

Tuning Of Conventional Pid And Fuzzy Logic Controller Using Different Defuzzification Techniques

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Abstract:- This paper presents a method for tuning of conventional PID controller. Simplicity, robustness, wide range of applicability and near-optimal performance are some of the reasons that have made PID control so popular in the academic and industry sectors. Recently, it has been noticed that PID controllers are often poorly tuned and some efforts have been made to systematically resolve this matter. Thus Fuzzy logic can be used in context to vary the parameters values during the transient response, in order to improve the step response performances. Simulation analysis has been carried out for the different processes by conventional and different defuzzification techniques and the results indicate that the values of percentage overshoot are reduced by using fuzzy logic mechanism.

1. INTRODUCTION

PID controllers are the most widely used type of controller for industrial applications. They are structurally simple and exhibit robust performance over a wide range of operating conditions. In the absence of the complete knowledge of the process these types of controllers are the most efficient of choices. However, because of their simple structure, PID controllers are particularly suited for pure first or second order processes, while industrial plants often present characteristics such as high order time delays, nonlinearities and so on [4]. Many tuning formulae that have been devised such as the Ziegler-Nichols one, assures a good load-disturbance attenuation, but often fail to achieve satisfactory performances, and therefore the operator has to use their experience and might fail to attain the best performances [4]. In this context, the use of fuzzy logic seems to be particularly appropriate, since it allows us to make use of the operator's experience and therefore to add some sort of intelligence to the automatic control. This paper has two main contributions. Firstly, a PID controller has been designed by using Ziegler-Nichols frequency response method and its performance has been observed. The Ziegler Nichols tuned controller parameters are fine tuned to get satisfactory closed loop performance. Secondly, for the same systems a fuzzy logic controller has been proposed. Performance comparison between Ziegler Nichols tuned PID controller, and the proposed fuzzy logic controller is presented for different defuzzification methods. This paper is organized as follows. Section 2 reviews the ZN tuning formula in the context of tuning of PID controller. Section 3 presents the tuning procedure, based on fuzzy logic controller. Simulation results are shown in section 4. Finally the conclusion follows in section 5.

2. TUNING OF CONVENTIONAL PID CONTROLLER

Proportional-Integral-Derivative (PID) controllers have been in existence for nearly two-thirds of a century. They remain a key component in industrial process control as over 90% of today's industrial processes are controlled by PID controllers. The PID controller is probably the most used feedback control design. PID controller has the general form

$$u(t) = K_p e(t) + K_i \int_0^t e(x). dx + K_d \frac{de(t)}{dt}$$

where $e(t) = y_{sp}(t) - y(t)$ is the system error (difference between the reference input and the system output), $u(t)$ the control variable, K_p the proportional gain, T_d the derivative time constant and T_i the integral time constant.

The value of controller parameters like K_p , K_d and K_i are reached by mainly trial and error method. But this method is very time consuming. Therefore we switch over to different tuning techniques which give more accurate results with less time. Ziegler-Nichols method is generally used for the purpose in which the parameters like ultimate gain K_u and ultimate period T_u is first calculated by Routh array criteria [10], and then K_p , K_i and K_d are calculated as shown below.

Table 1: Ziegler - Nichols tuning formulas based on ultimate gain (K_u) and ultimate period (T_u)

Controller	Gain (K_p)	Integral time(T_i)	Derivative time(T_d)
P	$0.5K_u$	-	-
PI	$0.45K_u$	$0.8T_u$	-
PID	$0.6K_u$	$0.5T_u$	$0.125T_u$

3. FUZZY LOGIC CONTROLLER

The ideas of fuzzy set and fuzzy control are introduced by Zadeh in an attempt to control systems that are structurally difficult to model. Since Mamdani did the first fuzzy control application fuzzy control has been one of the most active and fruitful research areas in fuzzy set theory, and many industrial applications are reported. Fuzzy logic controllers (FLCs) are increasingly applied to many systems with nonlinearity and uncertainty and it is based on experience

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of a human operator [7]. The structure of fuzzy system can be classified according to the different applications. One of the most popular type is the error feedback fuzzy controller, which is called conventional fuzzy logic controller (FLC). In linear control, there are proportional-derivative (PD), proportional Integral (PI) and proportional-Integral-derivative (PID) control. In conventional FLC, there are also PD type FLC (FZ-PD), PI type FLC (FZ-PI) and PID type FLC (FZ-PID) [7].

3.1 Principle of FLC

The FLC implementations incorporate the following stages:

- (1) Fuzzification: Fuzzification implies the process of transforming the crisp values of the inputs of a controller to the fuzzy domain.
- (2) Knowledge Base: The knowledge base of FLC consists of database and the rule base.
 - a. Data base: It is used to provide necessary information for functioning of fuzzification module, rule base and defuzzification module.
 - b. Rule base: The function of rule base is to represent in a structured way the control policy of an experienced control engineer. This is generally taken the expert in the field otherwise, the knowledge is acquired by having practical experience.
- (3) Fuzzy inference engine (system): A fuzzy inference engine has a simple input-output relationship. Input data from the external world is processed by the fuzzy inference engine to produce the data to be used back in the external world. The events taking place in this process are referred as the basic fuzzy inference algorithm [8]. Mamdani and Takagi-Sugeno fuzzy systems are the examples of fuzzy inference systems.
- (4) Defuzzification: is a process of transforming the fuzzy sets assigned to a control output variable into a crisp value. There are various method of defuzzification.
 - Centre of Area Method (COA)
 - Mean of Maximum Method (MOM)
 - Bisector Method
 - Largest of Maximum (LOM)
 - Smallest of Maximum (SOM)

3.2 Advantages of Using Fuzzy Logic Controller

- Computing with words also allow us to develop mathematical models of events articulated in language only.
- By fuzzifying crisp data obtained from measurements, fuzzy logic enhances the robustness of a system without fuzzification systems designed to act at certain input data points would not know what to do when data is somewhat corrupted.
- Representing a solution with fuzzy sets reduces computational burden. In some cases fuzzy technology makes a solution possible that would

be otherwise unthinkable due to cost of computing every single crisp data point.

3.3 Design and Tuning of Fuzzy Logic Controller

Fuzzy logic controllers (FLCs) are increasingly applied to many systems with nonlinearity and uncertainty and it is based on experience of a human operator. While controlling a plant a skilled human operator manipulates the output of the controller based on error and change in error with an aim to reduce the error with a shortest possible time.

The two types of structure of FLC have been studied so far: one is position-type fuzzy controller which generates control input (u) from error (e) and change in error, and the other is velocity-type fuzzy controller which generates incremental control input (Δu) from error and change in error. The former is called PD type FLC and the latter is called PI type FLC. PI type FLC is known to be more practical than PD type FLC [9].

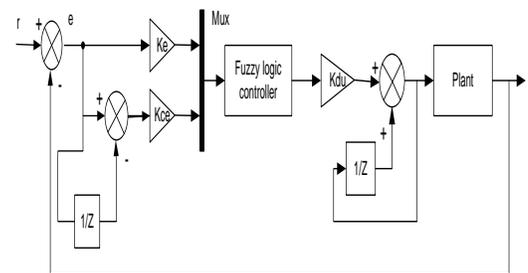


Fig 1: PI type Fuzzy Logic Controller

The error signal is defined as $e(k) = \text{Set point (kth sample time)} - \text{Output (kth sample time)}$. The change in error is defined as $\Delta e(k) = e(k) - e(k-1)$. The operation of PI type FLC can be described by $u(k) = u(k-1) + \Delta u(k)$ where, k is the sampling instant and Δu is the incremental change in controller output [9]. One of the well accepted rule base is the linear rule base which appears in many research work and applications. As the rule base conveys a general control policy, it should be sustained and leaves most of design and tuning work to the scaling gains. Each of the rules of FLC is characterized with an IF part called antecedent and then part called consequent. If the conditions of antecedents are satisfied, then consequents are applied. We shall call the error $e(k)$ and its change $ce(k)$ the inputs or antecedents and change of control $du(k)$ as the output or consequent of rule base. The scaling factors which describe the particular input normalization and output denormalization play a role similar to that of the gains of a conventional controller. Hence, they are very important with respect to controller stability and performance [9]. The set of rules which define the relation between the input and output of fuzzy controller can be found using the available knowledge in the area of designing the system. These rules are defined using the linguistic variables. All the 25 rules governing the mechanism for each output are explained in Table 2.

Table 2: Basic rules table for fuzzy inference system

e \ ce	NB	NS	Z	PS	PB
NB	NVB	NB	NM	NS	Z
NS	NB	NM	NS	Z	PS
Z	NM	NS	Z	PS	PM
PS	NS	Z	PS	PM	PB
PB	Z	PS	PM	PB	PVB

Table 3: Meaning of the linguistic variables in the fuzzy inference system

NVB	Negative very big
NB	Negative big
NM	Negative medium
NS	Negative small
Z	Zero
PS	Positive small
PM	Positive medium
PB	Positive big
PVB	Positive very big

4. SIMULATION RESULT

The performances of the different controllers have been evaluated on different plants. Here, the following transfer functions, with different values of the parameters, are considered:

$$G_1(s) = \omega_n^2 / s^2 + 2\omega\xi s + \omega_n^2; \quad \omega_n = 1, \quad \xi = 0.8.$$

$$G_2(s) = e^{-sL} / (1+sT)^2; \quad T = 1, \quad L = 0.1$$

$$G_3(s) = 1 / (1+s)^3;$$

After the tuning phase, accomplished the unit step responses have been simulated for both conventional and fuzzy logic controller with Matlab and Simuink. The Model used in Simulink/Matlab to analyze the effect of fuzzy logic controller, consists of fuzzy logic block and scaling factors. The unit step response is simulated by using MATLAB. A two-input and one output fuzzy controller is created and the membership functions and fuzzy rules are determined. Fuzzy logic block is prepared using FIS file in MATLAB. The resulting values of IAE, rise time, settling time and percentage overshoot are reported in Tables given below. The step responses of different system with the different controllers are plotted in Figs. 2, 3 and 4.

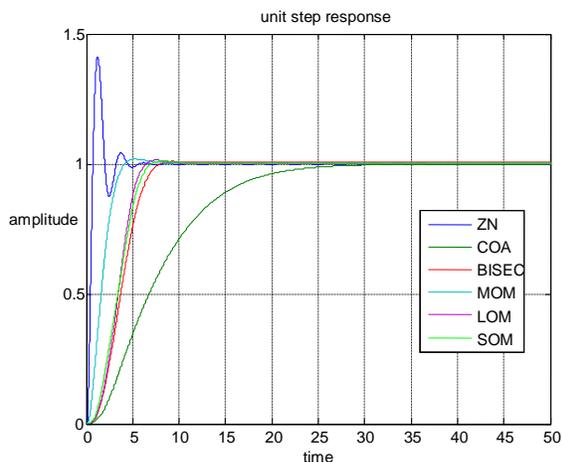


Fig .2: Step Response of $G_1(s)$ with ZN and Fuzzy logic controller

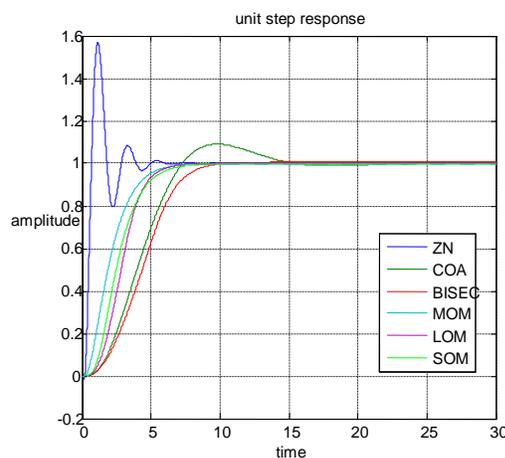


Fig. 3: Step Response of $G_2(s)$ with ZN and Fuzzy logic controller

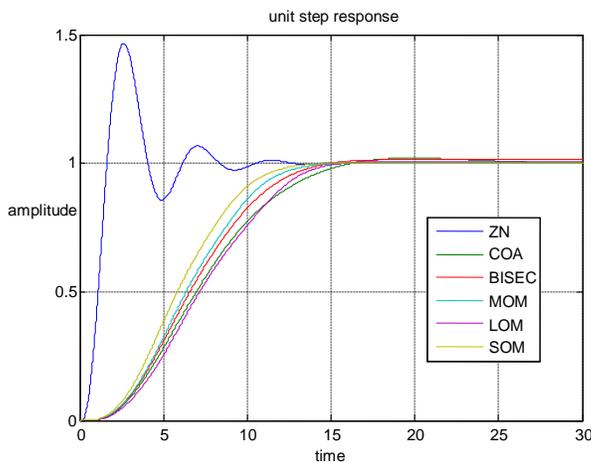


Fig .4: Step Response of $G_3(s)$ with ZN and Fuzzy logic Controller

Table 4: Value of IAE achieved by the examined controllers tuned by Ziegler-Nichols and Fuzzy Logic controller by using different Defuzzification Methods

Process	ZN	FLC (DIFFERENT DEFUZZIFICATION METHODS)				
		COA	BISECTOR	MOM	LOM	SOM
$G_1(s)$	0.8098	6.5485	3.7321	1.6857	3.3756	3.3802
$G_2(s)$	1.0009	4.1061	4.3341	2.1027	2.8704	2.6864
$G_3(s)$	1.9767	6.7508	6.4970	6.3366	6.9123	5.9122

Table 5: Value of Percentage Overshoot OS (%) achieved by the examined controllers tuned by Ziegler-Nichols and Fuzzy Logic controller by using different Defuzzification Method

Process	ZN	FLC (DIFFERENT DEFUZZIFICATION METHODS)				
		COA	BISECTOR	MOM	LOM	SOM
$G_1(s)$	40.9896	0	0.0629	1.5347	0.9437	0.4498
$G_2(s)$	56.8315	0.0284	0	0	0	0
$G_3(s)$	48.8019	0	0	0	0	0

Table 6: Value of Rise-time (t_r) achieved by the examined controllers tuned by Ziegler-Nichols and Fuzzy Logic controller by using different Defuzzification Methods

Process	ZN	FLC (DIFFERENT DEFUZZIFICATION METHODS)				
		COA	BISECTOR	MOM	LOM	SOM
$G_1(s)$	0.4527	6.5979	4.3923	2.4665	3.5112	4.0325
$G_2(s)$	0.3530	5.1739	4.9400	3.3514	3.1624	3.4605
$G_3(s)$	0.9534	6.1040	6.0768	6.0769	6.0353	5.9891

Table 7: Value of Settling-time (t_s) achieved by the examined controllers tuned by Ziegler-Nichols and Fuzzy Logic controller by using different Defuzzification Methods

Process	ZN	FLC (DIFFERENT DEFUZZIFICATION METHODS)				
		COA	BISECTOR	MOM	LOM	SOM
$G_1(s)$	4.1847	9.7298	7.4717	3.8511	6.0894	6.6099
$G_2(s)$	4.7060	8.4693	8.5862	5.9984	6.2155	6.5942
$G_3(s)$	8.2922	9.7802	9.7714	9.7705	9.8068	9.6902

5. Conclusion

In this paper, different methods regarding the tuning of conventional PID and fuzzy logic controller has been presented. Simulation was carried out using MATLAB version 2010a to get the output response of the system. Zeigler Nichols method is used for tuning of conventional PID controller. But this method is not satisfactory for many systems, as it gives high overshoot. In this context, the use of fuzzy logic seems to be particularly appropriate, since it allows us to make use of the operator's experience. According to the profiling results, the use of soft-computing technique resulted in a better outputs. The amount of overshoot for the output response was successfully decreased using the fuzzy logic controller. Certainly, the improved performance is at the cost of increased rise-time and settling-time. The performances of different processes using different defuzzification methods are also present in this paper. From the above results we can easily observe that except the value of the percentage overshoot Mean of Maximum defuzzification method gives better results as compared to other defuzzification techniques.

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