

Package ‘AHM’

January 20, 2025

Type Package

Title Additive Heredity Model: Method for the Mixture-of-Mixtures Experiments

Version 1.0.1

Maintainer Sumin Shen <sumin@vt.edu>

Description An implementation of the additive heredity model for the mixture-of-mixtures experiments of Shen et al. (2019) in Technometrics <doi:10.1080/00401706.2019.1630010>. The additive heredity model considers an additive structure to inherently connect the major components with the minor components. The additive heredity model has a meaningful interpretation for the estimated model because of the hierarchical and heredity principles applied and the nonnegative garrote technique used for variable selection.

License GPL-3

Encoding UTF-8

LazyData true

Imports mixexp, plgp, devtools, dplyr, tibble, tidyr, Matrix

Depends R (>= 2.10), quadprog, glmnet

Suggests knitr, rmarkdown, partitions

VignetteBuilder knitr

BuildVignettes yes

RoxygenNote 6.0.1

NeedsCompilation no

Author Sumin Shen [aut, cre],
Lulu Kang [aut],
Xinwei Deng [aut]

Repository CRAN

Date/Publication 2019-07-28 09:00:05 UTC

Contents

ahm	2
check_col_correlation	3
coating	4
coef.ahm	4
coef.cv.ahm	5
compute_aicc	6
cv.ahm	6
design_simplex_centroid_design_3_major_component	8
enlist	8
expand_interactions	9
find_condition_num	9
mapping_function	10
mymaximin	10
predict.ahm	12
predict.cv.ahm	13
pringles_candidates2search	14
pringles_fat	14
pringles_hardness	15
summary.ahm	15
Index	17

ahm	<i>This is one of the main functions. The function ahm computes the proposed additive heredity model.</i>
-----	-----------------------------------------------------------------------------------------------------------

Description

This is one of the main functions. The function ahm computes the proposed additive heredity model.

Usage

```
ahm(y, x, num_major = 3, dist_minor = c(2, 2, 1), type = "weak",
    alpha = 0, lambda_seq = seq(0, 5, 0.01), nfold = NULL,
    mapping_type = c("power"), powerh = 0, rep_gcv = 100)
```

Arguments

y	numeric vector
x	data.frame Note the column names of the x should be in the order of major components, minor components, and no interactions are needed.
num_major	number of major components
dist_minor	the allocation of number of minor components nested under major components
type	heredity type, weak heredity is the current support type
alpha	0 is for the ridge in glmnet https://web.stanford.edu/~hastie/glmnet/glmnet_alpha.html

lambda_seq	a numeric vector for the options of lambda used in ridge regression for estimating the initials
nfolds	used in cv.glmnet for initial value of parameters in the non-negative garrote method
mapping_type	the form of the coefficient function of major components in front of corresponding minor terms. Currently only support "power"
powerh	the power parameter used for the power function
rep_gcv	the number of choices of tuning parameter used in the GCV selection

Value

Return a list

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
out = ahm (y, x, num_major = 3, dist_minor = c(2,2,1),
          type = "weak", alpha=0, lambda_seq=seq(0,5,0.01), nfold = NULL,
          mapping_type = c("power"), powerh = h_tmp,
          rep_gcv=100)
summary(out)
```

check_col_correlation *Check column correlations*

Description

Check column correlations

Usage

```
check_col_correlation(dat)
```

Arguments

dat data.frame

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
check_col_correlation (dat=x)
```

coating	<i>Photoresist-coating experiment data</i>
---------	--------------------------------------------

Description

Photoresist-coating experiment data

Usage

```
data(coating)
```

Format

data.frame

References

Cornell, J.A. and Ramsey, P.J. (1998). A Generalized mixture model for categorized-components problems with an application to a photoresist-coating experiment. *Technometrics*, 40(1), 48-61. ([tandfonline](#))

Examples

```
data(coating)
print(coating)
```

coef.ahm	<i>Coefficient method for the fitted ahm object</i>
----------	-----------------------------------------------------

Description

Coefficient method for the fitted ahm object

Usage

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

Arguments

object	ahm object
...	not used

Value

a numerical vector

Examples

```

data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
out = ahm (y, x, num_major = 3, dist_minor = c(2,2,1),
          type = "weak", alpha=0, lambda_seq=seq(0,5,0.01), nfold = NULL,
          mapping_type = c("power"), powerh = h_tmp,
          rep_gcv=100)
coef(out)

```

coef.cv.ahm

Coefficient method for the fitted cv.ahm object

Description

Coefficient method for the fitted cv.ahm object

Usage

```

## S3 method for class 'cv.ahm'
coef(object, metric = "mse", ...)

```

Arguments

object	cv.ahm object
metric	"mse" or "aicc"
...	not used

Value

a numerical vector

Examples

```

data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
powerh_path = round(seq(0.001,2,length.out =15),3)
num_major = 3; dist_minor = c(2,2,1)
res = cv.ahm (y, x, powerh_path=powerh_path, metric = "mse", num_major, dist_minor, type = "weak"
, alpha=0, lambda_seq=seq(0,5,0.01), nfold=NULL, mapping_type = c("power"), rep_gcv=100)
coefficients = coef(res)

```

 compute_aicc

compute AICc

Description

compute AICc

Usage

```
compute_aicc(rss, n, p, type = "AICc")
```

Arguments

rss	residual sum of squares
n	number of observation
p	number of nonzero parameters
type	character "AICc"

References

[Calculating AIC “by hand” in R in Stack Overflow](#)

Examples

```
compute_aicc (rss=10, n=30, p=6, type = "AICc")
```

 cv.ahm

This is one of the main functions. It perform the cross validation on ahm models to select the optimal setting of hyper parameter h

Description

This is one of the main functions. It perform the cross validation on ahm models to select the optimal setting of hyper parameter h

Usage

```
cv.ahm(y, x, powerh_path = NULL, metric = c("mse", "AICc"), num_major = 3,
  dist_minor = c(2, 2, 1), type = "weak", alpha = 0, lambda_seq = seq(0,
  5, 0.01), nfolds = NULL, mapping_type = c("power"), rep_gcv = 100)
```

Arguments

y	numeric vector
x	data.frame Note the column names of the x should be in the order of major components, minor components, and no interactions between major or minor components are needed.
powerh_path	if NULL, then the default is the vector: round(seq(0.001,2,length.out =15),3)
metric	"mse" or "AICc" the metric used in cross validation where the minimum is selected as the optimal
num_major	number of major components
dist_minor	the allocation of number of minor components nested under major components
type	heredity type, weak heredity is the current support type
alpha	0 is for the ridge in glmnet https://web.stanford.edu/~hastie/glmnet/glmnet_alpha.html
lambda_seq	a numeric vector for the options of lambda used in ridge regression for estimating the initials
nfolds	used in cv.glmnet for initial value of parameters in the non-negative garrote method
mapping_type	the form of the coefficient function of major components in front of corresponding minor terms. Currently only support "power"
rep_gcv	the number of choices of tuning parameter used in the GCV selection

Value

Return a list

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
powerh_path = round(seq(0.001,2,length.out =15),3)
num_major = 3; dist_minor = c(2,2,1)
res = cv.ahm(y, x, powerh_path=powerh_path, metric = "mse", num_major, dist_minor, type = "weak",
, alpha=0, lambda_seq=seq(0,5,0.01), nfolds=NULL, mapping_type = c("power"), rep_gcv=100)
object = res$metric_mse
```

```
design_simplex_centroid_design_3_major_component
```

Design points for the simplex centroid design with 3 components

Description

Design points for the simplex centroid design with 3 components

Usage

```
data(design_simplex_centroid_design_3_major_component)
```

Format

```
data.frame
```

Examples

```
data(design_simplex_centroid_design_3_major_component)
print(design_simplex_centroid_design_3_major_component)
```

```
enlist
```

Create a list

Description

Create a list

Usage

```
enlist(...)
```

Arguments

... object to be included as elements in the list

Examples

```
item = c(1:10)
enlist(item)
```

expand_interactions	<i>Expand the interaction terms for each subset group, say x11, x12, or c1, c2, c3</i>
---------------------	----------------------------------------------------------------------------------------

Description

Expand the interaction terms for each subset group, say x11, x12, or c1, c2, c3

Usage

```
expand_interactions(dat, sel_names)
```

Arguments

dat	data frame
sel_names	characters

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
expand_interactions (dat=x, sel_names=c("c1", "c2", "c3"))
```

find_condition_num	<i>Compute the conditional number of design matrix</i>
--------------------	--------------------------------------------------------

Description

Compute the conditional number of design matrix

Usage

```
find_condition_num(x)
```

Arguments

x	matrix to be used in svd
---	--------------------------

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
find_condition_num (x)
```

mapping_function	<i>Mapping_function is a function to add the functional coefficients of major components in front of minor components terms</i>
------------------	---------------------------------------------------------------------------------------------------------------------------------

Description

Mapping_function is a function to add the functional coefficients of major components in front of minor components terms

Usage

```
mapping_function(x, num_major = 3, dist_minor = C(2, 2, 1),
  mapping_type = c("power"), powerh = 0)
```

Arguments

x	data.frame Note the column names of the x should be in the order of major components, minor components, and no interactions are needed.
num_major	number of major components
dist_minor	the allocation of number of minor components nested under major components
mapping_type	the form of the coefficient function of major components in front of corresponding minor terms. Currently only support "power"
powerh	the power parameter used for the power function

Value

data frame

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
mapping_function(x=x, num_major=3, dist_minor=c(2,2,1), mapping_type = c("power"), powerh=0)
```

mymaximin	<i>The mymaximin function generates the matrix of maximin design points. It uses the simplex centroid design as the base design, then in a stochastic way sample the candidate design points generated by the function partition.</i>
-----------	---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------

Description

This method is modified based on Prof. Bobby Gramacy's Computer Experiment lecture at Virginia Tech. [Prof. Gramacy's lecture website](#)

Usage

```
mymaximin(pool, n = 50, m = 3, iter = 1e+05, Xorig = NULL)
```

Arguments

pool	partition the base design points provided to the function
n	numeric, sample size
m	numeric, 3 stands for 3 components, i.e. c1, c2, and c3
iter	numeric, iterations used in the stochastic sampling
Xorig	matrix, initial design points

Value

Return a matrix of the design points for the major components

Examples

```
# The case of unconstrained experiments
library(mixexp)
num_size = 8 # num points in the design for the major component
Xorig = as.matrix(SCD(3))
# all possible combinations sum to 1
pool_3d = partitions::compositions(1000, 3, include.zero = TRUE)/1000
res_C = mymaximin(pool=pool_3d, n=num_size, m=3, iter=1e5, Xorig=Xorig)
DesignPoints(res_C, cornerlabs = c("c3", "c2", "c1"), axislabs=c("c1", "c2", "c3"))

# The case of constrained experiments
library(mixexp)
num_size = 8 # num points in the design for the major component
# all possible combinations sum to 1
pool_3d = partitions::compositions(1000, 3, include.zero = TRUE)/1000
c1_min=0.2
c1_max=0.45
c2_min=0.4
c2_max=0.6
c3_min=0.1
c3_max=0.25
tmp = Xvert(nfac=3, lc=c(c1_min, c2_min, c3_min), uc=c(c1_max, c2_max, c3_max), ndm=1, pseudo=FALSE)
Xorig=tmp[c(1:6, 13), c(1:3)]
colnames(Xorig)=c("V1", "V2", "V3")
pool_3d = t(dplyr::filter(as.data.frame(t(as.matrix(pool_3d))), t(pool_3d)[, 1] > c1_min &
    t(pool_3d)[, 1] <= c1_max &
    t(pool_3d)[, 2] > c2_min &
    t(pool_3d)[, 2] <= c2_max &
    t(pool_3d)[, 3] > c3_min &
    t(pool_3d)[, 3] <= c3_max
)
)
res_C = mymaximin(pool=pool_3d, n=num_size, m=3, iter=1e5, Xorig=Xorig)
DesignPoints(res_C, cornerlabs = c("c3", "c2", "c1"), axislabs=c("c1", "c2", "c3"))
```

```
,x1lower=c1_min,x2lower=c2_min,x3lower=c3_min
,x1upper=c1_max,x2upper=c2_max,x3upper=c3_max, pseudo=FALSE)
```

predict.ahm

Predict method for the fitted ahm object

Description

Predict method for the fitted ahm object

Usage

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

Arguments

object	ahm object
newx	Matrix of new values for x at which predictions are to be made.
...	not used

Value

predicted value(s) at newx

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
out = ahm (y, x, num_major = 3, dist_minor = c(2,2,1),
          type = "weak", alpha=0, lambda_seq=seq(0,5,0.01), nfold = NULL,
          mapping_type = c("power"), powerh = h_tmp,
          rep_gcv=100)
predict(out)
```

predict.cv.ahm	<i>Predict method for the fitted cv.ahm object</i>
----------------	----------------------------------------------------

Description

Predict method for the fitted cv.ahm object

Usage

```
## S3 method for class 'cv.ahm'
predict(object, newx, metric = "mse", ...)
```

Arguments

object	cv.ahm object
newx	Matrix of new values for x at which predictions are to be made.
metric	"mse" or "aicc"
...	not used

Value

Return a list

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
powerh_path = round(seq(0.001, 2, length.out = 15), 3)
num_major = 3; dist_minor = c(2, 2, 1)
res = cv.ahm(y, x, powerh_path=powerh_path, metric = "mse", num_major, dist_minor, type = "weak",
  , alpha=0, lambda_seq=seq(0, 5, 0.01), nfold=5, mapping_type = c("power"), rep_gcv=100)
pred = predict(res)
```

```
pringles_candidates2search
```

The candidate search points in the nonlinear optimization for the optimal value in the Pringles experiment

Description

The candidate search points in the nonlinear optimization for the optimal value in the Pringles experiment

Usage

```
data(pringles_candidates2search)
```

Format

matrix

Examples

```
data(pringles_candidates2search)
print(pringles_candidates2search)
```

```
pringles_fat
```

Pringles experiment data set with the percent of Fat as the response

Description

Pringles experiment data set with the percent of Fat as the response

Usage

```
data(pringles_fat)
```

Format

data.frame

References

Kang, L., Joseph, V.R. and Brenneman, W.A. (2011). Design and modeling strategies for mixture-of-mixtures experiments. *Technometrics*, 53(2), 125–36. ([tandfonline](#))

Examples

```
data(pringles_fat)
print(pringles_fat)
```

pringles_hardness	<i>Pringles experiment data set with the Hardness as the response</i>
-------------------	-----------------------------------------------------------------------

Description

Pringles experiment data set with the Hardness as the response

Usage

```
data(pringles_hardness)
```

Format

```
data.frame
```

References

Kang, L., Joseph, V.R. and Brenneman, W.A. (2011). Design and modeling strategies for mixture-of-mixtures experiments. *Technometrics*, 53(2), 125–36. ([tandfonline](#))

Examples

```
data(pringles_hardness)
print(pringles_hardness)
```

summary.ahm	<i>Summary method for the fitted ahm object</i>
-------------	-------------------------------------------------

Description

Summary method for the fitted ahm object

Usage

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

Arguments

object	fitted ahm object
...	not used

Examples

```
data("pringles_fat")
data_fat = pringles_fat
h_tmp = 1.3
x = data_fat[,c("c1", "c2", "c3", "x11", "x12", "x21", "x22")]
y = data_fat[,1]
out = ahm (y, x, num_major = 3, dist_minor = c(2,2,1),
          type = "weak", alpha=0, lambda_seq=seq(0,5,0.01), nfold = NULL,
          mapping_type = c("power"), powerh = h_tmp,
          rep_gcv=100)
summary(out)
```


Index

* datasets

- coating, [4](#)
- design_simplex_centroid_design_3_major_component,
[8](#)
- pringles_candidates2search, [14](#)
- pringles_fat, [14](#)
- pringles_hardness, [15](#)

ahm, [2](#)

check_col_correlation, [3](#)

coating, [4](#)

coef.ahm, [4](#)

coef.cv.ahm, [5](#)

compute_aicc, [6](#)

cv.ahm, [6](#)

design_simplex_centroid_design_3_major_component,
[8](#)

enlist, [8](#)

expand_interactions, [9](#)

find_condition_num, [9](#)

mapping_function, [10](#)

mymaximin, [10](#)

predict.ahm, [12](#)

predict.cv.ahm, [13](#)

pringles_candidates2search, [14](#)

pringles_fat, [14](#)

pringles_hardness, [15](#)

summary.ahm, [15](#)