

# Package ‘CEEMDANML’

April 7, 2023

**Type** Package

**Title** CEEMDAN Decomposition Based Hybrid Machine Learning Models

**Version** 0.1.0

**Author** Mr. Sandip Garai [aut, cre],  
Dr. Ranjit Kumar Paul [aut],  
Dr. Md Yeasin [aut]

**Maintainer** Mr. Sandip Garai <sandipnicksandy@gmail.com>

**Description** Noise in the time-series data significantly affects the accuracy of the Machine Learning (ML) models (Artificial Neural Network and Support Vector Regression are considered here). Complete Ensemble Empirical Mode Decomposition with Adaptive Noise (CEEMDAN) decomposes the time series data into sub-series and help to improve the model performance. The models can achieve higher prediction accuracy than the traditional ML models. Two models have been provided here for time series forecasting. More information may be obtained from Garai and Paul (2023) <[doi:10.1016/j.iswa.2023.200202](https://doi.org/10.1016/j.iswa.2023.200202)>.

**License** GPL-3

**Encoding** UTF-8

**Imports** stats, Rlibeemd, tseries, forecast, fGarch, aTSA, FinTS, LSTS,  
earth, caret, neuralnet, e1071, pso

**RoxygenNote** 7.2.1

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2023-04-07 08:30:02 UTC

## R topics documented:

carigaan . . . . .	2
carigas . . . . .	3

<b>Index</b>	<b>4</b>
--------------	----------

---

carigaan	<i>CEEMDAN Decomposition-Based ARIMA-GARCH-ANN Hybrid Modeling</i>
----------	--

---

### Description

CEEMDAN Decomposition-Based ARIMA-GARCH-ANN Hybrid Modeling

### Usage

```
carigaan(Y, ratio = 0.9, n_lag = 4)
```

### Arguments

Y	Univariate time series
ratio	Ratio of number of observations in training and testing sets
n_lag	Lag of the provided time series data

### Value

- Train\_fitted: Train fitted result
- Test\_predicted: Test predicted result
- Accuracy: Accuracy

### References

- Garai, S., & Paul, R. K. (2023). Development of MCS based-ensemble models using CEEMDAN decomposition and machine intelligence. *Intelligent Systems with Applications*, 18, 200202
- Garai, S., Paul, R. K., Rakshit, D., Yeasin, M., Paul, A. K., Roy, H. S., Barman, S. & Manjunatha, B. (2023). An MRA Based MLR Model for Forecasting Indian Annual Rainfall Using Large Scale Climate Indices. *International Journal of Environment and Climate Change*, 13(5), 137-150.

### Examples

```
Y <- rnorm(100, 100, 10)
result <- carigaan(Y, ratio = 0.8, n_lag = 4)
```

---

carigas	<i>CEEMDAN Decomposition-Based ARIMA-GARCH-SVR Hybrid Modeling</i>
---------	--

---

**Description**

CEEMDAN Decomposition-Based ARIMA-GARCH-SVR Hybrid Modeling

**Usage**

```
carigas(Y, ratio = 0.9, n_lag = 4)
```

**Arguments**

Y	Univariate time series
ratio	Ratio of number of observations in training and testing sets
n_lag	Lag of the provided time series data

**Value**

- Train\_fitted: Train fitted result
- Test\_predicted: Test predicted result
- Accuracy: Accuracy

**References**

- Garai, S., & Paul, R. K. (2023). Development of MCS based-ensemble models using CEEMDAN decomposition and machine intelligence. *Intelligent Systems with Applications*, 18, 200202
- Garai, S., Paul, R. K., Rakshit, D., Yeasin, M., Paul, A. K., Roy, H. S., Barman, S. & Manjunatha, B. (2023). An MRA Based MLR Model for Forecasting Indian Annual Rainfall Using Large Scale Climate Indices. *International Journal of Environment and Climate Change*, 13(5), 137-150.

**Examples**

```
Y <- rnorm(100, 100, 10)
result <- carigas(Y, ratio = 0.8, n_lag = 4)
```

# Index

carigaan, 2  
carigas, 3