Package 'metacart'

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Description Meta-CART integrates classification and regression trees (CART) into meta- analysis. Meta-CART is a flexible approach to identify interaction effects between modera- tors in meta-analysis. The method is described in Dussel- dorp et al. (2014) <doi:10.1037 hea0000018=""> and Li et al. (2017) <doi:10.1111 bmsp.12088="">.</doi:10.1111></doi:10.1037>
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metacart-package R package for meta-CART

Description

In meta-analysis, heterogeneity often exists between studies. To understand this heterogeneity, researchers search for study characteristics (i.e., potential moderators) that may account for the variance in study effect sizes. When multiple potential moderators are available (e.g., intervention characteristics), traditional meta-analysis methods often lack sufficient power to investigate interaction effects between moderators, especially high-order interactions. To solve this problem, meta-CART was proposed by integrating Classification and Regression Trees (CART) into meta-analysis. The method idenfities the interaction effects between influential moderators, partitions the studies into more homogeneous subgroups, and estimates summary effect size in each subgroup. The fixed effect or random effects assumption can be consistently taken into account in both tree-growing process and subgroup analysis.

Details

metacart
Package
2.0-0
2018-11-12
GPL

This method is suitable for identifying interaction effects between dichotomous, ordinal, continuous, and nominal moderators. The output of a REmrt object shows meta-CART analysis results based on the random effects model. And the output of a FEmrt object shows meta-CART analysis results based on the fixed effect model. The two objects display results for subgroup analysis including the Q-statistic and estimates for the subgroup effect sizes. Furthermore, the predict functions predict.REmrt and predict.FEmrt can be used to predict the effect size given the moderators. The plot functions plot.REmrt and plot.FEmrt show the interaction effects between identified moderators.

The core functions of the package are FEmrt and REmrt.

dat.balanced

Author(s)

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References

Dusseldorp, E., van Genugten, L., van Buuren, S., Verheijden, M. W., & van Empelen, P. (2014). Combinations of techniques that effectively change health behavior: Evidence from meta-cart analysis. *Health Psychology*, *33*(*12*), 1530-1540. doi: 10.1037/hea0000018.

Li, X., Dusseldorp, E., & Meulman, J. J. (2017). Meta-CART: A tool to identify interactions between moderators in meta-analysis. *British Journal of Mathematical and Statistical Psychology*, 70(1), 118-136. doi: 10.1111/bmsp.12088.

Therneau, T., Atkinson, B., & Ripley, B. (2014) rpart: Recursive partitioning and regression trees. R package version, 4-1.

The articles of our own work can be found at http://www.elisedusseldorp.nl/

See Also

FEmrt, REmrt, summary.FEmrt,summary.REmrt,plot.FEmrt,plot.REmrt,predict.FEmrt,predict.REmrt

dat.balanced	A simulated meta-analytic data set with balanced pure interaction ef-
	fects

Description

Data simulated from a true model with pure interactions between two moderators: x_1 , x_2 . If $x_1 = 0$ and $x_2 = 1$ or $x_1 = 1$ and $x_2 = 0$, the true effect size is 0.50. Otherwise, the true effect size is 0.

Usage

data(dat.balanced)

Format

A data frame of 60 studies with 4 moderators

- efk: The effect size of each study expressed as Hedges' g
- vark: The sampling variance of the effect size
- x1 to x4: Four randomly generated moderators. x1, x2, and x4 are dichotomous variables, x3 is a continuous variable generated from uniform distribution.

dat.BCT2009

Description

The complete data consist of 101 studies reporting 122 interventions targeted at physical activity and healthy eating. In this subset of the data, the interventions that include at least one of the motivation-enhancing behaviour change techniques (BCTs) were selected (N = 106).

Usage

data(dat.BCT2009)

Format

A data frame of 106 interventions with five motivation-enhancing behavior change techniques (BCTs).

- study: The name of the intervention.
- g: The effect size of each intervention.
- vi: The sampling variance of the effect size.
- T1: Indicating whether the BCT1 "Provide information about behavior-health link" was used by the intervention. "1" for used and "0" for not used.
- T2: Indicating whether the BCT2 "Provide information on consequences" was used by the intervention. "1" for used and "0" for not used.
- T3: Indicating whether the BCT3 "Provide information about other's approval" was used by the intervention. "1" for used and "0" for not used.
- T4: Indicating whether the BCT4 "Prompt intention formation" was used by the intervention. "1" for used and "0" for not used.
- T25: Indicating whether the BCT25 " Motivational interviewing" was used by the intervention. "1" for used and "0" for not used.

Details

IMPORTANT: for questions about these data contact Xinru Li: x.li@math.leidenuniv.nl.

References

If you use these data, please refer to: Michie, S., Abraham, C., Whittington, C., McAteer, J., & Gupta, S. (2009). Effective techniques in healthy eating and physical activity interventions: a meta-regression. *Health Psychology*, 28(6), 690.

An application of (a preliminary version of) meta-CART to this data set is given in: Dusseldorp, E., Van Genugten, L., van Buuren, S., Verheijden, M. W., & van Empelen, P. (2014). Combinations of techniques that effectively change health behavior: Evidence from Meta-CART analysis. *Health Psychology*, *33*(*12*), 1530.

FEmrt

Description

A function to fit fixed effect meta-trees to meta-analytic data. The model is assuming a fixed effect within subgroups and between subgroups. The tree growing process is equivalent to the approach described in Li et al. (2017) using fixed effect weights in the function rpart() developed by Therneau, Atkinson & Ripley (2014).

Usage

```
FEmrt(
  formula,
  data,
  vi,
  subset,
  c = 1,
  control = rpart.control(xval = 10, minbucket = 3, minsplit = 6, cp = 1e-04),
```

Arguments

)

formula	A formula, with an outcome variable (usually the effect size) and the potential moderator variables but no interaction terms.
data	A data frame of a meta-analytic data set, including the study effect sizes, sampling variance, and the potential moderators.
vi	sampling variance of the effect size.
subset	optional expression that selects only a subset of the rows of the data.
с	A non-negative scalar. The pruning parameter to prune the initial tree by the "c*standard-error" rule.
control	the control object (similar to rpart.control from the rpart package) that is used in the tree algorithm
	Additional arguments passed to the tree growing algorithm based on rpart .

Value

If (a) moderator effect(s) is(are) detected, the function will return a FEmrt object including the following components:

tree: The pruned tree that represents the moderator effect(s) and interaction effect(s) between them.

n: The number of the studies in each subgroup

Qb: The between-subgroups Q-statistic

df: The degree of freedoms of the between-subgroups Q test

- pval.Qb: The p-value of the between-subgroups Q test
- Qw: The within-subgroup Q-statistic in each subgroup
- g: The subgroup summary effect size, based on Hedges'g
- se: The standard error of the subgroup summary effect size
- zval: The test statistic of the subgroup summary effect size
- pval: The p-value for the test statistics of the subgroup summary effect size
- ci.lb: The lower bound of the confidence interval
- ci.ub: The upper bound of the confidence interval

call: The matched call

If no moderator effect is detected, the function will return a FEmrt object including the following components:

- n: The total number of the studies
- Q: The Q-statistic of the heterogeneity test
- df: The degrees of freedom of the heterogeneity test
- pval.Q: The p-value of the heterogeneity test
- g: The summary effect size for all studies
- se: The standard error of the summary effect size
- zval: The test statistic of the summary effect size
- pval: The p-value of the test statistic of the summary effect size
- ci.lb: The lower bound of the confidence interval for the summary effect size
- ci.ub: The upper bound of the confidence interval for the summary effect size
- call: The matched call

References

Dusseldorp, E., van Genugten, L., van Buuren, S., Verheijden, M. W., & van Empelen, P. (2014). Combinations of techniques that effectively change health behavior: Evidence from meta-cart analysis. *Health Psychology*, *33*(*12*), 1530-1540. doi: 10.1037/hea0000018.

Li, X., Dusseldorp, E., & Meulman, J. J. (2017). Meta-CART: A tool to identify interactions between moderators in meta-analysis. *British Journal of Mathematical and Statistical Psychology*, *70*(*1*), 118-136. doi: 10.1111/bmsp.12088.

Therneau, T., Atkinson, B., & Ripley, B. (2014) rpart: Recursive partitioning and regression trees. R package version, 4-1.

See Also

summary.FEmrt, plot.FEmrt, rpart, rpart.control

plot.FEmrt

Examples

```
data(dat.BCT2009)
library(Rcpp)
FEtree <- FEmrt(g ~ T1 + T2+ T4 + T25, vi = vi, data = dat.BCT2009, c = 0)
print(FEtree)
summary(FEtree)
plot(FEtree)</pre>
```

```
plot.FEmrt
```

Visualisation of a FE meta-tree

Description

Plot function for a FEmrt object. The plot shows the result of FEmrt. The plot function uses the plot method from the package **ggplot2**

Usage

S3 method for class 'FEmrt'
plot(x, ...)

Arguments

xA FEmrt object....additional arguments to pass

Details

For categorical variables we recommend to use short names for levels to avoid overlapping labels at split points.

```
plot.REmrt
```

Visualisation of a RE meta-tree

Description

Plot function for a REmrt object. The plot shows the result of REmrt. The plot function uses the plot method from the package **ggplot2**

Usage

```
## S3 method for class 'REmrt'
plot(x, ...)
```

Arguments

х	A REmrt object.
	Additional arguments to pass.

Details

For categorical variables we recommend to use short names for levels to avoid overlapping labels at split points.

predict.FEmrt

Predictions from a fitted metacart object

Description

Returns a data frame of predicted effect sizes and moderators from a fitted metacart object

Usage

```
## S3 method for class 'FEmrt'
predict(object, newdata, ...)
```

Arguments

object	fitted model object of class "FEmrt".
newdata	data frame containing the values at which predictions are required.
	Arguments that pass to other methods.

Value

A data frame containing the predicted effect size, the moderators, and the corresponding node lables in the fitted tree.

predict.REmrt Predictions from a fitted metacart object

Description

Returns a data frame of predicted effect sizes and moderators from a fitted metacart object

Usage

```
## S3 method for class 'REmrt'
predict(object, newdata, ...)
```

print.FEmrt

Arguments

object	fitted model object of class "REmrt".
newdata	data frame containing the values at which predictions are required.
	Arguments that pass to other methods.

Value

A data frame containing the predicted effect size, the moderators, and the corresponding node lables in the fitted tree.

print.FEmrt Print function for FEmrt

Description

Print the results of a FEmrt object

Usage

S3 method for class 'FEmrt'
print(x, ...)

Arguments

х	fitted tree of class FEmrt.
	additional arguments to be passed.

Details

The function returns the objects concerning the analysis results.

print.REmrt

Print function for REmrt

Description

Print the results of a REmrt object

Usage

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

Arguments

х	fitted tree of class FEmrt.
	additional arguments to be passed.

Details

The function returns the results (e.g., the value of the Q-between) after each split of the tree.

REmrt

Random effects meta-tree

Description

A function to fit a random effects meta-tree

Usage

```
REmrt(
    formula,
    data,
    vi,
    c = 1,
    maxL = 5,
    minsplit = 6,
    cp = 1e-05,
    minbucket = 3,
    xval = 10,
    lookahead = FALSE,
    ...
)
```

Arguments

formula	A formula, with a response variable (usually the effect size) and the potential moderator variables but no interaction terms.
data	A data frame of a meta-analytic data set, including the study effect sizes, sampling variance, and the potential moderators.
vi	sampling variance of the effect size.
С	A non-negative scalar. The pruning parameter to prune the initial tree by the "c*standard-error" rule.
maxL	the maximum number of splits
minsplit	the minimum number of studies in a parent node before splitting
ср	the stopping rule for the decrease of between-subgroups Q. Any split that does not decrease the between-subgroups Q is not attempted.

REmrt

minbucket	the minimum number of the studies in a terminal node
xval	the number of folds to perform the cross-validation
lookahead	an argument indicating whether to apply the "look-ahead" strategy when fitting the tree
	Additional arguments to be passed.

Value

If (a) moderator effect(s) is(are) detected, the function will return a list including the following objects:

tree: A data frame that represents the tree, with the Q-between and the residual heterogeneity (tau^2) after each split.

n: The number of the studies in each subgroup

moderators: the names of identified moderators

Qb: The between-subgroups Q-statistic

tau2: The estimate of the residual heterogeneity

df: The degrees of freedom of the between-subgroups Q test

pval.Qb: The p-value of the between-subgroups Q test

g: The subgroup summary effect size, based on Hedges'g

se: The standard error of subgroup summary effect size

zval: The test statistic of the subgroup summary effect size

pval: The p-value of the test statistic of the subgroup summary effect size

ci.lb: The lower bound of the confidence interval

ci.ub: The upper bound of the confidence interval

call: The matched call

cv.res: The cross-validation table

data: the data set subgrouped by the fitted tree

If no moderator effect is detected, the function will return a list including the following objects:

n: The total number of the studies

Q: The Q-statistics for the heterogeneity test

df: The degree of freedoms of the heterogeneity test

pval.Q: The p-value for the heterogeneity test

g: The summary effect size for all studies (i.e., the overall effect size)

se: The standard error of the summary effect size

zval: The test statistic of the summary effect size

pval: The p-value for the test statistic of the summary effect size

ci.lb: The lower bound of the confidence interval for the summary effect size

ci.ub: The upper bound of the confidence interval for the summary effect size

call: The matched call

See Also

summary.REmrt, plot.REmrt

Examples

```
data(dat.BCT2009)
library(Rcpp)
REtree <- REmrt(g ~ T1 + T2+ T4 +T25, vi = vi, data = dat.BCT2009, c = 0)
summary(REtree)
plot(REtree)</pre>
```

SimData

A simulated meta-analytic data set

Description

Data simulted from a true model with a three-way interaction between three moderators: m1, m2 and m3. If the values of the three moderators are all "B"s the true effect size will be 0.80. Otherwise, the true effect size is 0.

Usage

data(SimData)

Format

A data frame of 120 studies with 5 moderators

- efk: The effect size of each study expressed as Hedges' g
- vark: The sampling variance of the effect size
- m1 to m5: Five randomly generated moderators. m1 and m2 have two levels (A and B), whereas m3, m4 and m5 have three levels (A, B and C)

summary.FEmrt Summary of the results of a FE meta-tree object

Description

Summary of the results of a FE meta-tree object

Usage

```
## S3 method for class 'FEmrt'
summary(object, digits = 3, ...)
```

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summary.REmrt

Arguments

object	fitted tree of class FEmrt.
digits	specified number of decimals in the printed results.
	additional arguments to be passed.

Details

If no moderator effect is detected, the summary function will show the standard meta-analysis results. Otherwise, the summary function will show the subgroup meta-analysis results, with the significance test results for moderator effects, the split points of the moderators, and the estimated subgroup summary effect sizes.

summary.REmrt Summary of the results of a RE meta-tree object

Description

Summary of the results of a RE meta-tree object

Usage

```
## S3 method for class 'REmrt'
summary(object, digits = 3, ...)
```

Arguments

object	fitted tree of class REmrt.
digits	specified number of decimals in the printed results.
	additional arguments to be passed.

Details

If no moderator effect is detected, the summary function will show the standard meta-analysis results. Otherwise, the summary function will show the subgroup meta-analysis results, with the significance test results for moderator effects, the split points of the moderators, and the estimated subgroup summary effect sizes.

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