

Package ‘xrf’

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Title eXtreme RuleFit

Version 0.2.2

Description An implementation of the RuleFit algorithm as described in Friedman & Popescu (2008) <doi:10.1214/07-AOAS148>. eXtreme Gradient Boosting ('XGBoost') is used to build rules, and 'glmnet' is used to fit a sparse linear model on the raw and rule features. The result is a model that learns similarly to a tree ensemble, while often offering improved interpretability and achieving improved scoring runtime in live applications. Several algorithms for reducing rule complexity are provided, most notably hyperrectangle de-overlapping. All algorithms scale to several million rows and support sparse representations to handle tens of thousands of dimensions.

URL <https://github.com/holub008/xrf>

BugReports <https://github.com/holub008/xrf/issues>

Depends R (>= 3.1.0)

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Encoding UTF-8

RoxygenNote 7.1.1

Imports Matrix, glmnet (>= 3.0), xgboost (>= 0.71.2), dplyr, fuzzyjoin, rlang, methods

Suggests testthat, covr

NeedsCompilation no

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Repository CRAN

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coef.xrf	<i>Produce rules & coefficients for the RuleFit model</i>
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Description

Produce rules & coefficients for the RuleFit model

Usage

```
## S3 method for class 'xrf'  
coef(object, lambda = "lambda.min", ...)
```

Arguments

- object an object of class "xrf"
- lambda the lasso penalty parameter to be applied as in 'glmnet'
- ... ignored arguments

Examples

```
m <- xrf(Petal.Length ~ ., iris,  
         xgb_control = list(nrounds = 2, max_depth = 2),  
         family = 'gaussian')  
linear_model_coefficients <- coef(m, lambda = 'lambda.1se')
```

model.matrix.xrf	<i>Generate the design matrix from an eXtreme RuleFit model</i>
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Description

Generate the design matrix from an eXtreme RuleFit model

Usage

```
## S3 method for class 'xrf'  
model.matrix(object, data, sparse = TRUE, ...)
```

Arguments

object	an object of class "xrf"
data	data to generate design matrix from
sparse	a logical indicating whether a sparse design matrix should be used
...	ignored arguments

Examples

```
m <- xrf(Petal.Length ~ ., iris,
        xgb_control = list(nrounds = 2, max_depth = 2),
        family = 'gaussian')
design <- model.matrix(m, iris, sparse = FALSE)
```

predict.xrf	<i>Draw predictions from a RuleFit xrf model</i>
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Description

Draw predictions from a RuleFit xrf model

Usage

```
## S3 method for class 'xrf'
predict(
  object,
  newdata,
  sparse = TRUE,
  lambda = "lambda.min",
  type = "response",
  ...
)
```

Arguments

object	an object of class "xrf"
newdata	data to predict on
sparse	a logical indicating whether a sparse design matrix should be used
lambda	the lasso penalty parameter to be applied
type	the type of predicted value produced
...	ignored arguments

Examples

```
m <- xrf(Petal.Length ~ ., iris,
         xgb_control = list(nrounds = 2, max_depth = 2),
         family = 'gaussian')
predictions <- predict(m, iris)
```

print.xrf	<i>Print an eXtreme RuleFit model</i>
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Description

Print an eXtreme RuleFit model

Usage

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

Arguments

x	an object of class "xrf"
...	ignored arguments

Examples

```
m <- xrf(Petal.Length ~ ., iris,
         xgb_control = list(nrounds = 2, max_depth = 2),
         family = 'gaussian')
print(m)
```

summary.xrf	<i>Summarize an eXtreme RuleFit model</i>
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Description

Summarize an eXtreme RuleFit model

Usage

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

Arguments

object an object of class "xrf"
 ... ignored arguments

Examples

```
m <- xrf(Petal.Length ~ ., iris,
         xgb_control = list(nrounds = 2, max_depth = 2),
         family = 'gaussian')
summary(m)
```

xrf	<i>Fit an eXtreme RuleFit model</i>
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Description

S3 method for building an "eXtreme RuleFit" model. See [xrf.formula](#) for preferred entry point

Usage

```
xrf(object, ...)
```

Arguments

object an object describing the model to be fit
 ... additional arguments

Examples

```
m <- xrf(Petal.Length ~ ., iris,
         xgb_control = list(nrounds = 2, max_depth = 2),
         family = 'gaussian')
```

xrf.formula	<i>Fit an eXtreme RuleFit model</i>
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Description

See Friedman & Popescu (2008) for a description of the general RuleFit algorithm. This method uses XGBoost to fit a tree ensemble, extracts a ruleset as the conjunction of tree traversals, and fits a sparse linear model to the resulting feature set (including the original feature set) using glmnet.

Usage

```
## S3 method for class 'formula'
xrf(
  object,
  data,
  family,
  xgb_control = list(nrounds = 100, max_depth = 3),
  glm_control = list(type.measure = "deviance", nfolds = 5),
  sparse = TRUE,
  prefit_xgb = NULL,
  deoverlap = FALSE,
  ...
)
```

Arguments

object	a formula prescribing features to use in the model. transformation of the response variable is not supported. when using transformations on the input features (not suggested in general) it is suggested to set sparse=F
data	a data frame with columns corresponding to the formula
family	the family of the fitted model. one of 'gaussian', 'binomial', 'multinomial'
xgb_control	a list of parameters for xgboost. must supply an nrounds argument
glm_control	a list of parameters for the glmnet fit. must supply a type.measure and nfolds arguments (for the lambda cv)
sparse	whether a sparse design matrix should be used
prefit_xgb	an xgboost model (of class xgb.Booster) to be used instead of the model that xrf would normally fit
deoverlap	if true, the tree derived rules are deoverlapped, in that the deoverlapped rule set contains no overlapped rules
...	ignored arguments

References

Friedman, J. H., & Popescu, B. E. (2008). Predictive learning via rule ensembles. *The Annals of Applied Statistics*, 2(3), 916-954.

Examples

```
m <- xrf(Petal.Length ~ ., iris,
  xgb_control = list(nrounds = 2, max_depth = 2),
  family = 'gaussian')
```

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