

# Package ‘predtoolsTS’

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**Type** Package

**Title** Time Series Prediction Tools

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**Description** Makes the time series prediction easier by automatizing this process using four main functions: prep(), modl(), pred() and postp(). Features different preprocessing methods to homogenize variance and to remove trend and seasonality. Also has the potential to bring together different predictive models to make comparatives. Features ARIMA and Data Mining Regression models (using caret).

**License** GPL (>= 2)

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## Contents

modl . . . . .	2
modl.arima . . . . .	4
modl.dataMining . . . . .	5
modl.trControl . . . . .	6

modl.tsToDataFrame . . . . .	7
plot.pred . . . . .	7
plot.prep . . . . .	8
postp . . . . .	9
postp.deseason.differencing . . . . .	9
postp.detrend.differencing . . . . .	10
postp.detrend.s fsm . . . . .	11
postp.homogenize.boxcox . . . . .	11
postp.homogenize.log . . . . .	12
pred . . . . .	13
pred.arima . . . . .	14
pred.compareModels . . . . .	14
pred.dataMining . . . . .	16
prep . . . . .	16
prep.check.acf . . . . .	18
prep.check.adf . . . . .	19
prep.deseason.differencing . . . . .	19
prep.detrend.differencing . . . . .	20
prep.detrend.s fsm . . . . .	21
prep.homogenize.boxcox . . . . .	21
prep.homogenize.log . . . . .	22
print.modl . . . . .	23
print.pred . . . . .	23
print.prep . . . . .	24
summary.modl . . . . .	24
summary.pred . . . . .	25
summary.prep . . . . .	25
<b>Index</b>	<b>26</b>

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 modl

*Building predictive models*


---

## Description

This function give us the tools to build predictive models for time series.

## Usage

```
modl(tserie, method = "arima", algorithm = NULL, formula = NULL,
     initialWindow = NULL, horizon = NULL, fixedWindow = NULL)
```

**Arguments**

tserie	A ts or prep object.
method	A string. Current methods available are "arima" and "dataMining". Method "arima" is set as default.
algorithm	A string. In case method is "dataMining", pick the algorithm you want to use. There is a complete list of available algorithms here (only regression type allowed): <a href="http://topepo.github.io/caret/train-models-by-tag.html">http://topepo.github.io/caret/train-models-by-tag.html</a> .
formula	An integer vector. Contains the indexes from the time series wich will indicate how to extract the features. The last value will be the class index. Default value: c(1:16)
initialWindow	An integer. The initial number of consecutive values in each training set sample. Default value: 30.
horizon	An integer. The number of consecutive values in test set sample. Default value: 15.
fixedWindow	A logical: if FALSE, the training set always start at the first sample and the training set size will vary over data splits. Default value: TRUE.

**Details**

Returns an object modl which stores all the information related to the final chosen model (errors, parameters, model).

Currently this function covers two different methods: the widely know ARIMA and the "not so used for prediction" data mining. For the data mining we make use of the caret package.

The caret package offers plenty of data mining algorithms. For the data splitting here we use a rolling forecasting origin technique, wich works better on time series.

**Value**

A list is returned of class modl containing:

tserie	Original time serie.
tserieDF	Time serie converted to data frame.
method	Method used to build the model.
algorithm	If method is data mining, indicates wich algorithm was used.
horizon	Horizon for the splitting.
model	Model result from caret. It is a list, result of the caret::train function.
errors	Contains three different metrics to evaluate the model.

**Author(s)**

Alberto Vico Moreno

**References**

<http://topepo.github.io/caret/index.html>

**See Also**

[prep](#), [modl.tsToDataFrame](#), [modl.trControl](#), [modl.dataMining](#)

**Examples**

```
p <- prep(AirPassengers)
modl(p,method='arima')
modl(p,method='dataMining',algorithm='rpart')
```

---

modl.arima	<i>Automatic ARIMA model</i>
------------	------------------------------

---

**Description**

Assuming "tserie" is stationary, returns the best arima model

**Usage**

```
modl.arima(tserie)
```

**Arguments**

tserie      A ts object.

**Value**

ARIMA model.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
modl.arima(AirPassengers)
```

---

modl.dataMining	<i>Train the data</i>
-----------------	-----------------------

---

**Description**

Train the time serie(as data frame) to build the model.

**Usage**

```
modl.dataMining(form, tserieDF, algorithm, timeControl, metric = "RMSE",  
  maximize = FALSE)
```

**Arguments**

form	A formula of the form $y \sim x_1 + x_2 + \dots$
tserieDF	Data frame.
algorithm	A string. Algorithm to perform the training. Full list at <a href="http://topepo.github.io/caret/train-models-by-tag.html">http://topepo.github.io/caret/train-models-by-tag.html</a> . Only regression types allowed.
timeControl	trainControl object.
metric	A string. Specifies what summary metric will be used to select the optimal model. Possible values in caret are "RMSE" and "Rsquared". "RMSE" set as default. If you used a custom summaryFunction(see ?trainControl) your metrics will prevail over default.
maximize	A logical. Should the metric be maximized or minimized? Default is FALSE, since that is what makes sense for time series.

**Value**

train object

**Author(s)**

Alberto Vico Moreno

**Examples**

```
modl.dataMining(form=Class ~ .,  
  tserieDF=modl.tsToDataFrame(AirPassengers,formula=c(1:20)),  
  algorithm='rpart',  
  timeControl=modl.trControl(initialWindow=30,horizon=15,fixedWindow=TRUE))
```

---

modl.trControl      *Control the splitting to train the data*

---

### Description

Creates the needed caret::trainControl object to control the training splitting.

### Usage

```
modl.trControl(initialWindow, horizon, fixedWindow, givenSummary = FALSE)
```

### Arguments

initialWindow	An integer. The initial number of consecutive values in each training set sample. Default value: 30.
horizon	An integer. The number of consecutive values in test set sample. Default value: 15.
fixedWindow	A logical: if FALSE, the training set always start at the first sample and the training set size will vary over data splits. Default value: TRUE.
givenSummary	A logical. Indicates if it should be used the customized summaryFunction(?trainControl for more info) modl.sumFunction or not. Default is FALSE; this will use default caret metrics.

### Details

We always split using method "timeslice", wich is the better for time series. More information on how this works on <http://topepo.github.io/caret/data-splitting.html#data-splitting-for-time-series>.

### Value

trainControl object

### Author(s)

Alberto Vico Moreno

### Examples

```
modl.trControl(initialWindow=30,horizon=15,fixedWindow=TRUE,givenSummary=TRUE)
```

---

modl.tsToDataFrame      *Ts to data frame transformation*

---

**Description**

Transform a ts object into a data frame using the given formula.

**Usage**

```
modl.tsToDataFrame(tserie, formula = NULL)
```

**Arguments**

tserie	A ts object.
formula	An integer vector. Contains the indexes from the tserie wich will indicate how to extract the features. The last value will be the class index. Default value: c(1:16). Has to be length 6 minimum.

**Value**

the time serie as data frame

**Author(s)**

Alberto Vico Moreno

**Examples**

```
modl.tsToDataFrame(AirPassengers, formula=c(1,3,4,5,6,7))  
modl.tsToDataFrame(AirPassengers, formula=c(1:20))
```

---

plot.pred      *Generic function*

---

**Description**

Plots object prep

**Usage**

```
## S3 method for class 'pred'  
plot(x, ylab = "Values", main = "Predictions", ...)
```

**Arguments**

x	pred object
ylab	ylab
main	main
...	ignored

**Examples**

```
plot(pred(modl(prepare(AirPassengers))))
```

---

plot.prep

*Generic function*

---

**Description**

Plots object prep

**Usage**

```
## S3 method for class 'prep'  
plot(x, ylab = "Preprocessed time serie", xlab = "", ...)
```

**Arguments**

x	prep object
ylab	ylab
xlab	xlab
...	ignored

**Examples**

```
plot(prepare(AirPassengers),ylab="Stationary AisPassengers")
```

---

postp	<i>Post-processing of pre-processed data</i>
-------	--

---

**Description**

Using the prep data we undo the changes on a pred object.

**Usage**

```
postp(prd, pre)
```

**Arguments**

prd	A pred object.
pre	A prep object.

**Value**

A pred object with reverted transformations.

**Author(s)**

Alberto Vico Moreno

**See Also**

[pred.prep](#), [postp.homogenize.log](#), [postp.homogenize.boxcox](#), [postp.detrend.differencing](#), [postp.detrend.sfs](#), [postp.deseason.differencing](#)

**Examples**

```
preprocess <- prep(AirPassengers)
prediction <- pred(modl(preprocess),n.ahead=30)
postp.prediction <- postp(prediction,preprocess)
```

---

postp.deseason.differencing	<i>Undo deseason(differencing)</i>
-----------------------------	------------------------------------

---

**Description**

Uses inverse seasonal differences to reverse the changes

**Usage**

```
postp.deseason.differencing(tserie, nsd, firstseasons, frequency)
```

**Arguments**

tserie	A ts object.
nsd	Number of seasonal differences.
firstseasons	Values lost on the original differences
frequency	Frequency of the original time serie

**Value**

A ts object.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
p <- prep.deseason.differencing(AirPassengers)
postp.deseason.differencing(p$tserie,p$nsd,p$firstseasons,frequency(AirPassengers))
```

---

```
postp.detrend.differencing
      Undo detrend(differencing)
```

---

**Description**

Uses inverse differences to revert the changes

**Usage**

```
postp.detrend.differencing(tserie, nd, firstvalues)
```

**Arguments**

tserie	A ts object.
nd	Number of differences.
firstvalues	Values lost on the original differences

**Value**

A ts object.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
p <- prep.detrend.differencing(AirPassengers)
postp.detrend.differencing(p$tserie,p$nd,p$firstvalues)
```

---

postp.detrend.s fsm     *Undo detrend(substracting full-means method)*

---

**Description**

Undo detrend(substracting full-means method)

**Usage**

```
postp.detrend.s fsm(tserie, means, start, frequency)
```

**Arguments**

tserie	A ts object.
means	A numeric vector.
start	Start of original time serie
frequency	Frequency of the original time serie

**Value**

A ts object.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
p <- prep.detrend.s fsm(AirPassengers)
postp.detrend.s fsm(p$tserie,p$means,start(AirPassengers),frequency(AirPassengers))
```

---

postp.homogenize.boxcox  
                          *Undo Box-Cox transformation*

---

**Description**

Undo Box-Cox transformation

**Usage**

```
postp.homogenize.boxcox(tserie, lambda)
```

**Arguments**

tserie        A ts object.  
lambda        A numeric.

**Value**

A ts object.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
p <- prep.homogenize.boxcox(AirPassengers)
postp.homogenize.boxcox(p$tserie,p$lambda)
```

---

postp.homogenize.log    *Undo logarithmic transformation*

---

**Description**

Uses exponent to reverse the logarithm

**Usage**

```
postp.homogenize.log(tserie)
```

**Arguments**

tserie        A ts object.

**Value**

A ts object.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
postp.homogenize.log(prepare.homogenize.log(AirPassengers))
```

---

pred *Predictions*

---

**Description**

Performs predictions over a trained model.

**Usage**

```
pred(model = NULL, n.ahead = 20, tserie = NULL, predictions = NULL)
```

**Arguments**

model	A modl object. Contains the trained model we want to predict with.
n.ahead	Number of values to predict ahead of the end of the original time serie. Default value is 20. Must be lower than 100.
tserie	A ts object.
predictions	A ts object.

**Details**

Predicts future values over a "modl" object which can be ARIMA or data mining, and returns the predictions. Data mining predictions start right after the last value contained in the training data, so they overlap with the end of the original.

The object contains only two time series: the original one and the predictions. You can just set these series aswell.

**Value**

A list is returned of class pred containing:

tserie	Original time serie.
predictions	Time serie with the predictions.

**Author(s)**

Alberto Vico Moreno

**See Also**

[modl.pred.arima](#), [pred.dataMining](#), [pred.compareModels](#)

**Examples**

```
prediction <- pred(model=modl(prepare(AirPassengers)),n.ahead=25)
pred(tserie=prediction$tserie, predictions=prediction$predictions)
```

pred.arima

*Predicts for ARIMA*

---

**Description**

Performs predictions over an ARIMA model using the `stats::predict` function.

**Usage**

```
pred.arima(model, n.ahead)
```

**Arguments**

`model` An ARIMA model.  
`n.ahead` Number of values to predict.

**Value**

A `ts` object containing the predictions.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
pred.arima(forecast::auto.arima(prepare(AirPassengers)$tseries), n.ahead=30)
```

---

pred.compareModels*Compare different predictions*

---

**Description**

Plots the original time serie along with 2-5 predictive models.

**Usage**

```
pred.compareModels(originalTS, p_1, p_2, p_3 = NULL, p_4 = NULL,  
  p_5 = NULL, legendNames = NULL, colors = NULL, legend = TRUE,  
  legendPosition = NULL, yAxis = "Values", title = "Predictions")
```

**Arguments**

originalTS	A ts object
p_1	A ts object
p_2	A ts object
p_3	A ts object. Default is NULL.
p_4	A ts object. Default is NULL.
p_5	A ts object. Default is NULL.
legendNames	String vector with the names for the legend. Has to be same length as number of time series we are plotting(including the original one). Default is NULL.
colors	Vector with the colors. Has to be same length as number of time series we are plotting(including the original one). Default is NULL.
legend	A logical. Do we want a legend? Default is TRUE.
legendPosition	A string with the position of the legend (bottomright, topright, ...). Default is NULL.
yAxis	A string. Name for the y axis. "Values" as default.
title	A string. Title for the plot. Default is "Predictions".

**Details**

This function aims to ease the comparison between different predictive models by plotting them into the same graphic.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
data(AirPassengers)
#pre-processing
p <- prep(AirPassengers)
#modelling
arima.modl <- modl(p)
cart.modl <- modl(p,method='dataMining',algorithm='rpart')
#predicting
arima.pred <- pred(arima.modl,n.ahead=30)
cart.pred <- pred(cart.modl,n.ahead=45)
#post-processing
arima.pred <- postp(arima.pred,p)
cart.pred <- postp(cart.pred,p)
#visual comparison
pred.compareModels(AirPassengers,arima.pred$predictions,card.pred$predictions
,legendNames=c('AirPassengers','ARIMA','CART'),yAxis='Passengers',legendPosition = 'topleft')
```

---

pred.dataMining	<i>Predicts for data mining methods</i>
-----------------	---

---

**Description**

Performs predictions over a data mining model using the `caret::predict.train` function.

**Usage**

```
pred.dataMining(model, n.ahead)
```

**Arguments**

model	A modl object.
n.ahead	Number of values to predict.

**Value**

A `ts` object containing the predictions.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
m <- modl(prepare(AirPassengers),method='dataMining',algorithm='rpart')
pred.dataMining(m,n.ahead=15)
```

---

prep	<i>Automatic pre-preprocessing</i>
------	------------------------------------

---

**Description**

This function performs pre-processing on a time series object(`ts`) to treat heterocedasticity, trend and seasonality in order to make the serie stationary.

**Usage**

```
prep(tserie, homogenize.method = "log", detrend.method = "differencing",
     nd = NULL, deseason.method = "differencing", nsd = NULL,
     detrend.first = TRUE)
```

**Arguments**

<code>tserie</code>	A ts object.
<code>homogenize.method</code>	A string. Current methods available are "log" and "boxcox". Method "log" is set as default. If you don't want to perform this transformation, set method as "none".
<code>detrend.method</code>	A string. Current methods available are "differencing" and "sfsm". Method "differencing" is set as default. If you don't want to perform this transformation, set method as "none".
<code>nd</code>	A number. Number of differences you want to apply to the "differencing" detrending method. As default its value is NULL, which means nd will be calculated internally.
<code>deseason.method</code>	A string. Current methods available are "differencing". Method "differencing" is set as default. If you don't want to perform this transformation, set method as "none".
<code>nsd</code>	A number. Number of seasonal differences you want to apply to the "differencing" deseasoning method. As default its value is NULL, which means nsd will be calculated internally.
<code>detrend.first</code>	A boolean. TRUE if detrending method is applied first, then deseasoning. FALSE if deseasoning method is applied first. Default is TRUE.

**Details**

Returns an object `prep` which stores all data needed to undo the changes later on.

This function provides an automatic way of pre-processing based on unit root tests, but this is not the perfect way to do it. You should always check manually if the given time serie is actually stationary, and modify the parameters according to your thoughts.

**Value**

A list is returned of class `prep` containing:

<code>tserie</code>	Processed ts object.
<code>homogenize.method</code>	Method used for homogenizing.
<code>detrend.method</code>	Method used for detrending.
<code>nd</code>	Number of differences used on detrending through differencing.
<code>firstvalues</code>	First nd values of the original series.
<code>deseason.method</code>	Method used for deseasoning.
<code>nsd</code>	Number of seasonal differences used on deseasoning through differencing.
<code>firstseasons</code>	First nsd seasons of the original series.
<code>detrend.first</code>	Processed ts object

means	Vector of means used in "sfsm" detrending method.
lambda	Coefficient used in "boxcox" transformation.
start	Start of the original time serie.
length	Length of the original time serie.

**Author(s)**

Alberto Vico Moreno

**References**

<https://www.otexts.org/fpp/8/1>

**See Also**

[prep.homogenize.log](#), [prep.homogenize.boxcox](#), [prep.detrend.differencing](#), [prep.detrend.sfsm](#), [prep.deseason.differencing](#), [prep.check.acf](#), [prep.check.adf](#)

**Examples**

```
prep(AirPassengers)
prep(AirPassengers, homogenize.method='boxcox', detrend.method='none')
```

---

prep.check.acf      *Autocorrelation function*

---

**Description**

Plots the autocorrelation function to check stationarity

**Usage**

```
prep.check.acf(tserie)
```

**Arguments**

tserie      a ts or a prep object

**Details**

For a stationary time series, the ACF will drop to zero relatively quickly, while the ACF of non-stationary data decreases slowly. Also, for non-stationary data, the value is often large and positive.

**Examples**

```
prep.check.acf(AirPassengers)
prep.check.acf(prepare(AirPassengers))
```

---

```
prep.check.adf
```

*Augmented Dickey-Fuller test*

---

**Description**

Performs ADF test just as another tool to check stationarity.

**Usage**

```
prep.check.adf(tserie)
```

**Arguments**

tserie            a ts or a prep object

**Details**

Shows the results of an ADF test. A p-value<0.05 suggests the data is stationary.

**Examples**

```
prep.check.adf(AirPassengers)
prep.check.adf(prepare(AirPassengers))
```

---

```
prep.deseason.differencing
```

*Deseason with differencing method*

---

**Description**

Performs differencing with lag=frequency.

**Usage**

```
prep.deseason.differencing(tserie, nsd = NULL)
```

**Arguments**

tserie            a ts object

nsd                number of seasonal differences to apply. As default its value is NULL; in this case, the function will perform an automatic estimation of nsd.

**Details**

If no number of differences is specified, the function will make an estimation of the number of differences needed based on unit root test provided by `forecast::nsdiffs`

**Value**

A list is returned containing:

tserie	Transformed ts object.
nsd	Number of seasonal differencies applied.
firstseasons	Lost values after differencing.

**Examples**

```
prep.deseason.differencing(AirPassengers)
prep.deseason.differencing(AirPassengers,nsd=2)
```

---

```
prep.detrend.differencing
```

*Detrend with differencing method*

---

**Description**

Performs differencing with lag=1.

**Usage**

```
prep.detrend.differencing(tserie, nd = NULL)
```

**Arguments**

tserie	a ts object
nd	number of differences to apply. As default its value is NULL; in this case, the function will perform an automatic estimation of nd.

**Details**

If no number of differences is specified, the function will make an estimation of the number of differences needed based on unit root test provided by `forecast::ndiffs`

**Value**

A list is returned containing:

tserie	Transformed ts object.
nd	Number of differencies applied.
firstvalues	Lost values after differencing.

**Examples**

```
prep.detrend.differencing(AirPassengers)
prep.detrend.differencing(AirPassengers,nd=2)
```

---

prep.detrend.s fsm      *Detrend with "subtracting full-season means" method*

---

**Description**

Performs "subtracting full-season means" method to go for a totally automatic approach.

**Usage**

```
prep.detrend.s fsm(tserie)
```

**Arguments**

tserie            a ts object

**Details**

Under this detrending scheme, a series is first split into segments. The length of the segments is equal to the length of seasonality(12 for monthly). The mean of the historical observations within each of these segments is subtracted from every historical observation in the segment. To get the detrended serie we do:  $ds = xi - m$  Being  $xi$  the actual values on the time series and  $m$  the mean of the segment of  $xi$

**Value**

A list is returned containing:

tserie            Transformed ts object.  
means            Vector containing the historical means.

**Examples**

```
prep.detrend.s fsm(AirPassengers)
```

---

prep.homogenize.boxcox  
*Box-Cox transformation*

---

**Description**

Performs a Box-Cox transformation to a time serie.

**Usage**

```
prep.homogenize.boxcox(tserie)
```

**Arguments**

tserie            a ts object

**Value**

A list is returned containing:

boxcox            Transformed ts object.  
lambda            Lambda value.

**References**

Box-Cox transformation: [https://en.wikipedia.org/wiki/Power\\_transform#Box.E2.80.93Cox\\_transformation](https://en.wikipedia.org/wiki/Power_transform#Box.E2.80.93Cox_transformation)

**Examples**

```
prep.homogenize.log(AirPassengers)
```

---

prep.homogenize.log    *Logarithmic transformation*

---

**Description**

Performs a logarithmic transformation to a time serie.

**Usage**

```
prep.homogenize.log(tserie)
```

**Arguments**

tserie            a ts object

**Value**

ts object with transformed time serie

**Examples**

```
prep.homogenize.log(AirPassengers)
```

---

print.modl	<i>Generic function</i>
------------	-------------------------

---

**Description**

Prints object modl

**Usage**

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

**Arguments**

x	prep object
...	ignored

**Examples**

```
print(modl(prepare(AirPassengers)))
```

---

print.pred	<i>Generic function</i>
------------	-------------------------

---

**Description**

Prints object pred

**Usage**

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

**Arguments**

x	prep object
...	ignored

**Examples**

```
print(pred(modl(prepare(AirPassengers))))
```

print.prep

*Generic function*

---

**Description**

Prints object prep

**Usage**

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

**Arguments**

x	prep object
...	ignored

**Examples**

```
print(prepare(AirPassengers))
```

---

summary.modl

*Generic function*

---

**Description**

Summary of object modl

**Usage**

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

**Arguments**

object	prep object
...	ignored

**Examples**

```
summary(modl(prepare(AirPassengers)))
```

---

summary.pred	<i>Generic function</i>
--------------	-------------------------

---

**Description**

Summary of object pred

**Usage**

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

**Arguments**

object	prep object
...	ignored

**Examples**

```
summary(pred(modl(prep(AirPassengers))))
```

---

summary.prep	<i>Generic function</i>
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---

**Description**

Summary of object prep

**Usage**

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

**Arguments**

object	prep object
...	ignored

**Examples**

```
summary(prep(AirPassengers))
```

# Index

modl, [2](#), [13](#)  
modl.arima, [4](#), [4](#)  
modl.dataMining, [4](#), [5](#)  
modl.trControl, [4](#), [6](#)  
modl.tsToDataFrame, [4](#), [7](#)

plot.pred, [7](#)  
plot.prep, [8](#)  
postp, [9](#)  
postp.deseason.differencing, [9](#), [9](#)  
postp.detrend.differencing, [9](#), [10](#)  
postp.detrend.s fsm, [9](#), [11](#)  
postp.homogenize.boxcox, [9](#), [11](#)  
postp.homogenize.log, [9](#), [12](#)  
pred, [9](#), [13](#)  
pred.arima, [13](#), [14](#)  
pred.compareModels, [13](#), [14](#)  
pred.dataMining, [13](#), [16](#)  
prep, [4](#), [9](#), [16](#)  
prep.check.acf, [18](#), [18](#)  
prep.check.adf, [18](#), [19](#)  
prep.deseason.differencing, [18](#), [19](#)  
prep.detrend.differencing, [18](#), [20](#)  
prep.detrend.s fsm, [18](#), [21](#)  
prep.homogenize.boxcox, [18](#), [21](#)  
prep.homogenize.log, [18](#), [22](#)  
print.modl, [23](#)  
print.pred, [23](#)  
print.prep, [24](#)

summary.modl, [24](#)  
summary.pred, [25](#)  
summary.prep, [25](#)