> <li>• n The estimated population size n.</li> <li>• sigman The estimated standard deviation of the population size.</li> <li>• cin.l The estimated lower bound of a 95% confidence interval of n.</li> <li>• cin.u The estimated upper bound of a 95% confidence interval of n.</li> </ul>
N	The number of data points used in the estimation after removing rows with missing data.
ifvals	The estimated influence function values for the observed data.
nuis	The estimated nuisance functions (q12, q1, q2) for each element in funcname.
nuistmle	The estimated nuisance functions (q12, q1, q2) from tmle for each element in funcname.
idfold	The division of the rows into sets (folds) for cross-fitting.

### References

- Bickel, P. J., Klaassen, C. A., Bickel, P. J., Ritov, Y., Klaassen, J., Wellner, J. A., and Ritov, Y. (1993). Efficient and adaptive estimation for semiparametric models, volume 4. *Johns Hopkins University Press Baltimore*
- van der Vaart, A. (2002a). Part iii: Semiparameric statistics. *Lectures on Probability Theory and Statistics*, pages 331-457
- van der Laan, M. J. and Robins, J. M. (2003). Unified methods for censored longitudinal data and causality. *Springer Science & Business Media*

- Tsiatis, A. (2006). Semiparametric theory and missing data *springer. New York*
- Kennedy, E. H. (2016). Semiparametric theory and empirical processes in causal inference. *Statistical causal inferences and their applications in public health research*, pages 141-167. *Springer*
- Das, M., Kennedy, E. H., & Jewell, N.P. (2021). Doubly robust capture-recapture methods for estimating population size. *arXiv preprint arXiv:2104.14091*.

## Examples

```
data = simuldata(1000, l = 3)$data
qhat = popsize(data = data, funcname = c("logit", "gam"), nfolds = 2, margin = 0.005)
psin_estimate = popsize(data = data, getnuis = qhat$nuis, idfold = qhat$idfold)

data = simuldata(n = 6000, l = 3)$data
psin_estimate = popsize(data = data[,1:2])
#this returns the basic plug-in estimate since covariates are absent.

psin_estimate = popsize(data = data, funcname = c("gam", "rangerlogit"))
```

---

popsize_cond	<i>Estimate total population size and capture probability using user provided set of models conditioned on an attribute.</i>
--------------	--

---

## Description

Estimate total population size and capture probability using user provided set of models conditioned on an attribute.

## Usage

```
popsize_cond(
  data,
  K = 2,
  filterrows = FALSE,
  funcname = c("rangerlogit"),
  condvar,
  nfolds = 2,
  margin = 0.005,
  sl.lib = c("SL.gam", "SL.glm", "SL.glm.interaction", "SL.ranger", "SL.glmnet"),
  TMLE = TRUE,
  PLUGIN = TRUE,
  Nmin = 100,
  ...
)
```

**Arguments**

data	The data frame in capture-recapture format for which total population is to be estimated. The first K columns are the capture history indicators for the K lists. The remaining columns are covariates in numeric format.
K	The number of lists in the data. typically the first K rows of data.
filterrows	A logical value denoting whether to remove all rows with only zeroes.
funcname	The vector of estimation function names to obtain the population size.
condvar	The covariate for which conditional estimates are required.
nfolds	The number of folds to be used for cross fitting.
margin	The minimum value the estimates can attain to bound them away from zero.
sl.lib	Algorithm library for <code>qhat_sl()</code> . See <code>SuperLearner::listWrappers()</code> . Default library includes "gam", "glm", "glmnet", "glm.interaction", "ranger".
TMLE	The logical value to indicate whether TMLE has to be computed.
PLUGIN	The logical value to indicate whether the plug-in estimates are returned.
Nmin	The cutoff for minimum sample size to perform doubly robust estimation. Otherwise, Petersen estimator is returned.
...	Any extra arguments passed into the function. See <code>qhat_rangerlogit()</code> , <code>qhat_sl()</code> , <code>tmle()</code> .

**Value**

A list of estimates containing the following components for each list-pair, model and method (PI = plug-in, DR = doubly-robust, TMLE = targeted maximum likelihood estimate):

result	A dataframe of the below estimated quantities. <ul style="list-style-type: none"> <li>• psi The estimated capture probability.</li> <li>• sigma The efficiency bound.</li> <li>• n The estimated population size n.</li> <li>• sigman The estimated standard deviation of the population size.</li> <li>• cin.l The estimated lower bound of a 95% confidence interval of n.</li> <li>• cin.u The estimated upper bound of a 95% confidence interval of n.</li> </ul>
N	The number of data points used in the estimation after removing rows with missing data.
ifvals	The estimated influence function values for the observed data.
nuis	The estimated nuisance functions (q12, q1, q2) for each element in funcname.
nuistmle	The estimated nuisance functions (q12, q1, q2) from tmle for each element in funcname.
idfold	The division of the rows into sets (folds) for cross-fitting.

**References**

Das, M., Kennedy, E. H., & Jewell, N.P. (2021). Doubly robust capture-recapture methods for estimating population size. *arXiv preprint arXiv:2104.14091*.

**See Also**[popsize](#)**Examples**

```

data = simuldata(n = 10000, l = 2, categorical = TRUE)$data

psin_estimate = popsize_cond(data = data, funcname = c("logit", "gam"),
  condvar = 'catcov', PLUGIN = TRUE, TMLE = TRUE)
#this returns the plug-in, the bias-corrected and the tmle estimate for the
#two models conditioned on column catcov

```

---

popsize_simul	<i>Estimate the total population size and capture probabilities using perturbed true nuisance functions.</i>
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---

**Description**

Estimate the total population size and capture probabilities using perturbed true nuisance functions.

**Usage**

```

popsize_simul(
  data,
  n,
  K = 2,
  nfolds = 5,
  pi1,
  pi2,
  omega,
  alpha,
  margin = 0.005,
  iter = 100,
  twolist = TRUE
)

```

**Arguments**

data	The data frame in capture-recapture format for which total population is to be estimated. The first K columns are the capture history indicators for the K lists. The remaining columns are covariates in numeric format.
n	The true population size. Required to calculate the added error.
K	The number of lists in the data. typically the first K rows of data.
nfolds	The number of folds to be used for cross fitting.



pi1	The function to calculate the conditional capture probabilities of list 1 using covariates.
pi2	The function to calculate the conditional capture probabilities of list 2 using covariates.
omega	The standard deviation from zero of the added error.
alpha	The rate of convergence. Takes values in (0, 1].
margin	The minimum value the estimates can attain to bound them away from zero.
iter	An integer denoting the maximum number of iterations allowed for targeted maximum likelihood method.
twolist	The logical value of whether targeted maximum likelihood algorithm fits only two modes when $K = 2$ .

### Value

A list of estimates containing the following components:

psi	A matrix of the estimated capture probability for each list pair, model and method combination. In the absence of covariates, the column represents the standard plug-in estimate. The rows represent the list pair which is assumed to be independent conditioned on the covariates. The columns represent the model and method combinations (PI = plug-in, DR = bias-corrected, TMLE = targeted maximum likelihood estimate) indicated in the columns.
sigma2	A matrix of the efficiency bound $\sigma^2$ in the same format as psi.
n	A matrix of the estimated population size n in the same format as psi.
varn	A matrix of the variance for population size estimate in the same format as psi.
N	The number of data points used in the estimation after removing rows with missing data.

### References

Das, M., Kennedy, E. H., & Jewell, N.P. (2021). Doubly robust capture-recapture methods for estimating population size. *arXiv preprint arXiv:2104.14091*

### Examples

```

simulresult = simuldata(n = 2000, l = 2)
data = simulresult$data

psin_estimate = popsize_simul(data = data,
  pi1 = simulresult$pi1, pi2 = simulresult$pi2,
  alpha = 0.25, omega = 1)

```

---

qhat_gam	<i>Estimate marginal and joint distribution of lists j and k using generalized additive models.</i>
----------	---

---

## Description

Estimate marginal and joint distribution of lists j and k using generalized additive models.

## Usage

```
qhat_gam(List.train, List.test, K = 2, j = 1, k = 2, margin = 0.005, ...)
```

## Arguments

List.train	The training data matrix used to estimate the distribution functions.
List.test	The data matrix on which the estimator function is applied.
K	The number of lists in the data.
j	The first list that is conditionally independent.
k	The second list that is conditionally independent.
margin	The minimum value the estimates can attain to bound them away from zero.
...	Any extra arguments passed into the function.

## Value

A list of the marginal and joint distribution probabilities q1, q2 and q12.

## References

Trevor Hastie (2020). gam: Generalized Additive Models. *R package version 1.20*. <https://CRAN.R-project.org/package=gam>

## Examples

```
## Not run:  
qhat = qhat_gam(List.train = List.train, List.test = List.test, margin = 0.005)  
q1 = qhat$q1  
q2 = qhat$q2  
q12 = qhat$q12  
  
## End(Not run)
```

---

qhat_logit	<i>Estimate marginal and joint distribution of lists j and k using logistic regression.</i>
------------	---

---

### Description

Estimate marginal and joint distribution of lists j and k using logistic regression.

### Usage

```
qhat_logit(List.train, List.test, K = 2, j = 1, k = 2, margin = 0.005, ...)
```

### Arguments

List.train	The training data matrix used to estimate the distribution functions.
List.test	The data matrix on which the estimator function is applied.
K	The number of lists in the data.
j	The first list that is conditionally independent.
k	The second list that is conditionally independent.
margin	The minimum value the estimates can attain to bound them away from zero.
...	Any extra arguments passed into the function.

### Value

A list of the marginal and joint distribution probabilities q1, q2 and q12.

### Examples

```
## Not run:  
qhat = qhat_logit(List.train = List.train, List.test = List.test, margin = 0.005)  
q1 = qhat$q1  
q2 = qhat$q2  
q12 = qhat$q12  
  
## End(Not run)
```

---

qhat_mlogit	<i>Estimate marginal and joint distribution of lists j and k using multinomial logistic model.</i>
-------------	--

---

**Description**

Estimate marginal and joint distribution of lists j and k using multinomial logistic model.

**Usage**

```
qhat_mlogit(List.train, List.test, K = 2, j = 1, k = 2, margin = 0.005, ...)
```

**Arguments**

List.train	The training data matrix used to estimate the distribution functions.
List.test	The data matrix on which the estimator function is applied.
K	The number of lists in the data.
j	The first list that is conditionally independent.
k	The second list that is conditionally independent.
margin	The minimum value the estimates can attain to bound them away from zero.
...	Any extra arguments passed into the function.

**Value**

A list of the marginal and joint distribution probabilities q1, q2 and q12.

**References**

Croissant Y (2020). Estimation of Random Utility Models in R: The mlogit Package. *Journal of Statistical Software*, 95(11), 1-41. doi: 10.18637/jss.v095.i11 (URL: <https://doi.org/10.18637/jss.v095.i11>).

Venables, W. N. & Ripley, B. D. (2002) Modern Applied Statistics with S. Fourth Edition. *Springer*, New York. ISBN 0-387-95457-0

**Examples**

```
## Not run:
qhat = qhat_mlogit(List.train = List.train, List.test = List.test, margin = 0.005)
q1 = qhat$q1
q2 = qhat$q2
q12 = qhat$q12

## End(Not run)
```

---

qhat_ranger	<i>Estimate marginal and joint distribution of lists j and k using ranger.</i>
-------------	--

---

**Description**

Estimate marginal and joint distribution of lists j and k using ranger.

**Usage**

```
qhat_ranger(List.train, List.test, K = 2, j = 1, k = 2, margin = 0.005, ...)
```

**Arguments**

List.train	The training data matrix used to estimate the distribution functions.
List.test	The data matrix on which the estimator function is applied.
K	The number of lists in the data.
j	The first list that is conditionally independent.
k	The second list that is conditionally independent.
margin	The minimum value the estimates can attain to bound them away from zero.
...	Any extra arguments passed into the function.

**Value**

A list of the marginal and joint distribution probabilities q1, q2 and q12.

**References**

Marvin N. Wright, Andreas Ziegler (2017). ranger: A Fast Implementation of Random Forests for High Dimensional Data in C++ and R. *Journal of Statistical Software*, 77(1), 1-17. doi:10.18637/jss.v077.i01

**Examples**

```
## Not run:  
qhat = qhat_ranger(List.train = List.train, List.test = List.test, margin = 0.005)  
q1 = qhat$q1  
q2 = qhat$q2  
q12 = qhat$q12  
  
## End(Not run)
```

---

qhat_rangerlogit	<i>Estimate marginal and joint distribution of lists j and k using ensemble of ranger and logit.</i>
------------------	--

---

### Description

Estimate marginal and joint distribution of lists j and k using ensemble of ranger and logit.

### Usage

```
qhat_rangerlogit(
  List.train,
  List.test,
  K = 2,
  j = 1,
  k = 2,
  margin = 0.005,
  ...
)
```

### Arguments

List.train	The training data matrix used to estimate the distribution functions.
List.test	The data matrix on which the estimator function is applied.
K	The number of lists in the data.
j	The first list that is conditionally independent.
k	The second list that is conditionally independent.
margin	The minimum value the estimates can attain to bound them away from zero.
...	Any extra arguments passed into the function.

### Value

A list of the marginal and joint distribution probabilities q1, q2 and q12.

### References

Marvin N. Wright, Andreas Ziegler (2017). ranger: A Fast Implementation of Random Forests for High Dimensional Data in C++ and R. *Journal of Statistical Software*, 77(1), 1-17. doi:10.18637/jss.v077.i01

Polley, Eric C. and van der Laan, Mark J., (May 2010) Super Learner In Prediction. *U.C. Berkeley Division of Biostatistics Working Paper Series*. Working Paper 266. <https://biostats.bepress.com/ucbbiostat/paper266>

**Examples**

```
## Not run:
qhat = qhat_ranger(List.train = List.train, List.test = List.test, margin = 0.005)
q1 = qhat$q1
q2 = qhat$q2
q12 = qhat$q12

## End(Not run)
```

---

qhat_sl	<i>Estimate marginal and joint distribution of lists j and k using super learner.</i>
---------	---

---

**Description**

Estimate marginal and joint distribution of lists j and k using super learner.

**Usage**

```
qhat_sl(
  List.train,
  List.test,
  K = 2,
  j = 1,
  k = 2,
  margin = 0.005,
  sl.lib = c("SL.glm", "SL.gam", "SL.glm.interaction", "SL.ranger", "SL.glmnet"),
  num_cores = NA,
  ...
)
```

**Arguments**

List.train	The training data matrix used to estimate the distribution functions.
List.test	The data matrix on which the estimator function is applied.
K	The number of lists in the data.
j	The first list that is conditionally independent.
k	The second list that is conditionally independent.
margin	The minimum value the estimates can attain to bound them away from zero.
sl.lib	The functions from the SuperLearner library to be used for model fitting. See <a href="#">SuperLearner::listWrappers()</a> .
num_cores	The number of cores to be used for parallelization in Super Learner.
...	Any extra arguments passed into the function.

**Value**

A list of the marginal and joint distribution probabilities q1, q2 and q12.

**References**

Eric Polley, Erin LeDell, Chris Kennedy and Mark van der Laan (2021). SuperLearner: Super Learner Prediction. *R package version 2.0-28*. <https://CRAN.R-project.org/package=SuperLearner>

van der Laan, M. J., Polley, E. C. and Hubbard, A. E. (2008) Super Learner, *Statistical Applications of Genetics and Molecular Biology*, 6, article 25.

**Examples**

```
## Not run:
qhat = qhat_sl(List.train = List.train, List.test = List.test, margin = 0.005, num_cores = 1)
q1 = qhat$q1
q2 = qhat$q2
q12 = qhat$q12
# One can specify the number of cores to be used for parallel computing
qhat = qhat_sl(List.train = List.train, List.test = List.test, margin = 0.005, num_cores = 2)
q1 = qhat$q1
q2 = qhat$q2
q12 = qhat$q12

## End(Not run)
```

---

reformat

*A function to reorder the columns of a data table/matrix/data frame and to change factor variables to numeric.*

---

**Description**

A function to reorder the columns of a data table/matrix/data frame and to change factor variables to numeric.

**Usage**

```
reformat(data, capturelists)
```

**Arguments**

`data` The data table/matrix/data frame which is to be checked.

`capturelists` The vector of column names or locations for the capture history list columns.

**Value**

`data` With reordered columns so that the capture history columns are followed by the rest.



**Examples**

```
data = matrix(sample(c(0,1), 2000, replace = TRUE), ncol = 2)
x = matrix(rnorm(nrow(data)*3, 2, 1), nrow = nrow(data))

data = cbind(x, data)
result<- reformat(data = data, capturelists = c(4,5))
```

---

simuldata	<i>A function to reorder the columns of a data table/matrix/data frame and to change factor variables to numeric.</i>
-----------	---

---

**Description**

A function to reorder the columns of a data table/matrix/data frame and to change factor variables to numeric.

**Usage**

```
simuldata(n, l, categorical = FALSE, ep = 0, K = 2)
```

**Arguments**

n	The size of the population.
l	The number of continuous covariates.
categorical	A logical value of whether to include a categorical column.
ep	A numeric value to change the list probabilities.
K	The number of lists. Default value is 2. Maximum value is 3.

**Value**

A list of estimates containing the following components:

data	A dataframe in with K list capture histories and covariates from a population if true size n with only observed rows.
data_xstar	A dataframe in with two list capture histories and transformed covariates from a population if true size n with only observed rows.
psi0	The empirical capture probability for the set-up used.
pi1	The conditional capture probabilities for list 1.
pi2	The conditional capture probabilities for list 2.
pi3	The conditional capture probabilities for list 3 when K = 3.

**References**

Tilling, K., & Sterne, J. A. (1999). Capture-recapture models including covariate effects. *American journal of epidemiology*, 149(4), 392-400.

Kennedy, E. H. (2019). Nonparametric causal effects based on incremental propensity score interventions. *Journal of the American Statistical Association*, 114(526), 645-656.

**Examples**

```
data = simuldata(n = 1000, l = 2)$data
psi0 = simuldata(n = 10000, l = 2)$psi0
```

---

tmle	<i>Returns the targeted maximum likelihood estimates for the nuisance functions</i>
------	---

---

**Description**

Returns the targeted maximum likelihood estimates for the nuisance functions

**Usage**

```
tmle(
  datmat,
  iter = 250,
  margin = 0.005,
  stop_margin = 0.005,
  twolist = FALSE,
  K = 2,
  ...
)
```

**Arguments**

datmat	The data frame containing columns yj, yk, yjk, q10, q02 and q12.
iter	An integer denoting the maximum number of iterations allowed for targeted maximum likelihood method. Default value is 100.
margin	The minimum value the estimates can attain to bound them away from zero.
stop_margin	The minimum value the estimates can attain to bound them away from zero.
twolist	The logical value of whether targeted maximum likelihood algorithm fits only two modes when K = 2.
K	The number of lists in the original data.
...	Any extra arguments passed into the function.

**Value**

A list of estimates containing the following components:

error	An indicator of whether the algorithm ran and converged. Returns FALSE, if it ran correctly and FALSE otherwise.
datmat	A data frame returning datmat with the updated estimates for the nuisance functions q10, q02 and q12. This is returned only if error is FALSE.

## References

van der Laan, M. J. and Rubin, D. (2006). Targeted maximum likelihood learning. *The International Journal of Biostatistics*, 2(1)

Das, M., Kennedy, E. H., & Jewell, N.P. (2021). Doubly robust capture-recapture methods for estimating population size. *arXiv preprint* arXiv:2104.14091.

## Examples

```
data = matrix(sample(c(0,1), 2000, replace = TRUE), ncol = 2)
xmat = matrix(runif(nrow(data)*3, 0, 1), nrow = nrow(data))
datmat = cbind(data, data[,1]*data[,2], xmat)
colnames(datmat) = c("yj", "yk", "yjk", "q10", "q02", "q12")
datmat = as.data.frame(datmat)
result = tmle(datmat, margin = 0.005, stop_margin = 0.00001, twolist = TRUE)
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