

Package ‘ProSGPV’

October 12, 2022

Title Penalized Regression with Second-Generation P-Values

Version 1.0.0

Date 2021-08-06

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Description Implementation of penalized regression with second-generation p-values for variable selection. The algorithm can handle linear regression, GLM, and Cox regression. S3 methods `print()`, `summary()`, `coef()`, `predict()`, and `plot()` are available for the algorithm. Technical details can be found at Zuo et al. (2021) <[doi:10.1080/00031305.2021.1946150](https://doi.org/10.1080/00031305.2021.1946150)>.

Depends R (>= 3.5.0), glmnet, brglm2

Imports MASS, survival

License GPL-3

Encoding UTF-8

URL <https://github.com/zuoyi93/ProSGPV>

BugReports <https://github.com/zuoyi93/ProSGPV/issues>

LazyData true

RoxygenNote 7.1.1

Suggests rmarkdown, knitr

VignetteBuilder knitr

NeedsCompilation no

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Repository CRAN

Date/Publication 2021-08-06 21:40:02 UTC

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coef.sgpv	coef.sgpv: <i>Extract coefficients from the model fit</i>
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Description

S3 method coef for an S3 object of class sgpv

Usage

```
## S3 method for class 'sgpv'
coef(object, ...)
```

Arguments

object	An sgpv object
...	Other coef arguments

Value

Coefficients in the OLS model

Examples

```
# prepare the data
x <- t.housing[, -ncol(t.housing)]
y <- t.housing$V9

# run one-stage algorithm
out.sgpv <- pro.sgpv(x = x, y = y)
```

```
# get coefficients
coef(out.sgpv)
```

```
gen.sim.data
```

```
gen.sim.data: Generate simulation data
```

Description

This function can be used to generate autoregressive simulation data

Usage

```
gen.sim.data(
  n = 100,
  p = 50,
  s = 10,
  family = c("gaussian", "binomial", "poisson", "cox"),
  beta.min = 1,
  beta.max = 5,
  rho = 0,
  nu = 2,
  sig = 1,
  intercept = 0,
  scale = 2,
  shape = 1,
  rateC = 0.2
)
```

Arguments

n	Number of observations. Default is 100.
p	Number of explanatory variables. Default is 50.
s	Number of true signals. It can only be an even number. Default is 10.
family	A description of the error distribution and link function to be used in the model. It can take the value of <code>{gaussian}</code> , <code>{binomial}</code> , <code>{poisson}</code> , and <code>{cox}</code> . Default is <code>{gaussian}</code>
beta.min	The smallest effect size in absolute value. Default is 1.
beta.max	The largest effect size in absolute value. Default is 5.
rho	Autocorrelation level. A numerical value between -1 and 1. Default is 0.
nu	Signal to noise ratio in linear regression. Default is 2.
sig	Standard deviation in the design matrix. Default is 1.
intercept	Intercept of the linear predictor in the GLM. Default is 0.
scale	Scale parameter in the Weibull distribution. Default is 2.
shape	Shape parameter in the Weibull distribution. Default is 1.
rateC	Rate of censoring in the survival data. Default is 0.2.

Value

A list of following components:

X The generated explanatory variable matrix

Y A vector of outcome. If family is `{cox}`, a two-column object is returned where the first column is the time and the second column is status (0 is censoring and 1 is event)

index The indices of true signals

beta The true coefficient vector of length p

Examples

```
# generate data for linear regression
data.linear <- gen.sim.data(n = 20, p = 10, s = 4)

# extract x
x <- data.linear[[1]]

# extract y
y <- data.linear[[2]]

# extract the indices of true signals
index <- data.linear[[3]]

# extract the true coefficient vector
true.beta <- data.linear[[4]]

# generate data for logistic regression
data.logistic <- gen.sim.data(n = 20, p = 10, s = 4, family = "binomial")

# extract x
x <- data.logistic[[1]]

# extract y
y <- data.logistic[[2]]

# extract the indices of true signals
index <- data.logistic[[3]]

# extract the true coefficient vector
true.beta <- data.logistic[[4]]
```

get.candidate

get.candidate: *Get candidate set*

Description

Get the indices of the candidate set in the first stage

Usage

```
get.candidate(xs, ys, family)
```

Arguments

xs	Standardized independent variables
ys	Standardized dependent variable
family	A description of the error distribution and link function to be used in the model. It can take the value of <code>{gaussian}</code> , <code>{binomial}</code> , <code>{poisson}</code> , and <code>{cox}</code> .

Value

A list of following components:

candidate.index A vector of indices of selected variables in the candidate set

lambda The lambda selected by generalized information criterion

get.coef	get.coef: <i>Get coefficients at each lambda</i>
----------	--------------------------------------------------

Description

Get the coefficients and confidence intervals from regression at each lambda as well as the null bound in SGPVs

Usage

```
get.coef(xs, ys, lambda, lasso, family)
```

Arguments

xs	Standardized design matrix
ys	Standardized outcome
lambda	lambda in the lasso
lasso	An glmnet object
family	A description of the error distribution and link function to be used in the model. It can take the value of <code>{gaussian}</code> , <code>{binomial}</code> , <code>{poisson}</code> , and <code>{cox}</code> .

Value

A vector that contains the point estimates, confidence intervals and the null bound

get.var	get.var: <i>Get indices</i>
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Description

Get the indices of the variables selected by the algorithm

Usage

```
get.var(candidate.index, xs, ys, family, gvif)
```

Arguments

candidate.index	Indices of the candidate set
xs	Standardized independent variables
ys	Standardized dependent variable
family	A description of the error distribution and link function to be used in the model. It can take the value of <code>{gaussian}</code> , <code>{binomial}</code> , <code>{poisson}</code> , and <code>{cox}</code> .
gvif	A logical operator indicating whether a generalized variance inflation factor-adjusted null bound is used. Default is FALSE.

Value

A list of following components:

- out.sgpv** A vector of indices of selected variables
- null.bound.p** Null bound in the SGPV screening
- pe** Point estimates in the candidate set
- lb** Lower bounds of effect estimates in the candidate set
- ub** Upper bounds of effect estimates in the candidate set

gvif	gvif: <i>Get GVIF for each variable</i>
------	-----------------------------------------

Description

Get generalized variance inflation factor (GVIF) for each variable. See Fox (1992) doi: [10.1080/01621459.1992.10475190](https://doi.org/10.1080/01621459.1992.10475190) for more details on how to calculate GVIF.

Usage

```
gvif(mod, family)
```

Arguments

mod	A model object with at least two explanatory variables
family	A description of the error distribution and link function to be used in the model. It can take the value of <code>{gaussian}</code> , <code>{binomial}</code> , <code>{poisson}</code> , and <code>{cox}</code> .

Value

A vector of GVIF for each variable in the model

plot.sgpv	plot.sgpv: <i>Plot variable selection results</i>
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Description

S3 method plot for an object of class sgpv. When the two-stage algorithm is used, this function plots the fully relaxed lasso solution path on the standardized scale and the final variable selection results. When the one-stage algorithm is used, a histogram of all coefficients with selected effects is shown.

Usage

```
## S3 method for class 'sgpv'
plot(x, lpv = 3, lambda.max = NULL, short.label = T, ...)
```

Arguments

x	An sgpv object
lpv	Lines per variable. It can take the value of 1 meaning that only the bound that is closest to the null will be plotted, or the value of 3 meaning that point estimates as well as 95% confidence interval will be plotted. Default is 3.
lambda.max	The maximum lambda on the plot. Default is NULL.
short.label	An indicator if a short label is used for each variable for better visualization. Default is TRUE
...	Other plot arguments

Examples

```
# prepare the data
x <- t.housing[, -ncol(t.housing)]
y <- t.housing$V9

# one-stage algorithm
out.sgpv.1 <- pro.sgpv(x = x, y = y, stage = 1)
```

```

# plot the selection result

plot(out.sgpv.1)

# two-stage algorithm
out.sgpv.2 <- pro.sgpv(x = x, y = y)

# plot the fully relaxed lasso solution path and final solution
plot(out.sgpv.2)

# zoom in a little bit
plot(out.sgpv.2, lambda.max = 0.01)

# only plot one confidence bound
plot(out.sgpv.2, lpv = 1, lambda.max = 0.01)

```

predict.sgpv

predict.sgpv: *Prediction using the fitted model*

Description

S3 method predict for an object of class sgpv

Usage

```

## S3 method for class 'sgpv'
predict(object, newdata, type, ...)

```

Arguments

object	An sgpv objectect
newdata	Prediction data set
type	The type of prediction required. Can take the value of link, response, and terms. Default is response.
...	Other predict arguments

Value

Predicted values

Examples

```

# prepare the data
x <- t.housing[, -ncol(t.housing)]
y <- t.housing$V9

```



```
# run one-stage algorithm
out.sgpv <- pro.sgpv(x = x, y = y)

predict(out.sgpv)
```

<code>print.sgpv</code>	<code>print.sgpv</code> : <i>Print variable selection results</i>
-------------------------	-------------------------------------------------------------------

Description

S3 method `print` for an S3 object of class `sgpv`

Usage

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

Arguments

<code>x</code>	An <code>sgpv</code> object
<code>...</code>	Other <code>print</code> arguments

Value

Variable selection results

Examples

```
# prepare the data
x <- t.housing[, -ncol(t.housing)]
y <- t.housing$V9

# run one-stage algorithm
out.sgpv.1 <- pro.sgpv(x = x, y = y, stage = 1)

out.sgpv.1
```

 pro.sgpv

 pro.sgpv *function*

Description

This function outputs the variable selection results from either one-stage algorithm or two-stage algorithm.

Usage

```
pro.sgpv(
  x,
  y,
  stage = c(1, 2),
  family = c("gaussian", "binomial", "poisson", "cox"),
  gvif = F
)
```

Arguments

<code>x</code>	Independent variables, can be a matrix or a data.frame
<code>y</code>	Dependent variable, can be a vector or a column from a data.frame
<code>stage</code>	Algorithm indicator. 1 denotes the one-stage algorithm and 2 denotes the two-stage algorithm. Default is 2. When n is less than p, only the two-stage algorithm is available.
<code>family</code>	A description of the error distribution and link function to be used in the model. It can take the value of <code>{gaussian}</code> , <code>{binomial}</code> , <code>{poisson}</code> , and <code>{cox}</code> . Default is <code>{gaussian}</code>
<code>gvif</code>	A logical operator indicating whether a generalized variance inflation factor-adjusted null bound is used. Default is FALSE. See Fox (1992) doi: 10.1080/01621459.1992.10475190 for more details on how to calculate GVIF

Value

A list of following components:

var.index A vector of indices of selected variables

var.label A vector of labels of selected variables

lambda lambda selected by generalized information criterion in the two-stage algorithm. NULL for the one-stage algorithm

x Input data x

y Input data y

family family from the input

stage stage from the input

null.bound Null bound in the SGPV screening

pe.can Point estimates in the candidate set

lb.can Lower bounds of CI in the candidate set

ub.can Upper bounds of CI in the candidate set

See Also

- `print.sgpv()` prints the variable selection results
- `coef.sgpv()` extracts coefficient estimates
- `summary.sgpv()` summarizes the OLS outputs
- `predict.sgpv()` predicts the outcome
- `plot.sgpv()` plots variable selection results

Examples

```
# prepare the data
x <- t.housing[, -ncol(t.housing)]
y <- t.housing$V9

# run ProSGPV in linear regression
out.sgpv <- pro.sgpv(x = x, y = y)

# More examples at https://github.com/zuoyi93/ProSGPV/tree/master/vignettes
```

spine

Spine data

Description

Lower back pain can be caused by a variety of problems with any parts of the complex, interconnected network of spinal muscles, nerves, bones, discs or tendons in the lumbar spine. This dataset contains 12 biomechanical attributes from 310 patients, of whom 100 are normal and 210 are abnormal (Disk Hernia or Spondylolisthesis). The goal is to differentiate the normal patients from the abnormal using those 12 variables.

Usage

spine

Format

pelvic_incidence pelvic incidence
pelvic_tilt pelvic tilt
lumbar_lordosis_angle lumbar lordosis angle
sacral_slope sacral slope
pelvic_radius pelvic radius
degree_spondylolisthesis degree of spondylolisthesis
pelvic_slope pelvic slope
direct_tilt direct tilt
thoracic_slope thoracic slope
cervical_tilt cervical tilt
sacrum_angle sacrum angle
scoliosis_slope scoliosis slope
outcome 1 is abnormal (Disk Hernia or Spondylolisthesis) and 0 is normal

Source

<http://archive.ics.uci.edu/ml/datasets/vertebral+column>

summary.sgpv

summary.sgpv: *Summary of the final model*

Description

S3 method summary for an S3 object of class sgpv

Usage

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

Arguments

object An sgpv object
 ... Other arguments

Value

Summary of a model

Examples

```
# prepare the data
x <- t.housing[, -ncol(t.housing)]
y <- t.housing$V9

# run one-stage algorithm
out.sgpv <- pro.sgpv(x = x, y = y)

# get regression summary
summary(out.sgpv)
```

t.housing	<i>Tehran housing data</i>
-----------	----------------------------

Description

A dataset containing Tehran housing data. The data set has 372 observations. There are 26 explanatory variables at baseline, including 7 project physical and financial features (V2-V8) and 19 economic variables and indices (V11-V29). The outcome (V9) is the sales price of a real estate single-family residential apartment.

Usage

```
t.housing
```

Format

- V9** Actual sales price
- V2** Total floor area of the building
- V3** Lot area
- V4** Total Preliminary estimated construction cost based on the prices at the beginning of the project
- V5** Preliminary estimated construction cost based on the prices at the beginning of the project
- V6** Equivalent preliminary estimated construction cost based on the prices at the beginning of the project in a selected base year
- V7** Duration of construction
- V8** Price of the unit at the beginning of the project per square meter
- V11** The number of building permits issued
- V12** Building services index for preselected base year
- V13** Wholesale price index of building materials for the base year
- V14** Total floor areas of building permits issued by the city/municipality
- V15** Cumulative liquidity
- V16** Private sector investment in new buildings

- V17** Land price index for the base year
- V18** The number of loans extended by banks in a time resolution
- V19** The amount of loans extended by banks in a time resolution
- V20** The interest rate for loan in a time resolution
- V21** The average construction cost by private sector at the completion of construction
- V22** The average cost of buildings by private sector at the beginning of construction
- V23** Official exchange rate with respect to dollars
- V24** Nonofficial (street market) exchange rate with respect to dollars
- V25** Consumer price index (CPI) in the base year
- V26** CPI of housing, water, fuel & power in the base year
- V27** Stock market index
- V28** Population of the city
- V29** Gold price per ounce

Source

<http://archive.ics.uci.edu/ml/datasets/Residential+Building+Data+Set>

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