

# Package ‘splitSelect’

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**Type** Package

**Title** Best Split Selection Modeling for Low-Dimensional Data

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**Description** Functions to generate or sample from all possible splits of features or variables into a number of specified groups. Also computes the best split selection estimator (for low-dimensional data) as defined in Christidis, Van Aelst and Zamar (2019) <[arXiv:1812.05678](https://arxiv.org/abs/1812.05678)>.

**License** GPL (>= 2)

**Biarch** true

**Imports** multicool, glmnet, parallel, doParallel, foreach

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coef.cv.splitSelect    *Coefficients for splitSelect object*

---

## Description

coef.cv.splitSelect returns the coefficients for a cv.splitSelect for new data.

## Usage

```
## S3 method for class 'cv.splitSelect'
coef(object, optimal.only = TRUE, ...)
```

## Arguments

object	An object of class cv.splitSelect.
optimal.only	A boolean variable (TRUE default) to indicate if only the coefficient of the optimal split are returned.
...	Additional arguments for compatibility.

## Value

A matrix with the coefficients of the cv.splitSelect object.

## Author(s)

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

## See Also

[splitSelect](#)

## Examples

```
# Setting the parameters
p <- 4
n <- 30
n.test <- 5000
beta <- rep(5,4)
rho <- 0.1
r <- 0.9
SNR <- 3
# Creating the target matrix with "kernel" set to rho
target_cor <- function(r, p){
```

```

Gamma <- diag(p)
for(i in 1:(p-1)){
  for(j in (i+1):p){
    Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
  }
}
return(Gamma)
}
# AR Correlation Structure
Sigma.r <- target_cor(r, p)
Sigma.rho <- target_cor(rho, p)
sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
# Simulate some data
x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)

# Generating the coefficients for a fixed split

split.out <- cv.splitSelect(x.train, y.train, G=2, use.all=TRUE,
                           fix.partition=list(matrix(c(2,2),
                                                       ncol=2, byrow=TRUE)),
                           fix.split=NULL,
                           intercept=TRUE, group.model="glmnet", alphas=0, nfolds=10)

coef(split.out)

```

---

coef.splitSelect	<i>Coefficients for splitSelect object</i>
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---

## Description

coef.splitSelect returns the coefficients for a splitSelect object.

## Usage

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

## Arguments

object	An object of class splitSelect.
...	Additional arguments for compatibility.

## Value

A matrix with the coefficients of the splitSelect object.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**See Also**

[splitSelect](#)

**Examples**

```
# Setting the parameters
p <- 4
n <- 30
n.test <- 5000
beta <- rep(5,4)
rho <- 0.1
r <- 0.9
SNR <- 3
# Creating the target matrix with "kernel" set to rho
target_cor <- function(r, p){
  Gamma <- diag(p)
  for(i in 1:(p-1)){
    for(j in (i+1):p){
      Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
    }
  }
  return(Gamma)
}
# AR Correlation Structure
Sigma.r <- target_cor(r, p)
Sigma.rho <- target_cor(rho, p)
sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
# Simulate some data
x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)

# Generating the coefficients for a fixed partition of the variables

split.out <- splitSelect(x.train, y.train, G=2, use.all=TRUE,
  fix.partition=list(matrix(c(2,2), ncol=2, byrow=TRUE)), fix.split=NULL,
  intercept=TRUE, group.model="glmnet", alphas=0)

coef(split.out)
```

**Description**

cv.splitSelect performs the best split selection algorithm with cross-validation

**Usage**

```
cv.splitSelect(
  x,
  y,
  intercept = TRUE,
  G,
  use.all = TRUE,
  family = c("gaussian", "binomial")[1],
  group.model = c("glmnet", "LS", "Logistic")[1],
  alphas = 0,
  nsample = NULL,
  fix.partition = NULL,
  fix.split = NULL,
  nfolds = 10,
  parallel = FALSE,
  cores = getOption("mc.cores", 2L)
)
```

**Arguments**

x	Design matrix.
y	Response vector.
intercept	Boolean variable to determine if there is intercept (default is TRUE) or not.
G	Number of groups into which the variables are split. Can have more than one value.
use.all	Boolean variable to determine if all variables must be used (default is TRUE).
family	Description of the error distribution and link function to be used for the model. Must be one of "gaussian" or "binomial".
group.model	Model used for the groups. Must be one of "glmnet" or "LS".
alphas	Elastic net mixing parameter. Should be between 0 (default) and 1.
nsample	Number of sample splits for each value of G. If NULL, then all splits will be considered (unless there is overflow).
fix.partition	Optional list with G elements indicating the partitions (in each row) to be considered for the splits.
fix.split	Optional matrix with p columns indicating the groups (in each row) to be considered for the splits.
nfolds	Number of folds for the cross-validation procedure.
parallel	Boolean variable to determine if parallelization of the function. Default is FALSE.
cores	Number of cores for the parallelization for the function.

**Value**

An object of class cv.splitSelect.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**See Also**

[coef.cv.splitSelect](#), [predict.cv.splitSelect](#)

**Examples**

```
# Setting the parameters
p <- 4
n <- 30
n.test <- 5000
beta <- rep(5,4)
rho <- 0.1
r <- 0.9
SNR <- 3
# Creating the target matrix with "kernel" set to rho
target_cor <- function(r, p){
  Gamma <- diag(p)
  for(i in 1:(p-1)){
    for(j in (i+1):p){
      Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
    }
  }
  return(Gamma)
}
# AR Correlation Structure
Sigma.r <- target_cor(r, p)
Sigma.rho <- target_cor(rho, p)
sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
# Simulate some data
x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)

# Generating the coefficients for a fixed partition of the variables

split.out <- cv.splitSelect(x.train, y.train, G=2, use.all=TRUE,
  fix.partition=list(matrix(c(2,2),
    ncol=2, byrow=TRUE)),
  fix.split=NULL,
  intercept=TRUE, group.model="glmnet", alphas=0, nfolds=10)
```

---

generate\_partitions    *Generate Splits Partitions Possibilities*

---

### Description

generate\_partitions returns a matrix with the number of possible objects in each group using splits.

### Usage

```
generate_partitions(p, G, use.all = TRUE)
```

### Arguments

p	Number of variables or objects to split.
G	Number of groups into which the variables are split.
use.all	Boolean variable to determine if all variables must be used (default is TRUE).

### Value

A matrix or list with the number of possible objects in each group using splits.

### Author(s)

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

### Examples

```
# Generating the possible split partitions of 6 variables in 3 groups
# Using all the variables
split.3groups.all <- generate_partitions(6, 3)
split.3groups.all
# Without using all the variables
split.3groups <- generate_partitions(6, 3, use.all=FALSE)
split.3groups
```

---

generate\_splits      *Generate Splits Possibilities*

---

### Description

generate\_splits returns a matrix with the different splits of the variables in each row.

### Usage

```
generate_splits(p, G, use.all = TRUE, fix.partition = NULL, verbose = TRUE)
```

### Arguments

p	Number of variables or objects to split.
G	Number of groups into which the variables are split.
use.all	Boolean variable to determine if all variables must be used (default is TRUE).
fix.partition	Optional matrix with G columns (or list if more than one value of G) indicating the partitions (in each row) to be considered for the splits.
verbose	Boolean variable to determine if console output for cross-validation progress is printed (default is TRUE).

### Value

A matrix with the different splits of the variables in the groups.

### Author(s)

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

### Examples

```
# Generating the possible splits of 6 variables in 3 groups
# Using all the variables
split.3groups.all <- generate_splits(6, 3)
split.3groups.all
# Without using all the variables
split.3groups <- generate_splits(6, 3, use.all=FALSE)
split.3groups
```



---

nsplit	<i>Compute Total Number of Possible Splits</i>
--------	--

---

**Description**

nsplits returns the total number of possible splits of variables into groups.

**Usage**

```
nsplit(p, G, use.all = TRUE, fix.partition = NULL)
```

**Arguments**

p	Number of variables or objects to split.
G	Number of groups into which the variables are split.
use.all	Boolean variable to determine if all variables must be used (default is TRUE).
fix.partition	Optional matrix with G columns (or list if more than one value of G) indicating the partitions (in each row) to be considered for the splits.

**Value**

A numeric vector with the total number of possible splits.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**Examples**

```
# Compute the total number of possible splits of 6 variables into 3 groups
# We use all the variables
out.n.splits.all <- nsplit(p=6, G=3, use.all=TRUE)
out.n.splits.all
# We don't enforce using all the variables
out.n.splits <- nsplit(p=6, G=3, use.all=FALSE)
out.n.splits
```

---

`predict.cv.splitSelect`*Predictions for cv.splitSelect object*

---

**Description**

`predict.cv.splitSelect` returns the prediction for `cv.splitSelect` for new data.

**Usage**

```
## S3 method for class 'cv.splitSelect'  
predict(object, newx, optimal.only = TRUE, ...)
```

**Arguments**

<code>object</code>	An object of class <code>cv.splitSelect</code> .
<code>newx</code>	A matrix with the new data.
<code>optimal.only</code>	A boolean variable (TRUE default) to indicate if only the predictions of the optimal split are returned.
<code>...</code>	Additional arguments for compatibility.

**Value**

A matrix with the predictions of the `cv.splitSelect` object.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**See Also**

[cv.splitSelect](#)

**Examples**

```
# Setting the parameters  
p <- 4  
n <- 30  
n.test <- 5000  
beta <- rep(5,4)  
rho <- 0.1  
r <- 0.9  
SNR <- 3  
# Creating the target matrix with "kernel" set to rho  
target_cor <- function(r, p){  
  Gamma <- diag(p)  
  for(i in 1:(p-1)){  
    for(j in (i+1):p){
```

```

        Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
      }
    }
    return(Gamma)
  }
  # AR Correlation Structure
  Sigma.r <- target_cor(r, p)
  Sigma.rho <- target_cor(rho, p)
  sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
  # Simulate some data
  x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
  y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)
  x.test <- mvnfast::rmvn(n.test, mu=rep(0,p), sigma=Sigma.rho)
  y.test <- 1 + x.test %*% beta + rnorm(n.test, sd=sigma.epsilon)

  # Generating the coefficients for a fixed split

  split.out <- cv.splitSelect(x.train, y.train, G=2, use.all=TRUE,
                             fix.partition=list(matrix(c(2,2),
                                                         ncol=2, byrow=TRUE)),
                             fix.split=NULL,
                             intercept=TRUE, group.model="glmnet", alphas=0)
  predict(split.out, newx=x.test)

```

---

predict.splitSelect    *Predictions for splitSelect object*

---

## Description

predict.splitSelect returns the prediction for splitSelect for new data.

## Usage

```
## S3 method for class 'splitSelect'
predict(object, newx, ...)
```

## Arguments

object	An object of class splitSelect.
newx	A matrix with the new data.
...	Additional arguments for compatibility.

## Value

A matrix with the predictions of the splitSelect object.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**See Also**

[splitSelect](#)

**Examples**

```
# Setting the parameters
p <- 4
n <- 30
n.test <- 5000
beta <- rep(5,4)
rho <- 0.1
r <- 0.9
SNR <- 3
# Creating the target matrix with "kernel" set to rho
target_cor <- function(r, p){
  Gamma <- diag(p)
  for(i in 1:(p-1)){
    for(j in (i+1):p){
      Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
    }
  }
  return(Gamma)
}
# AR Correlation Structure
Sigma.r <- target_cor(r, p)
Sigma.rho <- target_cor(rho, p)
sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
# Simulate some data
x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)
x.test <- mvnfast::rmvn(n.test, mu=rep(0,p), sigma=Sigma.rho)
y.test <- 1 + x.test %*% beta + rnorm(n.test, sd=sigma.epsilon)

# Generating the coefficients for a fixed split

split.out <- splitSelect(x.train, y.train, G=2, use.all=TRUE,
  fix.partition=list(matrix(c(2,2), ncol=2, byrow=TRUE)), fix.split=NULL,
  intercept=TRUE, group.model="glmnet", alphas=0)
predict(split.out, newx=x.test)
```

**Description**

rsplit returns a matrix with random splits of the variables in groups.

**Usage**

```
rsplit(n, p, G, use.all = TRUE, fix.partition = NULL, verbose = TRUE)
```

**Arguments**

n	Number of sample splits.
p	Number of variables or objects to split.
G	Number of groups into which the variables are split.
use.all	Boolean variable to determine if all variables must be used (default is TRUE).
fix.partition	Optional matrix with G columns indicating the partitions (in each row) to be considered for the splits.
verbose	Boolean variable to determine if console output for cross-validation progress is printed (default is TRUE).

**Value**

A matrix or list with the number of possible objects in each group using splits.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**Examples**

```
# Generating sample splits of 6 variables in 3 groups
# Using all the variables
random.splits <- rsplit(100, 6, 3)
# Using fixed partitions
random.splits.fixed <- rsplit(100, 6, 3, fix.partition=matrix(c(2,2,2), nrow=1))
```

**Description**

splitSelect performs the best split selection algorithm.

**Usage**

```

splitSelect(
  x,
  y,
  intercept = TRUE,
  G,
  use.all = TRUE,
  family = c("gaussian", "binomial")[1],
  group.model = c("glmnet", "LS", "Logistic")[1],
  lambdas = NULL,
  alphas = 0,
  nsample = NULL,
  fix.partition = NULL,
  fix.split = NULL,
  parallel = FALSE,
  cores = getOption("mc.cores", 2L),
  verbose = TRUE
)

```

**Arguments**

x	Design matrix.
y	Response vector.
intercept	Boolean variable to determine if there is intercept (default is TRUE) or not.
G	Number of groups into which the variables are split. Can have more than one value.
use.all	Boolean variable to determine if all variables must be used (default is TRUE).
family	Description of the error distribution and link function to be used for the model. Must be one of "gaussian" or "binomial".
group.model	Model used for the groups. Must be one of "glmnet" or "LS".
lambdas	The shrinkage parameters for the "glmnet" regularization. If NULL (default), optimal values are chosen.
alphas	Elastic net mixing parameter. Should be between 0 (default) and 1.
nsample	Number of sample splits for each value of G. If NULL, then all splits will be considered (unless there is overflow).
fix.partition	Optional list with G elements indicating the partitions (in each row) to be considered for the splits.
fix.split	Optional matrix with p columns indicating the groups (in each row) to be considered for the splits.
parallel	Boolean variable to determine if parallelization of the function. Default is FALSE.
cores	Number of cores for the parallelization for the function.
verbose	Boolean variable to determine if console output for cross-validation progress is printed (default is TRUE).

**Value**

An object of class splitSelect.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>

**See Also**

[coef.splitSelect](#), [predict.splitSelect](#)

**Examples**

```
# Setting the parameters
p <- 4
n <- 30
n.test <- 5000
beta <- rep(5,4)
rho <- 0.1
r <- 0.9
SNR <- 3
# Creating the target matrix with "kernel" set to rho
target_cor <- function(r, p){
  Gamma <- diag(p)
  for(i in 1:(p-1)){
    for(j in (i+1):p){
      Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
    }
  }
  return(Gamma)
}
# AR Correlation Structure
Sigma.r <- target_cor(r, p)
Sigma.rho <- target_cor(rho, p)
sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
# Simulate some data
x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)

# Generating the coefficients for a fixed partition of the variables

split.out <- splitSelect(x.train, y.train, G=2, use.all=TRUE,
                        fix.partition=list(matrix(c(2,2),
                                                  ncol=2, byrow=TRUE)),
                        fix.split=NULL,
                        intercept=TRUE, group.model="glmnet", alphas=0)
```

---

splitSelect\_coef      *Split Selection for Regression - Coefficients Generation*

---

**Description**

splitSelect\_coef generates the coefficients for a particular split of variables into groups.

**Usage**

```
splitSelect_coef(  
  x,  
  y,  
  variables.split,  
  intercept = TRUE,  
  family = c("gaussian", "binomial")[1],  
  group.model = c("glmnet", "LS", "Logistic")[1],  
  lambdas = NULL,  
  alphas = 0  
)
```

**Arguments**

x	Design matrix.
y	Response vector.
variables.split	A vector with the split of the variables into groups as values.
intercept	Boolean variable to determine if there is intercept (default is TRUE) or not.
family	Description of the error distribution and link function to be used for the model. Must be one of "gaussian" or "binomial".
group.model	Model used for the groups. Must be one of "glmnet" or "LS".
lambdas	The shrinkage parameters for the "glmnet" regularization. If NULL (default), optimal values are chosen.
alphas	Elastic net mixing parameter. Should be between 0 (default) and 1.

**Value**

A vector with the regression coefficients for the split.

**Author(s)**

Anthony-Alexander Christidis, <anthony.christidis@stat.ubc.ca>



**Examples**

```

# Setting the parameters
p <- 6
n <- 30
n.test <- 5000
group.beta <- -3
beta <- c(rep(1, 2), rep(group.beta, p-2))
rho <- 0.1
r <- 0.9
SNR <- 3
# Creating the target matrix with "kernel" set to rho
target_cor <- function(r, p){
  Gamma <- diag(p)
  for(i in 1:(p-1)){
    for(j in (i+1):p){
      Gamma[i,j] <- Gamma[j,i] <- r^(abs(i-j))
    }
  }
  return(Gamma)
}
# AR Correlation Structure
Sigma.r <- target_cor(r, p)
Sigma.rho <- target_cor(rho, p)
sigma.epsilon <- as.numeric(sqrt((t(beta) %*% Sigma.rho %*% beta)/SNR))
# Simulate some data
x.train <- mvnfast::rmvn(30, mu=rep(0,p), sigma=Sigma.r)
y.train <- 1 + x.train %*% beta + rnorm(n=n, mean=0, sd=sigma.epsilon)
x.test <- mvnfast::rmvn(n.test, mu=rep(0,p), sigma=Sigma.rho)
y.test <- 1 + x.test %*% beta + rnorm(n.test, sd=sigma.epsilon)

# Generating the coefficients for a fixed split
splitSelect_coef(x.train, y.train, variables.split=matrix(c(1,2,1,2,1,2), nrow=1))

```

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