

Package ‘wast’

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Type Package

Title Subgroup testing in generalized linear models

Version 1.0.1

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Description Provide a method to calculate p-value of the test statistic for subgroup detecting in generalized linear models. In the paper Liu (2022), we consider hypothesis test of coefficients in the generalized linear models (GLM) to detect the existence of the subgroups, which can serve as the optimal individualized treatment recommendation in practice. Test we consider in this paper is one of the class of test problems when a part of parameters is not identifiable under the null. We propose a novel U-like statistic by taking the weighted average over the grouping parameter's space. The proposed test statistic not only improves significantly the power but also is computationally efficient.

License GPL (>= 2)

Depends R (>= 3.2.0)

LazyData true

NeedsCompilation yes

Repository github

URL <https://github.com/xliusufe/wast>

Encoding UTF-8

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wast-package

Subgroup testing in generalized linear models

Description

Provide a method to calculate p-value of the test statistic for subgroup detecting in generalized linear models. In the paper Liu (2022), we consider hypothesis test of coefficients in the generalized linear models (GLM) to detect the existence of the subgroups, which can serve as the optimal individualized treatment recommendation in practice. Test we consider in this paper is one of the class of test problems when a part of parameters is not identifiable under the null. We propose a novel U-like statistic by taking the weighted average over the grouping parameter's space. The proposed test statistic not only improves significantly the power but also is computationally efficient.

Details

Package: wast
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References

- Andrews, D. W. K. and Ploberger, W. (1994). Optimal tests when a nuisance parameter is present only under the alternative. *Econometrica*, 62(6):1383-1414.
- Fan, A., Rui, S., and Lu, W. (2017). Change-plane analysis for subgroup detection and sample size calculation. *Journal of the American Statistical Association*, 112(518):769-778.
- Huang, Y., Cho, J., and Fong, Y. (2021). Threshold-based subgroup testing in logistic regression models in two phase sampling designs. *Journal of the Royal Statistical Society: Series C*. 291-311.
- Liu, X. (2022). Subgroup testing in generalized linear models. Manuscript.

estglm

Estimation in Generalized Linear Models with subgroups

Description

Provide estimators of coefficients in generalized linear models with subgroups.

Usage

```
estglm(data, family = "gaussian", h = NULL, maxIter = 100, tol = 0.0001)
```

Arguments

data	A list, including Y (response), X (baseline variable), Z (grouping difference variable), and U (grouping variable).
family	Family for generalized linear models, including 'Gaussian', 'binomial', and 'Poisson'.
h	A numeric number, which is the bandwidth in the smooth function. Default is $h = \sqrt{\log(n)/n}$
maxIter	An integer, the maximum number of iterations. Default is <code>maxIter = 100</code> .
tol	Convergence threshold. Default is <code>tol = 0.0001</code> .

Details

Generalized linear models

$$f(\mathbf{V}_i; \boldsymbol{\alpha}, \boldsymbol{\beta}, \boldsymbol{\gamma}) = \exp \left\{ \frac{y_i \mu_i - c(\mu_i)}{a(\phi)} \right\} h(y_i),$$

where

$$\mu_i = \mathbf{X}_i^T \boldsymbol{\alpha} + \mathbf{Z}_i^T \boldsymbol{\beta} \mathbf{1}(\mathbf{U}_i^T \boldsymbol{\gamma} \geq 0).$$

Value

alpha	Estimator of the baseline parameter $\boldsymbol{\alpha}$.
beta	Estimator of the grouping difference parameter $\boldsymbol{\beta}$.
theta	Estimator of the grouping parameter $\boldsymbol{\theta}$.
delta	A vector with length n . Estimator of the indicator function $I(\mathbf{U}^T \boldsymbol{\gamma} \geq 0)$.

References

- Andrews, D. W. K. and Ploberger, W. (1994). Optimal tests when a nuisance parameter is present only under the alternative. *Econometrica*, 62(6):1383-1414.
- Fan, A., Rui, S., and Lu, W. (2017). Change-plane analysis for subgroup detection and sample size calculation. *Journal of the American Statistical Association*, 112(518):769-778.
- Huang, Y., Cho, J., and Fong, Y. (2021). Threshold-based subgroup testing in logistic regression models in two phase sampling designs. *Journal of the Royal Statistical Society: Series C*. 291-311.
- Liu, X. (2022). Subgroup detecting in generalized linear models. Manuscript.

Examples

```
data(simulatedData_gaussian)
fit <- estglm(data = data_gaussian, family = "gaussian")
fit$alpha

data(simulatedData_binomial)
fit <- estglm(data = data_binomial, family = "binomial")
fit$beta

data(simulatedData_poisson)
fit <- estglm(data = data_poisson, family = "poisson")
fit$alpha
fit$beta
```

exams

*Examples for Subgroup Test in Generalized Linear Models***Description**

Examples for Family 'Gaussian', 'binomial', and 'Poisson'.

Usage

```
exams(family = "gaussian", method = "wast", M = 1000, K = 1000)
```

Arguments

family	Family for generalized linear models, including 'Gaussian', 'binomial', and 'Poisson'.
method	There are three methods, including the proposed 'wast', 'sst', and 'slrt'.
M	An integer, the number of bootstrap samples.
K	An integer, the number of threshold values for 'sst' and 'slrt'.

Value

pvals	P-value of the corresponding test statistic.
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References

Andrews, D. W. K. and Ploberger, W. (1994). Optimal tests when a nuisance parameter is present only under the alternative. *Econometrica*, 62(6):1383-1414.

Fan, A., Rui, S., and Lu, W. (2017). Change-plane analysis for subgroup detection and sample size calculation. *Journal of the American Statistical Association*, 112(518):769-778.

Huang, Y., Cho, J., and Fong, Y. (2021). Threshold-based subgroup testing in logistic regression models in two phase sampling designs. *Journal of the Royal Statistical Society: Series C*. 291-311.

Liu, X. (2022). Subgroup detecting in generalized linear models. Manuscript.

Examples

```
pvals <- exams(family = "gaussian", method = "wast")
pvals

pvals <- exams(family = "binomial", method = "wast")
pvals

pvals <- exams(family = "poisson", method = "wast")
pvals
```

pval	<i>P-value for Subgroup Test in Generalized Linear Models</i>
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Description

Provide p-value for subgroup test in generalized linear models, including three methods 'wast', 'sst', and 'slrt'.

Usage

```
pval(data, family = "gaussian", method = 'wast', M=1000, K = 2000)
```

Arguments

data	A list, including Y (response), X (baseline variable), Z (grouping difference variable), and U (grouping variable).
family	Family for generalized linear models, including 'Gaussian', 'binomial', and 'Poisson'.
method	There are three methods, including the proposed 'wast', 'sst', and 'slrt'.
M	An integer, the number of bootstrap samples.
K	An integer, the number of threshold values for 'sst' and 'slrt'.

Details

Generalized linear models

$$f(\mathbf{V}_i; \boldsymbol{\alpha}, \boldsymbol{\beta}, \boldsymbol{\gamma}) = \exp \left\{ \frac{y_i \mu_i - c(\mu_i)}{a(\phi)} \right\} h(y_i),$$

where

$$\mu_i = \mathbf{X}_i^T \boldsymbol{\alpha} + \mathbf{Z}_i^T \boldsymbol{\beta} \mathbf{1}(\mathbf{U}_i^T \boldsymbol{\gamma} \geq 0).$$

The hypothesis test problem is

$$H_0 : \boldsymbol{\beta} = \mathbf{0} \quad \text{versus} \quad H_1 : \boldsymbol{\beta} \neq \mathbf{0}.$$

Value

pvals	P-value of the corresponding test statistic.
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References

- Andrews, D. W. K. and Ploberger, W. (1994). Optimal tests when a nuisance parameter is present only under the alternative. *Econometrica*, 62(6):1383-1414.
- Fan, A., Rui, S., and Lu, W. (2017). Change-plane analysis for subgroup detection and sample size calculation. *Journal of the American Statistical Association*, 112(518):769-778.
- Huang, Y., Cho, J., and Fong, Y. (2021). Threshold-based subgroup testing in logistic regression models in two phase sampling designs. *Journal of the Royal Statistical Society: Series C*. 291-311.
- Liu, X. (2022). Subgroup detecting in generalized linear models. Manuscript.

Examples

```
data(simulatedData_gaussian)
pvals <- pval(data = data_gaussian, family = "gaussian")
pvals

data(simulatedData_binomial)
pvals <- pval(data = data_binomial, family = "binomial")
pvals

data(simulatedData_poisson)
pvals <- pval(data = data_poisson, family = "poisson")
pvals
```

simulatedData	<i>Simulated data from generalized linear models</i>
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Description

Simulated data from generalized linear models, including family 'gaussian' (simulatedData_gaussian), 'binomial' (simulatedData_binomial), and 'poisson' (simulatedData_poisson).

Usage

```
data(simulatedData_gaussian)
```

Details

We simulated data generated from generalized linear models

$$f(\mathbf{V}_i; \boldsymbol{\alpha}, \boldsymbol{\beta}, \boldsymbol{\gamma}) = \exp \left\{ \frac{y_i \mu_i - c(\mu_i)}{a(\phi)} \right\} h(y_i),$$

where

$$\mu_i = \mathbf{X}_i^T \boldsymbol{\alpha} + \mathbf{Z}_i^T \boldsymbol{\beta} \mathbf{1}(\mathbf{U}_i^T \boldsymbol{\gamma} \geq 0).$$

- Y: the response, an n -vector
- X: the baseline variable with dimension $n \times p$
- Z: the grouping difference variable with dimension $n \times q$
- U: the grouping variable with dimension $n \times r$

References

Liu, X. (2022). Subgroup detecting in generalized linear models. Manuscript.

Examples

```
data(simulatedData_gaussian)

y <- data_gaussian$Y[1:5]
x <- dim(data_gaussian$X)
z <- dim(data_gaussian$Z)
u <- dim(data_gaussian$U)
```

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