Optimal Utilization of Transmission Capacity to Reduce Congestion with Distributed FACTS

Huijuan Li, Student Member, IEEE, Fangxing Li, Senior Member, IEEE, Pei Zhang, Senior Member, IEEE, and Xiayang Zhao

Abstract—The Flexible AC Transmission System (FACTS) device has been applied to enhance the controllability of power systems. A new generation of FACTS device called Distributed FACTS such as distributed series impedance or distributed static series compensator has recently received increasing interests for power system control and are expected to be broadly deployed. When technology of D-FACTS is being further advanced, there is an increasing interest to find the optimal locations of these devices. This paper presents a detailed formulation and algorithm to find the best location and size of D-FACTS to achieve the optimal utilization of transmission capacity to mitigate congestion. Simulation results are presented with the PJM 5-bus system.

Index Terms—D-FACTS; distributed series impedance; distributed static series compensator; optimal placement; OPF.

I. INTRODUCTION

The Flexible AC Transmission System (FACTS) device has been applied to enhance the controllability of power systems. However, the high cost and reliability concern restrict the wide deployment of the conventional FACTS devices. The emerging low-cost distributed FACTS (D-FACTS) supplies an alternative solution for the flexible control of the power systems. The typical D-FACTS device like distributed series impedance (DSI) or distributed static series compensator (DSSC), which can be easily deployed on the existing power lines to change the impedance of line, receives increasing interests and are expected to be broadly deployed to mitigate transmission stress [1]. When technology of D-FACTS is being further advanced, there is an increasing interest to find the optimal locations of these devices.

The techniques to find the good location and size of FACTS can be classified as sensitivity-based or optimization-based. The sensitivity-based approach typically use some indices based on sensitivity of a technical measure, such as a modal controllability index to damp inter-area oscillations [2] or the real power flow performance index with respect to control parameters [3-5]. The optimization-based approaches are more systematic because interaction of different FACTS device can be simultaneously considered. The objective functions in optimization model are typically represented by minimization of generation cost [6, 7], minimization of losses [3], maximization of Available Transfer Capability or Total Transfer Capability [5, 8, 9]. With the optimization model, different solution techniques can be applied. For instance, References [7] apply linearized network model for DC power flow with mixed integer linear programming technique. Reference [10] uses genetic algorithm to determine the location as well as the parameters of TCSC simultaneously. Reference [9] applies Particle Swarm Optimization (PSO) to achieve maximum system loadability with minimum cost of installation of FACTS devices.

Different from the previous optimization model that optimizes the installation of conventional FACTS, this work optimizes the new, simply structured, low-cost D-FACTS devices like DSI or DSSC. And, the technical goal of installing these D-FACTS is to mitigate transmission stress. So, we use the system-wide line flow margin as the goal to formulate our objective function. Also, linearized DC model is applied to easily model power flow as constraints since a large number of independent, line-flow-affecting variables must be considered due to the expected large-scale deployment of D-FACTS. In addition, compensators are modeled with susceptance to be easily included in the optimization. These are the unique features of this work.

This paper is organized as follows. Section II gives the generic problem formulation. Section III presents the solution techniques using deepest descent algorithm, including the case with inequality constraints. Section IV presents test results, and Section V concludes the paper.

II. PROBLEM FORMULATION

This section will formulate the proposed optimization model. In the discussion below, it is assumed that generation dispatch and load are known from typical or historical data. The problem here is for the transmission owners or operators to identify the best location and size of series compensators. This can be interpreted as solving this problem: Given one of more typical patterns of generation dispatch, where and how much should the series compensators be placed? Ideally, more than one typical patterns (such as peak load, shoulder load, or valley load) should be addressed. However, this paper considers only a single pattern (like peak load case) for easy
illustration. A combination of multiple patterns may be addressed in the future work.

The power flow is modeled with linearized DC model, which stems from the fast decoupled power flow and assumes voltage is 1.0 across the system and there is no real power loss in power lines. The reason for using linearized DC model is that it is robust and easy to solve, while using an AC-based optimization model may suffer from convergence problem and be sensitive to initial input. This is especially important when considering many distributed FACTS devices to be deployed in a large power system.

With the linearized DC power flow model, line flow at the kth line (connecting buses i and j) can be calculated using the following equation:

\[ F_k = F_{ij} = \frac{\delta_i - \delta_j}{x_{ij}} = b_{ij} \cdot (\delta_i - \delta_j) \]  

(1)

where

- \( F_k \) is the line flow at Line k, which connects two buses, i and j;
- \( \delta_i, \delta_j \) is the voltage angles at Bus i and Bus j, respectively;
- \( x_{ij} \) is the reactance of line k;
- \( b_{ij} \) is the susceptance of the (ij) entry in \( Y_{bus} \), which is equal to \( \frac{1}{x_{ij}} \) in value.

The optimization can be written in an easy-to-understand formulation as follows:

\[ \text{Min} \sum_{k=1}^{M} \left( F_k^2 - F_{k,\text{max}}^2 \right)^2 \]  

(2)

Subject to:

\[ [B] \cdot \delta = P \]  

(3)

\[ F_k^2 = F_{ij}^2 = \left( b_{ij} \cdot (\delta_i - \delta_j) \right)^2 \leq F_{k,\text{max}}^2 \]  

(4)

\[ b_{ij} = b_{ij,0} + b_{ij,c} \]  

(5)

\[ b_{ij,c}^{\text{min}} \leq b_{ij,c} \leq b_{ij,c}^{\text{max}} \]  

(6)

where

- \( N \) is the number of buses;
- \( M \) is the number of lines;
- \( F_k \) is the power flow through line k;
- \( F_{k,\text{max}} \) is the flow limit at line k;
- \([B]\) is the nodal admittance matrix;
- \( \delta \) is the vector of bus voltage angle, i.e., \([\delta_1, \delta_2, ..., \delta_N]^T\);
- \( P \) is the vector of net nodal injection, i.e., \([P_1, P_2, ..., P_N]^T\);
- \( b_{ij,0} \) is the susceptance of the original (uncompensated) line;
- \( b_{ij,c} \) is the equivalent susceptance of series compensator.

In the above formulation, the independent variables (unknowns) are line susceptance, \( b_{ij} \), which is the reciprocal of line reactance (or impedance in value since resistance is ignored). The susceptance is used instead of reactance to make the optimization model easy to solve. Since the uncompensated line has a fixed susceptance \( b_{ij,0} \), the independence variables are essentially \( b_{ij,c} \), i.e., the susceptance of series compensators. As shown in [1], the distributed series compensators consist of a set of switchable inductors or capacitors. So, it has minimum and maximum bounds as modeled in (6). It should be noted that here \( b_{ij} \) is represented as a continuous variable even though it is discrete in theory due to the possible step change characteristics of series compensators. This should be reasonable, especially at the planning stage. Since the series compensator may consist of many small inductors and capacitors, it can be roughly considered as a continuous variable. Also, the linearized DC model is suitable here to easily model line flows, because many independent, line-flow-affected variables \( (b_{ij,c}) \) must be considered due to the expected large-scale deployment of D-FACTS.

The goal in this formulation is to achieve the lowest congestion with installation of D-FACTS. To do so, an expression of absolute values may be used, but that will be more difficult to solve. Hence, Eq. (2) uses the squared sum of squared capacity margins, which is to address the possible direction problem, of all transmission lines. The choice of \( b_{ij,c} \) impacts the voltage angles, which determine the line flows. The solution shall give the best combination of \( b_{ij,c} \) to achieve the optimization objective function.

In equation (3), the nodal injection is obtained as the product of a row in \([B]\) and the vector \( \delta \). Different from a conventional OPF formulation for generation dispatch where \([B]\) is known and \( P \) is unknown, this model considers \( P \) is known and \([B]\) is unknown. The reason is that the generation dispatch is assumed to be known as a typical case like the peak-load case, as previously mentioned. Here, \([B]\) is unknown since the susceptance of each compensator is an independent unknown variable. Hence, (2)-(6) can be rewritten as follows, after putting (4) into (2) and combining (5) and (6).

\[ \text{Min} \sum_{k=1}^{M} \left( b_{ij} \cdot (\delta_i - \delta_j) \right)^2 - \left( F_{k,\text{max}}^2 \right)^2 \]  

(7)

Subject to:

\[ \sum_{j=1}^{N} b_{ij} \cdot (\delta_i - \delta_j) = P_i \]  

for \( i = 1, 2, ..., N \)  

(8)

\[ F_k^2 = F_{ij}^2 = \left( b_{ij} \cdot (\delta_i - \delta_j) \right)^2 \leq F_{k,\text{max}}^2 \]  

(9)

\[ b_{ij,0} + b_{ij,c}^{\text{min}} \leq b_{ij} \leq b_{ij,0} + b_{ij,c}^{\text{max}} \]  

(10)
III. SOLUTIONS WITH STEEEP DESCENT ALGORITHM

A. Solution techniques – unconstrained case

To solve the above formulation in (7-10), we first formulate the solution procedures with the steepest descent algorithm [11] by ignoring (9) and (10), i.e., considering unconstrained case only. To do so, it is necessary to formulate the augmented cost function as follows:

\[ C^* = f(b, \delta) - \lambda^T \times g(b, \delta) \]  
\[ f(b, \delta) = \sum_{k=1}^{M} \left( b_{ij} \cdot (\delta_i - \delta_j)^2 - (F_{ij}^{\max})^2 \right) \]  
\[ g(b, \delta) = \left( \sum_{j=1}^{N} b_{ij} \right) \delta_i - b_{ij} \cdot \delta_i - b_{i2} \cdot \delta_2 - \ldots - b_{iN} \cdot \delta_N - P_i \]

where \( b = \) the vector of all M lines (vector of \( b_{ij} \), not a row or column in \( |B| \)), which are the dependent variables; 
\( \delta = \) the vector of all N bus angles (vector of \( \delta_i \)), which are dependent state variables; 
\( \lambda = \) the vector of all N Lagrangian multipliers for \( g(b, \delta) \).

If we apply steepest descent algorithm, we have

\[ \nabla C^* = \left[ \frac{\partial C^*}{\partial b} \right] = \left[ \frac{\partial f}{\partial b} \right] - \left[ \frac{\partial g}{\partial b} \right]^T \left[ \frac{\partial g}{\partial \delta} \right]^T \left[ \frac{\partial f}{\partial \delta} \right] \]  

In (14), we have

\[ \frac{\partial f}{\partial b_{ij}} = 4b_{ij} (\delta_i - \delta_j)^2 (b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2}) \]  
\[ \frac{\partial f}{\partial \delta_i} = \sum_{j=1}^{M} 4b_{ij}^2 (\delta_i - \delta_j) (b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2}) \]

where \( M_i \) = number of branches connected with bus \( i \).

Also in (14), we have

\[ \frac{\partial g}{\partial \delta} = \begin{bmatrix} \frac{\partial g_1}{\partial \delta_1} & \frac{\partial g_1}{\partial \delta_2} & \ldots & \frac{\partial g_1}{\partial \delta_M} \\ \frac{\partial g_2}{\partial \delta_1} & \frac{\partial g_2}{\partial \delta_2} & \ldots & \frac{\partial g_2}{\partial \delta_M} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial g_M}{\partial \delta_1} & \frac{\partial g_M}{\partial \delta_2} & \ldots & \frac{\partial g_M}{\partial \delta_M} \end{bmatrix} \]

\[ \frac{\partial g}{\partial b} = \begin{bmatrix} \frac{\partial g_1}{\partial b_{line-1}} & \frac{\partial g_1}{\partial b_{line-2}} & \ldots & \frac{\partial g_1}{\partial b_{line-M}} \\ \frac{\partial g_2}{\partial b_{line-1}} & \frac{\partial g_2}{\partial b_{line-2}} & \ldots & \frac{\partial g_2}{\partial b_{line-M}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial g_M}{\partial b_{line-1}} & \frac{\partial g_M}{\partial b_{line-2}} & \ldots & \frac{\partial g_M}{\partial b_{line-M}} \end{bmatrix} \]

Assume Line \( k \) connects Buses \( i \) and \( j \). Then, \( \frac{\partial g_k}{\partial b_{line-k}} \) in (18) is equal to \( 0 \) if \( k \neq i \) and \( k \neq j \); \( \delta_i - \delta_j \) if \( k = i \); and \( \delta_j - \delta_i \) if \( k = j \).

Then common iterative approach of steepest descent algorithm can be applied to solve the unconstrained case. The procedure is as follows:

1. Let \( k = 0 \), start from an initial vector \( b^0 = b^0 \).
2. Given \( b^k \), solve the DC power flow ( \( g(b, \delta) = 0 \) ) to obtain \( \delta \), the angle vector of the buses.
3. Calculate \( C^{k+1} \) and \( \nabla C^{k+1} \). If \( \left\| C^{k+1} - C^k \right\| \) is less than the predefined tolerance, stop and the optimal solution is found. Otherwise go to 4.
4. Update \( b^{k+1} = b^k - \gamma \cdot \nabla C^k \), here \( \gamma \) is user defined step size. Go to Step 2 to continue the next iteration.

B. Solution techniques – considering capacity limits constraints cases

The system has two inequality constraints: inequality constraints on independent variables as (10), which represent the capacity limit of DSSC, and inequality constraints on dependent variables as (9), which represent the thermal capacity limit of the power lines. For the former one, we can set the independent variables to the active limit when the limit is violated and update the other variables. For the latter constraints, a quadratic penalty function method is applied.

\[ p(h_i(b, \delta)) = s \cdot (h_i(b, \delta))^2 \]
\[ s = \begin{cases} 0, & \text{when } h_i(b, \delta) < -t \\ p, & \text{when } h_i(b, \delta) \geq -t \end{cases} \]
\[ h_i(b, \delta) = b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2} \]

where \( t \) is user defined tolerance from the limit; \( p \) is the predefined penalty parameter.

Therefore, the augmented cost function can be written as:

\[ C^* = \sum_{k=1}^{M} \left( b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2} \right)^2 + \sum_{i=1}^{L} s \cdot \left( b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2} \right)^2 - \sum_{i=1}^{L} \lambda \left( \sum_{j=1}^{N} b_{ij} \delta_i - b_{ij} \delta_i - b_{i2} \delta_2 - \ldots - b_{iN} \delta_N - P_i \right) \]

where \( L \) is the number of branches whose line flow is very close to or exceeds the limit.

Compared (22) with (11)-(13), it is apparent that only \( f(b, \delta) \) has been changed, while \( g(b, \delta) \) keeps unchanged. So, applying steepest descent algorithm, we have the same
\( \frac{\partial g}{\partial b} \) and \( \frac{\partial g}{\partial \delta} \), while only \( \frac{\partial f}{\partial b} \) and \( \frac{\partial f}{\partial \delta} \) need to be changed as follows:

\[
\frac{\partial f}{\partial b_{ij}} = 4(1 + s)b_{ij} (\delta_i - \delta_j)^2 (b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2}) \tag{23}
\]

\[
\frac{\partial f}{\partial \delta_i} = \sum_{j=1}^{M} 4b_{ij}^2 (\delta_i - \delta_j) (b_{ij}^2 (\delta_i - \delta_j)^2 - F_{ij}^{\max^2}) + \sum_{k}^{M} \sum_{s} b_{ik}^2 (\delta_i - \delta_k) (b_{ik}^2 (\delta_i - \delta_k)^2 - F_{ik}^{\max^2}) \tag{24}
\]

The iteration procedure is very similar to what is presented in Section III(A).

It should be noted that since this is based on the linearized DC power flow model, the line flow is impacted by line susceptance only. Hence, there could be multiple solutions, especially if shunt admittances are all ignored. For instance, we have a solution with all line susceptance to be 1.0 p.u. Then, if all line susceptances are doubled, then the voltage angles will be only 50% of the values in the original solution. Since line flow is the voltage angle difference multiplied by line admittance, the line flow will be unchanged because we have \( 2b_{ij} \times 0.5(\delta_i - \delta_j) = b_{ij} \times (\delta_i - \delta_j) \). Fortunately, if we start with all \( b_{ij} \) as the base case \( b_{ij} \) (no series compensation at all), the solution should be the one closest to this initial condition. Therefore, it is avoided that the optimization model gives a large deviation in line susceptance which means unnecessary over series compensation on many lines.

**IV. TEST RESULTS**

The tests are carried out on the PJM-5-bus system, which is shown in Fig. 1. The uncompensated system data and the line flow are shown in Table 1. Two cases are tested. In Case 1, the only binding limits are the distributed series compensators’ capacity limits. In Case 2, both the distributed series compensators’ capacity limits and line thermal capacity limits are binding. It should be noted that the line flow limits in Case 3 are carefully set to lead a few binding transmission limits for testing purpose. Results as well as the input data are shown in Table 2 and Table 3.

![Fig.1. The PJM-5-bus system.](image-url)
flow away from their individual limit as much as possible to have a good mitigation of the overall congestion problem.

In addition, the optimum solution for the susceptance of each line is in the reasonable region without unnecessary over compensation problem.

V. CONCLUSIONS

This research work presents an optimization model that can achieve the best mitigation of transmission congestion with D-FACTS (such as distributed series impedances or distributed static series compensators), which are expected to be deployed in large scales. The proposed model is shown to be effective. Future work may lie in modeling the accurate AC power flow constraints and a comparison with this work.

VI. REFERENCES


VII. BIOGRAPHIES

Huijuan Li (S’07) is presently a Ph.D. student in electrical engineering at The University of Tennessee. She received her B.S.E.E. and M.S.E.E. in electrical engineering from North China Electrical Power University, China in 1999 and 2002 respectively. She previously worked as a research engineer at Shanghai Siyuan Electrical Company in China on the field of ungrounded distribution systems.

Fangxing (Fran) Li (M’01, SM’05) received the Ph.D. degree from Virginia Tech in 2001. He has been an Assistant Professor at The University of Tennessee (UT), Knoxville, TN and an adjunct researcher at ORNL since August 2005. Prior to joining UT, he worked at ABB, Raleigh, NC, as a senior and then a principal engineer for four and a half years. His current interests include energy market, reactive power, and distributed energy resources. Dr. Li is a registered Professional Engineer in North Carolina.

Pei Zhang (M’00 SM’05) is the Program Manager overseeing Grid Operation and Planning area at Electric Power Research Institute (EPRI). He received his Ph. D. degree from Imperial College of Science, Technology and Medicine, University of London, United Kingdom. His current research interests include application of probabilistic method to system planning, power system stability and control, power system reliability and security, and AI application to power system.

Xiayang Zhao received the Dr. -Ing degree from Institute of Power System and Power Economics, RWTH Aachen University, Germany in 2007. She is currently a project leader in the Energy Sector of Siemens AG. Previously, she worked at Hohai University in Nanjing, China as a lecturer for seven years and in Ministry of Water Resources in Beijing, China as a consultant for a year and a half.