Prediction of Life Expectancy in Maluku Province Using Artificial Neural Networks Backpropagation

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Abstract

Life Expectancy at Birth (LE) is defined as the average estimated number of years a person can live to since their birth. The purpose of LE is to represent the health rate of a community. Backpropagation is an algorithm in artificial neural networks (ANN) used to predict or forecast data. This study aims to predict Life Expectancy in Maluku Province. The results of this research obtained the forecasting success of 99.65% with the smallest error MAPE = 0.0035. Forecasting for the next 5 years shows that the Life Expectancy value tends to increase over the next 5 years from 2019-2023 at 65.7828 in 2019 and increasing to 66.6632 in 2023.

Keywords: Artificial neural network; Backpropagation; Life Expectancy.

1. Introduction

Artificial Neural Network is one of the methods resulting from the development of the world of information technology that is widely used to predict or forecast [2]. This method uses time series data analysis which is a time sequence observation of quantitative characteristics of events that occur [1, 4]. One of architecture in Artificial Neural Networks is Backpropagation. Backpropagation Artificial Neural Networks consist of a group of small processing units that are modeled in basis of human neural networks [11]. Generally, Backpropagation Architecture uses multi-layer network that aims to minimize errors that will be generated at the network output [10, 5]. The more complex the network is built, the better results of the resulting editions [3].

This research will use the Backpropagation Artificial Neural Network method for Prediction of Life Expectancy (LE) in Maluku Province in the future. Life expectancy data is one measure of life quality of a population [7]. The Indonesian government makes LE as an indicator of the assessment of the degree of public health [12]. This is used as an evaluation material for government work programs in the health sector. The higher of Life Expectancy a region has, the better its public health [6].

2. Method

The type of data used in this research is quantitative data. The data used for this research is the Maluku Province Life Expectancy (LE) data from 2010-2018. LE data used in the study were obtained from the Badan Pusat Statistika (BPS) of Maluku Province. Collecting data through direct observation methods in the Badan Pusat Statistika (BPS) of Maluku

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Province and literature studies obtained from literature studies on e-books, scientific journals, and websites. The method used is Backpropagation Neural Network method. The following is a flow chart of the training stages in Figure 1 and testing process stages in Figure 2 of the Backpropagation method.

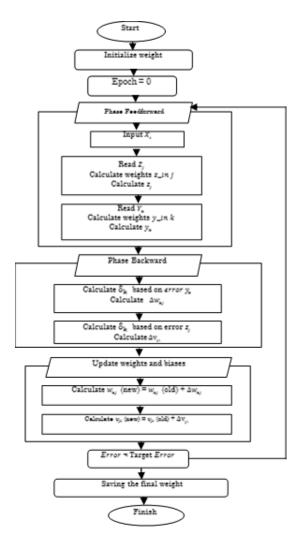


FIGURE 1. Training process flow chart on backpropagation networks[8]

3. Result and Discussion

3.1. **Normalization Data.** In this section, the data of Life Expectancy (LE) in Maluku Province that obtained from Badan Pusat Statistika (BPS) of Maluku Province shown in Table 1:

LE data in Table 1, contains LE Maluku Province data, which then will be normalized in the form of weighting so it can be used in the process of testing and training data using the Backpropagation method. This is the data preparation stage. Data normalized into interval [0,1; 0.9] using the formula [9], as follows:

$$x_{new} = \frac{0.8(x - x_{min})}{x_{max} - x_{min}} + 0.1 \tag{1}$$

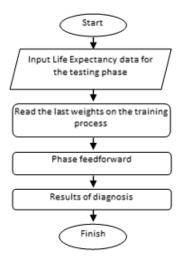


FIGURE 2. Testing process flow chart on backpropagation networks ([8])

Table 1. Life Expectancy (LE) Indicator Data

Year	Life Expectancy
2010	64.46
2011	64.61
2012	64.77
2013	64.93
2014	65.01
2015	65.31
2016	65.35
2017	65.4
2018	65.59

Table 2. Normalized life expectancy indicator data

Tahun	Life Expectancy
2010	0.1
2011	0.2062
2012	0.3195
2013	0.4327
2014	0.4894
2015	0.7018
2016	0.7301
2017	0.7655
2018	0.9

The data that has been transformed by normalization process above (Table 2) will be used as input parameters and the data has gone through the preprocessing stage. At this stage, the data is formed in time series data with a vulnerable time of 2 years as training data and the third year as target data. This data is then made into a pattern or scheme of data sharing in general as in Table 3, below:

Based on Table 3, we obtain 7 data from LE data which will be divided according to the annual data sharing scheme. This data will then be divided into training data and testing data.

X1	X2	Target
2010	2011	2012
Data	Data	Data
2011	2012	2013
Data	Data	Data
2015	2016	2017
Data	Data	Data
2016	2017	2018
Data	Data	Data

Table 3. Data division scheme each year

Table 4. Distributions of normalized annual life expectancy indicator data

X1	X2	Target
0.1	0.2062	0.3195
0.2062	0.3195	0.4327
0.3195	0.4327	0.4894
0.4327	0.4894	0.7018
0.4894	0.7018	0.7301
0.7018	0.7301	0.7655
0.7301	0.7655	0.9

The data in Table 4, above is then divided into training data and test data with a percentage of 80% (5 data) and 20% (2 data) respectively. Therefore, from the total data of 7 data, it is divided into 5 training data and 2 data of testing data for each LE data and the indicators are given in Table 5-6. This division was taken after going through several stages of testing the distribution of training data and testing data, the best distribution composition was obtained from the accuracy of the prediction results and the smallest error is 80% and 20%.

Table 5. Normalized annual life expectancy training data scheme

X1	X2	Target
0.1	0.2062	0.3195
0.2062	0.3195	0.4327
0.3195	0.4327	0.4894
0.4327	0.4894	0.7018
0.4894	0.7018	0.7301

Table 6. Normalized annual life expectancy testing data scheme

X1	X2	Target
0.7018	0.7301	0.7655
0.7301	0.7655	0.9

- 3.2. Research Data Processing Stage Training. The training stage of Life Expectancy data with ANN Backpropagation is given as follows:
 - (1) Network Architecture consists of input, two hidden layers, and output
 - (2) Each hidden layer has 3 neurons
 - (3) The initial weight of input to the first hidden layer $net.IW\{1,1\}=$

$$0.4 \quad 0.4 \\ 0.4 \quad 0.4 \\ 0.4 \quad 0.4$$

The initial weights of the first hidden layer to the second hidden layer $net.LW\{2,1\}=$

The initial weight of the second hidden layer to the output $net.LW\{3,2\}=$

$$0.4 \quad 0.4 \quad 0.4$$

The initial bias of the first hidden layer $net.b\{1,1\}$ = The initial bias of the first hidden layer

 $0.4 \\ 0.4 \\ 0.4$

 $net.b{2,1}$ = Initial bias output

 $0.4 \\ 0.4 \\ 0.4$

 $net.b{3,1}=0.4$

(4) Establish parameters for the training process

Learning rate=0.1

Target Error=0.0000001

Maximum epoch=500

The training process is carried out and stops at the 5th epoch with the following results:

3.3. Research Data Processing Stage Testing. After the analysis of training process, then in this part the data is given the analysis of testing stage by Artificial Neural Networks (ANN) Backpropagation: In Table 7, the following is the value of the testing data processing

Original Test Target	ANN Output Target
65.4	65.709
65.59	65.7515

results with MATLAB software. The measurement of forecasting/prediction accuracy is using the following mean absolute percentage error (MAPE) formula:

$$MAPE = \frac{\frac{\sum |e_i|}{x_{asti}} (100\%)}{n} \tag{2}$$

From the calculation is given the value of MAPE = 0.0035. It is means that the success of predictions/forecasting is 99.65%.

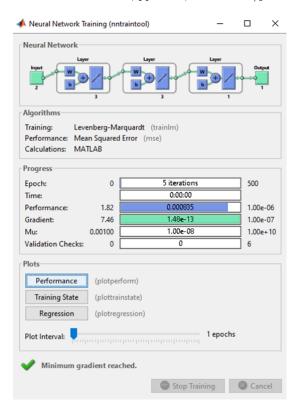


FIGURE 3. Training results graph (training) life expectancy data

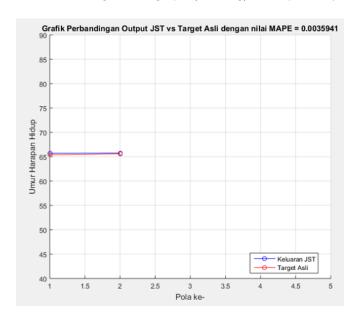


FIGURE 4. The testing result graph for life expectancy data (comparison of prediction result with original targets)

3.4. **LE Data Prediction Stages.** From the results of the testing stages in Table 7 and by using a network that has been trained, we can predict the Life Expectancy in Maluku Province for the next 5 years from 2019-2023. Prediction results can be obtained from the processing using Matlab Software, as follows:

 Year
 Prediction Result of Life Expectancy

 2019
 65.7828

 2020
 65.9924

 2021
 66.2027

 2022
 66.4344

 2023
 66.6632

Table 7. Predicted results of life expectancy data for years 2019-2023

Based on the prediction results in Table 8 can be concluded that the value of Life Expectancy tends to increase over the next 5 years started at 65.7828 in 2019 and will increasing to 66.6632 in 2023.

4. Conclusion

Based on the research it can be concluded that the average prediction/forecasting success is 99.65%, with MAPE = 0.0035, and the learning rate value is 0.1. The architecture of this research network uses input layer 2 neurons, hidden layer 3-3-1, and output layer 1 neurons, with epoch/iteration: 500. In addition, the Forecasting results Life expectancy (LE) tends to increase over 5 years from 2019-2023 at 65.7828 in 2019 and will increasing to 66.6632 in 2023. It means that the government of Maluku Province has improved the health status of the community, including improving access to and quality of health services.

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