

Optimal Placement of Multiple Distributed Generation to Improve Voltage Stability as well as Minimize Loss in Distribution System

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Abstract: The aim of paper is to present a technique based on Particle Swarm Optimization (PSO) to identify the switching operation plan for feeder reconfiguration and optimum size value of DG simultaneously. The key purpose is to diminish the real and reactive power losses and improve the bus voltage profile in the system while satisfying all the distribution constraints. A method based on PSO algorithm to find out the minimum formation is presented and their impact on the network real power losses and voltage profiles are investigated. The results are obtained through MATLAB coding.

Keywords: Radial Distribution System, Particle Swarm Optimization (PSO), Distributed Generation (DG), loss reduction and MATLAB.

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I. Introduction

The growing demand in the power system has posed a challenging task to power system engineers in maintaining a reliable and safe system cheaply. In the heavily loaded network, the load current drawn from the source would raise. This may lead to an increase in voltage drop and system losses. The performance of distribution system becomes inefficient due to the reduction in voltage magnitude and increase in distribution losses. Therefore, the operating cost will also increase. With this regard, changing environment of power systems design and operation have necessitated the need to consider active distribution network by incorporating Distributed Generation units (DGs) sources [18]. The incorporation of DGs in distribution system would prime to improving the voltage profile, reliability improvement such as service, refurbishment and uninterrupted power supply and increase energy efficiency. The distribution feeder reconfiguration (DFR) is one of the largely significant control arrangements in the distribution networks, which can be affected by the interconnection of DGs. Generally, the DFR is defined as varying the topological structure of distribution feeders by changing the open/closed status of section and tie switches so that the power losses is minimized, and the constraints are met. The analysis from [1] has suggested of employing a method based on heuristic algorithm to determine the configuration of radial distribution networks, which finally led to loss minimization. Mohammadi et al. [19] also described heuristic optimization technique for the reconfiguration of distribution networks to decrease their resistive line losses. In another approach, Lopez et al. [2] proposed a solution procedure by employing simulated annealing (SA) to search a satisfactory non-inferior solution. In [20], Sawa has proposed the new method in network reconfiguration that involves the discrete decimal mutant PSO and the fixed loop method. Jin et al [21] introduced a binary particle swarm optimization based reconfiguration methodology for the distribution system. The purpose of their configuration was load balancing. The reconfiguration methodology proposed in that work can only be applied in the power system with radial configuration. Zhou, et al [4] put forward a heuristic reconfiguration methodology for the distribution system to decrease the operating cost in a real time operation environment. In that work, the operation cost in the power system is the power loss in the distribution system. The operation cost reduction in that work is based on the long term operation of the power system. Another heuristic search based reconfiguration algorithm was proposed by Wu et al [3]. In that work, the reconfiguration methodology was applied to the radial power system for service restoration, load balancing, and repairs of the power system.

This paper proposes a network reconfiguration method for distribution network connected with DGs using the PSO algorithm. The proposed method is able to produce an optimum configuration in network distribution and at the same time yield the optimal size of DG and decrease power loss. The proposed PSO also improves convergence characteristics and less computation time as compared with GA technique. The effectiveness of the methodology is demonstrated by a practical sized distribution system consisting of 33-bus

system. The fine points of these algorithms are discussed in section II. Meanwhile, Section III shows the performance of this algorithm using standard test function. The results in terms of voltage profile and power loss are discussed in Section IV and finally the last section grants the conclusion of the study.

II. Problem Formulation

The purpose of distribution network reconfiguration is to find a radial operating structure that minimize the system power losses while fulfilling operating constraints. Thus the problem can be formulated as follows [21].

$$\text{Min } P_{Loss} = \sum_{i=1}^n |I_i|^2 k_i R_i^{iN} \quad (1)$$

Where is I_i = current in branch i , R_i = resistance of branch i , N is the total number of branches and k_i is the variable that represents the topology status of the branches (1=close, 0= open).

Subject to:

a) Radial network constraint: Distribution network should be composed of radial structure considering operational point view.

b) Node voltage constraint: Voltage magnitude V_i at each node must lie within their permissible ranges to maintain power quality

$$V_{\min} \leq V_{bus} \leq V_{\max} \quad (2)$$

The standard minimum voltage used is 0.95 and maximum voltage is 1.05 ($\pm 5\%$). The process of works begins with the initial population.

c) Generator operation constraints: All DG units are only permitted to operate within the acceptable limit where P_i^{\min} and P_i^{\max} are the lower and upper bound of DG output.

$$P_i^{\min} \leq P_g \leq P_i^{\max} \quad (3)$$

d) Feeder capability limits:

$$|I_k| \leq I_k^{\max} \quad k \in \{1, 2, 3, \dots, I\} \quad (4)$$

where I_k^{\max} = maximum current capability of branch k .

III. Fundamental Particle Swarm Optimization Algorithm (PSO)

The basic idea of the PSO is based on the social behavior (foraging) of organisms such as fish (schooling) and bird (flocking) [16-17]. The birds or fish will move to the food in certain speed or position. Their movement will depend on their own experience and experience from other 'friends' in the group (P_{best} and G_{best}).

The new velocity, V_j^{k+1} and the new position, X_j^{k+1} for the fish or birds are obtained using Eq.(5) and (6).

$$V_j^{k+1} = \omega \times V_j^k + C_1 \times rand_1 \times (P_{bestj}^k - X_j^k) + C_2 \times rand_2 \times (G_{bestj}^k - X_j^k) \quad (5)$$

$$X_j^{k+1} = X_j^k + V_j^{k+1} \quad (6)$$

where V_j^k is the velocity of particle j in iteration k , X_j^k is the position of particle j in iteration k , $rand_1$ and $rand_2$ are the random numbers between 0 and 1. P_{bestj}^k is the finest value of the fitness function that has been achieved

by particle j before iteration k . G_{best}^k is the best value of the fitness function that has been achieved so far by any particle. Constants C_1 and C_2 are weighting factors of the random acceleration terms which are usually set to 2.0. While small values allow particles to move away from the target region before they are pulled back, high values result in sharp movements toward the target region. The inertia weight ω is typically set according to the following equation:

$$\omega(t+1) = \omega_{\max} - \frac{\omega_{\max} - \omega_{\min}}{t_{\max}} \times t \quad (7)$$

In Eq.(7), t_{\max} is the maximum number of iterations and t is the current iteration number. ω_{\max} and ω_{\min} are maximum and minimum of the inertia weights, respectively. The process of implementation of PSO algorithm is as follows:

In this work, we only determined the optimal size of DG while the location of DG is fixed [12]. DG location in the network is fixed as a controlled measure in order to observe the responding changes of DG sizing. Furthermore, the DG location in practical is also depends on the suitability of the area.

Step 1: The input data including network configuration, line impedance and status of DGs and switches are to be read.

Step 2: Setup the set of parameters of PSO such as, number of particles N , weighting factors and C_1 , C_2 . The initial population is determined by selecting the tie switches and DG size randomly from the set of the original population. The variable for tie switches represented by S and as for DG size is represented by P_g . The proposed particles can be written as:

$$X_{particle} = \{S_1, S_2, \dots, S_\beta, P_{g1}, P_{g2}, \dots, P_{g\alpha}\} \quad (8)$$

Where β is the number of tie line and α is the number of DG.

Step 3: Calculate the power loss using distribution load flow based on the Newton - Raphson method.

Step 4: Randomly generates an initial population (array) of particles with random positions and velocities on dimension in the solution space. Set iteration counter $k=0$.

Step 5: For each particle if the bus voltage is within the limits, calculate the total loss using distribution load flow. Otherwise, that particle is infeasible.

Step 6: Record and update the best values. The two best values are recorded in the searching process. Each particle keeps track of its coordinate in the solution space that is associated with the best solution it has reached so far. This value is recorded as P_{best} . Another best value to be recorded is G_{best} , which is the overall best value obtained so far by any particle. P_{best} and G_{best} are the generations of switches, DG sizes and power loss. This step also updates P_{best} and G_{best} . At first, we compare the fitness of each particle with its P_{best} . If the current solution is better than its P_{best} , then replace P_{best} by the current solution then, the fitness of all particles is compared with G_{best} . If the fitness of any particle is better than G_{best} , then replace G_{best} .

Step 7: Update the velocity and position of the particles. Eq.(5) is applied to update the velocity of the particles. The velocity of a particle represents a movement of the switches. Meanwhile, Eq.(6) is applied to update the position of the particles.

Step 8: End conditions.

Check the end condition, if it is reached the algorithm stops, otherwise, repeat steps 3-7 until the end conditions are satisfied. In this work we only determine the optimal size of DG while the location of DG is fixed [12]. DG location in the network is fixed as a controlled measure in order to observe the responding changes of DG sizing. Furthermore, the DG location in practical is also depends on the suitability of the area.

IV. Case Studies

The test system for the case study consisting of the standard IEEE 33 bus radial distribution system is shown in fig(1). The system consists of one feeder, 32 normally closed tie line and five normally open tie lines (dotted line) and located on branch No. 33, 34, 35, 36 and 37. The system load is assumed to be constant and $S_{base} = 10.0\text{MVA}$ and $V_{base} = 12.66\text{KV}$. The line and load data details can be referred in [14]. The total load on the system is 3715kW and 2300kVAr. The maximum active output of DG in this study is set to 5MW. While, the size of the population for test systems is 50. The convergence value is taken as 0.0001. The minimum and maximum voltages are set at 0.95 and 1.05 p.u. respectively. All calculations for this method are carried out in the per-unit system. Four cases are considered:

Case 1: The system is without distributed generation and feeder reconfiguration (initial)

Case 2: The same as case 1 except that the feeders can be reconfigured by the available sectionalizing switched and tie switches.

Case 3: The same as case 1 except that there is, two DGs unit is installed and placed at bus number 12 and 28 respectively with feeder configuration.

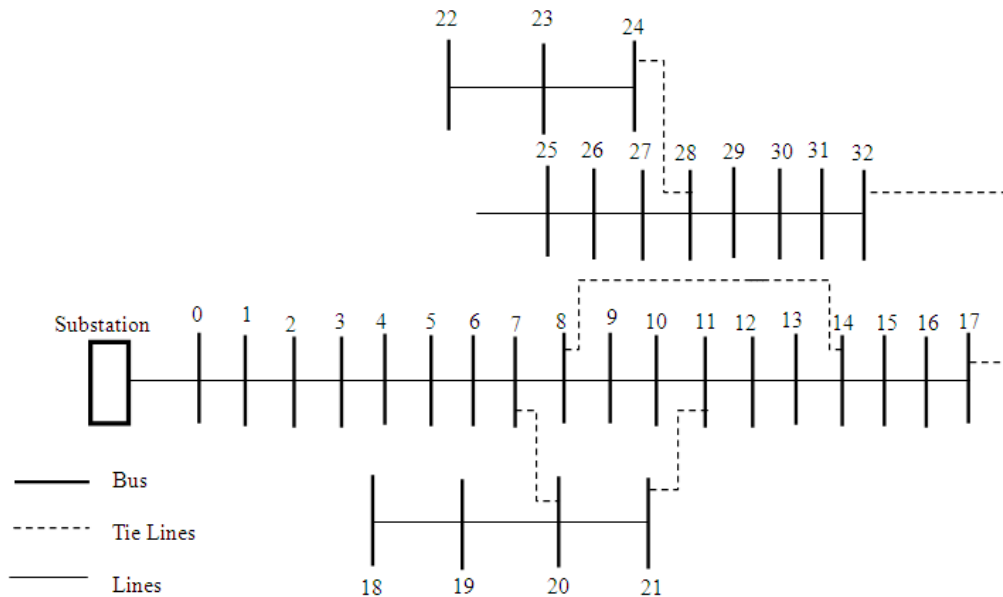


Figure 1: Initial Configuration of the 33 bus radial distribution system

V. Results And Discussion

Table – 1: The Size, Locations and Power Loss Reduction of 33-Bus System

Parameters	Ploss – No DG (kW)	Ploss – One DG (kW)	Ploss – Two DG (kW)
TotalSystemActivePower Loss	202.71	128.50	89.50
ActivePower LossReduction(%)	-----	36.61	55.84
OptimalLocations	-----	12	28
DG-size(kw)	-----	36.14 – (12)	36.14 – (12) 88.60 – (28)

V.1 Impact on Power Losses

After this algorithms is executed by using MATLAB software, only the minimum power loss with optimalDG size is selected. The results obtained consists of the total power loss and two optimal DGsizing. The numerical results for the four cases are summarized in Table 1. The results show the performance of PSO when tested using 33-bus distribution system. It is noticed a considerable decrease in the power loss values when the DGs are placed in the distribution system. It is confirmed from case 3 that the DG and feeder configuration helps to reduce the power loss after reconfiguration from 202.70 kW to 89.50 kW, or 55.84% of the reconfigurations with DG installation.

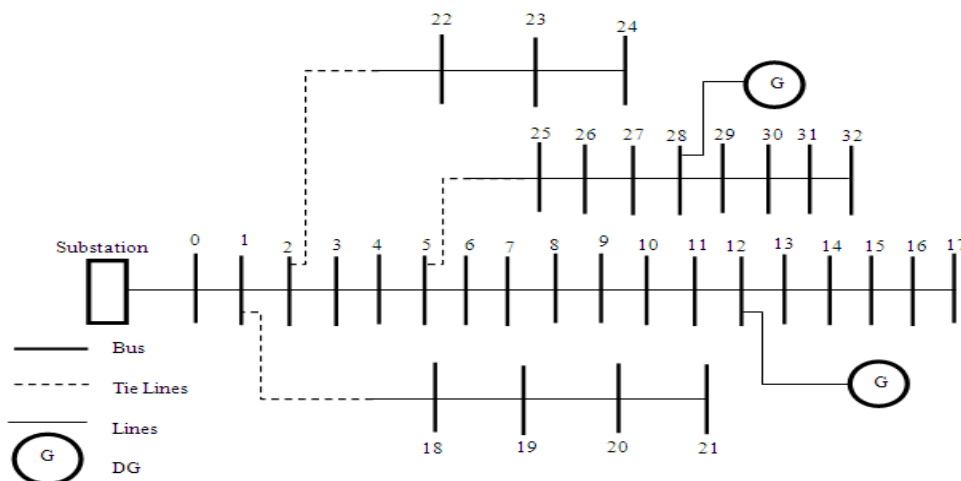


Figure 2: Network Configuration with 2 DGs of 33 bus radial

V.2 Impact on Voltage

The proposed method does not only give the lowest power losses, but also improves the overall voltage profile of the network reconfiguration. Figure 3 illustrates the results given an approx flat voltage profile achieved after reconfiguration.

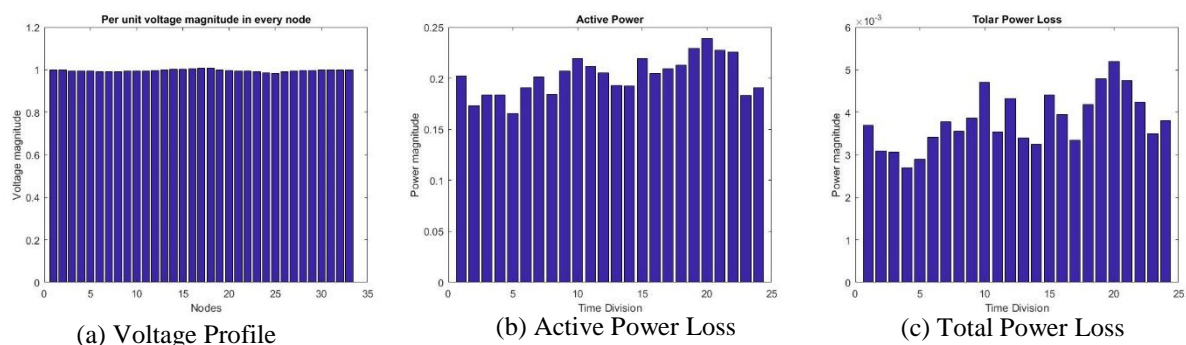


Figure 3: The Result for two DG installation

Meanwhile, Figure 3 depicted the results with reconfiguration and DG on voltage profile improvement achieved by the proposed algorithm. In this case, there are improvements on the voltage profile at every bus. While, the voltage improvement obviously raised to nearby 0.983612517p.u after reconfiguration. The PSO shows a great difference after reconfiguration with DG. Since the PSO gives the fastest solution compared to others and its performance is better than traditional methods, it can be concluded that PSO is a superior method in reconfiguration with DG process.

VI. Conclusion

PSO technique has been developed in this paper with presence of distributed generators for reconfiguration of the distribution system. The main objective of this method is to reduce the real power losses and determine the optimum value of DG size simultaneously on the distribution network reconfiguration. A 33-bus distribution system with two distributed generation is used to demonstrate the effectiveness of the proposed technique. In this paper, three cases are considered as explained in section IV. The results of the proposed algorithm which give the minimum power loss while keeping bus voltage magnitudes within the acceptable limits. Based on these reasons, it is strongly expected that PSO is capable of solving large-scale problems arose in network reconfiguration as compared to the existing methods.

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