Package 'pros'

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Title Penalized Regression on Steroids

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Description This is a project for STAT8053 at the University of Minnesota. Please see the README on github for compiler flags to improve performance.

Depends R (>= 3.5.0)

SystemRequirements C++11

License GPL-2

Encoding UTF-8

LazyData true

RoxygenNote 6.1.1

R topics documented:

`cv.pros` *Cross-validation*

Description

The `cv.pros` function is used for K-fold cross-validation.

Usage

```
cv.pros(X, y, K_fold = 10, alpha = c(1, 0, 0, 0, 0, 0),
  lambdas = c(), step_size, algorithm = "proximal_gradient_cd",
  max_iter = 10000, tolerance = 10^(-8), random_seed = 0)
```

Arguments

X	is an $n \times m$-dimensional matrix of the data.
y	is an $n \times m$-dimensional matrix of the data.
K_fold	is the number of folds in cross-validation.
alpha	is a 6-dimensional vector of the convex combination corresponding to the penalization: • α_1 is the l^1 penalty. • α_2 is the l^2 penalty. • α_3 is the l^4 penalty. • α_4 is the l^6 penalty. • α_5 is the l^8 penalty. • α_6 is the l^{10} penalty.
lambdas	is a vector of dual penalization values to be evaluated.
step_size	is a tuning parameter defining the step size. Larger values are more aggressive and smaller values are less aggressive.
algorithm	is the optimization algorithm • proximal_gradient_cd uses proximal gradient coordinate descent. • subgradient_cd uses subgradient coordinate descent.
max_iter	is the maximum iterations the algorithm will run regardless of convergence.
tolerance	is the accuracy of the stopping criterion.
random_seed	is the random seed used in the algorithms.

Value

A class cv_pros

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predict.cv_pros *Cross-validation Prediction*

Description

The prediction function for cv.pros.

Usage

```
## S3 method for class 'cv_pros'
predict(object, X_new, ...)
```

Arguments

object	an object of class cv_pros
X_new	is an $n \times m$-dimensional matrix of the data.
...	Other parameters (this is required by R)

Value

A vector of prediction values.

References

Zou, Hui. “Regularization and variable selection via the elastic net.” (2004).

Examples

```
n = 1000
X1 <- rnorm(n)
X2 <- rnorm(n)
y = 3 + 5 * X1 + 0 * X2
X = matrix(c(X1, X2), ncol = 2)
cv = cv.pros(X, y, step_size = .001)
predict(cv, X)
```

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predict.pros *Pros Prediction*

Description

The prediction function for pros.

Usage

```
## S3 method for class 'pros'
predict(object, X, ...)
```

Arguments

object	an object of class pros
X	is an $n \times m$-dimensional matrix of the data.
...	Other parameters (this is required by R)

Value

A vector of prediction values.

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pros *Pros*

Description

The pros function is used to fit a single regression model with a specified penalization.

Usage

```
pros(X, y, alpha = c(1, 0, 0, 0, 0, 0), lambda, step_size,
  algorithm = "proximal_gradient_cd", max_iter = 10000,
  tolerance = 10^(-8), random_seed = 0)
```

Arguments

`X`	is an $n \times m$-dimensional matrix of the data.
`y`	is an $n \times m$-dimensional matrix of the data.
`alpha`	is a 6-dimensional vector of the convex combination corresponding to the penalization: • α_1 is the l^1 penalty. • α_2 is the l^2 penalty. • α_3 is the l^4 penalty. • α_4 is the l^6 penalty. • α_5 is the l^8 penalty. • α_6 is the l^{10} penalty.
`lambda`	is the Lagrangian dual penalization parameter.
`step_size`	is a tuning parameter defining the step size. Larger values are more aggressive and smaller values are less aggressive.
`algorithm`	is the optimization algorithm • `proximal_gradient_cd` uses proximal gradient coordinate descent. • `subgradient_cd` uses subgradient coordinate descent.
`max_iter`	is the maximum iterations the algorithm will run regardless of convergence.
`tolerance`	is the accuracy of the stopping criterion.
`random_seed`	is the random seed used in the algorithms.

Value

A class `pros`

References

Zou, Hui. "Regularization and variable selection via the elastic net." (2004).

Examples

```
n = 1000
X1 <- rnorm(n)
X2 <- rnorm(n)
y = 3 + 5 * X1 + 0 * X2
X = matrix(c(X1, X2), ncol = 2)
fit = pros(X, y, lambda = 2, step_size = .001)
predict(fit, X)
```

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