TAHRI Mohamed University of Bechar, Algeria Mohamed BOUDIAF University of Science & Technology, Oran, Algeria

doi:10.15199/48.2020.03.30

# Protection System for Induction Motor based on Sugeno Inference

Abstract. The induction motor is a most important drive in the production area. These motors are used in various industrial applications. They can be protected from the different mechanical and electrical faults using different protection systems. The protection is very important to detect abnormal motor running conditions such as over current, over voltage, overload, over temperature, and unbalance conditions. In the classical protection systems, the time delay is adjusted constantly without considering the fault level. This paper presents protection system for induction motor based on Sugeno Inference. The time delay is computed by this intelligent protection for different faults. The obtained results are interesting and show the interest of the proposed intelligent protection.

Streszczenie. W artykule zaprezentowano system zabezpieczeń silnika indukcyjnego bazujący na Sugeno Inference. Określano opóźnienie czasowe dla różnych rodzajów błędów. W stosunku do istniejących systemów proponowany system umożliwia bardziej precyzyjne określenie błędów. Zabezpieczenie silnika indukcyjnego przy wykorzystaniu systemu Sugeno Inference

Keywords: Induction motor protection, Fuzzy logic, Sugeno inference– Overvoltage, Over current, Time delay Słowa kluczowe: silnik indukcyjny, detekcja błędów silnika, Sugeno Inference.

# Introduction

The induction motor is a most central drive in the modern era of automation. These motors are used in various industrial applications. But they can be protected from the different mechanical and electrical faults for helping their purposes. The induction motor experiences various kinds of electrical faults such as over voltage or under voltage, unbalanced voltage, overload, earth fault, phase reversing and single phasing. Due to these faults, the windings in the motor gets heated which lead to reduce the life of the motor. The faults in motor may occur due to faults in the motor or in the driven plant. The investigation of faults of induction motors focus on monitoring of thermal, vibration, electrical, noise, torque, and flux so far. The electricity-related faults are one of the important problems that demands special attention. Various sensors are used to detect the distinctive signals resulting from these faults. To extract the particular features from the faults, different types of signal processing techniques are applied to these sensor signals [1]. Nowadays, microcontroller-based methods can monitor the operating induction motor continuously that does not require human inspection to detect faults in motor. There has been extensive research on detecting abnormal conditions of induction machine using microcontroller-based systems [2-6].

# Sugeno inference for protection system

The protection system is to detect the failures in a rapid manner that occurs in the system. The intelligent protection is expected to perform the protection functions against the following conditions:

A- Over voltage: It is a condition in which voltage is higher than the rated level that is most often used to refer to voltage values in power lines. The problem can affect electric and electronics devices in all industries connected to the power line.

B- Over current: It is a condition in which current is higher than the rated current capacity that can be caused by a short circuit, loose connection, and ground fault.

C- Winding temperature is one of the most important parameters in induction motors since every rise in temperature greatly reduces the expected life span. The heat occurs in the winding of the motors if they operate for a long time. D-Voltage unbalance occurs in a three-phase system where the magnitudes of phase or line voltages are different and the phase angles differ from the balanced conditions, or voltage unbalance generally causes torque, speed variations, additional losses, current unbalance, and noise in three-phase motors.

E- Current unbalance leads to higher losses, circulating currents in windings, and increased heat in the motor. The current unbalance stems from voltage unbalance, large and unequal distribution of phase loads, and the circulating currents in windings can cause bearing failures.

F- Low voltage could occur when a motor is suddenly connected to a power supply and voltage falls to a level too low for safe operation of motor. The under voltage increases the motor current, the rated copper losses, and temperature of windings [7]. The term fuzzy means things which are not very clear or vague [8]. In real life, we may come across a situation where we cannot decide whether the statement is true or false. At that time, fuzzy logic offers very precious flexibility for reasoning. We can also consider the uncertainties of any situation. Fuzzy logic algorithm helps to solve a problem after considering all available data. Then it takes the best possible decision for the given the input. The fuzzy logic method imitates the way of decision making in a human which consider all the possibilities between digital values true and false. Fuzzy systems are robust because the system has been designed to control within some frame of uncertain conditions. Outputs of the system are used in formulating the system structure itself. Conventional systems analyze requires a model based on a collective set of assumptions needed to formulate a mathematical form. Since most industrial applications currently are more complex, fuzzy control should be developed about the control process exists by using a number of fuzzy rules. In this study, the proposed intelligent system for the induction motor protection systems was developed. In this intelligent protection, the time delay calculation used to stop motor was derived with fuzzy rules based on Sugeno inference. These values were determined according to characteristics of the chosen motor.

# Sugeno inference

Sugeno inference is similar to the Mamdani method in many respects. The first two parts of the fuzzy inference process, fuzzifying the inputs and applying the fuzzy operator, are exactly the same. The main difference between Mamdani and Sugeno is that the Sugeno output membership functions are either linear or constant[9-10].



Fig. 1. Sugeno inference with two inputs

A typical rule in a Sugeno fuzzy model has the form: If Input 1=x and Input 2=y, then Output is z=ax+by+c For a zero-order Sugeno inference, the output level z is a constant (a=b =0).

The final output is computed by the equation below:

(1) Final Output = 
$$\frac{\sum_{i=1}^{N} W_i z_i}{\sum_{i=1}^{N} W_i}$$

## Defuzzification

Defuzzification function converts the conclusions reached by the inference mechanism into the inputs to the plant. In the Sugeno inference, the defuzzification is implemented by two methods: the weighted average (Wtaver) and the weighted sum (Wtsum) which are represented by the equations below: A-Weighted average (Wtaver)

$$x^{*} = \frac{\sum_{i=1}^{n} \mu_{Ci}(x_{i}) x_{i}}{\sum_{i=1}^{n} \mu_{Ci}(x_{i})}$$

(2)

(3)

$$x^* = \sum_{i=1}^n \mu_{Ci}(x_i) x_i$$

where, C1, C2, ...Cn are the output fuzzy sets and (xi) is the input values. In this work, the weighted average is used to complete the defuzzification method.

## Simulation and discussion

The induction motor which is used in this work has the characteristics presented on table 1.

Table 1. Induction motor caracteristics

| Puissance 4.0<br>(kW) |         | Vitesse (tr/min) | 1430  |  |  |
|-----------------------|---------|------------------|-------|--|--|
| Tension (V)           | 220/380 | Couple (Nm)      | 12.73 |  |  |
| Current (A)           | 7.14    | Rendement (%)    | 84    |  |  |

The nominal phase current is 7.14 A, and the nominal power is 4.0 kW. The maximum permissible voltage

unbalance and current unbalance are  $\pm 10\%$ . The permissible under voltage is 200 V.

The maximum winding temperature in nominal operating mode is 140 °C. The limit values of the linguistic variables are depicted on table 2.

| Table 2. | limit values of | of input output | variables |
|----------|-----------------|-----------------|-----------|
|----------|-----------------|-----------------|-----------|

|                       | Min   | Max   |  |  |
|-----------------------|-------|-------|--|--|
| Overvoltage (V)       | 240   | >270  |  |  |
| Over current (A)      | 7.14  | >10.5 |  |  |
| Temperature (°C)      | 135   | >155  |  |  |
| Voltage unbalance (V) | 22    | >50   |  |  |
| Current unbalance (A) | 0.714 | >2    |  |  |
| Low voltage (V)       | 200   | <160  |  |  |
| Time delay (s)        | 0     | 3.5   |  |  |

The simulation of the induction motor protection system based on the Sugeno inference shown in the figure below is composed of two fuzzy controllers. The first fuzzy controller controls the input linguistic variables: Overvoltage, over current, and temperature. The second fuzzy controller controls the input linguistic variables: Voltage, current imbalance, and low voltage. Both controllers have the same output which represents the output.

The rule bases of the two controllers are presented on the table below:

Table 3. Rule Base samples for the first fuzzy controller

| Overvoltage | ervoltage Over current Temperature |    | Output |
|-------------|------------------------------------|----|--------|
| OVL         | OCL                                | TL | VL     |
| OVL         | OCL                                | ТМ | VL     |
| OVL         | OCM                                | TH | N      |
| OVL         | OCH                                | TL | LN     |
| OVM         | OCH                                | TH | VS     |
| OVH         | OCL                                | TL | LN     |
| OVL         | OCM                                | TL | LN     |
| OVL         | OCL                                | TL | VL     |
| NA          | OCL                                | TL | VL     |

where,

OVL, Overvoltage Low, OVM, Overvoltage Medium, OVH, Overvoltage High

OCL, Over current Low, Over current Medium, Over current High

TL, Temperature Low, TM, Temperature Medium, TH, temperature High

VL, Very long, N, normal, LN: long, VS, Very short

| able 4. Rule Bas | e samples | for the second | fuzz | y controller |
|------------------|-----------|----------------|------|--------------|
|------------------|-----------|----------------|------|--------------|

| Voltage   | Current   | Low voltage | Output |
|-----------|-----------|-------------|--------|
| unbalance | unbalance | g-          |        |
| VUL       | CUL       | LVL         | VL     |
| VUL       | CUL       | LVM         | VL     |
| VUL       | CUL       | LVH         | LN     |
| VUL       | CUM       | LVL         | VL     |
| VUL       | CUM       | LVM         | LN     |
| VUL       | CUM       | LVH         | N      |
| VUL       | CUH       | LVL         | LN     |
| VUM       | CUM       | NA          | N      |

where,

VUL, Voltage Unbalance Low, VUM, Voltage Unbalance Medium, VUH, Voltage Unbalance High

CUL, Current Unbalance Low, CUM, Current Unbalance Medium, CUH, Current Unbalance High

LVL, Low Voltage Low, LVM, Low Voltage Medium, LVH, Low Voltage High

The output membership function values of the both controllers are below:



Fig. 2. Simulation of the protection system for induction motor based on Sugeno Inference

The fuzzy inference system functions of the two controllers are built on Matlab fuzzy logic designer as is shown on the figures below:



Fig. 3. Fuzzy logic designer for the first fuzzy controller

Both controllers are the same fuzzy inference properties which are presented below:

### And method = min Or method = max

Defuzzification = wtaver

The simulation results of the first controller are presented on the table below:

The figure 5 shows the outputs of the test N° 2.

| Table 6 | 5. First | controller' | s results |  |
|---------|----------|-------------|-----------|--|
|         |          |             |           |  |

| N°   | Overvoltage | Over    | Temperature | Output1 |
|------|-------------|---------|-------------|---------|
| test | (V)         | current | (°C)        | (S)     |
|      |             | (A)     |             |         |
| 1    | 256.6       | 7.8     | 144.5       | 2.82    |
| 2    | 260.9       | 8.8     | 157.2       | 1.5     |
| 3    | 270.9       | 9.6     | 155.5       | 3       |
| 4    | 241.2       | 7.3     | 138.3       | 2.9     |
| 5    | 246.3       | 8.1     | 144.1       | 2.6     |
| 6    | 249.9       | 9.3     | 158.3       | 2.22    |
| 7    | 240         | 7.14    | 135         | 2.9     |



Fig. 4. Fuzzy logic designer for the second fuzzy controller



Fig. 5. Rule viewer of test N°2

Table 7. Second's controller results

| N°   | Voltage   | Current       | Low     | Output2 |
|------|-----------|---------------|---------|---------|
| test | unbalance | unbalance (A) | voltage | (S)     |
|      | (V)       |               | (V)     |         |
| 1    | 23.8      | 0.9           | 192.5   | 2.9     |
| 2    | 30.9      | 1.6           | 186.2   | 2.64    |
| 3    | 44.3      | 0.66          | 166.5   | 3       |
| 4    | 51.3      | 2.3           | 170.5   | 3       |
| 5    | 55.5      | 1.8           | 190.3   | 3       |
| 6    | 28.6      | 2.2           | 188.8   | 2.56    |
| 7    | 36.5      | 2.7           | 195.3   | 2.27    |





#### Table 8 Final time delay values

| N°   | Output1(s) | Output2(s) | Final time delay |
|------|------------|------------|------------------|
| test |            |            | (S)              |
| 1    | 2.82       | 2.9        | 2.82             |
| 2    | 1.5        | 2.64       | 1.5              |
| 3    | 3          | 3          | 3                |
| 4    | 2.9        | 3          | 2.9              |
| 5    | 2.6        | 3          | 2.6              |
| 6    | 2.22       | 2.56       | 2.22             |
| 7    | 2.9        | 2.27       | 2.27             |

The table above shows the final time delay which represents the minimal value of the two controller's responses according to the equation below:

Final time delay= min (Output1, Output2)

According to the obtained results presented on table 5, table 6 and table 7, we can conclude that the proposed protection system for induction motor based on Sugeno Inference is interesting and could be implemented in industrial instead of the classical protection system where the time delay is adjusted constantly manually. Also, the proposed protection could inform the operator about the default nature.

# Conclusion

This paper presents a protection system for induction motor based on Sugeno Inference detecting faulty operations for a three-phase induction motor. The proposed system provides an intelligent protection against overvoltage, over current, temperature, voltage unbalance, current unbalance, and under voltage. In the classical protection system, the time delay is adjusted manually according to the error type. The fuzzy protection system computes the time delay automatically according to the rules base. The proposed intelligent protection is easy to implement and can improves induction motors protection system.

#### Authors

Nadia Achou, University of Sciences and Technology of Oran (USTO), Algeria. E-mail: <u>ac nadia3@yahoo.fr</u>

Mohammed Nasser Tandjaoui, Dr. Ing., University of Sciences and Technology of Oran (USTO) E-mail: <u>tanjaoui 08@yahoo.fr</u> Challei, Banabaiha, Brafasar, University, of Sciences and

Chellali Benachaiba, Professor University of Sciences and Technology of Oran (USTO), E-mail: <u>chellali99@yahoo.fr</u>

Mokhtar Bendjebbar, University of Sciences and Technology of Oran, Algeria, E-mail: <u>bendjebb dz@yahoo.fr</u>

#### REFERENCES

- [1] Sin ML, Erugrul N (2003) Induction machine on-line condition monitoring and fault diagnosis—a survey. AUPEC2003, Australasian Universities Power Engineering Conference, Christchurch, New Zealand.
- [2] C, olak I, C, elik H, Sefa I, Demirbas, S, (2005), On line protection systems for induction motors. *Energy Convers Manag*' 46(17): 2773–2786.
- [3] C olak I , Bayindir R, Bektas A, Sefa I , Bal G (2007), 'Protection of induction motor using PLC. International Conference on Power Engineering, Energy and Electrical Drives', (POWERENG), pp 96–99, Portugal.
- [4] Bayindir R, Sefa I (2007) Novel approach based on microcontroller to online protection of induction motors. *Energy Convers Manag* 48(3):850–856.
- Convers Manag 48(3):850–856.
  [5] Bayindir R, Sefa I ', C , olak I ', Bektas A (2008), 'Fault detection and protection of induction motors using sensors. *IEEE Trans Energy Convers* 23(3):734–741.
- [6] C, unkas, M, Akkaya R, O ztu rk A (2000), 'Protection of AC motors by means of microcontrollers', 10th Mediterranean Electrotechnical Conference (MELECON), vol 3, pp 1093– 1096.
- [7] Okan Uyar, Mehmet C , unkas 'Fuzzy logic-based induction motor protection system', Neural Comput & Applic, Springer, 2012
- [8] Srinivas Gangisheeti, Tarakalyani Sandipamu 'Different Control Schemes for Sensor Less Vector Control of Induction Motor', International Journal of Power Electronics and Drive System (IJPEDS) Vol. 8, No. 2, June 2017, pp. 712~721 ISSN: 2088-8694, DOI: 10.11591/ijpeds.v8i2.pp712-721
- [9] Takagi, Tomohiro, and Michio Sugeno. "Fuzzy identification of systems and its applications to modeling and control." *IEEE transactions on systems, man, and cybernetics* 1 (1985): 116-132.
- [10] Nabil Farah, M. H. N. Talib, Z. Ibrahim, J. M. Lazi, Maaspaliza Azri, 'Self-tuning Fuzzy Logic Controller Based on Takagi-Sugeno Applied to Induction Motor Drives', *International Journal* of Power Electronics and Drive System (IJPEDS) Vol. 9, No. 4, December 2018, pp. 1967~1975 ISSN: 2088-8694, DOI: 10.11591/ijpeds.v9.i4.pp1967-1975.