

# Package: caretForecast (via r-universe)

October 12, 2024

**Title** Conformal Time Series Forecasting Using State of Art Machine Learning Algorithms

**Version** 0.1.1

**Description** Conformal time series forecasting using the caret infrastructure. It provides access to state-of-the-art machine learning models for forecasting applications. The hyperparameter of each model is selected based on time series cross-validation, and forecasting is done recursively.

**License** GPL (>= 3)

**URL** <https://github.com/Akai01/caretForecast>

**BugReports** <https://github.com/Akai01/caretForecast/issues>

**Depends** R (>= 3.6)

**Imports** forecast (>= 8.15), caret (>= 6.0.88), magrittr (>= 2.0.1), methods (>= 4.1.1), dplyr (>= 1.0.9), generics (>= 0.1.3)

**Suggests** Cubist (>= 0.3.0), knitr (>= 1.29), testthat (>= 2.3.2)

**Encoding** UTF-8

**LazyData** true

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.2.1

**Repository** <https://akai01.r-universe.dev>

**RemoteUrl** <https://github.com/akai01/caretforecast>

**RemoteRef** HEAD

**RemoteSha** d3b11594639bbc5174ba40b44421218f8756618c

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ARml	<i>Autoregressive forecasting using various Machine Learning models.</i>
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## Description

Autoregressive forecasting using various Machine Learning models.

## Usage

```
ARml(
  y,
  max_lag = 5,
  xreg = NULL,
  caret_method = "cubist",
  metric = "RMSE",
  pre_process = NULL,
  cv = TRUE,
  cv_horizon = 4,
  initial_window = NULL,
  fixed_window = FALSE,
  verbose = TRUE,
  seasonal = TRUE,
  K = frequency(y)/2,
  tune_grid = NULL,
  lambda = NULL,
  BoxCox_method = c("guerrero", "loglik"),
  BoxCox_lower = -1,
  BoxCox_upper = 2,
  BoxCox_biasadj = FALSE,
  BoxCox_fvar = NULL,
  allow_parallel = FALSE,
  ...
)
```

## Arguments

<code>y</code>	A univariate time series object.
<code>max_lag</code>	Maximum value of lag.
<code>xreg</code>	Optional. A numerical vector or matrix of external regressors, which must have the same number of rows as <code>y</code> . (It should not be a data frame.)

caret_method	A string specifying which classification or regression model to use. Possible values are found using names(getModelInfo()). A list of functions can also be passed for a custom model function. See <a href="http://topepo.github.io/caret/">http://topepo.github.io/caret/</a> for details.
metric	A string that specifies what summary metric will be used to select the optimal model. See ?caret::train.
pre_process	A string vector that defines a pre-processing of the predictor data. Current possibilities are "BoxCox", "YeoJohnson", "expoTrans", "center", "scale", "range", "knnImpute", "bagImpute", "medianImpute", "pca", "ica" and "spatialSign". The default is no pre-processing. See preProcess and trainControl on the procedures and how to adjust them. Pre-processing code is only designed to work when x is a simple matrix or data frame.
cv	Logical, if cv = TRUE model selection will be done via cross-validation. If cv = FALSE user need to provide a specific model via tune_grid argument.
cv_horizon	The number of consecutive values in test set sample.
initial_window	The initial number of consecutive values in each training set sample.
fixed_window	Logical, if FALSE, all training samples start at 1.
verbose	A logical for printing a training log.
seasonal	Boolean. If seasonal = TRUE the fourier terms will be used for modeling seasonality.
K	Maximum order(s) of Fourier terms
tune_grid	A data frame with possible tuning values. The columns are named the same as the tuning parameters. Use getModelInfo to get a list of tuning parameters for each model or see <a href="http://topepo.github.io/caret/available-models.html">http://topepo.github.io/caret/available-models.html</a> . (NOTE: If given, this argument must be named.)
lambda	BoxCox transformation parameter. If lambda = NULL If lambda = "auto", then the transformation parameter lambda is chosen using BoxCox.lambda.
BoxCox_method	BoxCox.lambda argument. Choose method to be used in calculating lambda.
BoxCox_lower	BoxCox.lambda argument. Lower limit for possible lambda values.
BoxCox_upper	BoxCox.lambda argument. Upper limit for possible lambda values.
BoxCox_biasadj	InvBoxCox argument. Use adjusted back-transformed mean for Box-Cox transformations. If transformed data is used to produce forecasts and fitted values, a regular back transformation will result in median forecasts. If biasadj is TRUE, an adjustment will be made to produce mean forecasts and fitted values.
BoxCox_fvar	InvBoxCox argument. Optional parameter required if biasadj=TRUE. Can either be the forecast variance, or a list containing the interval level, and the corresponding upper and lower intervals.
allow_parallel	If a parallel backend is loaded and available, should the function use it?
...	Ignored.

**Value**

A list class of forecast containing the following elements

- `x` : The input time series
- `method` : The name of the forecasting method as a character string
- `mean` : Point forecasts as a time series
- `lower` : Lower limits for prediction intervals
- `upper` : Upper limits for prediction intervals
- `level` : The confidence values associated with the prediction intervals
- `model` : A list containing information about the fitted model
- `newx` : A matrix containing regressors

**Author(s)**

Resul Akay

**Examples**

```
library(caretForecast)

train_data <- window(AirPassengers, end = c(1959, 12))

test <- window(AirPassengers, start = c(1960, 1))

ARml(train_data, caret_method = "lm", max_lag = 12) -> fit

forecast(fit, h = length(test)) -> fc

autoplot(fc) + autolayer(test)

accuracy(fc, test)
```

---

`conformalRegressor`     *Fit a conformal regressor.*

---

**Description**

Fit a conformal regressor.

**Usage**

```
conformalRegressor(residuals, sigmas = NULL)
```

**Arguments**

residuals      Model residuals.  
 sigmas         A vector of difficulty estimates

**Value**

A conformalRegressor object

**Author(s)**

Resul Akay

**References**

Boström, H., 2022. crepes: a Python Package for Generating Conformal Regressors and Predictive Systems. In Conformal and Probabilistic Prediction and Applications. PMLR, 179. [https://copa-conference.com/papers/COPA2022\\_paper\\_11.pdf](https://copa-conference.com/papers/COPA2022_paper_11.pdf)

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forecast.ARml	<i>Forecasting using ARml model</i>
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**Description**

Forecasting using ARml model

**Usage**

```
## S3 method for class 'ARml'
forecast(object, h = frequency(object$y), xreg = NULL, level = c(80, 95), ...)
```

**Arguments**

object         An object of class "ARml", the result of a call to ARml.  
 h              forecast horizon  
 xreg          Optionally, a numerical vector or matrix of future external regressors  
 level         Confidence level for prediction intervals.  
 ...           Ignored

**Value**

A list class of forecast containing the following elements

- x : The input time series
- method : The name of the forecasting method as a character string
- mean : Point forecasts as a time series

- lower : Lower limits for prediction intervals
- upper : Upper limits for prediction intervals
- level : The confidence values associated with the prediction intervals
- model : A list containing information about the fitted model
- newxreg : A matrix containing regressors

**Author(s)**

Resul Akay

**Examples**

```
library(caretForecast)

train_data <- window(AirPassengers, end = c(1959, 12))

test <- window(AirPassengers, start = c(1960, 1))

ARml(train_data, caret_method = "lm", max_lag = 12) -> fit

forecast(fit, h = length(test), level = c(80,95)) -> fc

autoplot(fc)+ autolayer(test)

accuracy(fc, test)
```

---

get_var_imp	<i>Variable importance for forecasting model.</i>
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**Description**

Variable importance for forecasting model.

**Usage**

```
get_var_imp(object, plot = TRUE)
```

**Arguments**

object	A list class of ARml or forecast object derived from ARml
plot	Boolean, if TRUE, variable importance will be plotted.

**Value**

A list class of "varImp.train". See [varImp](#) or a "trellis" plot.

**Author(s)**

Resul Akay

**Examples**

```
train <- window(AirPassengers, end = c(1959, 12))  
test <- window(AirPassengers, start = c(1960, 1))  
  
ARml(train, caret_method = "lm", max_lag = 12, trend_method = "none",  
pre_process = "center") -> fit  
  
forecast(fit, h = length(test), level = c(80,95)) -> fc  
  
autoplot(fc)+ autolayer(test)  
  
accuracy(fc, test)  
  
get_var_imp(fc, plot = TRUE)
```

---

`predict.conformalRegressor`  
*Predict a conformalRegressor*

---

**Description**

Predict a conformalRegressor

**Usage**

```
## S3 method for class 'conformalRegressor'  
predict(  
  object,  
  y_hat = NULL,  
  sigmas = NULL,  
  confidence = 0.95,  
  y_min = -Inf,  
  y_max = Inf,  
  ...  
)
```

**Arguments**

<code>object</code>	A conformalRegressor object
<code>y_hat</code>	Predicted values
<code>sigmas</code>	Difficulty estimates

confidence	Confidence level
y_min	The minimum value to include in prediction intervals
y_max	The maximum value to include in prediction intervals
...	Ignored

**Value**

Prediction intervals

**Author(s)**

Resul Akay

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retail	<i>Grouped sales data from an Australian Retailer</i>
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**Description**

A dataset containing 42 products' sales

**Usage**

retail

**Format**

A data class of "tbl\_df", "tbl", "data.frame" with 13986 rows and 3 columns:

**date** date

**item** products

**value** sales

**Source**

<https://robjhyndman.com/data/ausretail.csv>



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retail_wide	<i>Sales data from an Australian Retailer in time series format</i>
-------------	---

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**Description**

A dataset containing 42 products' sales

**Usage**

```
retail_wide
```

**Format**

An object of class `mts` (inherits from `ts`, `matrix`) with 333 rows and 43 columns.  
This data set is the wide format of `retail` data.

**Source**

<https://robjhyndman.com/data/ausretail.csv>

---

split_ts	<i>Split a time series into training and testing sets</i>
----------	---

---

**Description**

Split a time series into training and testing sets

**Usage**

```
split_ts(y, test_size = 10)
```

**Arguments**

<code>y</code>	A univariate time series
<code>test_size</code>	The number of observations to keep in the test set

**Value**

A list with train and test elements

**Author(s)**

Resul Akay

**Examples**

```
dlist <- split_ts(retail_wide[,1], test_size = 12)
```

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suggested\_methods      *Suggested methods for ARml*

---

**Description**

Suggested methods for ARml

**Usage**

```
suggested_methods()
```

**Value**

A character vector of Suggested methods

**Author(s)**

Resul Akay

**Examples**

```
suggested_methods()
```

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