

Electrical Daily Load Forecasting in Ramadhan Using Interval Type-2 Fuzzy Logic in Sulselrabar System

By Marhatang Marhatang

Electrical Daily Load Forecasting In Ramadhan Using Type-2 Fuzzy Logic In Sulselrabar System

Marhatang^{a1}, Muhammad Ruswandi Djalal^{a2}, Herman Nauwir^{a3}, Sonong^{a4}

21
^aEnergy Engineering, State Polytechnic of Ujung Pandang
Jalan Perintis Kemerdekaan km.10, Makassar
E-mail : ¹marhatang@gmail.com, ²wandi@poliupg.ac.id, ³hermannauwir@poliupg.ac.id,
⁴sonong@poliupg.ac.id,

Abstract

This study discusses the daily electricity load forecasting 24 hours on 150 kV electric power systems Sulselrabar. Forecasting electrical load requires the accuracy of the results with a small error. Peak load forecasting methods used to use smart methods Interval Type-1 Fuzzy Logic (IT1FL) and Interval Type-2 Fuzzy Logic (IT2FL) to predict the needs of the electrical load 1 Ramadhan 2016. As input data, it was used load data from 2012 through 2016 for the same day each 1st of Ramadhan each year, and as comparative data, it was used actual load data 1, 2016. For the Ramadhan input variable, it was used two of the data Variation Load Difference (VLD Max) 2015 as an input variable X, VLD Max 2016 as an input variable Y. From the simulation results obtained highly accurate results where each method produces a very small error, where for methods of using IT1FL of 1.607778264% while using IT2FL by, 1.344510913%.

9
Keywords: Type-1 Fuzzy Logic, Type-2 Fuzzy Logic, MAPE, Load Forecasting 22

1. Introduction

Electric load forecasting is an important part of power system operation in order to achieve optimal planning in operation of the systems [1]. Load forecasting is covering short-term, medium-term and long-term load forecasting. Short-term load forecasting required for controlling and scheduling the operation of power systems [2]. Medium and long-term load forecasting is required for maintenance, fuel purchases, plant development and planning of future distributions. Accurate load forecasting has a significant impact on the operation and production costs of electric utilities [3]. Research on load forecasting has spawned numerous papers and journals [4]. These publications have led to the development of various methods of forecasting. This method is classified into two categories: The classical approach (conventional method) and an artificial intelligence method.

10
The classical approach is based on statistical methods, which cannot accurately represent the complex nonlinear relationship between the load and a series of factors such as daily and weekly rhythms of time that can lead to high error in load forecasting [4]. Artificial intelligence method has the ability to provide better performance when dealing with nonlinear data. The advantages of artificial intelligence method compared to conventional method are computational technique and simple algorithm, structural simplicity and high accuracy performance without having to solve any nonlinear equations into mathematical equations. Therefore, the author in this research discusses the hybrid method in the load forecasting, which is a suggestion of earlier researchers. Thus the method of interval type 2 fuzzy inference system used in this research. Interval type-2 fuzzy inference system (IT2FIS) becomes a concern short-term load forecasting because it has a simple concept and high-performance identification.

11
IT2FIS is the formulation and mapping process from input to output using interval type 2 fuzzy logic [5-9]. One of the advantages of fuzzy logic is the knowledge and experience of experts can be easily used and applied. Interval Type-1 Fuzzy Logic and Interval Type-2 Fuzzy Logic is used in this research for load forecasting in Sulawesi Selatan, Tenggara dan Barat (Sulselrabar) system especially for 1 Ramadhan 2016. In the proposed method, we do not take environmental

factors as variable. The Sulsebarab electrical system is used because, this system has been growing, and requires further study on load forecasting. Several previous studies have been conducted and show satisfactory results [9-21].

2. Research Methods

The implementation of IT2FL for peak load forecasting on 1 Ramadhan 2016 is done by using three stages, namely the preparation stage (pre-processing), processing stage and final stage (post-processing) [4].

2.1. Pre-Processing

Preparation stage is the preparation of peak load data on 24 hours to look for load difference (LD), typical load difference (TLD), maximum weekdays (max WD) and variation load difference (VLD). Load difference (LD) for maximum load is a load difference within 4 days before the days which is given by [22]:

$$LD_{MAX}(i) = \frac{MaxSD(i) - MaxWD(i)}{MaxWD(i)} \times 100 \quad (1)$$

$$MaxWD(i) = \frac{WD(i)d-4 + WD(i)d-3 + WD(i)d-2 + WD(i)d-1}{4} \quad (2)$$

MaxSD (i) is the peak load on a special day and maxWD is the average of maximum load 4 days before the days. Then, looking for a distinctive characteristic of a typical peak load or typical load difference (TLD_{MAX} (i)) by averaging the peak load of similar LD_{MAX} (i) in previous years. After that, calculating the variation load difference, which is the difference between Load Difference (LD) and Typical Load Difference (TLD_{MAX} (i)) which can be seen by the following equation:

$$VLD_{max}(i) = LD_{max}(i) - TLD_{max}(i) \quad (3)$$

$$TLD_{max}(i) = \frac{LD_{max}(i-1) + LD_{max}(i-2) + LD_{max}(i-3)}{3} \quad (4)$$

Peak load data which used to calculate Max WD and LD max is based on (1) and (2) equations respectively and the results are presented in Table 1 and 2.

Table 1. Peak Load In 2016

WD(i) _{d-4}	WD(i) _{d-3}	WD(i) _{d-2}	WD(i) _{d-1}	MaxSD(i)
577.96	536.22	583.10	589.64	609.70
562.64	513.60	560.86	563.12	606.52
537.60	497.91	527.11	541.81	615.86
517.76	498.68	516.53	533.25	641.13
526.03	489.66	525.30	546.27	596.93
539.42	528.80	550.95	571.02	591.33
536.83	529.59	558.15	567.28	520.18
559.59	573.80	584.02	595.88	574.02
599.36	617.64	634.73	649.16	627.04
587.65	655.20	658.25	692.32	657.29
614.61	689.41	682.15	686.51	656.71
614.24	689.49	675.38	682.78	659.18
611.61	683.15	663.73	694.33	664.00
612.52	704.85	692.95	710.65	675.02
608.56	698.42	676.79	691.70	691.70
614.76	681.74	661.68	701.46	695.61
603.86	651.71	661.77	677.62	695.79
723.27	754.12	783.38	741.25	770.25
816.40	836.67	842.27	853.60	856.00
801.50	821.69	791.02	815.15	812.24

767.76	792.92	772.03	817.63	793.92
700.07	733.94	705.36	782.02	759.78
636.80	662.42	663.73	769.47	694.37
580.44	610.82	615.25	680.07	628.03

2.2. Processing

Fuzzyfication design of X and Y input is using IT2MF Editor. There are 11 membership functions is used [23], namely :

- Negative Very Big (NVB), range : [-48 -48 -40 -32.5 -48 -40 -28.5 -48]
- Negative Big (NB), range : [-40.5 -32 -24.5 -36.5 -32 -20.5]
- Negative Medium (NM), range : [-32.5 -24 -16.5 -28.5 -24 -12.5]
- Negative Small (NS), range : [-24.5 -16 -8.5 -20.5 -16 -4.5]
- Negative Very Small (NVS), range : [-16.5 -8 -2.5 -12.5 -8 2.5]
- Zero (ZE), range : [-8.5 0 4.5 -4.5 0 8.5]
- Positive Very Small (PVS), range : [-2.5 8 12.5 2.5 8 16.5]
- Positive Small (PS), range : [4.5 16 20.5 8.5 16 24.5]
- Positive Medium (PM), range : [12.5 24 28.5 16.5 24 32.5]
- Positive Big (PB), range : [20.5 32 36.5 24.5 32 40.5]
- Positive Very Big (PVB), range : [28.5 40 48 32.5 40 48 48]

31
Examples of fuzzy rules can be seen in Table 2.

Table 2. Fuzzy Rules

No. Rules	Antecedent X	Y	Consequent Z
1	NM	PS	PS
2	PVB	NS	PVB
3	NM	PM	PM
4	NM	PB	PB
5	NS	PM	PM
6	NS	PS	PS
7	NM	ZE	ZE
8	NM	PVS	PVS
9	NVB	ZE	ZE
10	NVB	ZE	ZE
11	NVB	NVS	NVS
12	NVB	ZE	ZE
13	NVB	ZE	ZE
14	NVB	ZE	ZE
15	NVB	PVS	PVS
16	NVB	PVS	PVS
17	NM	PS	PS
18	NM	PVS	PVS
19	NS	PVS	PVS
20	NS	PVS	PVS
21	NS	PVS	PVS
22	NS	PVS	PVS
23	NS	ZE	ZE
24	ZE	ZE	ZE

2.3. Post-Processing

After getting VLD_{MAX} forecasting value, then forecast load difference:

$$Forecast\ LD_{MAX}(i) = Forecast\ VLD_{MAX}(i) + TLD_{MAX}(i) \quad (5)$$

Peak load forecasting can be calculated:

$$P'_{MAX}(i) = MaxWD(i) + \frac{(ForecastLD_{MAX} \times MaxWD(i))}{100} \quad (6)$$

The smaller error obtained show the accuracy of the proposed method is higher. The absolute error can be expressed as follows:

$$Error = \left| \frac{P_{forecast} - P_{actual}}{P_{actual}} \right| \times 100\% \quad (7)$$

$$Error = \left| \frac{P_{MAX}(i) - MaxSD(i)}{MaxSD(i)} \right| \times 100\% \quad (8)$$

The research flowchart is shown in the following figure.

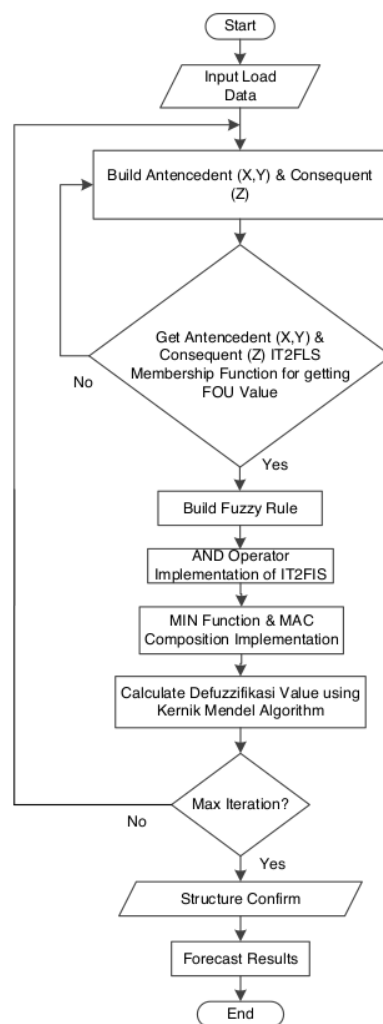


Figure 1. Flowchart IT2FL for Daily Peak Load Forecasting

3. Literature Review

3.1. Fuzzy Logic Type-2

The fuzzy type-2 set is a development of fuzzy type-1 which is re-defuzzy. The Fuzzy type-1 based-knowledge logic system is used to build the rules in an uncertainty fuzzy logic system (FLS). There are three reasons for uncertainty rules [6] :

- 1 Rules of antecedents and consequents can have different perception in different people.
- 2 Polling of group of experts on consequents is often different to the same rules as most experts do not agree on the rule.
- 3 The training data contains a lot of noise.

Type-2 fuzzy sets have their own membership levels are fuzzy. Rankings on type-2 fuzzy set can be on the subset of secondary membership. Similar with FLS Type-1, FLS Type-2 is also included FIS membership functions and defuzzification. The difference is that before the defuzzification process there is type-2 reduction process which has several methods; one of them is Kernik Mendel Algorithm (KMA). Interval Type-2 Fuzzy Logic (IT2FL) structure can be seen in Figure 2. Figure 2 shows the process of IT2FL from an input value of crisp x set into the output value of $Y=f(x)$ equation.

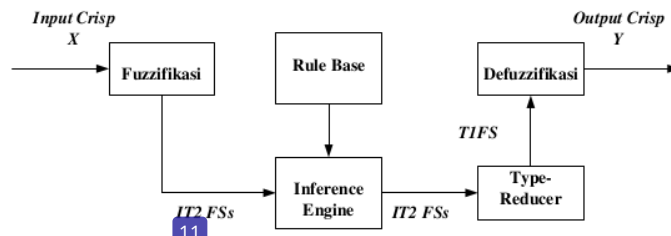


Figure 2. Type-2 Fuzzy Logic System (T2FLS) Structure

3.2. Interval Type-2 Fuzzy Set

An interval type-2 fuzzy set (IT2FS) is denoted \tilde{A} by the membership function $\mu_{\tilde{A}}$ with $x \in X$ and $u \in J_x \subseteq [0,1]$ its characteristic can be recognized on the following equation:

$$\tilde{A} = \int_{x \in X} \int_{u \in J_x} \frac{\mu_{\tilde{A}}(x, u)}{(x, u)} J_x \subseteq [0,1] \quad (9)$$

x is a primary variable; $u \in U$, secondary variable, have domain J_x for each $x \in X$; J_x is primary membership. certainty of \tilde{A} is the combination primary membership (footprint of uncertainty). The equation can be seen as follows:

$$FOU(\tilde{A}) = \bigcup_{x \in X} J_x = \{(x, u); u \in J_x \subseteq [0,1]\} \quad (10)$$

J_x is an interval with the following equation:

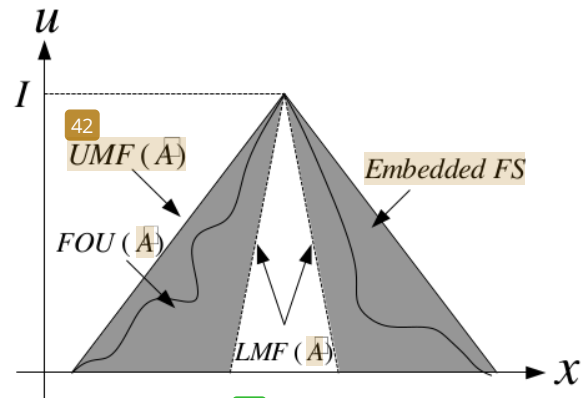
$$J_x = \{(x, u); u \in [\underline{\mu}_{\tilde{A}}(x), \bar{\mu}_{\tilde{A}}(x)]\} \quad (11)$$

From equation 2.5 $FOU(\tilde{A})$ can be expressed by the equation:

$$FOU(\tilde{A}) = \bigcup_{x \in X} [\underline{\mu}_{\tilde{A}}(x), \bar{\mu}_{\tilde{A}}(x)] \quad (12)$$

Where:

- J_x = Primary membership of x
- $\underline{\mu}_{\tilde{A}}$ = Lower Membership Function (LMF) of \tilde{A}
- $\bar{\mu}_{\tilde{A}}$ = Upper Membership Function (UMF) of \tilde{A}



29
Figure 3. FOU (dark color), LMF (dotted line), UMF (solid line) and Embedded FS (wavy line).

39 3.3. Interval Type-2 Fuzzy Membership Function Operations

24
Operation on fuzzy interval type-2 set is almost the same as fuzzy type-1 set; but on the IT2FL logic system, the operation performed on two intervals that are UMF (top) and LMF (below) at once. Operation on fuzzy interval type-2 membership function can be seen in Figure 4:

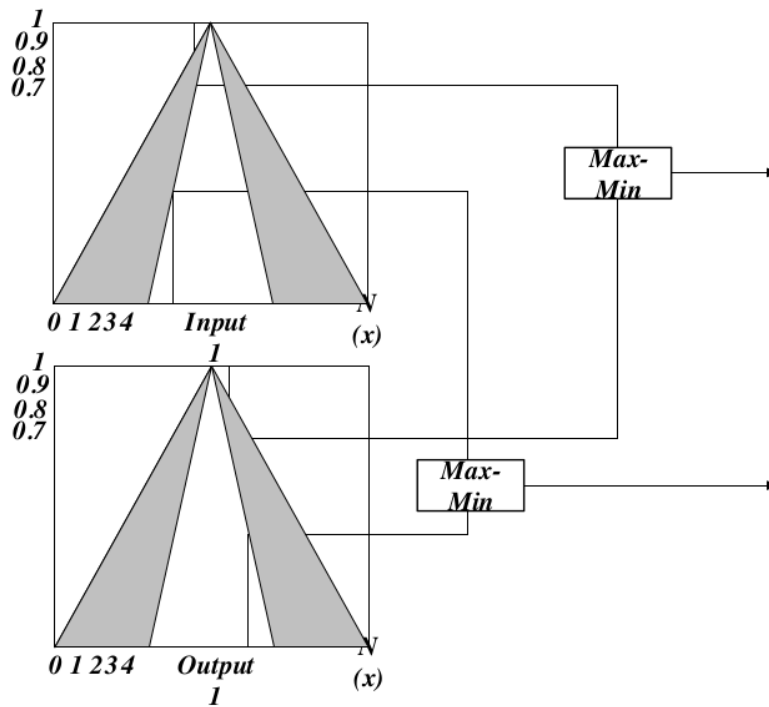


Figure 4. Operation fuzzy set interval type-2 (IT2FL)

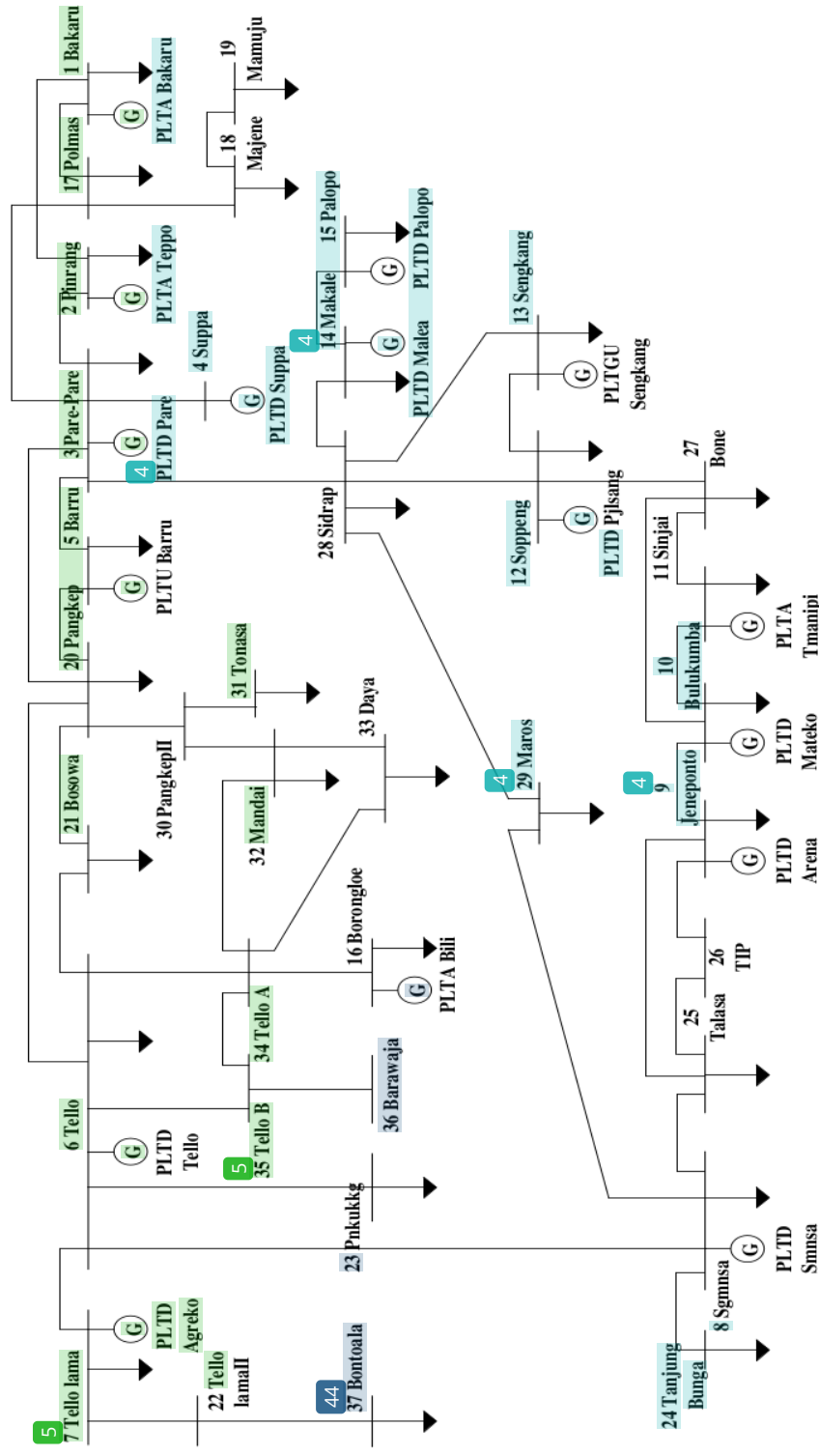


Figure 5. Sulselrabar System [10]

Table 3. Establishment Of Rule Base For Input X in 1st Ramadhan 2016

Hours	Variable	VLD max	Membership Function (μ)				Set of X
			NB	NM	PVS	PS	
01.00	X	-13.53477757	0.383694394	0.616305606			NM
	Y	7.837201199			0.0406997	0.9593003	PS
	Z	7.837201199			0.0406997	0.9593003	PS

Table 4. Result Of Variable Calculations X, Y, Z on 1st Ramadhan 2016

Hours	Input			Set		
	X	Y	Z	X	Y	Z
1:00	-13.53477757	7.837201199	7.837201199	NM	46	PS
2:00	38.15805202	-8.455067268	-8.455067268	PVB	NS	NS
3:00	-12.34897699	12.81561102	12.81561102	NM	PM	PM
4:00	-10.98277044	15.86782032	15.86782032	NM	PB	PB
5:00	-9.099179456	10.65770448	10.65770448	NS	PM	PM
6:00	-7.434924909	9.002263816	9.002263816	NS	PS	PS
7:00	-11.37068292	0.269638493	0.269638493	NM	ZE	ZE
8:00	-12.03990371	3.199737038	3.199737038	NM	26	PVS
9:00	-19.64995022	0.863689423	0.863689423	NVB	ZE	ZE
10:00	-19.60150714	1.756933675	1.756933675	NVB	ZE	ZE
11:00	-22.75197853	-2.897867872	-2.897867872	NVB	NVS	NVS
12:00	-20.16793366	-0.76171919	-0.76171919	NVB	ZE	ZE
13:00	-18.72397279	1.320215573	1.320215573	NVB	ZE	ZE
14:00	-23.01970881	0.154211021	0.154211021	NVB	ZE	40
15:00	-19.25924082	3.424255509	3.424255509	NVB	PVS	PVS
16:00	-18.30164779	4.941932377	4.941932377	NVB	PVS	PVS
17:00	-11.51601435	7.053862708	7.053862708	NM	PS	PS
18:00	-10.31446966	3.604203806	3.604203806	NM	PVS	PVS
19:00	-7.106373861	3.350674093	3.350674093	NS	PVS	PVS
20:00	-7.094262262	4.663216896	4.663216896	NS	PVS	PVS
21:00	-6.418655252	2.380709895	2.380709895	NS	PVS	PVS
22:00	-9.138939847	2.765256248	2.765256248	NS	7	PVS
23:00	-7.372420856	1.397545706	1.397545706	NS	ZE	ZE
0:00	1.060461042	1.539682191	1.539682191	ZE	ZE	ZE

Antecedent (X, Y) and consequent (Z) T2FIS figures as follows:

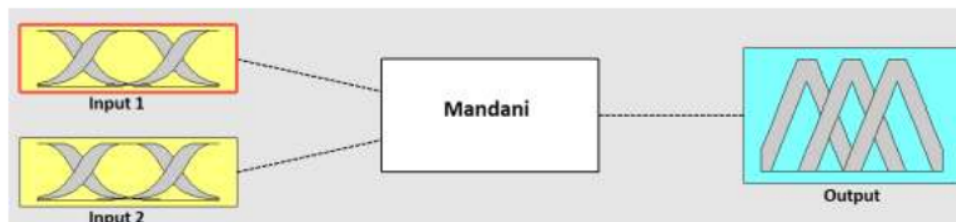


Figure 6. Design System

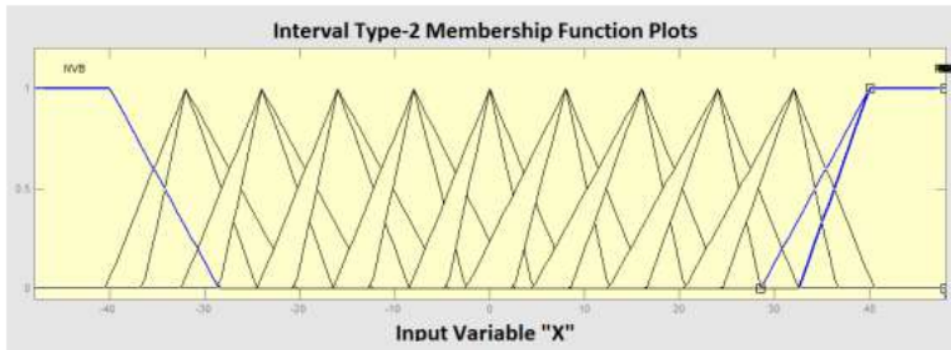


Figure 7. X,Y Input Design

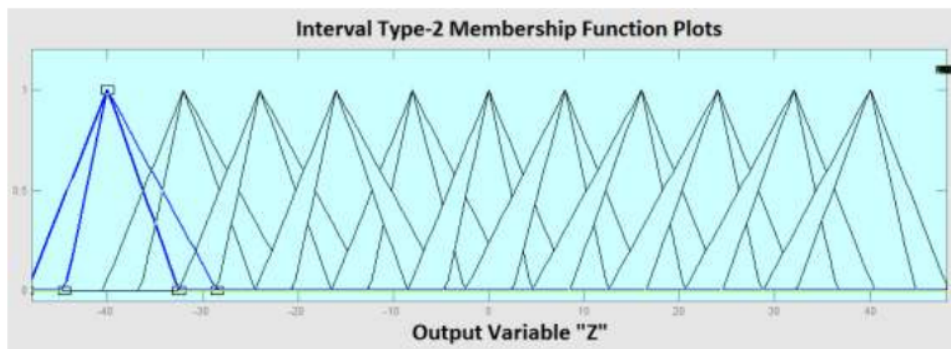


Figure 8. Z Output Design

4. Result & Analysis

The calculation of the input variable value X, Y, Z is to find the value of Load Difference Variable (VLD_{MAX}) by first calculating WD Max, LD Max, TLD_{max} each input data of 2012-2015, which is calculated based on equation 1-4. The results of the calculation of variables X, Y, Z can be seen in table 3 above.

Figure 5 shows the single line diagram of the sulselrabar system, where there are 37 Buses, each serving load centers in the sulselrabar system. Table 3 shows an example of the calculation of the membership function fuzzy logic for 01.00 hours, and Table 4 shows the complete result of the membership function calculation.

Figure 6-8 shows the membership design function type-2 fuzzy logic using Matlab. Where each uses 11 membership functions. While the image forecasting results shown in graphs 8 and 9. Graph 8 is the result of load forecasting and graph 9 is the error of forecasting results with the method of comparison of type-1 fuzzy logic.

The data used is the peak load data of Sulselrabar electricity system started in 2012-2015 by using Interval Type-1 Fuzzy Logic method and Interval Type-2 Fuzzy Logic (IT2FL) as a comparison. Then, the data is devoted to four days before and during 1 Ramadhan 2016.

The test results by using the proposed method as a proposed method for load forecasting showed excellent results, in which the Mean Absolute Percentage Error (MAPE) of VLD_{MAX} is 1.344510913%. By using IT1FL, MAPE is 1.607778264%. For complete results can be seen in figure 9-10.

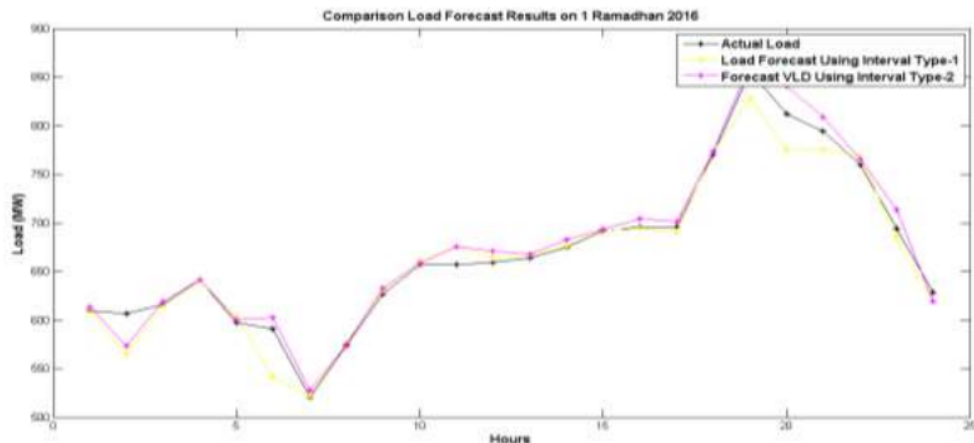


Figure 9. Results of Load Forecast for 1st Ramadhan in 2016

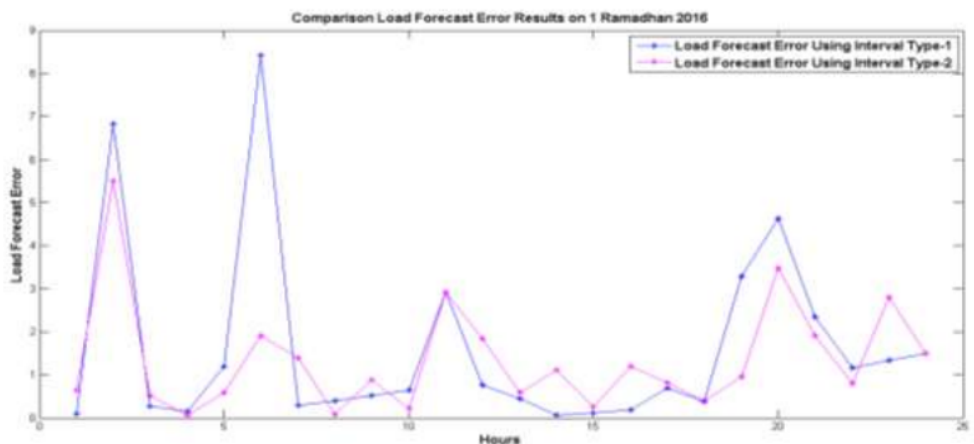


Figure 10. Results of Load Forecasting Error on 1st Ramadhan in 2016

5. Conclusions

Electrical Load Forecasting Day on the 1st of Ramadhan using intelligent methods based on Fuzzy Logic obtained very satisfactory results, with a very small error, this method is best used for short-term forecasting, medium and long-term. Error using Fuzzy Logic Type-2 of 1.607778264%, while using the proposed method Interval Type-2 Fuzzy Logic error is getting smaller in the amount of 1.344510913%. The application of intelligent methods for optimization of load forecasting is also highly recommended for any forecasting methods used by PT. Perusahaan Listrik Negara (PLN) also still produce a sizable error.

References

- [1] A. Srivastava, A. S. Pandey, and D. Singh, "Short-term load forecasting methods: A review," in *Emerging Trends in Electrical Electronics & Sustainable Energy Systems (ICETEESES), International Conference on*, 2016, pp. 130-138.
- [2] A. Jain, E. Srinivas, and S. kumar Kukkadapu, "Fuzzy based day ahead prediction of electric load using mahalanobis distance," in *Power System Technology (POWERCON), 2010 International Conference on*, 2010, pp. 1-6.
- [3] S. K. Panda, S. N. Mohanty, and A. K. Jagadev, "Long Term Electrical Load Forecasting: An Empirical Study across Techniques and Domains," *Indian Journal of Science and Technology*, vol. 10, 2017.

- [4] A. Ramadhani, Agus Dharma, & Imam Robandi, "Optimization FOU of Interval Type-2 Fuzzy Inference System Using Big Bang – Big Crunch Algorithm for Short Term Load Forecasting on National Holiday Case Study: South and Central Kalimantan-Indonesia," *International Review of Electrical Engineering (IREE)*, vol. 10, pp. 123-130, 2015.
- [5] A. Khosravi and S. Nahavandi, "Load forecasting using interval type-2 fuzzy logic systems: Optimal type reduction," *IEEE Transactions on Industrial Informatics*, vol. 10, pp. 1055-1063, 2014.
- [6] J. Zhao and L. Jiang, "Forecasting Of Type-2 Fuzzy Electric Power System Based On Phase Space Reconstruction Model," *network security*, vol. 8, 2015.
- [7] S. Hassan, A. Khosravi, J. Jaafar, and M. A. Khanesar, "Hybrid model for the training of interval type-2 fuzzy logic system," in *International Conference on Neural Information Processing*, 2015, pp. 644-653.
- [8] E. Kayacan, S. Coupland, R. John, and M. A. Khanesar, "Elliptic membership functions and the modeling uncertainty in type-2 fuzzy logic systems as applied to time series prediction," in *Fuzzy Systems (FUZZ-IEEE), 2017 IEEE International Conference on*, 2017, pp. 1-7.
- [9] S. Hassan, A. Khosravi, and J. Jaafar, "Training of interval type-2 fuzzy logic system using extreme learning machine for load forecasting," in *Proceedings of the 9th International Conference on Ubiquitous Information Management and Communication*, 2015, p. 87.
- [10] M. Y. Yunus, M. R. Djalal, and Marhatang, "Optimal Design Power System Stabilizer Using Firefly Algorithm in Interconnected 150 kV Sulselrabar System, Indonesia," *International Review of Electrical Engineering (IREE)*, vol. 12, pp. 250-259, 2017.
- [11] M. R. Djalal, D. Ajiatmo, A. Imran, and I. Robandi, "Desain Optimal Kontroler PID Motor DC Menggunakan Cuckoo Search Algorithm," *SENTIA 2015*, vol. 7, 2015.
- [12] M. R. Djalal, A. Imran, and I. Robandi, "Optimal placement and tuning power system stabilizer using Participation Factor and Imperialist Competitive Algorithm in 150 kV South of Sulawesi system," in *Intelligent Technology and Its Applications (ISITIA), 2015 International Seminar on*, 2015, pp. 147-152.
- [13] M. R. Djalal, H. Nawir, H. Setiadi, and A. Imran, "An Approach Transient Stability Analysis Using Equivalent Impedance Modified in 150 kV South of Sulawesi System," *Journal of Electrical and Electronics Engineering UMSIDA*, vol. 1, pp. 1-7, 2016.
- [14] M. R. Djalal, H. Setiadi, D. Lastomo, and M. Y. Yunus, "Modal Analysis and Stability Enhancement of 150 kV Sulselrabar Electrical System using PSS and RFB based on Cuckoo Search Algorithm," *International Journal on Electrical Engineering and Informatics*, vol. 9, pp. 800-812, 2017.
- [15] M. R. Djalal, M. Y. Yunus, H. Setiadi, and A. U. Krismanto, "Small-Signal-Stability Enhancement using a Power-System Stabilizer based on the Cuckoo-Search Algorithm against Contingency N-1 in the Sulselrabar 150-kV System," *Makara Journal of Technology*, vol. 22, pp. 1-8, 2018.
- [16] M. R. Djalal, M. Y. Yunus, H. Nawir, and A. Imran, "Optimal Design of Power System Stabilizer In Bakaru Power Plant Using Bat Algorithm," *2017*, vol. 1, p. 6, 2017-11-10 2017.
- [17] U. Umoh, I. Umoeka, M. Ntekop, and E. Babalola, "INTERVAL TYPE-2 FUZZY NEURAL NETWORKS FOR SHORT-TERM ELECTRIC LOAD FORECASTING: A COMPARATIVE STUDY."
- [18] N. Ammar, M. Sulaiman, and A. F. M. Nor, "Analysis Load Forecasting of Power System Using of Fuzzy Logic and Artificial Neural Network," *Journal of Telecommunication, Electronic and Computer Engineering (JTEC)*, vol. 9, pp. 181-192, 2017.
- [19] D. Ali, M. Yohanna, P. M. Ijasini, and M. B. Garkida, "Application of fuzzy–Neuro to model weather parameter variability impacts on electrical load based on long-term forecasting," *Alexandria Engineering Journal*, 2017.
- [20] D. Ali, M. Yohanna, P. M. Ijasini, and M. B. Garkida, "Application of fuzzy–Neuro to model weather parameter variability impacts on electrical load based on long-term forecasting," *Alexandria engineering journal*, vol. 57, pp. 223-233, 2018.

- [21] A. T. Ali, E. B. Tayeb, and Z. M. Shamseldin, "Short term Electrical Load Forecasting Using Fuzzy Logic," *International Journal Of Advancement In Engineering Technology, Management and Applied Science (IJAETMAS)*, vol. 3, 2016.
- [22] F. Tuaimah, "Iraqi Short Term Electrical Load Forecasting Based On Interval Type-2 Fuzzy Logic," *World Academy of Science, Engineering and Technology, International Science Index 92, International Journal of Electrical, Computer, Energetic, Electronic and Communication Engineering*, vol. 8, pp. 1255 - 1261, 2014.
- [23] M. R. Djalal and Faisal, "Intelligent Fuzzy Logic-Cuckoo Search Algorithm Method for Short-Term Electric Load Forecasting in 150 kV Sulselrabar System," *Lontar Komputer: Jurnal Ilmiah Teknologi Informasi*, vol. 8, pp. 154-165, 2017.

Electrical Daily Load Forecasting in Ramadhan Using Interval Type-2 Fuzzy Logic in Sulselrabar System

ORIGINALITY REPORT

25%

SIMILARITY INDEX

PRIMARY SOURCES

- | | | |
|---|--|----------------|
| 1 | eprints.uad.ac.id
Internet | 100 words — 3% |
| 2 | ijain.org
Internet | 72 words — 2% |
| 3 | Ezhas Ekawati Zarwono, Achmad Nizar Hidayanto. "Analysis and Design of Internal Information Systems of the APU-PPT Education and Training Center Using the User-Centered Design Method", 2020 International Conference on Informatics, Multimedia, Cyber and Information System (ICIMCIS), 2020
Crossref | 60 words — 2% |
| 4 | ijair.id
Internet | 44 words — 1% |
| 5 | media.neliti.com
Internet | 38 words — 1% |
| 6 | J Jamaaluddin, D Hadidjaja, I Sulistiyowati, E A Suprayitno, I Anshory, S Syahririni, A G Abdullah. "Very Sort Term Load Forecasting Using Interval Type - 2 Fuzzy Inference System (IT- 2 FIS) (Case Study: Java Bali Electrical System)", IOP Conference Series: Materials Science and Engineering, 2018
Crossref | 33 words — 1% |

-
- 7 C.J. Wu, A.H. Sung. "The application of fuzzy logic to JPEG", IEEE Transactions on Consumer Electronics, 1994
Crossref 27 words — 1%
-
- 8 id.123dok.com
Internet 27 words — 1%
-
- 9 newwsa.com
Internet 27 words — 1%
-
- 10 Yu-Hsiang Hsiao. "Household Electricity Demand Forecast Based on Context Information and User Daily Schedule Analysis From Meter Data", IEEE Transactions on Industrial Informatics, 2015
Crossref 18 words — 1%
-
- 11 Ahmad Nor Kasruddin Nasir, Ahmad Azwan Abdul Razak. "Opposition-based spiral dynamic algorithm with an application to optimize type-2 fuzzy control for an inverted pendulum system", Expert Systems with Applications, 2022
Crossref 16 words — < 1%
-
- 12 doi.org
Internet 16 words — < 1%
-
- 13 "Neural Information Processing", Springer Nature, 2015
Crossref 15 words — < 1%
-
- 14 T.W. Liao. "A procedure for the generation of interval type-2 membership functions from data", Applied Soft Computing, 2017
Crossref 15 words — < 1%
-
- 15 www.mdpi.com

15 words — < 1%

16 www.scielo.br

Internet

15 words — < 1%

17 docplayer.net

Internet

14 words — < 1%

18 Lin, F.-J.. "Interval type-2 fuzzy neural network control for X-Y-Theta motion control stage using linear ultrasonic motors", Neurocomputing, 200901

Crossref

13 words — < 1%

19 hdl.handle.net

Internet

13 words — < 1%

20 Chan-Uk Yeom, Keun-Chang Kwak. "Performance Evaluation of Automobile Fuel Consumption Using a Fuzzy-Based Granular Model with Coverage and Specificity", Symmetry, 2019

Crossref

12 words — < 1%

21 Yunus, A. M. Shiddiq, Ahmed Abu-Siada, and M. A. S. Masoum. "Improvement of LVRT capability of variable speed wind turbine generators using SMES unit", 2011 IEEE PES Innovative Smart Grid Technologies, 2011.

Crossref

12 words — < 1%

22 Dongxiao Niu. "A Short-Term Load Forecasting Model Based on LS-SVM Optimized by Dynamic Inertia Weight Particle Swarm Optimization Algorithm", Lecture Notes in Computer Science, 2009

Crossref

11 words — < 1%

23 Hamidreza Seiti, Ashkan Hafezalkotob, Luis Martinez. "-Sets, Comprehensive Fuzzy Sets Risk

10 words — < 1%

Modeling for Risk-Based Information Fusion and Decision-Making ", IEEE Transactions on Fuzzy Systems, 2021

Crossref

-
- 24 eprints.nottingham.ac.uk 10 words — < 1 %
Internet
-
- 25 yadda.icm.edu.pl 10 words — < 1 %
Internet
-
- 26 X. Z. Gao, S. J. Ovaska, X. Wang. "A simplified linguistic information feedback-based dynamical fuzzy system", Neural Computing and Applications, 2010 9 words — < 1 %
Crossref
-
- 27 ojs.umsida.ac.id 9 words — < 1 %
Internet
-
- 28 ribuni.uni.edu.ni 9 words — < 1 %
Internet
-
- 29 www.tandfonline.com 9 words — < 1 %
Internet
-
- 30 "Intelligent Fuzzy Logic - Cuckoo Search Algorithm Method for Short-Term Electric Load Forecasting in 150 kV Sulselrabar System", 'Universitas Udayana' 8 words — < 1 %
Internet
-
- 31 J Jamaaluddin, N M Nisak, R Astuti, S Syahririni, A Ahfas. "Application of Interval Type-1 Fuzzy Inference System to analyze the quality of memorization Qur'an", IOP Conference Series: Materials Science and Engineering, 2021 8 words — < 1 %
Crossref
-
- 32 J.M. Garibaldi, S. Musikasuwan, T. Ozen. "The Association between Non-Stationary and Interval 8 words — < 1 %

-
- 33 Mochammad Apriyadi Hadi Sirad, Muhammad Rais, Muhammad Ruswandi Djalal, Andi Nur Putri. "Optimization of grounding resistance to minimize transient currents at 150 kV SULSELBAR system", 2018 International Conference on Information and Communications Technology (ICOIACT), 2018
Crossref 8 words — < 1%
-
- 34 Zarandi, M.H.F.. "Interval type-2 fuzzy expert system for prediction of carbon monoxide concentration in mega-cities", Applied Soft Computing Journal, 201201
Crossref 8 words — < 1%
-
- 35 dergipark.org.tr
Internet 8 words — < 1%
-
- 36 ir.lib.uwo.ca
Internet 8 words — < 1%
-
- 37 www.coursehero.com
Internet 8 words — < 1%
-
- 38 www.researchgate.net
Internet 8 words — < 1%
-
- 39 Choi, B.I.. "Interval type-2 fuzzy membership function generation methods for pattern recognition", Information Sciences, 20090613
Crossref 7 words — < 1%
-
- 40 Gülçin Büyükoçkan, Celal Alpay Havle, Orhan Feyzioğlu, Fethullah Göçer. "A combined group 7 words — < 1%

decision making based IFCM and SERVQUAL approach for strategic analysis of airline service quality", Journal of Intelligent & Fuzzy Systems, 2020

Crossref

- 41 Jangkung Raharjo, Suyatno Budiharjo. "Short-Term Electric Load Forecasting Using Auto Regressive-Candidates Area Shifting Technique", International Journal on Engineering Applications (IREA), 2021

7 words — < 1%

Crossref

- 42 Mendel, J.M.. "New results about the centroid of an interval type-2 fuzzy set, including the centroid of a fuzzy granule", Information Sciences, 20070115

7 words — < 1%

Crossref

- 43 Mahmoud A. Hammad, Borut Jereb, Bojan Rosi, Dejan Dragan. "Methods and Models for Electric Load Forecasting: A Comprehensive Review", Logistics & Sustainable Transport, 2020

6 words — < 1%

Crossref

- 44 Muhammad Ruswandi Djalal, Herman Nawir, Herlambang Setiadi, Andi Imran. "An Approach Transient Stability Analysis Using Equivalent Impedance Modified in 150 kV South of Sulawesi System", Journal of Electrical and Electronic Engineering-UMSIDA, 2017

6 words — < 1%

Crossref

- 45 Nikhil R. Pal, Tandra Pal. "On rule pruning using fuzzy neural networks", Fuzzy Sets and Systems, 1999

6 words — < 1%

Crossref

- 46 Tarng, Y.S.. "Fuzzy control of feed rate in end milling operations", International Journal of Machine Tools and Manufacture, 199308

6 words — < 1%

Crossref

EXCLUDE QUOTES OFF

EXCLUDE BIBLIOGRAPHY ON

EXCLUDE SOURCES OFF

EXCLUDE MATCHES OFF