

Implementation of Neural Network Autoregression (NNAR) and Double Exponential Smoothing Holt Method in Forecasting PT. Telkom Indonesia Stock Price

Michela Sheryl Noven, Winita Sulandari , Respatiwulan

Department of Statistics, Faculty of Math and Science, Sebelas Maret University, Indonesia

Abbreviated abstract: Stock price is a financial time series that usually formed from non-dynamic linear process that can be used as an indicator that shows management succession in managing a company. The variability in this series is high so linear forecasting methods are not recommended for modeling. The aim of this study is implementing neural network autoregression (NNAR) method that is useful for forecasting non-linear series and Double Exponential Smoothing Holt method that is suitable for trend time series.

Related publications:

- W. Sulandari *et al*, AIP Conference Proceedings 2329 (1),06008 (2021)
- W. Sulandari and H. Utami, International Journal of Advances in Intelligent Informatics 2 (3), 131-139 (2016)



Problem and Challenge

Stock Price

Indicator of **success** in managing a company

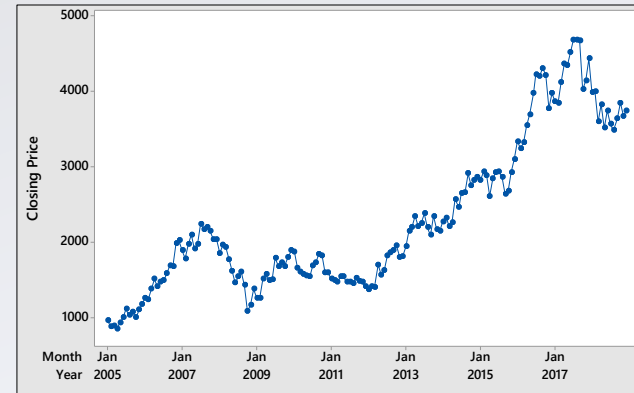


formed from a **dynamic non-linear process** with high variability (volatility) dependence



linear forecasting methods are **not recommended**

PT. Telkom Indonesia stock price is one of some popular stocks that have **great demand** from investors in Indonesia



Solution:

Double Exponential Smoothing Holt

- Focuses on exponential weight loss on the oldest observation
- Recent observations will be given a higher weight value than older observations
- Able to smooth trends and gradients by using smoothing constants for levels and trends
- Good for forecasting data with trend

Neural Network Autoregression (NNAR)

- Built based on an artificial neural network model on the performance of the central system neural network
- Good for estimating functions on large training data
- Able to model complex non-linear relationships without any underlying relationship assumptions

The aim of this study are **implementing and comparing** NNAR and Double Exponential Smoothing Holt method

michelasn33@student.uns.ac.id - 2

Methods

Splitting data to training and testing

Data exploration

NNAR modeling

Double Exponential Smoothing Holt modeling

Determining the auto-regression order in by examining non-linear relationship in training data

Determining the α and β constant

Training model

Training model

Residual diagnostic test

Residual diagnostic test

Evaluating error (RMSE,MAPE,MPE) in both training and testing model

No auto-correlations?

Normal distribution?

No auto-correlations?

Normal distribution?

Testing forecast

Testing forecast

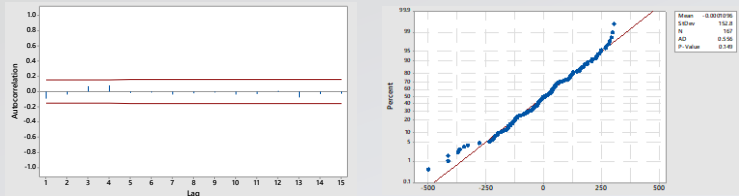
Comparing results

Choose the best model

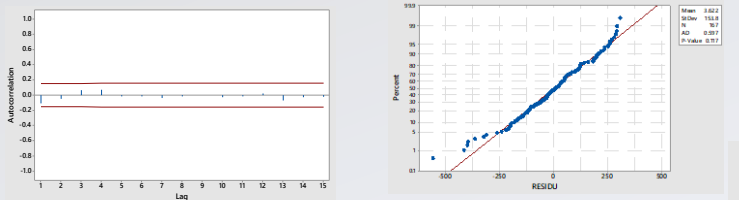


Results and Conclusions

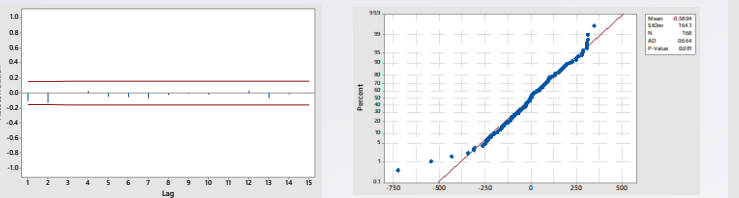
NNAR(1,1)



Without Box-Cox transformation



With Box-Cox transformation
Double Exponential Smoothing Holt

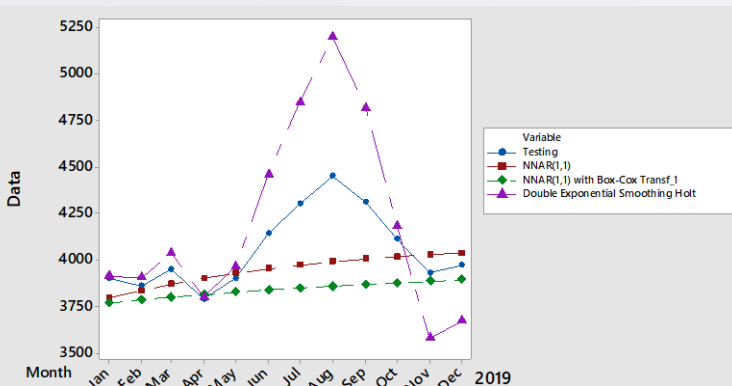


$\alpha=0.9$ $\beta=0.2$

micheln33@student.uns.ac.id - 4

MODEL	TRAINING			TESTING		
	RMSE	MAPE(%)	MPE (%)	RMSE	MAPE(%)	MPE (%)
NNAR(1,1)	3.119	5.420	-0.499	3.124	3.765	2.476
NNAR(1,1) with Box-Cox transform	3.122	5.411	-0.299	4.311	5.168	5.073
Double Exponential Smoothing Holt	3.317	5.669	-0.074	5.304	6.047	-3.291

Conclusion:



NNAR(1,1) offers **better performance** than Double Exponential Smoothing Holt method