

# Package: twoStageDesignTMLE (via r-universe)

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**Title** Targeted Maximum Likelihood Estimation for Two-Stage Study Design

**Version** 1.0

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**Description** An inverse probability of censoring weighted (IPCW) targeted maximum likelihood estimator (TMLE) for evaluating a marginal point treatment effect from data where some variables were collected on only a subset of participants using a two-stage design (or marginal mean outcome for a single arm study). A TMLE for conditional parameters defined by a marginal structural model (MSM) is also available.

**Depends** tmlle (>= 2.0)

**Suggests** dbarts (>= 0.9-18), glmnet

**License** GPL-3

**Encoding** UTF-8

**RoxygenNote** 7.3.1

**NeedsCompilation** no

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

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evalAugW	<i>.evalAugW calls TMLE to use super learner to evaluate preliminary predictions for <math>Q(0,W)</math> and <math>Q(1,W)</math> conditioning on stage 1 covariates</i>
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### Description

`.evalAugW` calls TMLE to use super learner to evaluate preliminary predictions for  $Q(0,W)$  and  $Q(1,W)$  conditioning on stage 1 covariates

### Usage

```
evalAugW(Y, A, W, Delta, id, family, SL.library)
```

### Arguments

Y	outcome vector
A	binary treatment indicator
W	covariate matrix
Delta	outcome missingness indicator
id	identifier of i.i.d. unit
family	outcome regression family
SL.library	super learner library for outcome regression modeling

### Value

W.Q, nx2 matrix of outcome predictions based on stage 1 covariates

---

```
print.summary.twoStageTMLE  
    print.summary.twoStageTMLE
```

---

### **Description**

print.summary.twoStageTMLE

### **Usage**

```
## S3 method for class 'summary.twoStageTMLE'  
print(x, ...)
```

### **Arguments**

x                    an object of class summary.twoStageTMLE  
...                   additional arguments (i)

### **Value**

print object

---

```
print.twoStageTMLE    print.twoStageTMLE
```

---

### **Description**

print.twoStageTMLE

### **Usage**

```
## S3 method for class 'twoStageTMLE'  
print(x, ...)
```

### **Arguments**

x                    an object of class twoStageTMLE  
...                   additional arguments (i)

### **Value**

print tmle results using print.tmle method from tmle package

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setV	<i>Utilities setV Set the number of cross-validation folds as a function of effective sample size See Phillips 2023 doi.org/10.1093/ije/dyad023</i>
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**Description**

Utilities setV Set the number of cross-validation folds as a function of effective sample size See Phillips 2023 doi.org/10.1093/ije/dyad023

**Usage**

```
setV(n.effective)
```

**Arguments**

n.effective      the effective sample size

**Value**

the number of cross-validation folds

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summary.twoStage	<i>summary.twoStageTMLE</i>
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**Description**

Summarizes estimation procedure for missing 2nd stage covariates

**Usage**

```
## S3 method for class 'twoStage'
summary(object, ...)
```

**Arguments**

object            An object of class twoStageTMLE  
 ...                Other arguments passed to the tmle function in the tmle package

**Value**

A list containing the missingness model, terms, coefficients, type,

---

summary.twoStageTMLE *summary.twoStageTMLE*


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**Description**

summary.twoStageTMLE

**Usage**

```
## S3 method for class 'twoStageTMLE'
summary(object, ...)
```

**Arguments**

```
object      an object of class twoStageTMLE
...         additional arguments (ignored)
```

**Value**

list summarizing the two-stage procedure components, summary of the twoStage missingness estimation summary of the tmle for estimating the parameter

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twoStageDesignTMLENews  
*twoStageDesignTMLENews* Get news about recent updates and bug fixes

---

**Description**

twoStageDesignTMLENews Get news about recent updates and bug fixes

**Usage**

twoStageDesignTMLENews(...)

**Arguments**

```
...         ignored
```

**Value**

invisible character string giving the path to the file found.

---

twoStageTMLE

*twoStageTMLE*


---

### Description

Inverse probability of censoring weighted TMLE for evaluating parameters when the full set of covariates is available on only a subset of observations.

### Usage

```
twoStageTMLE(
  Y,
  A,
  W,
  Delta.W,
  W.stage2,
  Z = NULL,
  Delta = rep(1, length(Y)),
  pi = NULL,
  piform = NULL,
  pi.SL.library = c("SL.glm", "SL.gam", "SL.glmnet", "tmle.SL.dbarts.k.5"),
  V.pi = 10,
  pi.discreteSL = TRUE,
  condSetNames = c("A", "W", "Y"),
  id = NULL,
  Q.family = "gaussian",
  augmentW = TRUE,
  augW.SL.library = c("SL.glm", "SL.glmnet", "tmle.SL.dbarts2"),
  rareOutcome = FALSE,
  verbose = FALSE,
  ...
)
```

### Arguments

Y	outcome
A	binary treatment indicator
W	covariate matrix observed on everyone
Delta.W	binary indicator of missing second stage covariates
W.stage2	matrix of second stage covariates observed on subset of observations
Z	optional mediator of treatment effect for evaluating a controlled direct effect
Delta	binary indicator of missing value for outcome Y
pi	optional vector of missingness probabilities for W.stage2
piform	parametric regression formula for estimating pi

pi.SL.library	super learner library for estimating pi
V.pi	number of cross validation folds for estimating pi using super learner
pi.discreteSL	Use discrete super learning when TRUE, otherwise ensemble super learning
condSetNames	Variables to include as predictors of missingness in W. stage2, any combination of Y, A, and either W (for all covariates in W), or individual covariate names in W
id	Identifier of independent units of observation, e.g., clusters
Q.family	Regression family for the outcome
augmentW	When TRUE include predicted values for the outcome the set of covariates used to model the propensity score
augW.SL.library	super learner library for preliminary outcome regression model (ignored when augmentW is FALSE)
rareOutcome	When TRUE specifies less ambitious SL for Q in call to tmle (discreteSL, glm, glmnet, bart library, V=20)
verbose	When TRUE prints informational messages
...	other parameters passed to the tmle function (not checked)

### Value

object of class 'twoStageTMLE'.

tmle	Treatment effect estimates and summary information
twoStage	IPCW weight estimation summary, pi are the probabilities, coef are SL weights or coefficients from glm fit, type of estimation procedure, discreteSL flag indicating whether discrete super learning was used
augW	Matrix of predicted outcomes based on stage 1 covariates only

### See Also

- [tmle::tmle\(\)](#) for details on customizing the estimation procedure
- [twoStageTMLEmsm\(\)](#) for estimating conditional effects
- S Rose and MJ van der Laan. A Targeted Maximum Likelihood Estimator for Two-Stage Designs. *Int J Biostat.* 2011 Jan 1; 7(1): 17. [doi:10.2202/15574679.1217](https://doi.org/10.2202/15574679.1217)

### Examples

```
n <- 1000
W1 <- rnorm(n)
W2 <- rnorm(n)
W3 <- rnorm(n)
A <- rbinom(n, 1, plogis(-1 + .2*W1 + .3*W2 + .1*W3))
Y <- 10 + A + W1 + W2 + A*W1 + W3 + rnorm(n)
d <- data.frame(Y, A, W1, W2, W3)
# Set 400 with data on W3, more likely if W1 > 1
n.sample <- 400
p.sample <- 0.5 + .2*(W1 > 1)
```

```

rows.sample <- sample(1:n, size = n.sample, p = p.sample)
Delta.W <- rep(0,n)
Delta.W[rows.sample] <- 1
W3.stage2 <- cbind(W3 = W3[Delta.W==1])
#1. specify parametric models and do not augment W (fast, but not recommended)
result1 <- twoStageTMLE(Y=Y, A=A, W=cbind(W1, W2), Delta.W = Delta.W,
  W.stage2 = W3.stage2, piform = "Delta.W~ I(W1 > 0)", V.pi= 5,verbose = TRUE,
  Qform = "Y~A+W1",gform="A~W1 + W2 +W3", augmentW = FALSE)
summary(result1)

#2. specify a parametric model for conditional missingness probabilities (pi)
# and use default values to estimate marginal effect using \code{tmle}
result2 <- twoStageTMLE(Y=Y, A=A, W=cbind(W1, W2), Delta.W = Delta.W,
  W.stage2 = cbind(W3)[Delta.W == 1], piform = "Delta.W~ I(W1 > 0)",
  V.pi= 5,verbose = TRUE)
result2

```

---

twoStageTMLEmsm

*twoStageTMLEmsm*


---

## Description

Inverse probability of censoring weighted TMLE for evaluating MSM parameters when the full set of covariates is available on only a subset of observations, as in a 2-stage design.

## Usage

```

twoStageTMLEmsm(
  Y,
  A,
  W,
  V,
  Delta.W,
  W.stage2,
  Delta = rep(1, length(Y)),
  pi = NULL,
  piform = NULL,
  pi.SL.library = c("SL.glm", "SL.gam", "SL.glmnet", "tmle.SL.dbarts.k.5"),
  V.pi = 10,
  pi.discreteSL = TRUE,
  condSetNames = c("A", "V", "W", "Y"),
  id = NULL,
  Q.family = "gaussian",
  augmentW = TRUE,
  augW.SL.library = c("SL.glm", "SL.glmnet", "tmle.SL.dbarts2"),
  rareOutcome = FALSE,
  verbose = FALSE,
  ...
)

```



**Arguments**

Y	outcome of interest (missingness allowed)
A	binary treatment indicator
W	matrix or data.frame of covariates measured on entire population
V	vector, matrix, or dataframe of covariates used to define MSM strata
Delta.W	Indicator of inclusion in subset with additional information
W.stage2	matrix or data.frame of covariates measured in subset population
Delta	binary indicator that outcome Y is observed
pi	optional vector of sampling probabilities
piform	optional parametric regression model for estimating pi
pi.SL.library	optional SL library specification for estimating pi (ignored when piform or pi is provided)
V.pi	optional number of cross-validation folds for super learning (ignored when pi-form or pi is provided)
pi.discreteSL	flag to indicate whether to use ensemble or discrete super learning (ignored when pi-form or pi is provided)
condSetNames	variables to condition on when estimating pi. Default is covariates in V and W.Can optionally also condition on A and/or Y.
id	optional indicator of independent units of observation
Q.family	outcome regression family, "gaussian" or "binomial"
augmentW	set to TRUE to augment W with predicted outcome values when A = 0 and A = 1
augW.SL.library	super learner library for preliminary outcome regression model (ignored when augmentW is FALSE)
rareOutcome	when TRUE sets V.Q = 20, Q.discreteSL = TRUE, Q.SL.library includes glm, glmnet, bart
verbose	when TRUE prints informative messages
...	other arguments passed to the tmleMSM function

**Value**

Object of class "twoStageTMLE"

Treatment effect estimates and summary information from call to tmleMSM function

**tmleStage** IPCW weight estimation summary, pi are the probabilities,coef are SL weights or coefficients from glm fit, type of estimation procedure, discreteSL flag indicating whether discrete super learning was used

**augW** Matrix of predicted outcomes based on stage 1 covariates only

**See Also**

- [tmle::tmleMSM\(\)](#) for details on customizing the estimation procedure
- [twoStageTMLE\(\)](#) for estimating marginal effects

**Examples**

```

n <- 1000
set.seed(10)
W1 <- rnorm(n)
W2 <- rnorm(n)
W3 <- rnorm(n)
A <- rbinom(n, 1, plogis(-1 + .2*W1 + .3*W2 + .1*W3))
Y <- 10 + A + W1 + W2 + A*W1 + W3 + rnorm(n)
Y.bin <- rbinom(n, 1, plogis(-4.6 - 1.8* A + W1 + W2 -.3 *A*W1 + W3))
# Set 400 obs with data on W3, more likely if W1 > 1
n.sample <- 400
p.sample <- 0.5 + .2*(W1 > 1)
rows.sample <- sample(1:n, size = n.sample, p = p.sample)
Delta.W <- rep(0,n)
Delta.W[rows.sample] <- 1
W3.stage2 <- cbind(W3 = W3[Delta.W==1])

# 1. specify parametric models, misspecified outcome model (not recommended)
result1.MSM <- twoStageTMLEmsm(Y=Y, A=A, V= cbind(W1), W=cbind(W2),
Delta.W = Delta.W, W.stage2 = W3.stage2, augmentW = FALSE,
piform = "Delta.W~ I(W1 > 0)", MSM = "A*W1", augW.SL.library = "SL.glm",
Qform = "Y~A+W1",gform="A~W1 + W2 +W3", hAVform = "A~1", verbose=TRUE)
summary(result1.MSM)

# 2. Call again, passing in previously estimated observation weights,
# note that specifying a correct model for Q improves efficiency
result2.MSM <- twoStageTMLEmsm(Y=Y, A=A, V= cbind(W1), W=cbind(W2),
Delta.W = Delta.W, W.stage2 = W3.stage2, augmentW = FALSE,
pi = result1.MSM$twoStage$pi, MSM = "A*W1",
Qform = "Y~ A + W1 + W2 + A*W1 + W3",gform="A~W1 + W2 +W3", hAVform = "A~1")
cbind(SE.Qmis = result1.MSM$tmle$se, SE.Qcor = result2.MSM$tmle$se)

#Binary outcome, augmentW, rareOutcome
result3.MSM <- twoStageTMLEmsm(Y=Y.bin, A=A, V= cbind(W1), W=cbind(W2),
Delta.W = Delta.W, W.stage2 = W3.stage2, augmentW = TRUE,
piform = "Delta.W~ I(W1 > 0)", MSM = "A*W1", gform="A~W1 + W2 +W3",
Q.family = "binomial", rareOutcome=TRUE)

```

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