Load Frequency Control of Multi-Source Power System with Redox Flow Batteries: An Analysis

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ABSTRACT

This paper presents the Load frequency control (LFC) of a multi-source power system with Redox flow batteries (RFB). The power system comprises non-reheat thermal, hydro and gas generating units. RFB is integrated to the LFC system to improve the dynamic responses. The proportional plus integral (PI) controller is designed using the PSO technique. Dynamic responses are obtained by giving 1% step load perturbation (SLP) in the area. Further, to show the robustness of the controller, the dynamic responses are obtained by varying the SLP from 1% to 3%. Dynamic responses obtained in all the cases, satisfy the LFC requirements. PSO technique gives good convergence characteristics and promising computational results.

General Terms

Frequency deviation, Multi-source

Keywords

Controller, Particle swarm Optimization, Load Frequency Control

1. INTRODUCTION

The size and complexity of the power system is increasing and its structure is also changing [1]. LFC is one of the important ancillary services in the power system operation [1-3]. The successful operation of any power system requires the matching of total real power generation (MW) with the total load demand plus losses [1-5]. The operating point of the power system changes with time, and hence power systems may experience deviation in frequency and tie line power exchanges [5]. These deviations may yield undesirable effects. The LFC plays an important role in maintaining the frequency and tie line power at their nominal values [1-5].

LFC problem has been studied by many researchers as it plays a vital role in modern power systems [5]. Optimal control theory based controllers are presented for the LFC study of multi area interconnected power systems [6-8]. A MISO PID controller is used for LFC of decentralized system [8].

Artificial intelligence (AI) and Evolutionary algorithms based controllers have been applied to LFC problem by a wider section of researchers [5, 9-16]. Modern intelligent methods and Evolutionary algorithms, such as Genetic algorithms(GAs), Artificial neural networks (ANNs), Fuzzy logic, Particle swarm optimization (PSO), Modified PSO and its hybrids based optimization have helped in the design of variety of controllers for the LFC systems [5, 9-17]. GA/particle swarm intelligence based optimization technique has been applied to the automatic generation control [13]. The GA optimization method faces some deficiencies like premature convergence which degrades its efficiency and the search capability [5,17]. LFC study of multi area interconnected power systems using type-2 fuzzy approach [14] and Variable structure fuzzy gain scheduling [15] has also been appeared in the literature. PSO is very important and powerful tool to solve the various engineering problems [5, 13, 17].

A fast acting battery energy storage system can effectively reduce the electromechanical oscillations in a power system by providing storage capacity in addition to the kinetic energy of generator rotor [5]. Battery energy storage [5, 20] is very helpful in meeting the sudden load requirements and is effective in improving transient response of frequency and tie line power deviations. A variety of energy storage systems like SMES, BES, CES, RFB have been used in the LFC systems [18-25]. SMES has been applied to LFC problem by many researchers [18, 19, 21]. Rechargeable batteries such as RFB are equivalent to SMES and gaining momentum in research due to their fast and outstanding performance [5, 22-25]

Representing a control area of power systems with multisources like hydro, thermal, gas, nuclear etc is more pragmatic [4,17, 28-30]. Most probably, Ramakrishna and Bhatti [28] have introduced first the LFC analysis using multi-source power generation in a control area. Further, many researchers have used and extended the concept of multi-source power system models for LFC analysis [29-31].

Challa et al. [30] studied the LFC of two area interconnected power system with hydro, thermal and gas based units in each control area. Further, Parmar et al. [31] overcome the difficulty of full state feedback controllers by replacing them with output feedback controllers (few states are required for feedback) and implemented in a single area multi-source power system.

In this paper, the concept of multi-source modeling [4, 17, 28, 30] is extended to present a power system with RFB [22-23] for LFC study using PSO based PI controller. The presented power system comprises non-reheat thermal, hydro and gas based units in the control area. The dynamics of non-reheat thermal, hydro and gas based units are taken from the literature [2, 3, 34-36] which are referred by most of the researchers for LFC study. The model presented for LFC system using PSO technique is more practical as it comprises multi-source of power generation with RFB.

2. PSO AN OVERVIEW

In 1995, Kennedy and Eberhart [37] first introduced the Particle swarm optimization (PSO) technique. This method is motivated by social behaviour of organisms such as fish schooling and bird flocking. A brief theory is presented for the ready reference; more detail on PSO is contained in [37].

The PSO tool [37] provides a population based search technique in which individuals defined as particles change their position with time. In a PSO system, particles fly around in a multi-dimensional search space. During the flight, each particle adjusts its position based on its own experience and the experience of its neighbouring particles, making use of the best fittest position met by it and neighbouring particles [37-39]. The swarm direction of a particle is defined by the set of particles neighbouring the particle and its past historic experience. Each individual in PSO flies in the search space with a velocity which is dynamically adjusted according to its own flying experience.

Let p denote a particle's position and v denote its corresponding flight velocity in a search space [37-39]. Each i_{th} particle is treated as a volume less particle and represented as $p_i=(p_{i1},p_{i2},\ldots,p_{id})$ in the d-dimensional space. The best previous position of the i_{th} particle is recorded and represented as $p_{besti}=(p_{besti1}, p_{besti2},\ldots, p_{bestid})$. The index of the best particle among all the particles is treated as global best particle and represented as p_{besti} . The velocity for the i_{th} particle is represented as $v_i=(v_{i1}, v_{i2}, \ldots, v_{id})$.

The modified velocity and position of each particle can be computed using the current velocity and the distance from p_{besti} to g_{bestd} as shown in the following equations [37-39].

$$v_{id}^{(t+1)} = v_{id}^{t} * w + c_1 * rand() * (p_{bestid}^{t} - p_{id}^{t}) + c_2 * Rand() * (g_{bestd}^{t} - p_{id}^{t})$$
(1)

$$p_{id}^{(t+1)} = p_{id}^t + v_{id}^{(t+1)}$$
(2)

In the above equation, c_1 and c_2 are defined as the acceleration coefficients that pull each particle towards the p_{best} and g_{best} positions. Rand() and rand() are the uniform random numbers between 0 and 1. The term rand()*(p_{bestid}^t - p_{id}^t) is called the cognitive component. The term Rand()*(g_{bestd}^t - p_{id}^t) is called the social component [37-39]. w is the inertia weight factor. Low values of acceleration coefficients allow particles to roam far from the target regions before being tugged back and high values result in abrupt movement towards, or past target regions. Hence, the acceleration constants c_1 and c_2 are usually set to be 2.0 according to past experiences.

A large inertia weight factor enhances global exploration while a low inertia weight factor helps in local search. As originally developed [39], w often decreases linearly from about 0.9 to 0.4 during a run. In general, the inertia weight w is set according to the following equation [37-39]:

$$w = w_{max} - [(w_{max} - w_{min})^* ITR] / (ITR_{max})$$
(3)

Here w_{max} is the maximum inertia weight, w_{min} is the minimum inertia weight, ITR is current no. of iterations, ITR_{max} is maximum no. of iterations.

The performance index is evaluated for each of the particles in the current population. The PSO then updates the particle coordinates based on the equations (1) and (2). This evaluation and updating the coordinates are repeated from population to population until the search termination criterion is met.

3. REDOX FLOW BATTERIES SYSTEM AND MODELLING

RFB are rechargeable batteries and becoming very popular due to their quick responses [5, 22, 23]. RFB are not aged by frequent charging and discharging and have outstanding function during overload [5, 22, 23]. Simple operating principle, long service life, suitability for high capacity systems, quick start up and ease in maintenance are the salient features of these batteries [22, 23].

Recently, RFB have been integrated in LFC scheme to improve the system response in wake of small load disturbances [24, 25]. A basic diagram [22, 23] of the integration of RFB to the power system is as shown in Fig. 1. During the low load periods, battery charges and delivers the energy back to the system during the peak load demands or sudden load changes [22-25]. The dual converter performs both AC-DC and DC-AC conversions.



Fig. 1: Integration of Energy storage system to a power system

A simplified transfer block of the RFB [22-25] is given in the Fig. 2. Area Control Error (ACE) is used as the input command signal for the RFB in controlling the output response in the LFC system [24].



Fig. 2: Simplified transfer function block diagram of the RFB system

The RFB transfer function block is suitably integrated [24,25] to the LFC system as shown in the Fig. 3.

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Fig. 3: MATLAB Simulation of a multi-source power system with Redox flow batteries system

4. POWER SYSTEM SIMULATION AND PARAMETER OPTIMIZATION

A multi-source power system with RFB is presented for LFC system using the PSO based PI controller. The power system is shown in Fig.3. The power system comprises non-reheat thermal, hydro and gas based units in the control area. PSO based PI controller is used in the secondary loop.

The RFB transfer function shown in Fig. 2 is integrated with LFC system. The polarity of the RFB output control signals is kept as per the LFC requirement.

The signal fed into the proposed PI controller is referred to as area control error (ACE). For the single area power system, the ACE can be defined [1-3] in terms of frequency only as there is no interconnection with neighbouring areas.

$$ACE = B\Delta f \tag{4}$$

Where, B is frequency bias parameter (pu MW/Hz) and assumed unity.

The speed changer (Governor) commands for control area can be given as

$$\Delta P_{ref} = -K_p \left(ACE \right) - K_i \int (ACE) dt$$
⁽⁵⁾

Where,

 K_p and K_i are the proportional and integral gains of the PI controller, respectively.

The minus signs are included since each area of the power system should increase its generation level if its frequency error increment is negative [1-3].

In the isolated power systems, the main aim of the controller in the secondary loop is to minimize the frequency deviations and to bring the frequency at its nominal value. The performance index for the PI controller is defined using the Integral square error (*ISE*) criteria as follow:

$$J = \int_{0}^{T} (ACE^{2}) dt = \int_{0}^{T} (\Delta f^{2}) dt$$
 (6)

Where, T is the simulation time.

The controller gains K_p and K_i have been optimized by minimizing the performance index (J) using the efficient PSO technique as described in section 2.

PSO parameters used are as follow:

Population size=40,

Maximum number of iterations=100, $c_1=2$; $c_2=2$,

 $w_{max}=0.9; w_{min}=0.4,$

Simulation time=100s.

The MATLAB code based on PSO algorithm has been run in MATALB Environment [40]. The convergence characteristic of the PSO algorithm is shown in Fig. 4. The minimum value of the *J* becomes 0.00021223. The optimized values of the controller gains are computed as: $K_p = 1.4814$, $K_i = 0.5657$.

5. SIMULATION RESULTS AND DISCUSSION

The optimized gains of PI controller computed have been used in the controller simulation of the LFC system. The parameters of RFB are taken from [22-25] and given in the Appendix A. The nominal power system parameters [1-3, 31] used are given in Appendix A. Nominal area load is taken 1740 MW with 4 % regulation parameters.

The power system simulation is run by giving 1% SLP in the control area. The dynamic responses obtained are shown in the Figs 5-8. The dynamic responses have been obtained with and without consideration of RFB in the LFC system. As

observed, application of RFB improves the transient responses greatly; however settling time remains approximately same. The frequency deviation response of the power system is shown in Fig. 5. The percentage improvement in the frequency overshoot becomes 72. The generating companies either increase or decrease their generation to match the total power generation with the load demand by sensing the incremental change in frequency.



Fig. 4: Convergence characteristics of PSO technique





Fig. 6: Power output of thermal generating unit



Fig. 9: Frequency deviation response for 1%-3% SLP variation

The incremental change in generator power outputs of thermal, hydro and gas units are shown in Figs 6, 7 and 8, respectively. In response to 0.01 pu MW SLP in the area, the incremental change in the generation (pu MW) of thermal, hydro and gas units become 0.006643, 0.00216, and 0.001208, respectively. The total sum of the generation at steady state matches the desired value 0.01 pu MW. Further, dynamic responses have been obtained by varying the SLP levels from 1% to 3%. The frequency deviation response is shown in Fig. 9. The overshoot increases with the increase in SLP level and the settling time remains same. The generator responses are shown in Figs 10-12, an extensive analysis

reveals that generators respond according to the SLP levels in order to match the total power generation with load demand. Thus, all the dynamic responses are satisfactory and PSO based PI controller is robust.



Fig. 10: Power output of thermal generating unit for 1%-3% SLP variation



Fig. 11: Power output of hydro generating unit for 1%-3% SLP variation



Fig. 12: Power output of gas based generating unit for 1%-3% SLP variation

6. CONCLUSION

LFC of a multi-source power system with RFB is presented using efficient PSO technique. PSO based PI controller is used in the secondary loop of the LFC system. The dynamic responses are obtained in wake of 1% SLP in the area. The dynamic responses have been obtained with and without consideration of RFB in the LFC system. Examination of dynamic responses shows that application of RFB improves the transient responses greatly; however settling time remains approximately same. The percentage improvement in the frequency overshoot becomes 72. The total sum of the generation at steady state matches the desired value 0.01 pu MW. Further, dynamic responses have been obtained by varying the SLP levels from 1% to 3%. The overshoot increases with the increase in SLP level and the settling time remains same. The PSO based PI controller is robust and performs well under the wide range of load disturbances. All the simulation results are as per the requirements of the LFC system.

APPENDIX A

Power System Parameters

$P_{rr} = 2000 \ MW$ (Rated capacity of the area)
$P_L^0 = 1740 \text{ MW}$ (Nominal load of the area)
f = 60 Hz,
H = 5 MWs/MVA
D = 0.0145 pu MW/Hz,
$K_{PS} = 68.9655 \text{ Hz/pu MW},$
$T_{PS} = 11.49$ s,
$T_{sG} = 0.06 \text{ s}, T_T = 0.3 \text{ s},$
$R_{TH} = R_{HY} = R_G = 2.4 \text{ Hz/puMW},$
$T_W = 1.1 \mathrm{s}, T_{RS} = 5 \mathrm{s},$
$T_{RH} = 28.75 \mathrm{s},$
$T_{GH} = 0.2 \mathrm{s}, X_G = 0.6 \mathrm{s},$
$Y_G = 1$ s,
$c_g = 1, b_g = 0.05 \mathrm{s}, T_F = 0.23 \mathrm{s},$
$T_{CR} = 0.01 \text{s}, T_{CD} = 0.2 \text{s}$
$K_{RF} = 1.8, T_D = 0s$

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