Package 'HSTree'

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Type Package
Title Hierarchical Shrinkage: improving the accuracy and interpretability of tree-based methods
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Description Hierarchical shrinkage is an extremely fast post-hoc regularization method which works on any decision tree (or tree-based ensemble, such as Random Forest). It does not modify the tree structure, and instead regularizes the tree by shrinking the prediction over each node towards the sample means of its ancestors (using a single regularization parameter). Experiments over a wide variety of datasets show that hierarchical shrinkage substantially increases the predictive performance of individual decision trees and decision-tree ensembles. ``Hierarchical shrinkage" is first defined in Agarwal et al. (2022) https://proceedings.mlr.press/v162/agarwal22b.html >
Imports rpart, randomForest, gbm
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<pre>URL https://github.com/wanghaoxue0/HSTree</pre>
BugReports https://github.com/wanghaoxue0/HSTree/issues
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R topics documented:
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2 HSTreeClassifier

HSTreeClassifier

HSTree fit for classification

Description

modify the decision tree(or each tree in ensemble) structure based on hierarchical shrinkage regularization

Usage

HSTreeClassifier(X, y, reg_param=1, max_leaf_nodes=20, interaction.depth=1, estimator="CART", shri

Arguments

X the design matrix

y the response vector

reg_param Higher is more regularization (can be arbitrarily large, should not be < 0)

max_leaf_nodes the maximum number of leaf nodes, the default is 20

interaction.depth

Integer specifying the maximum depth of each tree (i.e., the highest level of variable interactions allowed). A value of 1 implies an additive model, a value

of 2 implies a model with up to 2-way interactions, etc. Default is 1.

estimator CART decision tree or tree ensemble model (e.g. RandomForest or Gradient-

Boosting) Defaults to CART Classification Tree with 20 max leaf nodes Note:

this estimator will be directly modified and keep its original functions

shrinkage shrinkage methods, default is "node_based", options are: 1. node_based shrinks

based on number of samples in parent node 2. leaf_based only shrinks leaf nodes based on number of leaf samples 3. constant shrinks every node by a constant

lambda

Value

call: the input setting

fit: keep all the output as the orginal estimator, only replace the value in each node based on shrinkage methods

regularization: chr "HSTree"

shrinkage: the hierarchical shrinkage method used in this model

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HSTreeClassifierCV

HSTree fit for classification with cross validation

Description

modify the decision tree(or each tree in ensemble) structure with the best regularization parameter chosen by cross validation

Usage

HSTreeClassifierCV(X, y, cv=4, verbose=FALSE, reg_param=c(0.1, 1, 10, 50, 100, 500), max_leaf_nodes

Arguments

X the design matrix

y the response vector

cv cross validation level, default is 4 (divide the data into 4 portions, train:test=3:1)

verbose whether to print the cross validation process and result

reg_param numerical array of regularization parameter, default is c(0.1, 1, 10, 50, 100, 500)

max_leaf_nodes the maximum number of leaf nodes, the default is 20

interaction.depth

Integer specifying the maximum depth of each tree (i.e., the highest level of variable interactions allowed). A value of 1 implies an additive model, a value

of 2 implies a model with up to 2-way interactions, etc. Default is 1.

estimator CART decision tree or tree ensemble model (e.g. RandomForest or Gradient-

Boosting) Defaults to CART Classification Tree with 20 max leaf nodes Note:

this estimator will be directly modified and keep its original functions

shrinkage methods, default is "node_based", options are: 1. node_based shrinks

based on number of samples in parent node 2. leaf_based only shrinks leaf nodes based on number of leaf samples 3. constant shrinks every node by a constant

lambda

Value

call: the input setting

fit: keep all the output as the orginal estimator, only replace the value in each node based on shrinkage methods

regularization: chr "HSTree"

shrinkage: the hierarchical shrinkage method used in this model

4 HSTreeRegressor

HSTreeRegressor

HSTree fit for regression

Description

modify the decision tree(or each tree in ensemble) structure based on hierarchical shrinkage regularization

Usage

HSTreeRegressor(X, y, reg_param=1, max_leaf_nodes=20, interaction.depth=1, estimator="CART", shrir

Arguments

X the design matrix

y the response vector

reg_param Higher is more regularization (can be arbitrarily large, should not be < 0)

max_leaf_nodes the maximum number of leaf nodes, the default is 20

interaction.depth

Integer specifying the maximum depth of each tree (i.e., the highest level of variable interactions allowed). A value of 1 implies an additive model, a value

of 2 implies a model with up to 2-way interactions, etc. Default is 1.

estimator CART decision tree or tree ensemble model (e.g. RandomForest or Gradient-

Boosting) Defaults to CART Classification Tree with 20 max leaf nodes Note:

this estimator will be directly modified and keep its original functions

shrinkage shrinkage methods, default is "node_based", options are: 1. node_based shrinks

based on number of samples in parent node 2. leaf_based only shrinks leaf nodes based on number of leaf samples 3. constant shrinks every node by a constant

lambda

Value

call: the input setting

fit: keep all the output as the orginal estimator, only replace the value in each node based on shrinkage methods

regularization: chr "HSTree"

shrinkage: the hierarchical shrinkage method used in this model

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HSTreeRegressorCV	HSTree fit for regression with cross validation

Description

modify the decision tree(or each tree in ensemble) structure with the best regularization parameter chosen by cross validation

Usage

```
HSTreeRegressorCV(X, y, cv=4, verbose=FALSE, reg_param=c(0.1, 1, 10, 50, 100, 500), max_leaf_nodes=
```

Arguments

X	the design matrix
У	the response vector
cv	cross validation level, default is 4 (divide the data into 4 portions, train:test=3:1)
verbose	whether to print the cross validation process and result

 $\label{eq:reg_param} \textit{numerical array of regularization parameter, default is } c(0.1, 1, 10, 50, 100, 500)$

max_leaf_nodes the maximum number of leaf nodes, the default is 20

interaction.depth

Integer specifying the maximum depth of each tree (i.e., the highest level of variable interactions allowed). A value of 1 implies an additive model, a value of 2 implies a model with up to 2-way interactions, etc. Default is 1.

estimator CART decision tree or tree ensemble model (e.g. RandomForest or Gradient-

Boosting) Defaults to CART Classification Tree with 20 max leaf nodes Note:

this estimator will be directly modified and keep its original functions

shrinkage shrinkage methods, default is "node_based", options are: 1. node_based shrinks

based on number of samples in parent node 2. leaf_based only shrinks leaf nodes based on number of leaf samples 3. constant shrinks every node by a constant

lambda

Value

call: the input setting

fit: keep all the output as the orginal estimator, only replace the value in each node based on shrinkage methods

regularization: chr "HSTree"

shrinkage: the hierarchical shrinkage method used in this model

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