Power System Design using Firefly Algorithm for Dynamic Stability Enhancement

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Abstract

Utilising additional devices in power systems have been developed by industry. Devices such as a Power System Stabilizer (PSS) and a Superconducting Magnetic Energy Storage (SMES) are commonly employed in industry. This work investigated the coordination of a PSS and SMES applied to a power system to enhance dynamic stability. To obtain optimal coordination, the parameters of the PSS and SMES are tuned using the Firefly Algorithm (FA). The simulation of the power system, PSS, and SMES has been performed using MATLAB and Simulink, and the FA run in Matlab. For testing the small signal stability, the eigenvalue of the system will be investigated, while for dynamic stability the system will be given an external disturbance. The rotor angle and frequency deviation of the power system are compared without a controller, with a PSS and SMES included, and with the PSS and SMES tuned by FA. The simulation results show that the proposed system can improve not only small signal stability (steady state stability) but also dynamic stability.

Keywords: Power System Stabilizer, Superconducting Magnetic Energy Storage, Firefly Algorithm, Dynamic Stability

1. Introduction

Electricity demand in recent times has become a liability and is more than likely to increase rapidly in the future. This growing demand for power requires electricity providers to increase their generating capacity and to expand their distribution network, thus making the entire system large. On large systems there are common problems associated with system stability. Disturbances that affect the stability of the system can be transient disturbance or dynamic disturbance. Large transient disturbance is one that occurs as a result of a broken transmission line, whilst dynamical disturbances are minor ones, such as load changes [1].

In the case of dynamic stability, the system is often impaired owing to the effect of load changes, which cause the system to have frequency oscillation. According to Goshal [2] the oscillation frequency that occurs is relatively low, and within range of 0.2 to 0.3 Hz. If this oscillation not well damped, magnitude of these oscillation may keep growing until system loss synchronism. To overcome these problems, additional control equipment should be installed on the system, where the Power System Stabilizer (PSS) [3] is one approach that is often used. PSS is an additional control device which serves to dampen the oscillation frequency and voltage on the generator. Since the system is larger, the PSS on its own is not enough to overcome the problem, hence other devices such as energy storage systems are required. In this era, there are a lot of energy storage approaches that have been employed, systems such as Battery Energy Storage (BES) [4, 5, 6], Redox Flow Batteries (RFB) [7, 8], Capacitive Energy Storage (CES) [6, 9] and Superconducting Magnetic Energy Storage (SMES) [6, 10-13]. SMES is one energy storage approach that is commonly used in a power plant that gives active power temporary in the load. In Ansari and Velusami [13] SMES has been used to improve the performance of a wind-diesel hybrid power system.

The use of additional equipment on the electric power system such as PSS and SMES that is not appropriate in stabilizing the system can lead to issues. The problems often arise owing to errors in taking the reference signal, or the parameters of the devices are not optimal. Adding a controller that increases system performance can make the system more unstable. Owing to these issues, an optimization method should be used to overcome the problems.

Nature always give us inspiration, and there are a number of computation or iteration methods inspired from the apparent intelligence shown in nature [14]. Smart computation can
be categorized into three groups: physically inspired such as the Simulated Annealing Algorithm; social inspired like the Tabu Search Algorithm, or Imperialist Competitive Algorithm, and lastly biologically inspired approaches such as Genetic Algorithms, Particle Swarm Optimization, Artificial Immune System, Ant Colony Optimization, Bat Algorithm and the Firefly Algorithm [14]. The Firefly Algorithm is one of the newer nature inspired algorithms, and is inspired by the behaviour of fireflies. Application of the Firefly Algorithm (FA) in optimization of power systems have been developed. There are a number of applications such as in the research of Niknam [15]. In that research a FA is used to find the optimal generation level for a power system, achieving the lowest price with a reserve constraint. Another application of the FA is as a tool to tune the parameters of a PI controller applied to speed control of a DC series motor [16]. The Firefly Algorithm has been used to address the economic emissions in a power system [17]. The FA has not only been used in power systems but also in cultivation control, for example Roeva O, Slavov [18] used a FA to tune a PID controller for glucose concentration control. All of the research proven that FA is one of the metaheuristic algorithm that very good to solve optimization problem.

Some research have been proposed to solve dynamic stability enhancement. Like in Parimi and friends research [19], they try to enhance power system stability by using one of Flexible Alternating Current Transmissions System (FACTS) devices called Interline Power Flow Controller (IPFC). Another research to enhance dynamic stability of power system have proposed by Mohammad Kashki [20], they try to improve dynamic stability in power system by combining PSS and one FACT devices called Static Phase Shifter (SPS). Kashki used Combinatorial Discrete and Continuous Action Reinforcement Learning Automata to tune PSS and SPS parameter.

Usman research [21], try to enhance the stability by controlling excitation system of power system using Automatic Voltage Regulator (AVR) combine by PSS, this time Particle Swarm Optimization have used by them to optimized parameter of AVR and PSS. Different from others, Hakim and friends [22], try to improve dynamic stability by designing PSS based of Fuzzy PID. They combine 2 Artificial Intelligent which were Fuzzy and Genetic Algorithm.

This research proposed in involves the coordination of a Power System Stabilizer (PSS) and an energy storage device called s Superconducting Magnetic Energy Storage (SMES), to improve the dynamic stability of a power system. To get improved performance of the devices, the parameters of the PSS and SMES will be tuned using one of the nature inspired algorithms, namely the Firefly Algorithm (FA). To find the best solution the objective function utilised within the Firefly Algorithm is the Comprehensive Damping Index (CDI). It is expected that by using this method the dynamic stability of the power system can be enhanced.

2. Research Method

2.1. Power System Model

Generally the power plant system can be described as shown in Figure 1. The governor is a part of a generating unit that serves to regulate the provision of fuel (steam or water) in the generating system in order to obtain a stable rotor speed. If there is a change at the output of the governor owing to a load change, then there will be a feedback function that is set by the governor to readjust the rotation of the rotor.

![Figure 1. General Power Plant Model](image-url)
The excitation system is a control system for the generator output, such as voltage, current or power factor. If there is a change in the generator output, then the excitation system serves to control the generator in order to adjust and find a new balance point. Both parts of the control system, the governor and the excitation systems, have different response times. The governor has a slow response, while the excitation system has a fast response. In the work presented here, the power system model that will be used is one that is connected to the infinite bus. The power system will be modelled in Laplace domain as shown in Figure 2. This model first introduce in [23] journal, and now this model are commonly used to solve the small signal stability and dynamic stability because in this model power system have been modeled to linear model. All of the value in this model is asum in per unit value. So this model limited in per unit value, there are no var, watt, ampere, or voltage value.

\[
\begin{align*}
\Delta U_i &+ \frac{K_w}{1+sT_w} \Delta T, \\
\Delta T &+ \frac{1}{1+T_m} \Delta T_m, \\
\frac{1}{Ms+D} \Delta \omega &+ \frac{2\pi f}{s} \Delta \delta
\end{align*}
\]

2.2. Power System Stabilizer

The block diagram of the Power System Stabilizer (PSS) is shown in Figure 3. The PSS block diagram consists of a gain block, washout block, two blocks of lead-lag and a limiter. The block gain function is to adjust the amount of reinforcement that is obtained in accordance with the desired torque. The washout block serves to provide a steady state bias output PSS that will modify the terminal voltage of the generator. The PSS is expected to only respond the transient variation of the rotor speed signal to the signal generator and DC offset [25].

\[
\begin{align*}
\frac{K_w}{1+sT_w} \Delta V, \\
\frac{1}{K_e+sT_e} \Delta V, \\
\frac{1}{1+sT_w K_e} \Delta E_{pq}, \\
\frac{1}{1+sT_m K_e} \Delta E_{pq}
\end{align*}
\]

Figure 2. Power System Dynamic Model (Taken from [24])

Figure 3. Block Diagram of PSS (Taken from [25])
The lead-lag functions as phase-lead producing characteristics appropriate to compensate for the phase lag between the input excitation and torque generator. The limiter function restricts the PSS action on the AVRS. For example, in the event of load shedding, the AVR acts to reduce the terminal voltage of the generator when the PSS generates a control signal to raise the voltage (because the generator rotor speed increase in size during load shedding). In this condition it is necessary to disable the PSS. It should be noted that, the high negative limit values can disrupt the stability of the first swing [25].

2.3. Superconducting Magnetic Energy Storage
A Superconducting Magnetic Energy Storage (SMES) system within an electric power system is used to control the balance of power in the synchronous generator during periods of dynamic change. The SMES is installed in terminal bus generator on the model of the power system, as shown in the block diagram of Figure 4. All of the calculation in here asum and limit and Laplace domain.

\[
\Delta E_d = \frac{1}{1 + sT_{dc}} [k_d \Delta \omega_1 - k_{id} \Delta I_d] \quad (1)
\]

\[
\Delta I_d = \frac{1}{L_s} \Delta E_d \quad (2)
\]

Where \(\Delta E_d\) is DC voltage applied to the inductor, \(\Delta \omega_1\) is the rotor deviation, \(\Delta I_d\) is the current flowing through the inductor, \(k_{id}\) is the gain for feedback \(\Delta I_d\), \(T_{dc}\) is the converter time delay, \(k_d\) is a gain constant and \(L\) is the inductance of the coil. The deviation in the inductor real power of the SMES unit is expressed in the time domain as follows [12]:

\[
\Delta P_{smes}(t) = \Delta I_d \Delta E_d + \Delta I_d \Delta E_d \quad (3)
\]

\(\Delta P_{smes}(t)\) is the real power that will be given to the grid. The energy stored in the SMES at any instant in time \(W_{smes}(t)\), is given by [12]:

\[
W_{smes}(t) = \frac{L I_d^2}{2} \quad (4)
\]

2.4. Firefly Algorithm
Firefly Algorithm is a nature inspired algorithm that derives from the characteristics of fireflies. This algorithm was defined by Dr Xin-She Yang from the University of Cambridge, UK in 2008. In this algorithm there are three baseline considerations. All of the fireflies are unisex so that a firefly will be attracted to any other firefly regardless of gender [26]. The attractiveness
is proportional to the brightness of a firefly’s flickering light. Therefore, for every two firefly’s blinking, one of them would move to another light. Their brightness decreases as the distance between the fireflies increases. If there is no brightest firefly within the population, then all fireflies will move randomly [27]. The brigness of a firefly is influenced or determined by the objective function [27].

The objective function used to test the system is the Comprehensive Damping Index (CDI) [28, 29]. Furthermore the dynamic model of power system will be converted into state matrix in equation (5) and (6). By using the state matrix is easier to find the eigenvalues of a higher order system, the eigenvalue itself can be presented as equation (7). To find the damping value, then equation (8) is used. Finally the CDI objective function can be determined by equation (9). The variable that will be optimized by the FA is the PSS parameter and SMES parameter, there will be range of value of each parameter. The reason why this parameter is chosen is because it represents the PSS and SMES value.

\[
\Delta \dot{x} = A\Delta x + B\Delta u \tag{5}
\]

\[
\Delta y = C\Delta x + D\Delta u \tag{6}
\]

Where \( \Delta \dot{x} \) is state matrix, \( \Delta y \) is the output variable matrix, \( u \) is the input matrix, \( A \) is the system matrix, \( B \) is the input matrix, \( C \) is the measurement matrix, and \( D \) is the input to the output matrix.

\[
\lambda_i = \sigma_i + j\Omega_i \tag{7}
\]

\[
\xi_i = \frac{\sigma_i}{\sqrt{\sigma_i^2 + \Omega_i^2}} \tag{8}
\]

\[
CDI = \sum_{i=1}^{n} (1 - \xi_i) \tag{9}
\]

Where \( \lambda_i \) is the eigenvalue, \( \sigma_i \) is the real component of the eigenvalue, \( \Omega_i \) is the imaginary component of the eigenvalue, and \( \xi_i \) is the damping ratio of the system.

3. Results and Analysis

Simulation was performed using MATLAB 2010, where the power system dynamic model, Power System Stabilizer (PSS) and Superconducting Magnetic Energy Storage (SMES) were created within Simulink. The Firefly Algorithm (FA) was implemented using an M-file. Results and analysis are presented in two sections, first section shows the eigenvalue analysis and results, while the second section provides information about the dynamic performance of the system by showing the frequency and rotor angle response. Table 1 provides the parameters of the PSS and SMES that were optimized using the FA, while Figure 5 shows the convergence graph for the Firefly Algorithm (FA) as it finds the optimal value, or meets the criteriation of the objective function.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Value</th>
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</thead>
<tbody>
<tr>
<td>Kpss</td>
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</tr>
<tr>
<td>Tw</td>
<td>10.5</td>
</tr>
<tr>
<td>T1</td>
<td>0.465</td>
</tr>
<tr>
<td>T2</td>
<td>0.2965</td>
</tr>
<tr>
<td>T3</td>
<td>0.3845</td>
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<tr>
<td>T4</td>
<td>0.2257</td>
</tr>
<tr>
<td>Ko</td>
<td>60</td>
</tr>
<tr>
<td>Kid</td>
<td>5.9143</td>
</tr>
</tbody>
</table>
3.1. Small Signal Stability Analysis

Small signal stability usually called steady state stability [1], relates to the stability of the system where there are no external disturbance, or the stability of the system itself. This stability is very important is because by finding this stability, we can know whether this system is stable or not, if we know the system is stable than we can start to run the system.

The stability of the system can be determined by finding the eigenvalue of the system. If the real parts of the eigenvalue are negative then the system is stable.
Figures 6 to 8 show the eigenvalue plot for the system without a controller, the system with conventional PSS and SMES, and the system with PSS and SMES tuned by the FA respectively. Based on [1] the system can be called stable if the real parts of the eigenvalue are negative, where the larger the magnitude then the faster the stability. From Figure 6, it can be seen that the system without a controller is in itself stable. However, for an improved performance, the system needs to have more negative magnitude. Figures 7 and 8 show that the stability of the system increases when the system has the PSS and SMES installed. Furthermore, enhanced results are obtained when the PSS and SMES have their parameters tuned by the FA.

Another important parameter for stability is the damping ratio. If the damping ratio has a negative value then there will be a positive eigenvalue that will make the system unstable. According to Rogers [30], a power system with damping ratio greater than 0.05 is satisfactory. For this work all of the damping ratio values are greater than 0.05, as shown in figure 6 to 8. By installing the PID and SMES in the power system, the damping ratio of the system is increased: the higher the damping ratio then the oscillation level decays.

3.2. Dynamic Stability Analysis

Dynamic stability is the ability of the system to maintain a synchronised condition after the first swing happens until the system reaches a new equilibrium steady state condition [31]. This stability is where there is an external disturbance such as load change.

To analyse the dynamic behaviour of the power system, the system needs to be modelled into a linear model (that is through Laplace transformation) and given a load change 0.01 pu: the reason why 0.01 is chosen because 0.01-0.05 pu is the normal way to simulate the load changing in power system. In this simulation the load change is represented as a step signal in Simulink, where the final value of the step signal is 0.01 pu. Analysis is conducted through comparing the frequency response and rotor angle response of the power system without controller, with the inclusion of a PSS and SMES, and also with the PSS and SMES tuned by the FA. Figures 9 and 10 show the frequency response and rotor angle of the power system for all three configurations, while Tables 5 and 6 illustrate the overshoot and settling time for both frequency and rotor angle for the three system arrangements.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Uncontrolled</th>
<th>PSS SMES</th>
<th>PSS SMES (FA tuned)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Overshoot (pu)</td>
<td>-8.622e5</td>
<td>-6.46e5</td>
<td>-5.743e5</td>
</tr>
<tr>
<td>Settling Time (Sec)</td>
<td>11.57</td>
<td>7.53</td>
<td>5.33</td>
</tr>
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</table>
Tables 5 and 6 show that PSS and SMES can decrease the overshoot and reduce the settling time of the frequency and the rotor angle in a power system. The best result for the frequency response is that when the PSS and SMES is tuned by the FA. When the system is installed with the PSS and SMES then the system overshoot is reduced to 0.00002979 pu and the settling time is 6.24 seconds, which is faster than the system without a controller. For the rotor angle, the system including the PSS and SMES can dampen the overshoot, however the time settling is larger than that for the system without a controller. As shown in the graph, the system without a controller has quite an oscillatory response which is not appropriate because if this oscillation rises again and again, then the system will be out of synchronization and it will collapse. That is why using an additional controller is required. The PSS is used to control the excitation of the power system and the SMES works as energy storage to help the governor. Owing to that response frequency and rotor angle is better than system without controller. The FA itself helps an engineer to find the best parameter of the system: that is why the system with a PSS and SMES tuned by the FA have the best parameters.
4. Conclusion

Including a Power System Stabilizer (PSS) and Superconducting Magnetic Energy Storage (SMES) can improve the eigenvalue of a power system, thus making the overall system more stable. As well as the eigenvalue, the dynamic stability of the power system improves when the system has the PSS and SMES installed, as shown in the response of the frequency and rotor angle. The Firefly Algorithm (FA) can be employed to tune the PSS and SMES parameters so that it is easier to find the more appropriate values, although the FA itself might never find the best parameter as the FA is a stochastic method. The simulation results show that the system without a controller have high oscillation levels that continue for quite a long time. This is not good for the rotor of the generator and also not advantageous for the prime mover: the generator could go out of synchronization, while the prime mover (turbine) could break. That is why the inclusion of an additional controller is needed. In this work that fact is clearly shown.

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