

DIFFERENTIAL SEARCH ALGORITHM IN MULTI MACHINE POWER SYSTEM STABILIZERS FOR DAMPING OSCILLATIONS

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ABSTRACT

Power system oscillations, a major problem in power system, is suppressed employing power system stabilizers (PSSs). Proper optimization of PSSs is a complex design problem. In this paper, a bio-inspired metaheuristic optimization technique named as differential search algorithm (DSA) is presented to solve the optimization problem of multi machine PSSs. The optimization of PSSs is converted as a cost function then DSA is applied to tune the optimal parameters for PSSs by minimizing the cost function. PSSs are optimized in order to achieve adequate damping for local and inter-area modes of growing oscillations in a multi machine power system. Simulations are conducted in linear and non-linear models of power system to verify the robustness of proposed algorithm. A comprehensive investigation is conducted to compare the performance of DSA based PSSs with the tuned PSSs using bacterial foraging optimization algorithm (BFOA) and particle swarm optimization (PSO) in terms of convergence, improvements of electromechanical modes and system damping over oscillations. The obtained results show that the presented DSA technique is efficient for PSS optimization for the safety of multi machine power system.

Keywords: *Differential search algorithm (DSA), Power system stabilizer (PSS), Power system oscillations, Damping controller, Multi machine power system*

1. INTRODUCTION

Power system stability is a serious concern for the power system security and reliability. Large interconnection of power system may cause some severe and complex problem such as power system electromechanical oscillations [1]. The oscillations eventually may keep growing to lead complete system collapse if none damping control is undertaken [2-*]. However, in order to suppress growing oscillations, the damping scheme by means of power system stabilizers (PSSs) have been applied mainly to enhance system dynamic stability [3]. PSS is an auxiliary control device installed on synchronous generators that can manage the stability by providing an additional control signal to excitation system. In general, rotor speed signal is taken as the input to the PSSs control loop. In addition, the fixed lead-lag structure of conventional PSS (CPSS) is

recommended by major power utilities to carry out an extensive overall power system dynamic performance [4]. It was concluded that proper selection of CPSS parameters can dramatically improve system damping and enhance stability. There are several techniques have been used to determine appropriate settings of PSS parameters such as root locus techniques [5], deterministic optimization techniques [6] and heuristic optimization techniques [7]. The tuning of PSS parameters is generally very difficult to solve using traditional techniques. Conversely, eigenvalue based heuristic optimization techniques are easy to implement and solve complex power system problem for PSS design procedures. Therefore, finding the optimal parameters for PSSs is one of the optimization problems of stability studies.

In last few years, there are many bio-inspired meta-heuristic algorithms such as BAT algorithm [8], and genetic algorithm (GA) [9] have been

utilized to simplify the difficulties of multi machine PSSs design. The meta-heuristic algorithm like particle swarm optimization (PSO) has been introduced for multi machine PSSs design [10-11]. Unfortunately, PSO has some control variables which affects convergence if they are not properly selected. In addition, sometimes it gets trapped in local minima easily. Bacterial foraging optimization algorithm (BFOA) was also proposed for coordination control of PSSs with other devices in order to achieve sufficient damping over different modes of oscillations by minimizing single time domain cost function [12-13]. Furthermore, to improve convergence rate and better damping performance, research was suggested using eigenvalue based multi-objective (D-shaped) cost functions to suppress different oscillating modes [14]. In that case, previously used meta-heuristic algorithms may not succeed to obtain optimum design solution of PSS tuning.

Generally, tuning PSS parameters is a multimodal optimization problem. Therefore, most of the existing algorithms get easily trapped and failed to obtain the reasonably accurate solution especially in large system. In addition, the boundary conditions of optimizing parameters make it harder to find optimal solution. To overcome all the issues above, a new meta-heuristic technique named differential search algorithm (DSA) can be applied in PSS design [15]. The uniqueness of the DSA concept is that it is simpler and easy to implement.

In this paper, DSA algorithm is presented for efficient optimization of PSSs in a multi machine power system. PSS parameters are effectively tuned with DSA so that sufficient damping is ensured over different modes of oscillations. The optimization problem of PSSs is converted into a multimodal search problem (cost function). Later on, the formulated cost function was minimized using the proposed DSA technique for optimum design of PSSs that concurrently improving damping over oscillations. Thus, a system safety is ensured from unwanted incidents [16-18]. Although, the test system taken in this research is a medium sized test power system. The sufficient damping by the damping controllers is the main objective of this research using an efficient optimization technique.

2. MULTI MACHINE POWER SYSTEM

In this research, 2 areas 4 machines system is considered as the multi machine test power system

to optimize PSSs using DSA. The test system is displayed as a single line diagram in the Figure 1. This is a benchmark multi machine power system to study multimode oscillations inherited in interconnected system [19]. The system consists of two generation area and two load area interconnected between bus 9 and bus 10 through double circuit tie lines. There are two synchronous generators in each area rated at 900 MVA, 20 kV. Each generator is linked via transformer to the 230 kV power line. The entire system is quite heavily loaded (at Bus 11, 976MW and 100MVar; at Bus 12, 1765MW and 100MVar) and area 1 is transferring 400MW to area 2. The turbines, governor and excitation systems are identical in all generators. Excitation system contains static exciter equipped with generic PSS. Detailed dynamic properties and data about bus, line, exciter, turbine, governors, loads, and machines are adopted from [20]. Each generator of multi machine system is modeled by six first-order nonlinear ordinary differential equations. This test system shows the basic electromechanical oscillations inherent in multi machine interconnected power system. In this system, generators are tempted to oscillate against other generators for local and inter-area modes of oscillations after subjected to disturbances [21]. Local mode and inter-area mode of oscillations are due to the generators of same area (G1-G2 and G3-G4) and the generators for other areas (G1-G4 and G2-G3) respectively.

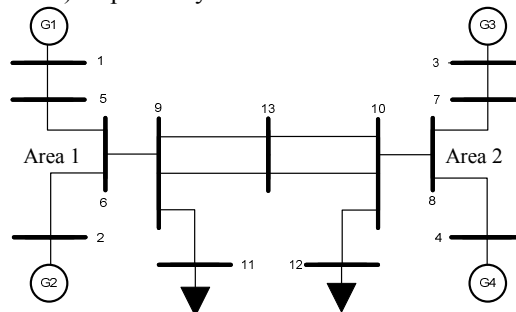


Figure 1: 2 areas 4 machines power system.

3. POWER SYSTEM STABILIZER

A PSS is an additional control devices installed at excitation system of a synchronous generator. The main purpose of PSS is to maintain power system stability by damping electromechanical oscillations providing additional control signal to the input of excitation system. Local stabilizing signal such as generator terminal frequency, rotor angle deviation, rotor speed can be taken as input

signal to PSS. In this case, the rotor speed of the synchronous machine is taken as PSS input.

The well-known generic structure of the PSS taken in this research is shown in the Figure 2. It is composed of a gain block; a washout block and two stages lead-lag blocks. Gain block estimate the amount of damping, washout block acts as a high pass filter and the lead lag block determine the phase lead or lag required to compensate between exciter input and the generator electrical torque.

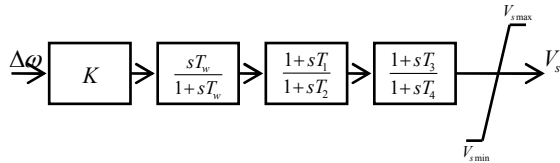


Figure 2: Lead-Lag Structure of Power System Stabilizer.

4. POWER SYSTEM MODEL

The dynamics of a power system can be represented by a set of non-linear ordinary differential equations (ODE).

$$\begin{aligned} \dot{x} &= f(x,u) \\ y &= g(x,u) \end{aligned} \quad (1)$$

where x is the state vector, and u is the input vector.

For system simplification, linear time invariant (LTI) system theory can be applied to streamline non-linear equations into state space (SS) model forming a set of coupled first order linear differential equations.

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx + Du \end{aligned} \quad (2)$$

where, A , B , C and D are the state matrix, input matrix, output matrix, and feed-forward matrix. Later, LTI state space model is characterized in frequency domain which simplify to analyses of the dynamic behavior of power system around an operating point. In the analysis, the stability is determined by observing eigenvalues ($\lambda_i = \sigma_i \pm i\omega_i$) of matrix A . The system damping factors and damping ratios are determined from the real (σ_i) and imaginary part (ω_i) of eigenvalues as follow-

$$\text{Damping factor, } \sigma_i = \text{real}(\lambda_i) \quad (3)$$

$$\text{Damping ratio, } \zeta_i = -\frac{\sigma_i}{\sqrt{\sigma_i^2 + \omega_i^2}} \quad (4)$$

where $i = 1, 2, 3, \dots, i$ and i is the total number of eigenvalues of the system.

5. COST FUNCTION

To achieve fast and sufficient damping over growing oscillation due to disturbance in the power system, the eigenvalue based multi-objective cost function (shown in Equation 5) involving damping factors and damping ratios of poorly and unstable electromechanical modes is preferred [14].

$$f = \sum_{j=1}^{np} \sum_{\sigma_{ij} \geq \sigma_0} (\sigma_0 - \sigma_{ij})^2 + a \sum_{j=1}^{np} \sum_{\zeta_{ij} \leq \zeta_0} (\zeta_0 - \zeta_{ij})^2 \quad (5)$$

where, a is the weight factor and $j = 1, 2, 3, \dots, np$ and np is the number of operating points considered during PSS design. This multi-objective cost function ensure system stability by forcing electromechanical modes to be placed into a D-shaped region due to multiple system operating points. If the expected damping factor and damping ratio are σ_0 and ζ_0 respectively then the D-shaped region for stability is formed imposing the condition $\sigma_{i,j} \leq \sigma_0$ and $\zeta_{i,j} \geq \zeta_0$ as displayed in the Figure 3.

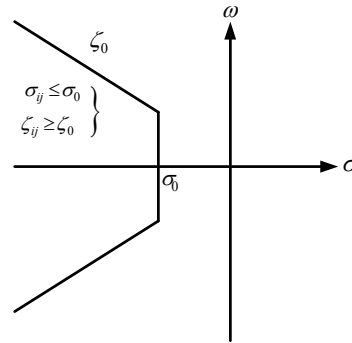


Figure 3: System stability criteria defined by D-shaped region in S-plane.

In that case, the requirements $\sigma_{i,j} \leq \sigma_0$ and $\zeta_{i,j} \geq \zeta_0$ represent the damping factor and damping ratio for i -th eigenvalue of j -th operating condition respectively. Therefore, the design problem of PSSs can be formulated to minimize the cost function, f subject to:

$$K_{\min} \leq K \leq K_{\max}$$

$$T_{1,\min} \leq T_1 \leq T_{1,\max}$$

$$T_{2,\min} \leq T_2 \leq T_{2,\max}$$

$$T_{3,\min} \leq T_3 \leq T_{3,\max}$$

$$T_{4,\min} \leq T_4 \leq T_{4,\max}$$

The final value of cost function will become zero [14] if all electromechanical modes move into the D-shaped stability zone.

6. DIFFERENTIAL SEARCH ALGORITHM AND IMPLEMENTATION

DSA is a nature inspired meta-heuristic optimization algorithm developed by Civicioglu [15], and it is described as an effective technique for multimodal optimization solution. The concept behind its development is the seasonal migration of different species of nature for the search of fruitful living. This means discovering better stopover sites during the movement of organisms. In this algorithm, the individual search organisms together form a bigger population known as superorganism. The superorganism move towards the target of donors in order to discover stopover sites. The donors are assumed to be in fertile area. During their movement for discovery, superorganism checks whether some randomly chosen locations meet their transitory criteria or not. If any location is appropriate for their lays over temporarily during the journey, the individuals of the superorganism ($x_{\text{superorganism}}$) that revealed the stopover, right away settle in that location and carry on their journey from that location. The discovery of stopover sites (x_{stopover}) is determined using the Equation 6.

$$x_{\text{stopover}} = x_{\text{superorganism}} + \text{Scale}(x_{\text{donor}} - x_{\text{superorganism}}) \quad (6)$$

The successful migration of the superorganism depends on the mechanism of stopover site discovery. The *Scale* factor is estimated using gamma random function (*randg*) for each generation as shown in Equation 7. Donors are made up of reshuffling individuals of superorganism as shown in Equation 8.

$$\text{Scale} = \text{randg}[2 \cdot \text{rand}] \cdot (\text{rand} - \text{rand}) \quad (7)$$

$$x_{\text{donor}} = (x_{\text{superorganism}})_{\text{random_reshuffle}} \quad (8)$$

It is worth to mention that some randomly selected participants involve in the search of

stopoversite discovery and they are considered to discover global minima point of the problem.

In DSA, exploration and exploitation concept are used simultaneously to escape local minima stagnation and to narrow down the search space for actual optimal solution, respectively. This algorithm has only two control parameters and both of them are stochastic variables. Both parameters are used to randomize the selection process of search participants. As they don't have any direct influence over organism's movement, the initial selection of them does not affect the solution at all. Moreover, the convergence rate is very swift due to minimal factors to be considered during migration. In addition, DSA has no inclination to global best, personal best solution. Therefore, noisy-stagnation for over-acceleration or under acceleration is easily avoided. All these unique specialties are recommended to solve complex multimodal optimization problems. Interested reader are referred to the original research paper of this algorithm in [15].

In this paper, DSA is applied to optimize parameters of PSSs for its robust performance over different modes of power system oscillations. The flowchart of DSA based PSS design in multi machine power system is portrayed in the Figure 4.

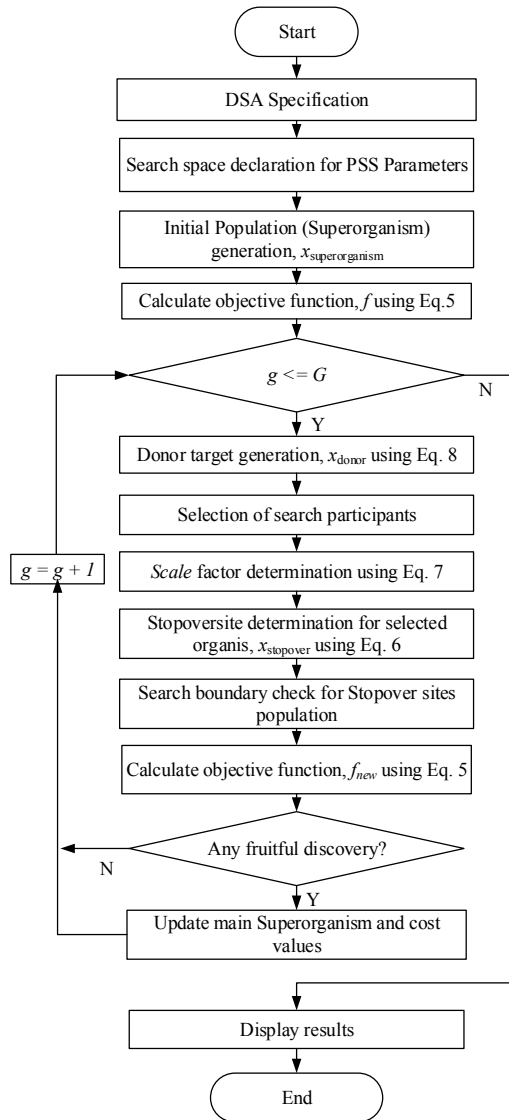


Figure 4: Flowchart for DSA based PSS design.

7. RESULTS AND DISCUSSIONS

In order to validate the applicability and effectiveness of proposed DSA to design PSSs in multi machine power system, a comprehensive analysis that includes simulations in linear model as well as non-linear model is conducted. The purpose of eigenvalue based linear model optimization is to minimize the formulated cost function that concurrently improves system inferior electromechanical modes. Finally obtained optimized parameters are justified in nonlinear time domain simulation. For both type of simulations, obtained results are compared against BFOA and PSO based design performance. The weight factor, a (in the Equation 5) was set to 10

[14]. The expected damping factor and damping ratios were selected to -1.5 and 0.2 respectively. The settings for optimization algorithms are given in Appendix section. In majority of preceding literatures, T_1 and T_3 were typically considered variables to be optimized [8, 13]. For this research, T_2 and T_4 were included in the optimization to extend complexity of optimization problem. There are four PSSs in 2 areas 4 machines power system and for each generator, the number of optimizing parameters are five ($K, T_1, T_2, T_3,$ and T_4). The time constant T_w was set to 10s. Therefore, total 20 parameters were optimized in search problem using proposed DSA technique. The search boundary limits are listed in Table 1.

Table 1: Boundaries of search space

Parameters	K	T_1	T_2	T_3	T_4
Minimum	0.01	0.01	0.01	0.01	0.01
Maximum	100	2	2	2	2

Simulations were continued in the linear model of power system 20 times for each optimization technique. The Figure 5 is the comparison of convergence curves obtained from best optimization results of DSA, BFOA and PSO techniques. From the Figure 5, the convergence of DSA is faster as compared to both BFOA and PSO techniques. Thus, DSA shows rapid convergence for PSS optimization in complex power system.

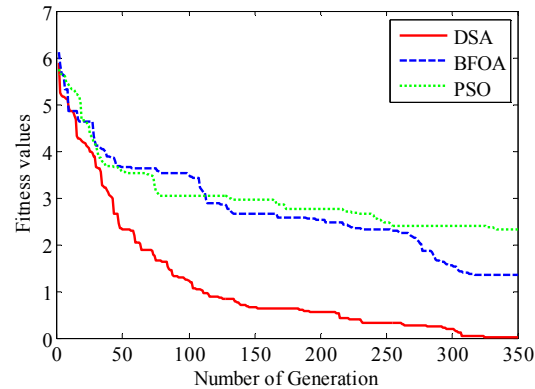


Figure 5: Comparative convergence curve for DSA, BFOA and PSO achieved during best solution

In the optimization, system stability is associated with the location of eigenvalues in complex-s plane. The improvement of system eigenvalues after optimization is shown in the Figure 6 for DSA, BFOA and PSO techniques. Only for DSA-based techniques, the system eigenvalues have achieved the minimum damping ratio as considered in the design according to the Figure 6. On the other hand, lightly damped

eigenvalues using the DSA-based techniques are nearest to the minimum damping factor as compared to BFOA-based and PSO-based techniques shown in the Figure 6.

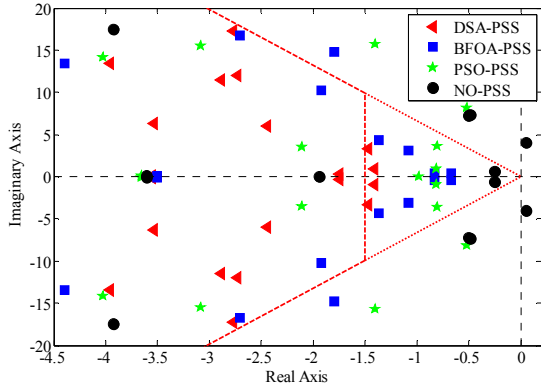


Figure 6: Comparative progress of electromechanical modes in D-shaped region for best optimization solution

The non-linear model of test system were executed using the optimized parameters. A three-phase fault was created to one tie circuit between bus 9 and 13 near at bus 9 of 2-area power system shown in the Figure 1. The fault was applied at 100ms of simulation time (9s) for 150ms. After that the system was operating with one tie line only and oscillations were grown in the system. Figure 7 and Figure 8 are the damping responses over local as well as inter-area modes of oscillations, respectively. From the figures, the system is totally unstable without PSSs because of the three-phase faults. However, the installation of PSSs after optimization bring back the stability for all optimization techniques as shown in figures 7-8. From both figures, as compared to BFOA-based and PSO-based designed PSSs, DSA-based designed PSSs have significantly reduced overshoots and settling times for both type of oscillating modes. Thus, system stability is improved notably in a disturbed system using DSA-based designed PSSs.

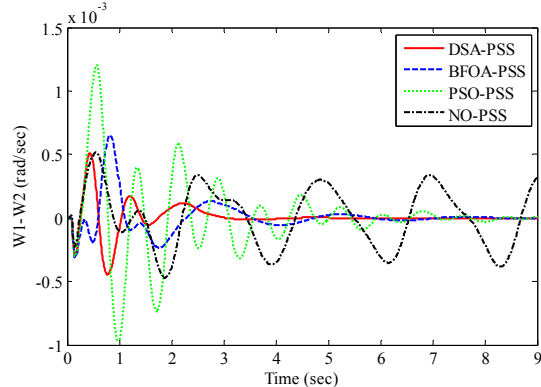


Figure 7: Local mode between G1 and G2 (W1-W2) for three phase fault at bus 9

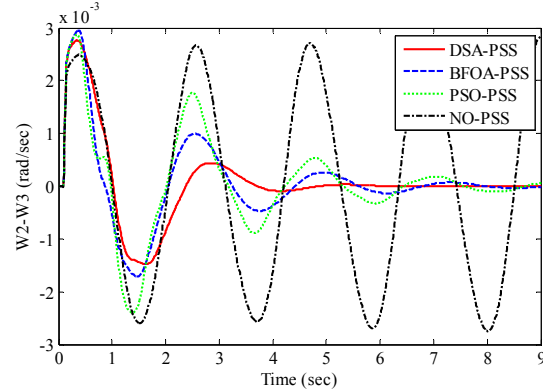


Figure 8: Inter-area mode between G2 and G3 (W2-W3) for three phase fault at bus 9.

8. CONCLUSION

This paper proposes DSA for optimization of PSSs controller in order to provide adequate damping restraining power system oscillation. The optimization of PSSs is formulated as an optimization problem. After that, DSA is applied to find for the best optimized parameter values. To verify the performance of the proposed optimization algorithm, simulation is conducted in linear model as well as non-linear model and then compared with other popular meta-heuristic optimization algorithms such as BFOA, PSO. The comprehensive analyses are based on the convergence rate, improvement of electromechanical modes in a visual D-shaped s-plane, and the non-linear time domain simulation. In order to design PSSs efficiently, the efficiency of DSA designed PSSs is remarkable than that of BFOA and PSO techniques in multi machine power system. The eigenvalues are more stable in the proposed design, which indicates efficient search techniques of the proposed optimization algorithm. In the time domain simulation, enough damping is achieved to stabilize system from disturbances and its consequence incidents. Therefore, the presented DSA optimization technique has great potential to solve complex design problem of coordinated PSSs for the safety of multi machine power system.

APPENDIX:

Optimization algorithm settings:

- (a) DSA Settings: dimension size, $D = 20$; population size, $N = 50$; generation, $G =$



- 350; control parameters, $P1 = P2 = 0.3 \times rand$.
- (b) BFOA Settings: dimension size, $D = 20$; population size, $N = 50$; chemotactic steps, $N_c = 70$; reproduction steps, $N_{re} = 5$; elimination-dispersal steps, $N_{ed} = 5$; $p_{ed} = 0.25$; $C = 0.05$; $m_{ar} = 0.1$; $w_{at} = 0.2$; $w_{re} = 10$.
- (c) PSO Settings: dimension size, $D = 20$; population size, $N = 50$; generation, $G = 350$; cognitive constant, $C_1 = 2$; social constant, $C_2 = 2$; $w_{min} = 0.4$; $w_{max} = 0.9$.
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