PSO Optimized FLC For The Design Of Load Frequency Control

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Abstract: Load frequency control problem is considered as one of the mostimportant issues in the design & operation of power systems. Due to lack of goodefficiency in parameters variation conditions, working conditions of system and non-linear factors, a simple PI controller is not suitable in industrial applications. Instead, fuzzy controllers can be used in order to enhance the performances of thesystems. In this paper, the use of the optimized type-1 fuzzy logic controller using Particle Swarm Optimization (PSO) algorithm is proposed to solve the load frequency control problem. To the best of our knowledge, the PSO optimization offuzzy type-1 controller in order to solve load-frequency control problem, has notbeen investigated so far. The proposed controller has good performance and iscapable to solve the load-frequency control problem in conditions of widevariations of system parameters and nonlinear factors such as generation rateconstraint. Simulation results show that the optimized fuzzy controller proposed in this paper exhibits better performance compared to PI controller in damping of system deviations.

Key words: power system, load frequency control, type-1 fuzzy logiccontroller, PSO algorithm.

I. INTRODUCTION

The main aim of power systems is to provide the required power supplyto customers with a given voltage and frequency. Considering the growingenergy demand of the customers, the stability and reliability of power systems are important. The load-frequency control problem in a power system dealswith sudden disturbances that disrupt normal conditions of system operationand occur due to outage and connection of loads on different hours in power system. Any change in active power demand in a point of a power system isreflected throughout the system as a frequency change (Panda, 1998).

The problem of output active power control of a power system inresponse to power system frequency changes and between regional powers of system lines in a specified range is known as the load-frequency control problem. For optimal performance and optimal operation of power systems, the frequency changes must be maintained within certain limits. Many processcontrol systems, such as computers, are sensitive to changes in the frequency and their operation is impaired. For such systems, their frequency must be regulated and controlled (Chaturvediet al., 1999; Moon et al., 2002). Therefore, an adequate supplementary controller to regulate and prevent frequency deviations must be used in the main control center.

The purpose of load frequency control problem is to maintain a uniformfrequency and to adjust and control converted power between areas of thepower system in a planned value. In other words, we can say that the aim ofload-frequency control problem is keeping the steady-state error of the systemon zero. In previous studies on solving the load-frequency control problem,various methods have been used. In the proposed methods, the PI controller ismost widely used in industry. The PI controller has a fixed gain that is designedin rated operating conditions and its utilization is simple, but frequencyoscillations can also appear in this case. This means that the PI controllerindicates a poor dynamic performance against of system parameters variationand non-linear conditions such as generation rate constraint (Hameedet al.,2010; Chang & Fu, 1997). Different types of fixed gain controllers are designedin rated operating conditions while they are unable to achieve performance inpractice under many changes in the operating conditions of the system.

In order to solve the load frequency control problem and to minimizepower system deviations, operating conditions and system parameters variationsmust be considered; genetic algorithms and other intelligent methods are used to improve the PI controller performance and to optimize controller coefficients. As for the non-linearity of the power system and inability to extract its precisemathematical model, in recent years the use of the fuzzy method in design of controller was proposed (Ying, 2000; Mendel, 2000; Hidalgo et al., 2009; Panda et al., 2009; Ying, 2006). In some studies, fuzzy logic is used to adjust PI controller parameters.

In this paper the type-1 fuzzy logic controller optimized using ParticleSwarm Optimization (PSO) algorithm is used to damp oscillations of frequencyin a single area power system. In designing the proposed controller, the nonlinearfactor of generationrate constraint is considered and the results of theoptimized type-1 fuzzy logic controller using PSO algorithm are compared tothat of a classical PI controller.

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THE SINGLE AREA POWER SYSTEM II.

The load frequency model of the studied single area power system is shown in Fig. 1. In solving the load-frequency control problem of the studiedsystem, the generation rate constraint is considered. The nonlinear turbinemodel considering the generation rate constraint is shown in Fig. 2.

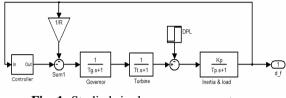


Fig. 1: Studied single area power system.

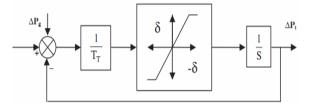


Fig. 2 : Non-linear turbine model considering the generation rate constraint.

The studied single area power system parameters are defined with theadding limiter to the turbine that are listed in Table 1. In the studied powersystem, the area control error (ACE) and its variation (ΔACE) are consideredas fuzzy controller inputs.

Studied Single Area Power System Parameters					
Parameter	Value				
Tg	0.08				
Tt	0.3				
Тр	20				
Кр	120				
R	2.4				
DPL	0.01				

Table 1

PI AND TYPE-1 (T1) FUZZY LOGIC CONTROLLERS III.

The PI controller block-diagram and its parameters are illustrated inFig. 3. The fuzzy controller uses fuzzy logic rules in control applications. Fuzzyrules have been established based on control laws. Fuzzy logic systems are notdesigned based on mathematical models. Fuzzy controllers implement thehuman logic using fuzzy logic based on membership functions, fuzzy rules and membership rules.

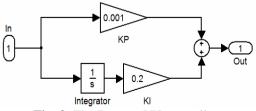


Fig. 3: The Proposed PI controller.

A fuzzy system is described by a set of IF-THEN rules and uses diversemembership functions. To solve the load-frequency control problem in fuzzy controller, controller inputs, $\Box \Delta ACE$, ACE, and its output are considered as the control signal. The fuzzy type-1 controller structure is shown in Fig. 4.

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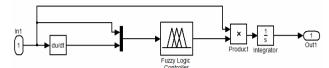


Fig. 4 : The Proposed fuzzy type-1 controller.

The design of the type-1 fuzzy controller has 4 steps that are presented as follows:

Step 1: The first step should be to determine the system dynamicbehavior and characteristics. At this stage, the controller input and outputvariables and their variations range should be determined. In addition, suitableinput signals should be determined for load-frequency control problem. Inputsignals are considered to create the base rules as the IF part and the outputsignal of the fuzzy controller are considered as the THEN part.

Step 2: At this stage, the appropriate fuzzy sets and membershipfunctions are determined. At this stage, the degree of fuzzy membershipfunctions associated with each variable or input, output signals are determined and the process of fuzzification is completed. The membership functions of fuzzy type-1 controller are illustrated in Fig. 5.

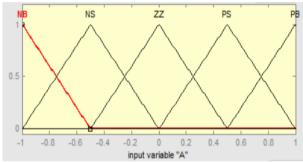


Fig. 5 The membership functions of proposed fuzzy type-1 controller.

Step 3: The inference engine is described. In this step, fuzzy rules areformed by the control rules that system performance is based on that rules.

Step 4: At this stage, the defuzzification process is performed. Fuzzyrules are combined and the defuzzification process is performed on the output and the output of the fuzzy inference engine becomes crisp.

In this paper, inputs and outputs membership functions are the same andthe 5-segments triangular membership functions are used. Since each input hasfive membership functions, the number of base fuzzy rules is 25 that arepresented in Table 2.

Fuzzy Rules of the Proposed Fuzzy Type-1 Controller									
e	NB	NS	ZZ	PS	PB				
NB	S	S	М	М	В				
S	S	М	М	В	VB				
ZZ	М	М	В	VB	VB				
PS	М	М	VB	VB	VVB				
PB	В	VB	VB	VVB	VVB				

Table 2

IV. **OPTIMIZED TYPE-1 FUZZY LOGIC CONTROLLER USING PSO ALGORITHM**

In this paper, to design the self-tuning adaptive and fuzzy loadfrequencycontrol system and to achieve a better performance of type-1 fuzzycontroller, the PSO algorithm is used to optimize the membership functions parameters with the aim of minimizing system frequency deviations. In thestudied system, ACE and \triangle ACE are considered as the input signals and as for the input signals, the output signals are used to change the set points in the loadfrequencycontrol problem.

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4.1. PSO Algorithm

In this paper, a PSO algorithm is applied to optimize the membershipfunctions of fuzzy type-1 controller. The optimization problem is defined as follows:

 $\min f(x)$

subject tox_i \in X_i, i=1,2,3,... N \rightarrow (1)

In (1), f(x) is the objective function and x is the collection of the decision variables x_i . The set X_i is the collection of possible range of variables, and N is the number of variables. The PSO optimization algorithm (Taher, 2008; Kennedy & Eberhart, 1995) is one of the latest and strongest methods of optimization and has been used in solving complex problems up to now.

The PSO algorithm starts to work with a random population (i.e.particles) and then searches the optimal solution (particle) of the problem witheach updated generation. Particle i is defined by its spatial position S_i and its velocity V_i . At each stage of population movement, each particle is updated by tracking the two best positions.

The first one is the local best position, in terms of competency, which isobtained separately for each particle up to now. This value is P_{best} . The second one is the global best position, obtained by the all of the particles among the

population, up to now. This value is G_{best} . After finding the values of the P_{best} and G_{best} , each particle updates its new velocity and position based on the following equations:

$$V_{i}^{k+1} = WV_{i}^{k} + C_{1}r_{1}(P_{best_{i}} - S_{i}^{k}) + C_{2}r_{2}(G_{best_{i}} - S_{i}^{k}) \rightarrow (2)$$
$$S_{i}^{k+1} = S_{i}^{k} + V_{i}^{k+1} \rightarrow (3)$$

The problem convergence is dependent on the PSO algorithmparameters such as W, C₁ and C₂. W is the updating factor of the particlesvelocity. C₁ and C₂ are the acceleration factors, which are the same and are in the range of [0, 2]. r₁ and r₂ are two random variables uniformly distributed in the range of [0, 1]. W is the inertia weight and is constant in the PSO algorithm. In modified PSO, to strengthen the PSO algorithm performance and solve its disadvantages with updating W, for obtaining the best reply in terms of the convergence velocity and accuracy in the optimization problem, the following equation is used: $\frac{W^{k} - W}{W} = \frac{k}{(W} - \frac{W}{W} - \frac{W}{W}$

$$W^{\wedge} = W_{\max} + k \left((W_{\min} - W_{\max}) / iterMax \right) \rightarrow (4)$$

In (4), Wmin and Wmax are the minimum and maximum values of theinertia weight, the iterMaxis the maximum number of the algorithm iterations, and k is the current iteration of the algorithm. The inertia weight is varied by (4) and causes the convergence, which is defined as a variable in the range of [0.2 0.9]. Because of updating the inertia weight with updating the particles velocity, the PSO algorithm has a perfect performance. In solving the optimization problem, the number of algorithm iterations has been reduced and the convergence power has been increased under the conditions of increased community members. Finally the optimization algorithm is ended by the particles convergence to a certain extent. The optimal parameters of PSO algorithm used in this study are presented in Table 3.

Swarm Size	C_1	C_2	W	iterMax
20	2	2	0.4-0.9	100

 Table 3

 The Optimal Parameters of PSO Algorithm Used in Optimization Problem

4.2. Implementation of the Proposed Method

In the proposed fuzzy typ-1 controller for better performance, the processor parts of controller (K_1 , K_2 and K_3) are optimized in a specified range. The fuzzy type-1 controller consists of processor Parts is shown in Fig. 6.

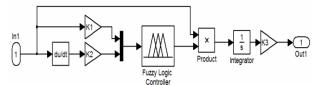


Fig. 6 – The fuzzy type-1 controller with processor parts.

In proposed fuzzy type-1 controller, the parameters K_1 , K_2 and K_3 areused to tune the controller for better performance in damping system frequencydeviations.

The optimal and accurate performance of the fuzzy controller is determined by the membership functions. That means that if the bestmembership functions are selected for controller then the best performance of the controller is achieved. Thus, the PSO algorithm is used to determine the parameters of K_1 , K_2 and K_3 considering the objective function of optimization problem in Fig. 7.

The objective function is chosen as the Integral of Time multiplied bySquared Error (ITSE) as follow:

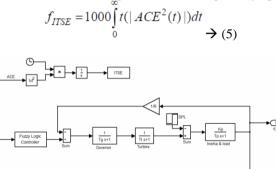
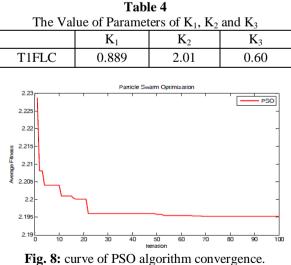


Fig. 7: Application of PSO algorithm considering objective function for optimization of proposed fuzzy type-1 controller.

V. SIMULATION RESULTS

Using PSO algorithm the optimal parameters of K_1 , K_2 and K_3 are determined as Table 4. The curve of PSO algorithm convergence is presented ifFig. 8.



After determination of parameters of K_1 , K_2 and K_3 the single areapower system is simulated using PI and optimized fuzzy type-1 controllers and the result of simulation is shown in Fig. 9.

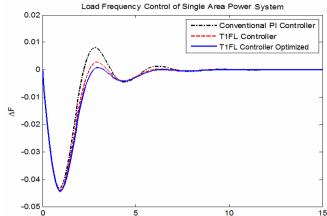


Fig. 9: Load frequency control of single area power system using PI andoptimized fuzzy

Fig. 9 shows that the frequency deviations of the single area powersystem are damped by using the proposed optimized fuzzy type-1 controller inless settling time and maximum overshoot versus the PI controller. So the proposed fuzzy controller shows better performance in load frequency controllof a single area power system.

VI. CONCLUSIONS

1. In this paper, the optimized type-1 fuzzy logic controller was proposed to solve the load frequency control problem of single area power system.

2. Using PSO algorithm the parameters of the membership function of type 1 fuzzy controller were optimized.

3. Simulation results show that the proposed type-1 fuzzy controller indamping of frequency deviations of power system has a better performance than the PI controller.

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