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Optimal Control of PID-FUZZY based on Gravitational Search Algorithm for Load Frequency Control

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Abstract:- In this paper, an efficient and effectual optimization algorithm called Gravitational Algorithm (GSA) is suggested to optimally tune the PID-FUZZY parameters toward accurate frequency control of two-area power system. An effective and precise optimization algorithm i.e., GSA has been constructed based on the gravity law and mass interactions. In this algorithm, the explorer operators are a masses collection which interacts with each other according to the motion laws and Newtonian gravity. To appropriately and accurately confirm the dynamic performance of suggested GSA-based PID-Fuzzy, four performance indexes, i.e.: ISE, ITSE, IAE and ITAE have been considered as objective functions. Furthermore, the simulations results by the suggested GSAbase PID-FUZZY have been compared with the simulation results based on Particle Swarm Optimization (PSO) and Artificial Bee Colony (ABC) algorithms which are generally realized as conventional heuristic algorithms and extensively applied by the researchers and scholars. Finally, optimal GSA-based PID-FUZZY presents a high dynamic performance as compared to PSO-based PID-FUZZY and ABC-based PID-FUZZY.

Keywords: Power system dynamic stability, low frequency oscillation, automatic generation control, optimization algorithms, gravitational search algorithms.

1. INTRODUCTION

The value of consumed power around the world exhibits the state of the art level of countries that shows the importance of power quality, voltage regulation and system frequency stability issues. That is to say, all criterions of power quality have been studied and analyzed on frequency and voltage of power system. Since the system frequency control is highly longer than the voltage control, it is regarded as salient benchmark in the power system. Thus, system frequency stability should be principally investigated and involved in many researches and projects [1]. Integration of many different power sources or areas has constructed the interconnected power system. Any power variations in power source and load which happen in these areas can disturb other connected areas. The tie-line characteristic between these areas is one other important item in impressing the frequency stability. During the high variations in system frequency which go beyond the defined limits, they can create severe power system instability, which will stop the connected power plants and even shut down the entire system in the later step. In this instance, the fed areas of the system remain without energy and cause vast economic damages [2]. Consequently, considering these vast economic damages due to system collapse, the frequency control of loads is regarded a salient problem which must be considered to hinder such damages.

Automatic Generation Control (AGC) has decreased the deviations of tie-line power and frequency in order to swiftly restore its nominal frequency [3-5]. The conventional controller such as: classical Proportional Integral Derivative (PID) controller has been widely used for AGC in interconnected power systems. But, this controller cannot effectively suppress the fluctuations tie-line power and frequency during severe perturbations. Considering the benefits of PID controller, it has been integrated with Fuzzy controller to construct PID-Fuzzy controller aimed at suppression of low frequency oscillations. As reported in various literatures, Fuzzy controller provides good performance during the normal, unbalance and transient conditions [6-8]. The parameters of PID-Fuzzy can be tuned in real time in accordance with load perturbations to enhance its accuracy and robustness toward acquiring the best control aims.

In recent years, the evolutionary optimization algorithms such as Bacteria Foraging (BF), Particle Swarm Optimization (PSO) and Artificial Bee Colony (ABC), Ant Colony Optimization (ACO) have been widely used by researchers to solve the various objective functions in different problems [9-13]. Although aforementioned algorithms seem to be excellent to optimally tune the parameters of controller, they introduce a slow convergence with local minimums. Gravitational Search Algorithm (GSA), a new evolutionary optimization algorithm, presents a fast convergence and accurate performance in these aspects [14-16]. In this regard, it is applied to optimally tune the parameters of PID-Fuzzy for stabilizing the interconnected power system.

In this study, the parameters of PID-Fuzzy have been tuned by GSA, PSO and ABC so that its performances can be well compared and appraised. The relevant studies have been performed in two-area interconnected power system. Severe load perturbation has affected the interconnected power system to deal with the dynamic stability criterions with presence of PID-FUZZY. Finally, optimal GSA-based PID-FUZZY presents a high dynamic performance as compared to PSO-based PID-FUZZY and ABC-based PID-FUZZY.

2. INTERCONNECTED POWER SYSTEM STRUCTURE

Kinetic and potential energy has been transformed into mechanical energy using turbines, following that, the mechanical energy has been transformed into electrical energy using power plant's generators. Simple motion equation based on the mentioned concept can be presented by:

$$T_m - T_e = J \frac{d\omega}{dt} \tag{1}$$

Where, T_m and T_e are respectively mechanical and electrical torque. J and ω are respectively the inertia and angular speed.

The understudy power system contains two interconnected areas via tie-line. Note that, all available generators in each area have a combined construction. Occurrence of any perturbations in each area of power system leads to frequency deviation in other areas. Then, this deviation creates fluctuation in tie-line power. In this regard, both the tie-line power and frequency deviations must be damped to maintain the dynamic stability of interconnected power system. The tie-line power can be presented by:

$$P_{line\,12} = \frac{|V_1| \cdot |V_2|}{X_{12}} \sin(\delta_1 - \delta_2) \tag{2}$$

Where, V_1 and V_2 are respectively voltage of first and second areas. δ_1 and δ_2 are respectively machine angle of first and second areas. X_{12} indicates tie-line impedance. The machine angle deviation of each area can be presented by:

$$\Delta \delta = 2\pi \int \Delta f dt \tag{3}$$

The tie-line power deviation can be presented by:

$$\Delta P_{12} = \frac{|V_1| \cdot |V_2|}{X_{12}} \cos(\delta_1 - \delta_2)(\Delta \delta_1 - \Delta \delta_2) \tag{4}$$

Where, the synchronous coefficient is written as follows:

$$T_{12} = \frac{|V_1| \cdot |V_2|}{X_{12}} \cos(\delta_1 - \delta_2) \tag{5}$$

The tie-line power deviation is rewritten by:

$$\Delta P_{tie} = T_{12}(\Delta \delta_1 - \Delta \delta_2) \tag{6}$$

Thereupon load deviation ΔP_L , tie-line power deviation, frequency deviation and Area Control Error (ACE) of each area can be given by following equations:

$$\Delta P_{tie} = \frac{-\Delta P_{L1}(\frac{1}{R_2} + D_2)}{\frac{1}{R_1} + \frac{1}{R_2} + D_1 + D_2}$$
(7)

$$\Delta f = \frac{-\Delta P_{L1}}{\frac{1}{R_1} + \frac{1}{R_2} + D_1 + D_2} \tag{8}$$

$$ACE_1 = \Delta P_{tie} + B_1 \cdot \Delta f_1 \tag{9}$$

$$ACE_2 = B_2 \cdot \Delta f_2 + a_{12} \cdot \Delta P_{tie} \tag{10}$$

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AREA 1 B₁ 1/R₁ $\Delta pL_1(s)$ Δf $G_{g1}(s)$ G_{ps1}(s) $G_{c1}(s)$ Grh1(s) G+1(s) Power System 1 Governor 1 Controller 1 Re-heater 1 Turbine 1 ΔP_{ti} $G_{tl}(s)$ Tie line a₁₂ a₁₂ ACE₂ Δf_2 $G_{c2}(s)$ $G_{g2}(s)$ Grha(s) $G_{t2}(s)$ $G_{ps2}(s)$ Controller 2 Governor 2 Re-heater 2 Turbine 2 Power System 2 $\Delta pL_2(s)$ AREA 2 B₂ 1/R₂

Figure 1. Two-area interconnected power system

Figure 1 presents the linearized model of two-area interconnected power system. The following equations give the transfer function of each block:

Governor:

$$G_{g1}(s) = G_{g2}(s) = \frac{K_h}{T_h \cdot s + 1} \tag{11}$$

Reheater:

$$G_{rh1}(s) = G_{rh2}(s) = \frac{(K_{r12} \cdot T_{r11}) \cdot s + 1}{T_{r1} \cdot s + 1}$$
(12)

Turbine:

$$G_{t1}(s) = G_{t2}(s) = \frac{K_t}{T_t \cdot s + 1}$$
(13)

Power system:

$$G_{ps1}(s) = G_{ps2}(s) = \frac{K_g}{T_g \cdot s + 1}$$
 (14)

Tie-line:
$$G_{tt}(s) = \frac{2\pi \cdot T_{12}}{s}$$
(15)

The understudy two areas of interconnected power system are thermal plant. Rate of each area is 2000 MW with 1000 MW (nominal load). This system has been widely studied in different literatures which propose, design and analyze the required controller for AGC to enhance the power system dynamic stability [17-19].

CONSTRUCTION OF PID-FUZZY CONTROLLER

A PID-Fuzzy controller enhances the dynamic performance of PID controllers, because Fuzzy logic can appropriately control the changes in parameters of power systems or operating point via auto-tuning gains of PID controller [20]. Precise design of Fuzzy controller has been performed by choosing an appropriate membership functions and rule formulation. So, general membership functions and rules

have been selected for input/output while the parameters of PID controller can be optimally tuned to enhance the overall dynamic performance of PID-Fuzzy controller [21]. Twodimensional rule base PID-Fuzzy controller has been designed by the error signal along with its derivative for input signal PID controller for output signal [22. As mentioned above, PID-Fuzzy controller which is shown in Fig. 2 (a) is designed for AGC to enhance the dynamic performance of power system. K1 and K2 are input gains of PID-Fuzzy controller while K_P, K_I and K_D are its output gains. The fuzzy rules and its input membership function are respectively presented in Table 1 and Fig. 2(b). The membership functions are defined by: PB (Positive Big), PM (Positive Medium), PS (Positive Small), ZE (Zero), NS (Negative Small), NM (Negative Medium) and NB (Negative Big) as fuzzy measure.

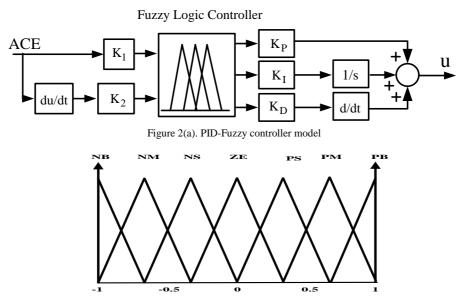


Figure 2(b). Normalized membership function for inputs

Table 1. Lookup table of fuzzy rules

ACE ΔACE	NB	NM	NS	ZE	PS	PM	PB
NB	NB	NB	NB	NB	NM	NS	ZE
NM	NB	NM	NM	NM	NS	ZE	PS
NS	NB	NM	NS	NS	ZE	PS	PM
ZE	NB	NM	NS	ZE	PS	PM	PB
PS	NM	NS	ZE	PS	PS	PM	PB
PM	NS	ZE	PS	PM	PM	PM	PB
PB	ZE	PS	PM	PB	PB	PB	PB

4. GRAVITATIONAL SEARCH ALGORITHM

GSA is a heuristic stochastic evolutionary algorithm proposed by Rashedi et al in 2009 [17]. It is inspired by the mass interactions and gravitation newton principle. All objects are absorbed by a gravitational force, which forces all objects to move toward heavy masses as shown in Fig. 3(a). It exhibits a natural behavior between earth and the moon. The weightiest object suggests the problem solution. The Flowchart of GSA process is given in the Fig. 3(b), and also

its general Pseudo code is provided in Fig. 3(c). According to GSA procedure, each mass includes four statements: position, inertial mass, active gravitational mass, and passive gravitational mass [18]. The mass position related to the problem solution, and its gravitational and inertial masses have been defined via an objective function. That is to say, each mass gives a result that GSA is navigated via correct tuning the gravitational and inertia masses [19]. By time span, it is expected that masses to be pulled by the weightiest mass which gives an optimal response in the search space.

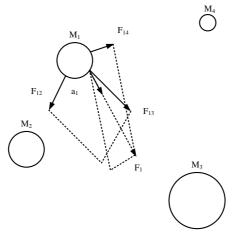


Figure 3(a). Every mass accelerate towards to the resultant force

absolute error (IAE) and integral of time multiplied absolute error (ITAE) [23-25]. To appropriately and accurately confirm the dynamic performance of suggested GSA-based

PID-Fuzzy, all aforementioned have been taken into account

These objective

for designed optimization problem.

functions are given as follows:

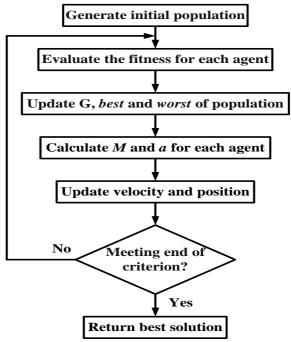


Figure 3(b). Flowchart of GSA procedure

Random initialization of the population Find the best and worst solutions in the initial population while (stop criterion) **for** i=1:N (for all elements) update G(t), best(t), worst(t) and $M_i(t)$ for i=1,2,...N calculate the mass of individual $M_i(t)$ calculate the gravitational constant G(t)calculate the acceleration $a_i^h(t)$ update the velocity and position of each individual v_i^h , x_i^h end for i=1:N (for all elements) Find the best individual end while (stop criterion) Display the best individual as the solution

Figure 3(b). GSA Pseudo code

5. SIMULATIONS AND RESULTS

The performance of control systems has been assessed based on an especial performance index depends on the required objective function. The performance index is alternatively used to design of controllers which are commonly known by: integral of squared error (ISE), integral of time multiplied squared error (ITSE), integral of

$$ISE = \int_{0}^{T} \left\{ \Delta F_{1}^{2} + \Delta F_{2}^{2} + \Delta P_{tie,12}^{2} \right\} dt$$

$$ITSE = \int_{0}^{T} \left\{ \Delta F_{1}^{2} + \Delta F_{2}^{2} + \Delta P_{tie,12}^{2} \right\} t. dt$$

$$IAE = \int_{0}^{T} \left\{ \Delta F_{1} | + |\Delta F_{2}| + |\Delta P_{tie,12}| \right\} dt$$

$$ITAE = \int_{0}^{T} \left\{ \Delta F_{1} | + |\Delta F_{2}| + |\Delta P_{tie,12}| \right\} t. dt$$

$$(18)$$

$$ITSE = \int_{0}^{T} \left\{ \Delta F_{1}^{2} + \Delta F_{2}^{2} + \Delta P_{tie,12}^{2} \right\} t.dt$$
 (17)

$$IAE = \int_{0}^{T} \left\{ \Delta F_{1} \left| + \left| \Delta F_{2} \right| + \left| \Delta P_{tie,12} \right| \right\} dt$$
 (18)

$$ITAE = \int_{0}^{T} \left\{ \Delta F_1 \right| + \left| \Delta F_2 \right| + \left| \Delta P_{tie,12} \right| \right\} t.dt \tag{19}$$

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The PID-Fuzzy parameters have been optimally tuned by GSA in order to alleviate the tile-line power and frequency deviations. Furthermore, the controller parameters have been optimized by PSO and ABC to more validate the dynamic performance of GSA-based PID-Fuzzy controller. The operational constraints of controller parameters are given as follows:

$$K_{1}^{\min} \leq K_{1} \leq K_{1}^{\max}$$

$$K_{2}^{\min} \leq K_{2} \leq K_{2}^{\max}$$

$$K_{P}^{\min} \leq K_{P} \leq K_{P}^{\max}$$

$$K_{I}^{\min} \leq K_{I} \leq K_{I}^{\max}$$

$$K_{D}^{\min} \leq K_{D} \leq K_{D}^{\max}$$
(20)

In the simulations, a step load perturbation occurs in first area of the interconnected power system to affect the dynamic stability. The optimization problem has been performed with GSA, PSO and ABC considering all chosen performance indexes. The load perturbation occurs at t=10 s with $\Delta P_L=0.01$. The optimal parameters of PID-Fuzzy controller optimized by GSA, PSO and ABC are tabulated in Table 2. Meanwhile, the values of defined performance indexes, i.e.: ISE, ITSE, IAE and ITAE are presented in Table 3. Also, three stability benchmarks i.e.: settling time, overshoot and undershoot are tabulated in Table 4.Both the time-varying frequency and tie-line power deviations are respectively presented in Fig. 4(a-c).

Table 2. The parameters of PID-Fuzzy optimized by GSA, PSO and ABC

	GSA		ABC		PSO	
	Area 1	Area 2	Area 1	Area 2	Area 1	Area 2
K ₁	1.1854	1.6210	1.5021	1.8428	1.8029	2.2367
\mathbf{K}_2	0.5649	0.7431	0.6712	0.9142	0.8291	1.0056
K _P	0.6094	0.7885	0.7003	0.9462	0.9906	1.2049
K _I	0.7951	1.0359	0.9615	1.3410	1.2986	1.372
Kp	0.4932	0.6372	0.58960	0.7743	0.5765	0.9211

Table 3. The optimal value of dynamic performance indexes achieved by GSA, PSO and ABC

Parameter indexes	GSA	ABC	PSO
ISE	2.7025	4.0962	4.7643
ITSE	0.1496	0.1924	0.2241
IAE	2.5618	3.4577	3.9553
ITAE	71.7942	110.4983	130.3509

Table 4. The optimal value of dynamic performance indexes achieved by GSA, PSO and ABC

Stability benchmarks	GSA	ABC	PSO
Overshoot	0.0116	0.0148	0.0156
Undershoot	0.0192	0.01984	0.0208
Settling time	55.31	62.12	68.76

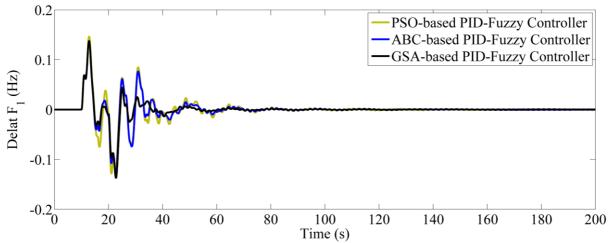


Figure 4(a). Change in frequency deviation of area-1

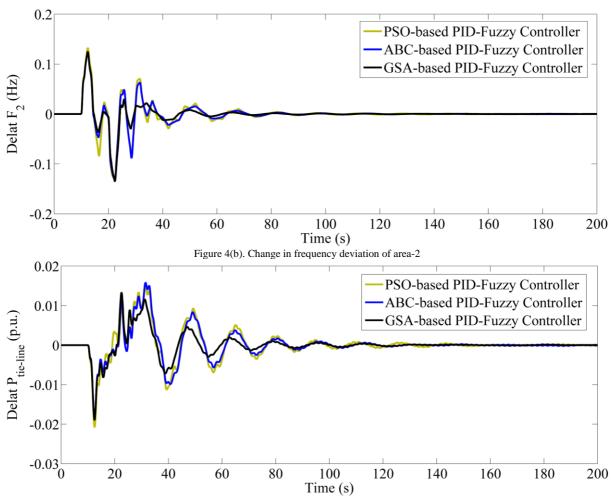


Figure 4 (c). Change in tie-line power deviation between area-1 and area-2

As for the Table 2-4, all performance indexes optimized by GSA present less values as compared to corresponding indexes optimized by PSO and ABC. The presented results prove that the overall dynamic stability of interconnected power system has been highly improved by GSA-based PID-Fuzzy more than two others. The time-varying curves presented in Fig. 4(a-c) have better portrayed the dynamic performance of GSA-based PID-Fuzzy.

6. CONCLUSION

In this paper, PID-Fuzzy controller is suggested for AGC appropriately enhance the dynamic stability of two-area interconnected power system. GSA algorithm has been complementary applied to optimize the parameters of PID-Fuzzy so that its maneuverability to be increased. Hence, PSO and ABC have been taken part in optimization problem to clear the accuracy and capability of GSA. Dynamic stability analysis has been more carried out with consideration of four performance indexes i.e.: ISE, ITSE, IAE and ITAE. The pertinent dynamic stability studies have been performed by affecting the interconnected power system caused by load perturbation occurrence in area-1. Eventually, the simulation results confirm the high dynamic performance of GSA-based PID-Fuzzy via damping the low frequency oscillations of tie-line power and frequency deviations.

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