

Package ‘lambdaTS’

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Type Package

Title Variational Seq2Seq Model with Lambda Transformer for Time Series Analysis

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Author Giancarlo Vercellino

Maintainer Giancarlo Vercellino <giancarlo.vercellino@gmail.com>

Description

Time series analysis based on lambda transformer and variational seq2seq, built on 'Torch'.

License GPL-3

Encoding UTF-8

LazyData true

RoxygenNote 7.1.1

Depends R (>= 3.6)

Imports purrr, abind, ggplot2, readr, stringr, lubridate, narray, bestNormalize, fANCOVA, imputeTS, modeest, scales, tictoc, bizdays, torch

NeedsCompilation no

Repository CRAN

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bitcoin_gold_oil *bitcoin_gold_oil data set*

Description

A data frame with different time series (prices and volumes) for bitcoin, gold and oil.

Usage

```
bitcoin_gold_oil
```

Format

A data frame with 18 columns and 1827 rows.

Source

Yahoo Finance

lambdaTS *lambdaTS: Variational Seq2Seq Lambda Transformer Model for Time Series Analysis*

Description

Time series analysis based on Lambda Transformer and Variational Seq2Seq, built on 'Torch'.

Usage

```
lambdaTS(  
  data,  
  target,  
  future,  
  past = future,  
  ci = 0.8,  
  deriv = 1,  
  yjt = TRUE,  
  shift = 0,  
  smoother = FALSE,  
  k_embed = 30,  
  r_proj = ceiling(k_embed/3) + 1,  
  n_heads = 1,  
  n_bases = 1,  
  activ = "linear",  
  loss_metric = "elbo",  
  optim = "adam",
```

```

    epochs = 30,
    lr = 0.01,
    patience = epochs,
    verbose = TRUE,
    sample_n = 100,
    seed = 42,
    dev = "cpu",
    starting_date = NULL,
    dbreak = NULL,
    days_off = NULL,
    min_set = future,
    holdout = 0.5,
    batch_size = 30
  )

```

Arguments

<code>data</code>	A data frame with ts on columns and possibly a date column (not mandatory)
<code>target</code>	String. Time series names to be jointly analyzed within the seq2seq model
<code>future</code>	Positive integer. The future dimension with number of time-steps to be predicted
<code>past</code>	Positive integer. The past dimension with number of time-steps in the past used for the prediction. Default: future
<code>ci</code>	Confidence interval. Default: 0.8
<code>deriv</code>	Positive integer. Number of differentiation operations to perform on the original series. 0 = no change; 1: one diff; 2: two diff, and so on.
<code>yjt</code>	Logical. Performing Yeo-Johnson Transformation on data is always advisable, especially when dealing with different ts at different scales. Default: TRUE
<code>shift</code>	Vector of positive integers. Allow for target variables to shift ahead of time. Zero means no shift. Length must be equal to the number of targets.
<code>smoother</code>	Logical. Perform optimal smooting using standard loess. Default: FALSE
<code>k_embed</code>	Positive integer. Number of Time2Vec embedding dimensions. Minimum value is 2. Default: 30
<code>r_proj</code>	Positive integer. Number of dimensions for the reduction space (to reduce quadratic complexity). Must be largely less than k_embed size. Default: ceiling(k_embed/3) + 1
<code>n_heads</code>	Positive integer. Number of heads for the attention mechanism. Computationally expensive, use with care. Default: 1
<code>n_bases</code>	Positive integer. Number of normal curves to build on each parameter. With more than one base you can model be Default: 1
<code>activ</code>	String. The activation function for the linear transformation of the attention matrix into the future sequence. Implemented options are: "linear", "leaky_relu", "celu", "elu", "gelu", "selu", "softplus", "bent", "snake", "softmax", "softmin", "softsign", "sigmoid", "tanh", "tanhshrink", "swish", "hardtanh", "mish". Default: "linear".

loss_metric	String. Loss function for the variational model. Two options: "elbo" or "crps". Default: "crps".
optim	String. Optimization methods available are: "adadelat", "adagrad", "rmsprop", "rprop", "sgd", "asgd", "adam". Default: "adam".
epochs	Positive integer. Default: 30.
lr	Positive numeric. Learning rate. Default: 0.01.
patience	Positive integer. Waiting time (in epochs) before evaluating the overfit performance.
verbose	Logical. Default: TRUE
sample_n	Positive integer. Number of samples from the variational model to evaluate the mean forecast values. Computationally expensive, use with care. Default: 100.
seed	Random seed. Default: 42.
dev	String. Torch implementation of computational platform: "cpu" or "cuda" (gpu). Default: "cpu".
starting_date	Date. Initial date to assign temporal values to the series. Default: NULL (progressive numbers).
dbreak	String. Minimum time marker for x-axis, in liberal form: i.e., "3 months", "1 week", "20 days". Default: NULL.
days_off	String. Weekdays to exclude (i.e., c("saturday", "sunday")). Default: NULL.
min_set	Positive integer. Minimum number for validation set in case of automatic resize of past dimension. Default: future.
holdout	Positive numeric. Percentage of time series for holdout validation. Default: 0.5.
batch_size	Positive integer. Default: 30.

Value

This function returns a list including:

- prediction: a table with quantile predictions, mean and std for each ts
- history: plot of loss during the training process for the joint-transformed ts
- plot: graph with history and prediction for each ts
- learning_error: errors for the joint-transformed ts (rmse, mae, mdae, mpe, mape, smape, rrse, rae)
- feature_errors: errors for each ts (rmse, mae, mdae, mpe, mape, smape, rrse, rae)
- pred_stats: some stats on predicted ts (average iqr, iqr ratio t/1, approx upside probability)
- time_log

Author(s)

Giancarlo Vercellino <giancarlo.vercellino@gmail.com>

Examples

```
## Not run:  
lambdaTS(bitcoin_gold_oil, c("bitcoin_Close", "gold_close", "oil_Close"), 30, 30, deriv = 2)  
  
## End(Not run)
```

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