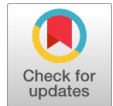


The Prediction Of Tourist Visiting With Average Based Fuzzy Time Series Method



Joan Angelina Widians, Novianti Puspitasari, Andi Famela Anggita Sari

ABSTRACT--- This research aims to build a system which able to predict tourist visiting in the Province of East Kalimantan with Average Based Fuzzy Time Series method. Data collection in this research conducted by direct observation and interview with the Secretary on Duty of the East Kalimantan Province Tourism Department. This research uses the total numbers of tourist that visited the Province of East Kalimantan from 2008 through the year 2016 to predict tourist visit has the following years. In this research, value of MAPE is used to count the value of error or the imprecision of prediction result, so that the result of prediction will reach close to. The result of the building system is the prediction result for international tourist visiting with the MAPE value of 0,77375%, which is included in the criteria of 'very good', therefore it can be concluded that this system can be used to predict tourist visiting the Province of East Kalimantan.

Keywords: Average Based; Fuzzy Time Series; MAPE value; Prediction; Tourist

I. INTRODUCTION

East Kalimantan Province is one of tourist destination that has interesting potential tourism which compared to Indonesia. The number of tourists who visiting the province of East Kalimantan has increased every year. The increasing of the number of tourist visiting, the government and the business community in tourism need to prepare various facilities and public infrastructure for tourists. It is done for the convenience of the tourists in order they will interested to come back visit to East Kalimantan Province. The prediction of the number of tourist who visit East Kalimantan Province needs to be accomplished, so the provincial government can anticipate the burst of tourist numbers. Therefore, it requires a system that is able to help Department of Tourism to predict tourist visiting for the future time. The Average Based Fuzzy Time Series method has been widely used as a tool to provide predictions in various purposes. For example the use of method is forecasting inbound tourism demand to Istanbul [1], to the tourism demand forecasting in China [2], forecasting tourist arrivals [3] and many more [4], [5]. The method in this

study is Based Fuzzy Time Series which is a simple calculation method that has a simple structure. [6]–[8].

II. EXPERIMENTAL DETAILS

In this study, the Fuzzy Time Series method is a data forecasting method using fuzzy principles as the basis for which the length of the interval is determined by the average based method [9]–[11]. The algorithm in the Average Based Fuzzy Time Series method used is as follows: (a) Input followings the historical data that will be predicted. (b) Determine the maximum data value (D_{max}) and minimum value data value (D_{min}) from historical data. Then, find the values $D1$ and $D2$ where $D1$ and $D2$ are two appropriate and appropriate positive numbers. (c) Define Universe of Discourse (U) or set of universe with rule $U = [D_{min} - D1, D_{max} + D2]$. Furthermore, determine the length interval (I) of the Universe of Discourse (U) using the Average Based method. The next step is calculate the number of intervals formed (m) using the formula (1).

$$m = \frac{(D_{max} + D1 - D_{min} + D2)}{I} \quad (1)$$

As pointed in (1), m = the number of intervals, L = length of interval. (d) Divide the Universe of Discourse (U) into ($U_1, U_2, U_3, \dots, U_m$) with the same interval length (I) as much as the interval formed (m). (e) Determine the fuzzy membership set (A_1, A_2, \dots, A_m) of the entire Universe of Discourse (U) defined by the number of intervals (m) defined in formula (2).

$$\begin{aligned} A_1 &= \frac{1}{u_1} + \frac{0.5}{u_2} + \frac{0}{u_3} + \frac{0}{u_4} + \dots + \frac{0}{u_m} & A_2 &= \frac{0.5}{u_1} + \frac{1}{u_2} + \frac{0.5}{u_3} \\ &+ \frac{0}{u_4} + \dots + \frac{0}{u_m} & A_3 &= \frac{0}{u_1} + \frac{0.5}{u_2} + \frac{1}{u_3} + \frac{0.5}{u_4} + \dots + \frac{0}{u_m} \\ \dots & & A_m &= \frac{0}{u_1} + \frac{0}{u_2} + \dots + \frac{0}{u_{m-2}} + \frac{0.5}{u_{m-1}} + \frac{1}{u_m} \end{aligned} \quad (2)$$

The next step is perform fuzzification historical data, by giving the fuzzy membership set for historical data. For example, if the actual data lies on the U_t range, then the data belongs to the fuzzy A_t membership set, then Fuzzy Logical Relationship (FLR) is determined based on existing historical data. Determine Fuzzy Logical Relationship (FLR) for all historical data, where Fuzzy A_k means if the value of Fuzzy Set in year is Logical Relationship A_j , A_j then in year $i + 1$ is A_k . A_j as the left side of the relationship is called the current state and A_k as the right side of the relationship is called the next state. If there is a recurrence of the relationship then still counted once. Then, Build Fuzzy Logical Relationship Groups (FLRG). Namely by dividing Fuzzy Logical Relationship (FLR) that has been obtained into several sections or groups based on the same left side as follows: $A_j \rightarrow A_{k1}, A_j \rightarrow A_{k2}$ Then it will be $\rightarrow A_{k1}, A_{k2}$. The last, calculate the predicted output. At this

Manuscript published on 30 May 2019.

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stage will be tested validation of the calculation results to ensure the results of the system has been in accordance with the needs of the Department of Tourism using Mean Absolute Percentage Error (MAPE) method. MAPE measures absolute error as average absolute error in the number of actual data periods. This can avoid problems in the interpretation of measurements of accuracy relative to the magnitude of the actual value and predictive value. The calculation of Mean Absolute Percentage Error can be seen in the following formula: $MAPE = \frac{1}{n} (\sum Absolute\ Percent\ Error)$ Where: N = amount of actual data. The resulting value through this evaluation indicates the ability of forecasting. Where, Absolute Percentage Error value below 20% is good, and if the value less than 10% is declared very good.

III. RESULT AND DISCUSSION

The method of testing of Average Based Fuzzy Time Series that tested on the data of tourists visiting the province of East Kalimantan. This study uses the data of foreign tourists visiting from 2008 to 2016 as test data. The fuzzyfication results denoted into linguistic numbers can be seen in Table 1. Furthermore, in process of forecasting with fuzzy time series models, fuzzy logical relationship (FLR) is one of the most critical factors that influence the forecasting accuracy [6]. Fuzzy logical relationship (FLR) is the stage after data fuzzification, FLR in this research is $A_1 \rightarrow A_4$, $A_4 \rightarrow A_5$, $A_5 \rightarrow A_{10}$, $A_{10} \rightarrow A_9$, $A_9 \rightarrow A_{13}$, $A_{13} \rightarrow A_{34}$, $A_{34} \rightarrow A_{30}$, $A_{30} \rightarrow A_{51}$. Then Fuzzy Logical Relationship (FLR) which has been formed, can be arranged into several groups based on the same left side (current state). The formed Fuzzy Logical Relationship Groups (FLRG) can be seen in Figure 1.

Table 1: Fuzzyfication of data on the number of visits of foreign tourists

Year	Actual Total	Fuzzy Set
2008	20.142	A1
2009	23.768	A4
2010	24.410	A5
2011	29.768	A10
2012	28.273	A9
2013	32.973	A13
2014	53.257	A34
2015	49.285	A30
2016	70.997	A51

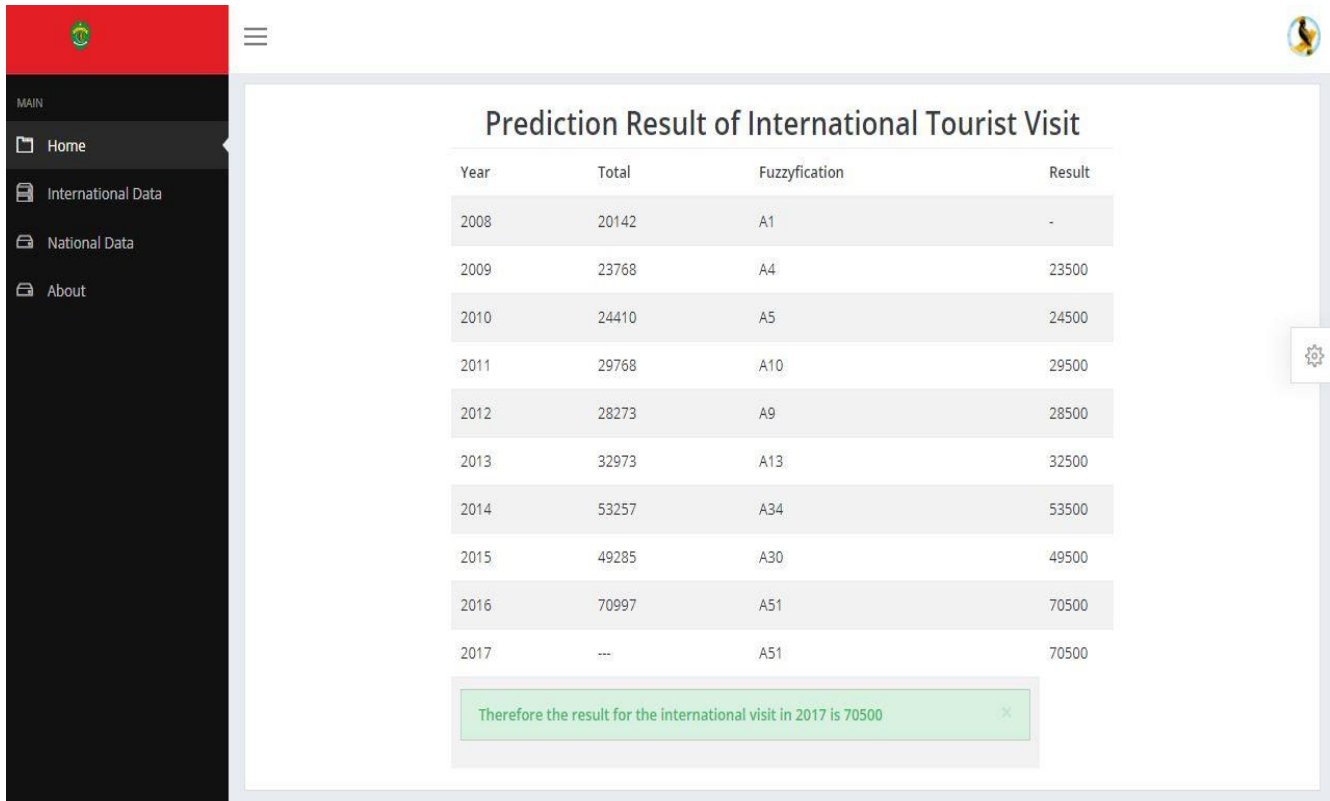
Group 1	$A1 = A1 \rightarrow A4$
Group 2	$A4 = A4 \rightarrow A5$
Group 3	$A5 = A5 \rightarrow A10$
Group 4	$A9 = A9 \rightarrow A13$
Group 5	$A10 = A10 \rightarrow A9$
Group 6	$A13 = A13 \rightarrow A34$
Group 7	$A30 = A30 \rightarrow A51$
Group 8	$A34 = A34 \rightarrow A30$
Group 9	$A51 = \emptyset$

Fig. 1: Fuzzy logical relationship groups (FLRG Aboard)

The next step is calculate the prediction result. In 2017, due to the fuzzyfication result of the number of foreign tourist arrivals in 2016 in Table 1 is A51, and from Figure 1 it can be seen that A51 does not have Fuzzy Logical Relationship Group (FLRG) with current state that formed A51:A51 the maximum membership value for the A51 fuzzy set occurs at the interval U51, with $U51 = [70,000, 71,000]$, and the middle value of the U51 interval is 70,500. As a result the prediction of the number of foreign tourists who visit the province of East Kalimantan in 2017 is 70,500 can be seen in Table 2. The prediction result for foreign tourist visiting on prediction system can be seen on Figure 2. Then, calculate Absolute Percent Error (APE) from foreign data for each year that has actual number and predicted results from 2009 to 2016. In 2008 it cannot be calculated, as it has no predicted results. After obtaining the Absolute Percent Error result from the foreign data, then the MAPE value obtained for foreign data with the amount of data $n = 8$ is 0.77375%. This indicates that prediction results for foreign data are included in very good criteria based on Table 2 because the MAPE value is obtained below 10%.

Table 2: Prediction result

Year	Actual Total	Prediction Result
2008	20.142	-
2009	23.768	23.500
2010	24.410	24.500
2011	29.768	29.500
2012	28.273	28.500
2013	32.973	32.500
2014	53.257	53.500
2015	49.285	49.500
2016	70.997	70.500
2017	-	70.500



IV. CONCLUSION

The conclusion obtained from this research are: 1) This system is able to give very good prediction result in the prediction process of tourist visit to East Kalimantan Province using method of Average Based Fuzzy Time Series so that predictions obtained more precise, computerized and prevent decision making which is subjective. 2) Using the Average Based Fuzzy Time Series method, the MAPE value is 0.77375% for foreign data and 4.26% for the archipelago data, both of which are included in the criteria very well. Because the MAPE result is less than 10%.

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