

# 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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<b>modl</b>	<i>Building predictive models</i>
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---

**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 which 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 starts 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 known 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, which 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 which 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.arima](#), [modl.tsToDataFrame](#), [modl.trControl](#), [modl.dataMining](#)

**Examples**

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

---

**modl.arima**

*Automatic ARIMA model*

---

**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))
```

<code>modl.trControl</code>	<i>Control the splitting to train the data</i>
-----------------------------	--

## Description

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

## Usage

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

## Arguments

<code>initialWindow</code>	An integer. The initial number of consecutive values in each training set sample. Default value: 30.
<code>horizon</code>	An integer. The number of consecutive values in test set sample. Default value: 15.
<code>fixedWindow</code>	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.
<code>givenSummary</code>	A logical. Indicates if it should be used the customized <code>summaryFunction(?trainControl</code> for more info) <code>modl.sumFunction</code> or not. Default is FALSE; this will use default <code>caret</code> metrics.

## Details

We always split using method "timeslice", which 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 which 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**

<code>x</code>	pred object
<code>ylab</code>	ylab
<code>main</code>	main
<code>...</code>	ignored

**Examples**

```
plot(pred(mod1(prep(AirPassengers))))
```

**plot.prep***Generic function***Description**

Plots object prep

**Usage**

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

**Arguments**

<code>x</code>	prep object
<code>ylab</code>	ylab
<code>xlab</code>	xlab
<code>...</code>	ignored

**Examples**

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

---

postp

*Post-processing of pre-processed data*

---

## 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.sfsm](#), [postp.deseason.differencing](#)

## Examples

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

---

postp.deseason.differencing

*Undo deseason(differencing)*

---

## Description

Uses inverse seasonal differences to reverse the changes

## Usage

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

**Arguments**

<code>tserie</code>	A <code>ts</code> object.
<code>nsd</code>	Number of seasonal differences.
<code>firstseasons</code>	Values lost on the original differences
<code>frequency</code>	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**

<code>tserie</code>	A <code>ts</code> object.
<code>nd</code>	Number of differences.
<code>firstvalues</code>	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.sfsm	<i>Undo detrend(substracting full-means method)</i>
--------------------	---

**Description**

Undo detrend(substracting full-means method)

**Usage**

```
postp.detrend.sfsm(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.sfsm(AirPassengers)
postp.detrend.sfsm(p$tserie,p$means,start(AirPassengers),frequency(AirPassengers))
```

postp.homogenize.boxcox	<i>Undo Box-Cox transformation</i>
-------------------------	------------------------------------

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

---

pred	<i>Predictions</i>
------	--------------------

---

## 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 ve 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(prep(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**

<code>model</code>	An ARIMA model.
<code>n.ahead</code>	Number of values to predict.

**Value**

A `ts` object containing the predictions.

**Author(s)**

Alberto Vico Moreno

**Examples**

```
pred.arima(forecast::auto.arima(prep(AirPassengers)$tserie), 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 comparation 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$predictions,cart.pred$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

<code>model</code>	A <code>modl</code> object.
<code>n.ahead</code>	Number of values to predict.

### Value

A `ts` object containing the predictions.

### Author(s)

Alberto Vico Moreno

### Examples

```
m <- modl(prep(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

<code>means</code>	Vector of means used in "sfsm" detrending method.
<code>lambda</code>	Coefficient used in "boxcox" transformation.
<code>start</code>	Start of the original time serie.
<code>length</code>	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**

<code>tserie</code>	a <code>ts</code> or a <code>prep</code> 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(prep(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(prep(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:

<code>tserie</code>	Transformed ts object.
<code>nsd</code>	Number of seasonal differences applied.
<code>firstseasons</code>	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**

<code>tserie</code>	a ts object
<code>nd</code>	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:

<code>tserie</code>	Transformed ts object.
<code>nd</code>	Number of differences applied.
<code>firstvalues</code>	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:

<code>tserie</code>	Transformed <code>ts</code> object.
<code>means</code>	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:

<b>boxcox</b>	Transformed ts object.
<b>lambda</b>	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

*Generic function*

---

### Description

Prints object modl

### Usage

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

### Arguments

x	prep object
...	ignored

### Examples

```
print(modl(prep(AirPassengers)))
```

---

---

print.pred

*Generic function*

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### Description

Prints object pred

### Usage

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

### Arguments

x	prep object
...	ignored

### Examples

```
print(pred(modl(prep(AirPassengers))))
```

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print.prep	<i>Generic function</i>
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**Description**

Prints object prep

**Usage**

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

**Arguments**

x	prep object
...	ignored

**Examples**

```
print(prep(AirPassengers))
```

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summary.modl	<i>Generic function</i>
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**Description**

Summary of object modl

**Usage**

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

**Arguments**

object	prep object
...	ignored

**Examples**

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

---

summary.pred

*Generic function*

---

### 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

*Generic function*

---

### Description

Summary of object prep

### Usage

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

### Arguments

object	prep object
...	ignored

### Examples

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

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