ATC using FACTS Devices in Deregulated Power Systems

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ABSTRACT

A Repeated Power Flow with FACTS devices including ATC is used to evaluate the feasible ATC value within real and reactive power generation limits, line thermal limits, voltage limits and FACTS operation limits. An IEEE-30 bus system is used to demonstrate the effectiveness of the algorithm as an optimization tool to enhance ATC. A Genetic Algorithm technique is used for validation purposes. Introduction of FACTS devices in a right combination of location and parameters could enhance ATC and Ant Colony optimization can be efficiently used for this kind of nonlinear integer optimization.

Keywords: RPF, FACTS, Available Transfer Capability, Ga, ACO, OPF

Introduction

In recent years, power system operation faces new challenges due to deregulation and restructuring of the electricity markets. The old system known as monopoly based are substituted by a competitive marketplace. Hence the new structures of power system become more complex. These new structures have to deal with problem raised by the difficulties in building new transmission lines and the significant increase in power transactions associated to competitive electricity markets. Thus a large interconnected system has been built in order to be able to obtain a high operational efficiency and network security. In this situation, one of the possible solutions to improve the system operation is the use of flexible ac transmission technologies (FACTS).

The implementation of the FACTS devices extends the possibility that current through a line can be controlled at a reasonable cost, enabling large potential of increasing the capacity of existing lines, and use of one of the FACTS devices to enable corresponding power to flow through such lines under normal and contingency conditions.

Several studies [5, 49, 79, and130] have found that FACTS technology not only provides solutions for efficiently increasing transmission system capacity but also increases ATC, relieve congestion, improve reliability and enhances operation and control. However, it is hard to determine the optimal allocation and parameters of FACTS devices due to the complicated combinatorial optimization. Thus, attention is paid in this current work to study a technique to optimally allocate the devices to enhance ATC. The task of calculating ATC is one of main concerns in power system operation and planning. ATC is determined as a function of increase in power transfers between different systems through prescribed interfaces. Various OPF methods are detailed discussed in this paper.

In [2], Evolutionary Programming was used to determine the optimal allocation of FACTS for maximizing TTC between sources and sink area. Four types of FACTS are included in the studies; TCSC, UPFC, TCPS and SVC. The inequality constraints are power generation limit, voltage limit, line flow limit and facts operation limit. Three important aspects of FACTS to be considered throughout the optimization were the types of FACTS controller used, rated value and its location. The algorithm demonstrates the effectiveness of FACTS to improve system load ability. It did not compare the proposed method with other optimization methods to show the robustness of the proposed methods. In[136] presented a Parallel Tabu Search (PTS) based method for determining optimal allocation of FACTS devices in competitive power systems. Available Transfer Capability (ATC) was maximized with the FACTS devices. UPFC was modeled and concept of incremental load rate was used. The proposed method was compared with Simulated Annealing, GA and Tabu Search methods. It is 1.95 and 2.68 times
faster than TS and GA respectively. It is not affected by the initial conditions and gave higher quality solutions.

In PSO technique is used to find optimal location of FACTS to achieve maximum system load ability with minimum cost of installations of FACTS. The combinations of multi-type of FACTS were considered with line thermal limit and bus voltage limit as their inequality constraints. While in only STATCOM has been used to perform the effectiveness of PSO in allocating the devices. In this paper, the STATCOM is placed to improve voltage profile, minimize power system total losses and maximize system load ability with respect to the size of it. Furthermore, no comparison has been made in showing the robustness of the method with other existing methods.

In this paper, novel method using Ant Colony Optimization technique is proposed to determine the optimal allocation of FACTS devices for maximizing the Available Transfer Capability (ATC) of power transactions between source and sink areas in the deregulated power system. The algorithm simultaneously searches the FACTS location, FACTS parameters and FACTS types. Two types of FACTS are simulated in this study namely Thyristor Controlled Series Compensator (TCSC) and Static Var Compensator (SVC). The FACTS devices to enable corresponding power to flow through such lines under normal and contingency conditions. Several studies have found that FACTS technology not only provides solutions for efficiently increasing transmission system capacity but also increases ATC, relieve congestion, improve reliability and enhances operation and control. However, it is hard to determine the optimal allocation and parameters of FACTS devices due to the complicated combinatorial optimization. Thus, attention is paid in this current work to study a technique to optimally allocate the devices to enhance ATC. The task of calculating ATC is one of main concerns in power system operation and planning. ATC is determined as a function of increase in power transfers between different systems through prescribed interfaces.

Mathematical Model of FACTS Devices

TCSC: The TCSC changes the line reactance. It can be inductive or capacitive compensation respectively by modifying the line reactance. The reactance of TCSC is adjusted directly based on the reactance of the transmission line.

\[ X_{ij} = X_{\text{line}} + r \cdot X_{\text{TCSC}} \]

Where \( X_{\text{line}} \) is the reactance of the transmission line, \( r \cdot X_{\text{TCSC}} \) represents the reactance contributed by TCSC and \( r \) represents the degree of compensation of TCSC. The working range of TCSC (\( X_{\text{MIN}} \sim X_{\text{MAX}} \)) is set between -0.7 Xline and 0.2 Xline.

\[
P_{ij}^c = V_i^2 G_y - V_i V_j (G_y \cos \delta_y + B_y \sin \delta_y) \\
Q_{ij}^c = -V_i^2 (B_y + B_{\text{sh}}) - V_j V_j [G_y \sin \delta_y - B_y \cos \delta_y] \\
P_{ji}^c = V_j^2 G_y - V_i V_j (G_y \cos \delta_y - B_y \sin \delta_y) \\
Q_{ji}^c = -V_j^2 (B_y + B_{\text{sh}}) + V_i V_j [G_y \sin \delta_y + B_y \cos \delta_y]
\]
The change in the line flow due to series capacitance can be represented as a line without series capacitance with power injected at the receiving and sending ends of the line as shown in Fig. 3. The real power injections at bus-i \( (P_i) \) and bus-j \( (P_j) \) can be expressed, using equations

\[
P_{i} = V_{i}^{2}G_{i} + V_{j}^{2}G_{j} - 2 V_{j} V_{j} G_{j} \cos \delta_{j} \quad (5)
\]

\[
P_{j} = -V_{i}^{2}G_{i} - V_{j}^{2}G_{j} \cos \delta_{j} + \Delta B_{j} \cos \delta_{j} \quad (6)
\]

\[
Q_{i} = -V_{i}^{2}B_{i} + V_{j}^{2}B_{j} \cos \delta_{j} \quad (7)
\]

\[
Q_{j} = -V_{i}^{2}B_{i} - V_{j}^{2}B_{j} \cos \delta_{j} \quad (8)
\]

\[
P_{j} = 0; \quad Q_{j} = -|V_j|^2 B_{SVC} \quad (12)
\]

Also power mismatch are given as

\[
\begin{bmatrix}
\Delta P_j \\
\Delta Q_j
\end{bmatrix} =
\begin{bmatrix}
0 & 0 \\
0 & \frac{\partial Q_j}{\partial B_{SVC}}
\end{bmatrix}
\begin{bmatrix}
\Delta \delta_j \\
\Delta B_{SVC}
\end{bmatrix} \quad (13)
\]

At the end of iteration P, the variable shunt susceptance is corrected as

\[
B_{SVC}^{(p+1)} = B_{SVC}^{(p)} + \Delta B_{SVC}^{(p)} \quad (14)
\]

This may further be written as

\[
P_{SVC}^{(p+1)} = B_{SVC}^{(p)} + (\Delta B_{SVC} / B_{SVC}^{(p)})^{(p+1)} B_{SVC}^{(p)} \quad (15)
\]

The changing the susceptance represents the total SVC susceptance necessary to maintain the nodal voltage magnitude at the specified value. SVC compensation may also be computed in terms of thyristor
firing angle. However, the additional calculation requires an iteration solution as the SVC susceptance and thyristor firing angle are non-linearly related.

The steady-state susceptance of SVC can be obtained from the relation,

\[ B_{SVC} = B_t - B_{tcx} = \frac{1}{X_c} \left[ X_j - X_c \right] \frac{1}{\Pi} \left[ 2(\Pi - \alpha) + \sin(2\alpha) \right] \]  

(16)

Where \( X_c = \alpha L \) and \( X_c = \frac{1}{\alpha} \).

Since, \( Q_j = -V_j^2 \cdot B_{SVC} \), we can write

\[ Q_j = -\frac{V_j^2}{X_c} \left[ X_j - X_c \right] \frac{1}{\Pi} \left[ 2(\Pi - \alpha) + \sin(2\alpha) \right] \]  

(Assuming that \( Q_j \) is the jth bus reactive power injected due to SVC installation at the jth bus). The linearized SVC equation is then given by,

\[
\begin{bmatrix}
\Delta P_j \\
\Delta Q_j
\end{bmatrix}^{(p+1)} =
\begin{bmatrix}
0 & 0 \\
0 & \frac{2V_j^2}{\Pi X_c} [-\cos(2\alpha) - 1]
\end{bmatrix}^{(p+1)}
\begin{bmatrix}
\Delta \delta_j \\
\Delta \alpha
\end{bmatrix}^{(p+1)}
\]  

(18)

\( \alpha \) is the firing angle of SVC. Therefore, at the end of iteration, the variable firing angle \( \alpha \) is updated by the equation by the equation.

\[ \alpha^{(p+1)} = \alpha^{(p)} + \Delta \alpha^{(p+1)} \]  

(19)

The bus with which the SVC is connected (i.e. the jth bus) becomes a voltage-controlled bus where the voltage magnitude and active and reactive powers are specified along with either the SVC firing angle (\( \alpha \)) or the SVC equivalent susceptance (\( B_{SVC} \)). If \( \alpha \) or \( B_{SVC} \) is within limits, the specified voltage magnitude is attained and the bus remains as voltage. Controlled (PV) bus. However, is \( \alpha \) or \( B_{SVC} \) go beyond limits, then these variables are fixed at the violated limits and the bus becomes a pure load (PQ) bus.

**Problem Formulation**

The main objectives of this work is to determine the optimal location and the optimal parameter setting of the FACTS device in the power network to maximize the available power that can be transferred from a specific set of generators in a source area to loads in sink area, subject to real and reactive power generation limits, voltage limits, line thermal limits and FACTS devices operation limits.

The RPF with FACTS devices is used to evaluate the feasible ATC value of the power transactions. RPF is based on generalized search method, where successive power flow solution is performed until it reaches the maximum transfer capability. The RPF enables transfers by increasing complex load in the sink area and injected real power generation in the source area in incremental steps until any violation incurred.

Mathematically, the calculation of TTC with TCSC or SVC is installed in the network can be defined as:

\[ \text{Max } F(x) = O_{ATC} \]  

(20) Subject to:
\[ E(f,g) = 0 \]
\[ PG_i^{\text{min}} \leq PG_i \leq PG_i^{\text{max}} \tag{21} \]
\[ QG_i^{\text{min}} \leq QG_i \leq QG_i^{\text{max}} \tag{22} \]
\[ X_{TCSC}^{\text{min}} \leq X_i \leq X_{TCSC}^{\text{max}} \tag{23} \]
\[ Q_{SVC}^{\text{min}} \leq Q_{SVC} \leq Q_{SVC}^{\text{max}} \tag{25} \]

Where,
\( O_{ATC} \): ATC maximized function
\( E(f,g) \): Conventional power flow equation
\( PG_i, QG_i \): Real and reactive power generation at bus \( i \)
\( f \): The variables of FACTS devices
\( g \): Operating state of the power system
\( V_i^{\text{min}}, V_i^{\text{max}} \): Lower and upper limit of voltage magnitude at bus \( i \)
\( S_i^{\text{min}}, S_i^{\text{max}} \): Thermal limit of line \( i \)
\( Q_{SVC}, Q_{SVC}^{\text{max}} \): reactive power limit
\( X_{TCSC}, X_{TCSC}^{\text{max}} \): TCSC line reactance limit

For calculating TTC and ATC, the injected \( P_{Gi} \) at source area, and \( P_{Di}, Q_{Di} \) at sink area are increased in function of \( \lambda \) in which;
\[ P_{Gi} = P_{Gi}^0 (1 + \lambda K_{Gi}) \tag{26} \]
\[ P_{Di} = P_{Di}^0 (1 + \lambda K_{Di}) \tag{27} \]
\[ Q_{Di} = Q_{Di}^0 (1 + \lambda K_{Di}) \tag{28} \]

Where \( P_{Gi}^0, P_{Di}^0, Q_{Di}^0 \) are the base case injection at bus-i and \( K_{Gi}, K_{Di} \) are the constant used to specify the rate of changes in load as \( \lambda \) varies. In order to maintain a zero balance, the incremental power losses resulting from increases in transfer power are allocated by a given formula. At PV buses, the reactive power is maintain at the base case value. However, in sink area, the reactive power demand \( Q_{Di} \) is incremented accordingly to real power in order to keep a constant value of power factor. The rate of \( \lambda \) change from \( \lambda = 0 \) corresponds to no transfer (base case) to \( \lambda = \lambda_{\text{max}} \) corresponds to the largest value of transfer power that causes no limit violations. \( PD(\lambda_{\text{max}}) \) is the sum of load in sink area when \( \lambda = \lambda_{\text{max}} \) while \( PoD \) refers to the sum of load when \( \lambda = 0 \). Therefore, the sum of real power loads in sink area at the maximum power transaction in (normal or contingency case) represents the TTC value.

\[ TTC = \sum_{i=1}^{Nd_{SNE}} P_{Di}(\lambda_{\text{max}}) - \sum_{i=1}^{Nd_{SNE}} P_{Di}^0 \tag{29} \]

**Proposed Methodology of ANT Colony Optimization**

The general algorithm ACO operators for the implementation of OPF. In the process involves initialization, state transition rule, local updating rule, fitness evaluation and global updating rule.

**Step 1**: Initialization; during the initialization process \( n, m, t_{\text{max}}, d_{\text{max}}, \beta, \rho, \alpha \) and \( g \) are specified.

Where
\( n \): no. of nodes
\( m \): no. of ants
$t_{max}$: maximum iteration  
$d_{ma}$: maximum distance for every ants tour  
$\beta$: parameter, which determines the relative importance of pheromone versus instance ($\beta > 0$)  
$\rho$: heuristically defined coefficient (0 < $\rho$ < 1)  
$\alpha$: pheromone decay parameter (0 < $\alpha$ < 1)  
$q_0$: parameter of the algorithm (0 < $q_0$ < 1)  
$\tau_0$: initial pheromone level 

Every parameter requires to be set for limiting the search range in order to avoid large computation time.  
$d_{max}$ can be calculated using the following formula:

$$d_{\text{max}} = \max\left[\sum_{i=1}^{n} d_i\right]$$

(30)

Where:

- $r$: current node  
- $u$: unvisited node  
- $d$: distance between two nodes

**Step 2:** Generate first node randomly; the first node will be selected by generating a random number according to a uniform distribution, ranging from 1 to $n$.

**Step 3:** Apply state transition rule; in this step the ant located at node $r$ (current node) will choose the nodes $s$ (next node) based on the following rule.

$$s = \begin{cases} 
\arg\max_s \left\{ \tau(r,u)\eta(r,u)^{\beta} \right\}, & \text{if } q \leq q_0 (\text{exploration}) \\
\text{s, otherwise (biased exploration)} & 
\end{cases}$$

(31)

Where:

- $q$: random number uniformly distributed in [0…1]  
- $S$: random variable selected according to the probability distribution given in equation (31)  

The probability for an ant $k$ at node $r$ to choose the next node $s$, is calculated using the following equation

$$p_k(r,s) = \frac{\tau(r,s)^{\beta} \eta(r,u)^{\beta}}{\sum_{s \in J_{k(r)}} \tau(r,s)^{\beta} \eta(r,u)^{\beta}}, \text{ if } s \in J_{k(r)}, 0, \text{ otherwise}$$

(32)

Where:

- $\tau$: Pheromone  
- $J_{k(r)}$: set of nodes that remain to be visited by ant $k$ positioned on node (to make the solution feasible)  
- $\eta$: 1/$\delta$, is the inverse of the distance $\tau(r,s)$

Ants that have the highest fitness are chosen as “selected ants” ($m$ nodes) and path visited by them are chosen for neighborhood search.

**Step 4:** Apply local updating rule; while constructing a solution of UPFC optimization, ants visit edges and change their pheromone level by applying the local updating rule of equation 33

$$\tau(r,s) \leftarrow (1 - \rho)\tau(r,s) + \rho\Delta\tau(r,s)$$

(33)

Where:

- $\rho$: heuristically defined coefficient (0 < $\rho$ < 1)  
- $\Delta\tau(r,s) = \tau_0$
**Step 5:** Determine tuned parameters; two variables \((x_1, x_2)\) required to represent the UPFC parameters (i.e. UPFC voltage constant, \(V_s\) and angle, \(\phi_s\)) and are selected within the specified ranges from RSA method.

**Step 6:** Fitness evaluation; it is performed after all ants have completed their tours. In this step, the control variable is computed using the following equation:

\[
x = \frac{d}{d_{\text{max}}} \times x_{\text{max}}
\]

(34)

where:
- \(d\): distance for every ants tour
- \(x_{\text{max}}\): maximum \(x\)

The values of \(x\) will be assigned for UPFC parameters. The fitness is computed by performing an ac load flow program. This program is called repeatedly into the ACO main program for the whole process.

**Step 7:** Apply global updating rule; to simplify the problem, this step is applied to edges belonging to the best ant tour which give the best fitness among all ants. The pheromone level is updated by applying the global updating rule in equation 35

\[
\tau(r,s) \leftarrow (1-\alpha)\tau(r,s) + \alpha \Delta \tau(r,s)
\]

(35)

Where,

\[
\Delta(r,s) = \begin{cases} 
(L_{gb})^{-1}, & \text{if } (r,s) \text{ is global best tour} \\
0, & \text{otherwise}
\end{cases}
\]

(36)

\((L_{gb})\): The length of the globally best tour from the beginning of the trial

**Step 8:** End condition; the algorithms stop the iteration when a maximum number of iterations have been performed otherwise, repeat step 3. Every tour that was visited by ants should be evaluated. If a better path is discovered in the process, it will be kept for next reference. The best path selected between all iterations engages the optimal scheduling solution to UPFC optimal parameters problem.
Simulation Results and Discussion

To illustrate the applicability of the proposed method to allocate the FACTS devices in large scale system, a modified IEEE 30 bus system is used. Figure.3 shows the sample test system. The system is divided into three areas with two generators in each area. And FACTS device allocated to these area for calculation of ATC. In this work, ant colony optimization based algorithm has been proposed for ATC calculation. To verify and validity of the results of the proposed algorithm with a Genetic Algorithm method is prepared as shown in Table I. Two FACTS devices, TCSC and SVC are installed simultaneously in the system in order to enhance ATC. The ACO algorithm and GA could allocate the devices in certain combination of types, rated value and location. The results of the simulation are shown in Table I. The Six cases are studied where the power is transferred from 1 to 2, 1 to 3, 2 to 1, 2 to 3, 3 to 1 and 3 to 2. Using the RPF method without FACTS devices, the ATC to transfer the power from Area 1 to area 2 is 50.86 MW and the limiting condition as expected is the generation upper limit at bus 1, PG1, if further transfers take place. When FACT Devices are installed using GA, the ATC to transfer power from Area 1 to Area 2 is increased 52.21 MW and the limiting condition is the upper limit of the same generator, PG1. Using GA, two SVCs need to be placed at line 12-15 with 35.74MVAR and at line 9 to 10 with 59.74MVAR respectively. However, using ACO, the ATC is increased to 52.33MW. The combination changes to one TCSC with rated value of -28.674% Xline at location 6 to 9 and one SVC with size of 74.24MVAR at location 10 to 20.

Results for IEEE-30 bus system

<table>
<thead>
<tr>
<th>Transfer From Area</th>
<th>From To</th>
<th>ATC without FACTS (MW)</th>
<th>Limit Condition</th>
<th>Allocation Technique</th>
<th>ATC with FACTS Devices</th>
<th>Facts Type</th>
<th>Size (rated value)</th>
<th>Location</th>
<th>ATC (MW)</th>
<th>Limit Condition</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>2</td>
<td>50.86</td>
<td>PG1</td>
<td>GA</td>
<td>51.21</td>
<td>SVC</td>
<td>35.74MVAR 95.74MVAR</td>
<td>LINE 12-15</td>
<td></td>
<td>PG1</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>TCSC</td>
<td>-28.64% Xline 74.24MVAR</td>
<td>LINE 6-9</td>
<td>51.33</td>
<td>PG1</td>
</tr>
<tr>
<td>1</td>
<td>3</td>
<td>48.31</td>
<td>PG22</td>
<td>GA</td>
<td>50.86</td>
<td>SVC</td>
<td>29.68MVAR 34.98MVAR</td>
<td>LINE 8-28</td>
<td></td>
<td>PG1</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>TCSC</td>
<td>-28.71% Xline 65.94MVAR</td>
<td>LINE 6-9</td>
<td>50.93</td>
<td>PG1</td>
</tr>
<tr>
<td>2</td>
<td>1</td>
<td>9.07</td>
<td>PG23</td>
<td>GA</td>
<td>9.43</td>
<td>TCSC</td>
<td>-50.29% Xline 38.3MVAR</td>
<td>LINE 6-9</td>
<td></td>
<td>PG23</td>
</tr>
</tbody>
</table>
### Fig 4:
The convergences characteristics of the fitness of the ACO and GA

It clearly shown that ACO could effectively allocate the A device as GA does. Moreover ACO could reach the optimum solution 4 times faster than GA for nearly the same results. For all of the cases, ACO could find better location to install the devices to enhance ATC slightly better than GA. Figure 4 shows the comparison of rapid convergences characteristics of the proposed method and GA. For other transfer between different control areas, with and without FACT devices, test results indicate that optimally placed FACTS devices can significantly enhance ATC of the system as shown in Table I

### Conclusion

New optimization algorithm to optimally allocate FACTS devices to enhance ATC. The simulation results tremendously prove that the proposed algorithm has remarkable robustness, in maximizing the
Furthermore, the result also shows the effectiveness of the new approach in simultaneously optimized the FACTS location, rated values and FACTS types. It is a practical method for the allocation of FACTS devices in large power systems. The Ant colony algorithm converged to the maximum without becoming trapped at local optima. The algorithm generally outperformed the GA techniques that were compared with it in terms of speed of optimization and accuracy of the results obtained. The main advantage of ACO is that it does not require external parameters such as cross over rate and mutation rate etc, as in case of genetic algorithms these are hard to determine in prior. The other advantage is that the global search ability in the algorithm is implemented by introducing neighborhood source production mechanism which is a similar to mutation process. However, one of the drawbacks of the algorithm is the number of tunable parameters used. Nevertheless, it is possible to set the parameter values by conducting a small number of trials. The program has been developed in Object Oriented Programming (OOP) advanced Java language and the has been tested on latest Core 2 duo processor.

References


