

Package ‘EMDANNhybrid’

January 20, 2025

Type Package

Title Empirical Mode Decomposition Based Artificial Neural Network Model

Version 0.2.0

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Depends EMD,forecast

Description

Application of empirical mode decomposition based artificial neural network model for nonlinear and non stationary univariate time series forecasting. For method details see (i) Choudhury (2019) <<https://www.indianjournals.com/ijor.aspx?target=ijor:ijee3&volume=55&issue=1&article=013>>; (ii) Das (2020) <<https://www.indianjournals.com/ijor.aspx?target=ijor:ijee3&volume=56&issue=2&article=002>>.

Encoding UTF-8

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Suggests knitr, rmarkdown, testthat (>= 3.0.0)

VignetteBuilder knitr

Config/testthat/edition 3

NeedsCompilation no

Repository CRAN

Date/Publication 2023-09-14 10:32:42 UTC

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 EMDANNhybrid

New Ensemble Hybrid Machine Learning Model

Description

The researchers can use this package to fit empirical mode decomposition and artificial neural network (also known as TDNN) based hybrid model for nonlinear and non stationary time series data. It will also provide you with accuracy measures along with an option to select the proportion of training and testing data sets. User can get to choose appropriate lag with tuning parameter like maximum iterations for training the neural model.

Usage

```
EMDANNhybrid(data,k,l,n,r,m)
```

Arguments

data	A univariate time series data
k	Partition value for splitting the data set into training and testing sets
l	The lag length for fitting neural network model
n	Size of the hidden node for fitting neural network model
r	Number of networks to fit with different random starting weights
m	Maximum number of iterations for fitting neural network model

Details

The package implements an ensemble hybrid approach for forecasting nonlinear and nonstationary time series data proposed by Choudhary et al. (2019), Das (2019) and Das et al. (2020,2022,2023). In this method, EMD is to disintegrate a non-stationary and nonlinear time series data into several simple modes (IMFs and residue). Each of these modes further forecasted using artificial neural network model. finally the all forecasted values are aggregated for final forecast value.

Value

Prediction_Accuracy_EMDANN

List of performance measures of the fitted EMDANN model.

Final_Prediction_EMDANN

Final forecasted value of the VMD based ANN model. It is obtained by combining the forecasted value of all individual IMF and residue.

Author(s)

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References

- Das,P., Jha, G. K., Lama, A., Parsad, R. and Mishra, D. (2020). Empirical Mode Decomposition based Support Vector Regression for Agricultural Price Forecasting. Indian Journal of Extension Education, 56(2): 7-12. (<https://www.indianjournals.com/ijor.aspx?target=ijor:ijee3&volume=56&issue=2&article=002>).
- Das, P., Jha, G. K. and Lama, A. (2023). Empirical Mode Decomposition Based Ensemble Hybrid Machine Learning Models for Agricultural Commodity Price Forecasting. Statistics and Applications. 21(1),99-112.(https://ssca.org.in/media/9_SA31042022_R1_SA_17042022_FINAL_Finally_Pankaj_Das_Empirical_
- Das, P., Jha, G. K., Lama, A. and Bharti (2022). EMD-SVR Hybrid Machine Learning Model and its Application in Agricultural Price Forecasting. Bhartiya Krishi Anusandhan Patrika. (DOI: 10.18805/BKAP385)
- Das, P. (2019). Study On Machine Learning Techniques Based Hybrid Model for Forecasting in Agriculture. Unpublished Ph.D. Thesis.
- Choudhury, K., Jha, G. K., Das, P. and Chaturvedi, K. K. (2019). Forecasting Potato Price using Ensemble Artificial Neural Networks. Indian Journal of Extension Education, 55(1):71-77. (<https://www.indianjournals.com/ijor.aspx?target=ijor:ijee3&volume=55&issue=1&article=013>).

See Also

EMD,nnfor,EMDSVRhybrid

Examples

```
set.seed(6)
data = rnorm(300,6.6,.36)
EMDANNhybrid(data,0.7,1,5,20,120)
```

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