Optimal Location and Sizing of Static Var Compensator (SVC) by Particle Swarm Optimization (PSO) Technique for Voltage Stability Enhancement and Power Loss Minimization

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ABSTRACT: In recent years, the transmission lines are operated under the heavily stressed condition, hence there is a risk of consequent voltage instability in the power network. There is a multi-functional control device which can be effectively control the load flow distribution and the power transfer capability is the flexible alternating current transmission system (FACTS) device. The facts device performance is depend upon its location and parameter setting. In this paper the optimal location and optimal sizing of static var compensator (SVC) is studied on the basis of particle swarm optimization (PSO) technique to minimize load voltage magnitude deviations and network losses using particle swarm optimization have been presented. Particle swarm optimization (PSO) is one of the artificial intelligent search approaches which have the potential to solve such problems. For this study, static var compensator (SVC) is chosen as the compensator device. Validation through the implementation on the IEEE 14 bus system shows that the PSO is found feasible to achieve the task.

Keywords:

FACTS Device, optimal location and sizing, Particle swarm optimization (PSO), Voltage deviation, power loss, Static var compensator (SVC) etc...

1. INTRODUCTION:

The power network is more difficult to operate and more insecure due to the ever increasing demand for the electrical power.

On the other hand, Flexible AC transmission system (FACTS) device, which can provide direct and flexible control of power transfer and are very helpful in the operation of power network. The power system performance and the power system stability can be enhanced by using FACTS device. [1] Static var compensator (SVC) is one of the most effective measure device for enhancing the power stability and power transfer capability of transmission network, for this the SVC should be properly installed in the

system with appropriate parameter setting. The some factors considering for optimal installation and the optimal parameter of SVC, Which are the Stability margin improvement, power loss reduction, power blackout prevention and the power transmission capacity enhancement.

In last 20 years, there are algorithm have been developed for optimal power flow incorporating with SVC device and for the optimal placement of SVC; such as are Newtons-Raphson load flow algorithm, Genetic algorithm and the Particle swarm optimization technique for optimal location of the FACTS device. [2], [3] & [4]. It is an actual and important subject to appropriately select the suitable location of the FACTS device installation at the viewpoint of the voltage stability enhancement and power loss minimization. The world wide researchers in the power system have retained the interest in this problem. The various method and criteria were proposed and used to optimal allocation of FACTS devices in power network. [5]

In this paper the application of particle swarm optimization (PSO) for the optimal location and optimal sizing of the SVC with consideration of active power loss reduction and voltage deviation minimization in the power system is highlighted.

2. PROBLEM FORMULATION:

For minimizing the load voltage magnitude deviation and loss of power system the determination of the optimal location and the optimal parameter setting of the SVC in the power network is the main objective. For this the performance index is selected:

$$\label{eq:min} \begin{split} Min \ F &= F_1 + F_2 = P_{loss} + VD \end{split} \tag{1} \end{split}$$
 Where,

Ploss = network real power loss
VD = voltage deviation
F₁=P_{loss}=
$$\sum_{k=1}^{nl} gk[V_i^2 + V_j^2 - 2V_iV_j\cos\theta_{ij}]$$
 (2)
F₂ = VD = $\sum_{k=1}^{nl} N_{PQ} + V_i - V_{ref.} + \frac{1}{N_{PQ}}$ (3)

Subjected to this there is some equality and inequality constraints are as follows:

3.1 Equality constraints:

$$P_{\text{Gi}} - P_{\text{Di}} - V_i \sum_{j=1}^{Ng} V_j (\text{Gij Cos} (\theta_i - \theta_j) + B_{ij} \text{Sin} (\theta_i - \theta_j)) = 0$$

$$i = 1, 2, ---- N_B$$
 ----- (4)

 $Q_{Gi} - Q_{Di} - V_i \sum_{j=1}^{Ng} V_j (G_{ij} \operatorname{Sin} \theta_{ij} - B_{ij} \operatorname{Cos} \theta_{ij}) = 0$

$$i = 1, 2, ---- N_B$$
 ---- (5)

3.2 Inequality constraints:

$$P_{gi}^{\min} \leq P_{gi} \leq P_{gi}^{\max}$$
(6)

$$Q_{gi}^{\min} \leq Q_{gi} \leq Q_{gi}^{\max}$$
(7)

$$V_{i}^{\min} \leq V_{i} \leq V_{i}^{\max}$$
(8)

$$X_{svc}^{\min} \leq X_{svc} \leq X_{svc}^{\max}$$
(9)

where,

F is the objective function.

 P_{gi} is the acrive power generation at bus i. Q_{gi} is the active power load at bus i. P_{di} is the active power load at i.

 Q_{di} is the reactive power load at bus i.

 V_i is the voltage magnitude at bus i.

 V_i is the voltage magnitude at bus j.

 V_{ref} is the reference voltage magnitude

Gij, Bij are mutual conductance and susceptance between bus i and bus j

 X_{svc} is the reactance of SVC

 θ_{ij} is voltage angle difference between bus i and j N_B is total number of buses excluding slack bus N_{PO} is number of PQ buses

3. STATIC VAR COMPENSATOR

Static var compensator (SVC) is a simplest, low cost FACTS device which is connected in shunt at load buses as a VAR generator or absorber, whose output is adjusted to enhance capacitive or inductive current so as to control specific parameters of electric power system, typically a bus voltage. The performance characteristics and construction are details in [12]. SVC can be used as susceptance model to adjust the susceptance of the transmission line and as VAR compensator to control the reactive power of the buses. for the connection of steady state model of SVC in a particular load bus ' i ' then the exchanged reactive power at that bus is $Q_i = Q_{SVC}$. The equivalent circuit of the SVC is shown in figure 1.

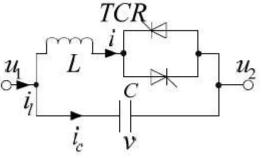


Fig 1: Equivalent circuit of SVC

3. METHODOLOGY FOR OPTIMAL LOCATION AND SIZING OF SVC. 3.1PARTICAL SWARM OPTIMIZATION (PSO):

Mr. Kennedy and Mr. Eberhart first introduced the PSO in the year of 1995.[6] PSO has its roots in artificial life and social psychology as well as in engineering and computer science. It utilizes a population of individuals, called particles, which fly through the problem hyperspace with some given initial velocities. In each iteration the velocities of the particles are stochastically adjusted considering the historical best position of the particles and their neighborhood best position; where these positions are determined according to some predefined fitness function. Then, the movement of each particle naturally evolves to an optimal or at least nearoptimal solution. In PSO algorithm, the particles fly in a multidimensional search space. During the flight each particle adjust its position according its own experience, and the experience of the neighboring particles, making use of the best position encountered by itself and its neighbors as shown in fig.2. as follows.

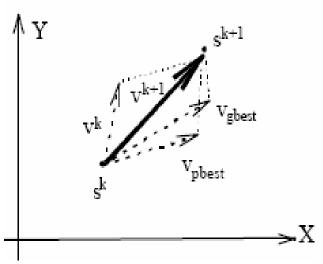


Fig.2: The swarm direction of a particle is defined by the set of particle neighboring the particle and its history experience.

 S^k : current searching point, S^{k+1} : modified searching point, V^k : current velocity, V^{k+1}: modified velocity, V_{pbest}: velocity based on Pbest, V_{gbest}: velocity based on Gbest.

Each particle keeps track of its coordinates in the problem space which are associated with the best solution (fitness) it has achieved so far. The fitness value is also stored. This value is called Pbest. When a particle takes all the population as its topological neighbors, the best value is a global best and is called Gbest. After finding the two best values, the particle updates its velocity and positions with following equations.

$$v_{id}^{(t+1)} = w \cdot v_{id}^{(t)} + c_1 * rand0 * (p_{id}^{best} - x_{id}^{(t)} + c_2 * rand1 * (g_d^{best} - x_{id}^{(t)})$$

$$x_{id}^{(t+1)} = x_{id}^{(t)} + v_{id}^{(t+1)}$$
(11)

$$p_{id}^{best} \text{ is the particle best of agent i.}$$

$$p_{id}^{best} \text{ is the particle best of agent i.}$$

$$p_{d}^{best} \text{ is the globle best.}$$
n is the number of particles in a group.

m is the number of members in a particle.

t is the pointer of iterations(generations) w is the inertia weight factor.

 c_1 and c_2 are two acceleration constants .

rand0 and rand1 are two uniform random values in the range [0,1].

 $v_i^{(t)}$ is the velocity of particle i at iteration t,

$$v_d^{min} \le v_{id}^{(0)} \le v_d^{mas}$$

 $x_i^{(t)}$ is the current position of particle i at iteration i.

3.1.1. PSO Parameters:

The performance of the PSO is greatly affected by its parameter values. Therefore, a way to find a suitable set of parameters has to be chosen. In this case, the selection of the PSO parameters follows the strategy of considering different values for each particular parameter and evaluating its effect on the PSO performance.

3.1.2. Number of Particles:

There is a trade-off between the number of particles and the number of iterations of the swarm and each particle fitness value has to be evaluated using a power flow solution at each iteration, thus the number of particles should not be large because computational effort could increase dramatically. Swarms of 5 and 25 particles are chosen as an appropriate population sizes.

3.1.3. Inertia Weight:

The inertia weight is linearly decreased. The purpose is to improve the speed of convergence of the results by reducing the inertia weight from an initial value of 0.9 to 0.1 in even steps over the maximum number of iterations as shown in equation

$$w_i = w_{max} - \frac{wmax - wmin}{itermax} x \text{ iter}$$
(12)

(10)

Wi = The inertia weight at iteration i.*iter* = the iteration number. max*iter* = The maximum number of iterations.

4. SIMULATION RESULTS

On the IEEE-14 bus test system (shown in Fig-3) the proposed PSO algorithm technique have been tested. The data for the mentioned system is taken from [7]. A MATBAB code for PSO algorithm was developed for simulation purposes.

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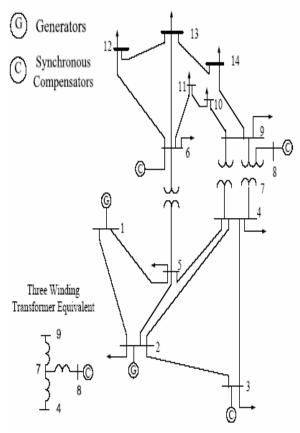


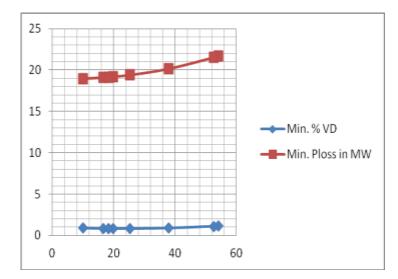
Fig-3: The IEEE-14 Bus system.

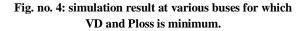
It is observed from the obtained results that the unique advantages of these particle swarm optimization (PSO) technique, on one hand, there capability of finding the global optimal solution to the optimal location, parameter settings and sizing of SVC problem, on the other hand, they don't suffer from the extant computational complexity and other limiting mathematical assumptions that the traditional optimization techniques suffer from.

The program is tested for the various buses of the IEEE 14 test bus system, for which the voltage deviation as well as power loss is minimum. The table no.1 and figure no.4 shows the simulation results for various buses.

		SVC]
		SIZE IN		
SR. NO.	BUS NO.	(MVA)	MIN. % VD	

1	4	54.1346	1.1108	21.6862
2	5	52.6567	1.0857	21.5179
3	9	37.9272	0.9063	20.1491
4	10	25.28	0.8505	19.398
5	11	16.5799	0.8616	19.0914
6	12	10	0.8952	18.9657
7	13	19.751	0.8531	19.1842
8	14	18.3438	0.8562	19.1404





5. CONCLUSION:

In this paper, the optimal location and optimal sizing of SVC device is find out to minimize voltage deviation and the active power losses in the power system network using particle swarm optimization (PSO) technique.

With the above proposed algorithm technique, it is possible for utility to place the TCSC device in the transmission grid such that proper power planning can be achieved with minimum system losses. The result obtained from the IEEE-14 bus system test, the power system shows that the PSO algorithm can

easily find out the optimal location and the best **MIN** all size of the static var compensator (SVC) for **PLOSS** the voltage deviation and power loss is **IN** nimum. And It is clearly seen that the bus No. 12 (**MSWi**) or e feasible and even economical at which the percentage voltage deviation as well as power loss is minimum.

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