

# **Network Reconfiguration for Real power Loss Reduction Using DPSO**

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**ABSTRACT:** This paper presents the Discrete particle swarm optimization (DPSO) algorithm for solving the optimal distribution system reconfiguration problem for real power loss minimization. The DPSO is a relatively new and powerful intelligence evolution algorithm for solving optimum reconfiguration problem. It is a population based approach. The proposed DPSO in this paper is introduced with some modification such as using an inertia weight that decreases linearly during the simulation and also the particle position is rounded to the nearest value satisfying the position constraints. This allows the DPSO to explore a large area at the start of the simulation,. Also, a modification in the number of iterations and the population size is presented. The proposed algorithm is applied to IEEE-33 bus system and the obtained results are compared with conventional methods.

**KEYWORDS:** Reconfiguration, Discrete Particle Swarm Optimization, Real power Loss Reduction and Load flow analysis.

## **I.INTRODUCTION**

The subject of minimizing the distribution system losses has gained a great deal of attention due to the high cost of electrical energy. There are many alternatives are available for reducing losses at the distribution level: 1. reconfiguration, 2. capacitor installation, 3. load balancing, 4. introduction of higher voltage levels. This research focuses on the reconfiguration alternative. Primary distribution systems have two type of switches. They are normally closed switches (sectionalizing switches) and normally open switches (tie switches). Those two type switches are designed for both protection and configuration management. Network reconfiguration is the process of changing the topology of distribution systems by altering the open/closed status of switches. The distribution network reconfiguration is a complicated non differentiable constrained optimization problem. The network reconfiguration is achieved by opening or closing of these two types of switches in such a way that the radiality of the network is maintained. The analysis from [1] has suggested of employing a method based on heuristic algorithm to determine the configuration of radial distribution networks, for loss minimization. The paper [2] introduced a binary particle swarm optimization based reconfiguration methodology for the distribution system. The main objective of the reconfiguration is load balancing. The reconfiguration methodology proposed work can only be applied in the power system with radial.

A heuristic based approach for feeder reconfiguration was proposed [3]. The purpose of reconfiguration is to reduce the operating cost in the real time operation environment. The algorithm emphasized timely finding of the solution with a small no of switching operation as possible. Feeder reconfiguration as well as coordination with other distribution automation applications such as relay coordination is also addressed here. The test result proves the performance of the algorithm is efficient and robust.

This paper proposes a DPSO algorithm for distribution system reconfiguration with a new variable expression design to overcome the drawbacks in the previous method. The effectiveness of the methodology is demonstrated by a practical sized distribution system consisting of 33-bus system. Distribution system radiality checking algorithm is explained in section III. Meanwhile, the details of DPSO algorithms are discussed in section IV. Section V summarized the methodology used for the reconfiguration problem. The simulation results in term of power loss and voltage profile are

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discussed in Section VI and finally the last section presents the conclusion of the study.

## II. PROBLEM FORMULATION

The purpose of distribution network reconfiguration is to find a radial operating structure that minimizes the system power losses while satisfying operating constraints. The problem can be formulated as follows.

$$\text{Min } P_{\text{losses}} = \sum_i^n |I_i|^2 R_i \quad i=1, 2, \dots, n \tag{1}$$

Where  $P_{\text{losses}}$  = objective function (kW),  $I_i$  = current in branch  $i$ ,  $R_i$  = resistance of branch  $i$ ,  $n$  is the total number of branches.

Subject to:

- a) Radial network constraint:  
Distribution network should be composed of radial structure considering operational point view.  
 $\det(A) = 1$  or  $-1$  radial system  
 $\det(A) = 0$  for not radial  
where  $A$ : bus incidence matrix
- b) Node voltage constraint:

Voltage magnitude  $V_i$  at each node must be within their permissible ranges to maintain power quality

$$V_{\min} \leq V_{\text{bus}} \leq V_{\max} \tag{2}$$

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) Power flow equation  
Total active power generation must be equal to the sum of total active power losses and total active load. Similarly, total reactive power generation must be equal to the sum of total reactive power losses and total reactive load as given by the following equations.  
 $P_G = P_L + L_P$  (3)  
 $Q_G = Q_L + L_Q$  (4)  
where,  
 $P_G$  – Total active power generation,  $Q_G$  – Total reactive power generation,  $P_L$  – Total active load  
 $Q_L$  – Total reactive load,  $L_P$  – Total active power loss,  $L_Q$  – Total reactive power loss.

## III. RADIALITY ALGORITHM

In this section a new algorithm based on the bus incidence matrix  $A$  is proposed for checking the radiality of trial solutions. The flow chart of the algorithm is shown in Fig. 1. A graph may be described in terms of a connection or incidence matrix. Of particular interest is the branch-to-node incidence matrix  $A$ , which has one row for each branch and one column for each node with an entry  $a_{xy}$  in row  $x$  and column  $y$  according to the following rules:

$$a_{xy} = 0 \quad \text{if branch } x \text{ is not connected to node } y \tag{4}$$

$$a_{xy} = 1 \quad \text{if branch } x \text{ is directed away from node } y \tag{5}$$

$$a_{xy} = -1 \quad \text{if branch } x \text{ is directed toward node } y \tag{6}$$

These rules formalize the procedure used to set up the coefficient of  $A$ . In network calculation, a reference node have

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to be chosen. The column corresponding to the reference node is omitted from A and the resultant matrix is denoted by A. If the number of branches is equal to the number of nodes then, applying the previous rules a square branch-to-node matrix is obtained. The A matrix has the row-column dimension  $B \times N$  for any net-work with B branches and N nodes excluding the reference node. By assuming, that there is a branch between this reference node and the root of the network; this will lead to a square matrix if the initial structure of the network is radial. The proposed method is based on the value of the determinant of A. It is found that, if the determinant of A is equal to 1 or -1, then the system is radial. Else if the determinant of A is equal to zero, meaneither the system is not a radial or a group of loads are disconnected from the service.

It has the following advantages, they are

- It reduces the computation time.
- It restricts each trial solution to be radial network in distribution network reconfiguration.
- It can be used to determine the branches of each loop formed by closing a tie line.

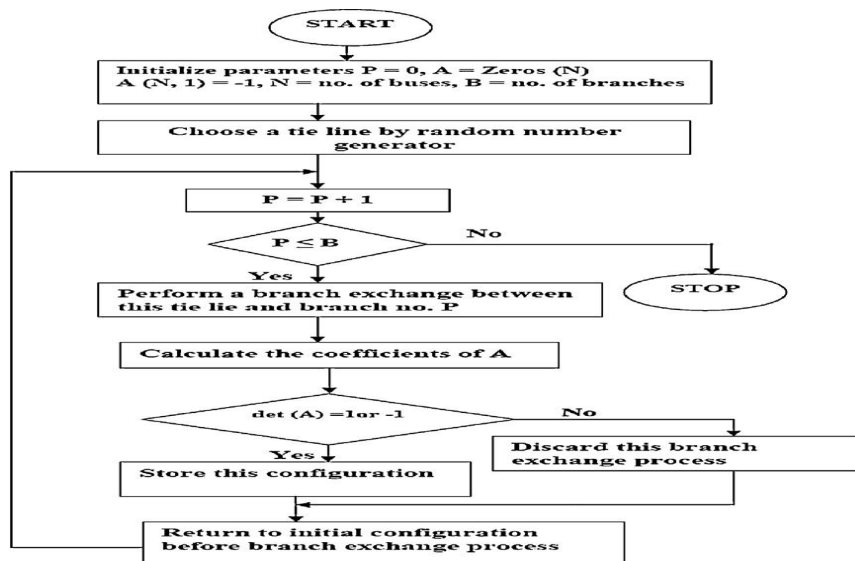


Fig. 1. Flow chart of the checking system radiality algorithm.

## IV. OVERVIEW OF THE DISCRETE PARTICLE SWARM OPTIMIZATION ALGORITHM (DPSO)

The PSO algorithm is developed by simulating human social behavior and individuals of a swarm. PSO has roots in two main component methodologies. It has been noticed that members within a group seem to share information among themselves-a fact that leads to increased efficiency of the group. The PSO algorithm searches in parallel using a group of individuals similar to other artificial intelligence (AI)-based heuristic optimization techniques. DPSO is similar to PSO, the particle position is rounded to the nearest discrete value satisfying the position constraints. The DPSO algorithm searches in parallel using number of individuals. An individual in a swarm approaches the optimum through its present velocity, previous experience, and the experience of its neighbors. In a search space, the position and velocity of an individual are represented as the vectors  $X_i=(X_{i1},X_{i2},\dots,X_{in})$  and  $V_i=(V_{i1},V_{i2},\dots,V_{in})$ . Let  $Pbest=(X_{i1}^{Pbest}, X_{i2}^{Pbest}, \dots, X_{in}^{Pbest})$  and  $Gbest=(X_{i1}^{Gbest}, X_{i2}^{Gbest}, \dots, X_{in}^{Gbest})$  respectively, be the best position of and individual and the best position of its neighbors. The updated velocity of an individual is modified under the following equation in the DPSO algorithm.

$$V_i^{k+1}=W(V_i^k+C_1rand1*(Pbest_i^k-X_i^k)+C_2rand2*(Gbest_i^k-X_i^k)) \quad (7)$$

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Where  $C_1$  and  $C_2$  are positive constants, called cognitive and social parameters, respectively, And both are equal to two in general cases;  $W$  is the inertia weight factor (a large weight factor facilitates a global search, while a Small inertia weight facilitates a local search; in general, the inertia weight starting at 0.9, linearly decreasing to 0.2 during a run, is adopted to give DPSO a better Performance);

$V_i^k$  is the velocity of individual  $i$  at iteration  $k$ ;

rand1 and rand2 are random numbers between [0,1];

$X_i^k$  is the position of individual  $i$  at iteration  $k$ ;

$Pbest_i^k$  is the best position of individual  $i$  up to iteration  $k$ ; and

$Gbest_i^k$  is the best position of the group up to iteration  $k$ .

Each individual moves from the current position to the next one by using the modified velocity equation (Eq. (7)) in Eq. (8),

$$X_i^{k+1} = \text{round}(X_i^k + V_i^{k+1}) \quad (8)$$

The process of the DPSO algorithm can be summarized as follows:

Step 1: Randomly initialize the group while satisfying constraints.

Step 2: Velocity and position updates while satisfying constraints.

Step 3: Update of Pbest and Gbest.

Step 4: Go to Step 2 until satisfying stopping criteria.

Step 5: Display the results of global best values (Gbest).

## Initialization and structure of Individuals

In the initialization process, a set of individuals is created at random. A tie switch and some sectionalizing switches with the feeders form a loop. A particular switch of each loop is selected to open and to make the loop radial, such that the selected switch naturally becomes a tie switch. The network reconfiguration problem is identical to the problem of selecting an appropriate tie switch for each loop to minimize the power loss. A coding scheme for each individual that recognizes the positions of the tie switches is proposed. The total number of tie switches is kept constant, regardless of any change in the system's topology or the tie switches' positions. Figure 2 shows an individual that is composed of tie switch position. Different switches from a loop are selected for cutting the loop circuit and trying to become a tie switch. After each loop is made radial a desired configuration is obtained.



Fig2. Composition of an individual

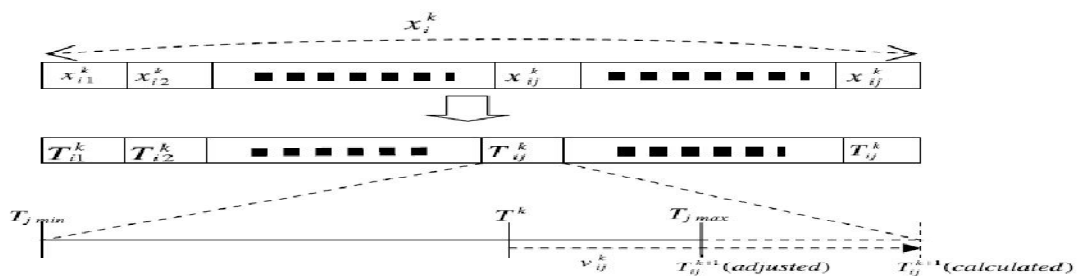


Fig3. Adjustment strategy for an individual position within a boundary

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As shown in Figure 3, the position of the individual at iteration 0 can be represented as the vector of  $X_i^0 = (T_{i1}^0, \dots, T_{in}^0)$ , where  $n$  is the number of tie switches. The velocity of an individual  $i$  (i.e.,  $V_i^0 = (V_{i1}^0, \dots, V_{in}^0)$ ). The velocity corresponds to the tie switch update quantity, selected randomly, covering all possible values for tie switches. The following procedure is suggested for any individual in a group:

- Step 1: Set  $j = 1$ .
  - Step 2: Select a tie switch from loop  $j$  at random.
  - Step 3: If  $j = n$ , then go to Step 4; otherwise  $j = j + 1$  and go to Step 2.
  - Step 4: Stop the initialization process
- After creating the initial position of each individual is also created at random.

**Velocity Update** To modify the position of each individual, it is necessary to calculate the velocity of each individual in the next stage. In this velocity updating process, the values of parameters  $k$ ,  $C1$ , and  $C2$  should be determined in advance. The values of  $C1$  and  $C2$  are considered equal, which implies the same weights are Given between  $Pbest$  and  $Gbest$  in the evolution processes. The velocity of each particle in the next stage can be updated by using Eq. (7).

**Position Modification.** The position of each individual is modified by Eq. (8). The resulting position of an individual is not always guaranteed to satisfy the inequality constraints due to over/under velocity.

If any element of an individual violates its boundary condition due to over/under speed, then the position of the individual is fixed to its maximum/minimum operating point. Therefore this can be formulated as

$$T_{ij}^{k+1} = T_{ij}^{k+1} + V_{ij}^{k+1} \quad \text{if } T_{ij, \min} \leq (T_{ij}^{k+1} + V_{ij}^{k+1}) \leq T_{ij, \max}$$

$$T_{ij, \min} \quad \text{if } T_{ij}^{k+1} + V_{ij}^{k+1} < T_{ij, \min}$$

$$T_{ij, \max} \quad \text{if } T_{ij}^{k+1} + V_{ij}^{k+1} > T_{ij, \max}$$

Where  $T_{ij}^{k+1}$  represents the position of the particle, and  $T_{ij, \min}$  and  $T_{ij, \max}$  are the boundary values. If the value  $(T_{ij}^{k+1} = T_{ij}^k + V_{ij}^{k+1})$  crosses the boundary, then it is set to boundary value. The aforementioned method always produces the position of each individual, satisfying the boundary condition of tie switch position for each loop.

**Update of Pbest and Gbest.** The  $Pbest$  of each individual  $i$  at iteration  $k+1$  is Updated as follows

$$Pbest_i^{k+1} = X_i^{k+1} \quad \text{if } TC_i^{k+1} < TC_i^k$$

$$Pbest_i^{k+1} = Pbest_i^{k+1} \quad \text{if } TC_i^{k+1} > TC_i^k$$

$$Gbest_i^{k+1} = \text{best}(Pbest_i^{k+1})$$

Where  $TC_i$ , at the position of each individual  $i$ , the objective function, is evaluated.  $Gbest$  at Iteration  $k+1$  is set as the best evaluated position among  $Pbest_i^{k+1}$

**Stopping Criteria,** The DPSO is terminated if the iteration approaches the predefined no of Maximum iteration.

**Parameter selection** There exist several parameters to be determined for the Implementation of PSO. In this article, these parameters have been determined through the experiments for the 16-node radial distribution system for loss reduction and the 69-node system for load balancing. The procedures and strategies are adopted as follows:

- 1) The values of  $C1$  and  $C2$  have the same value, which implies the same weights are given between  $Pbest$  and  $Gbest$  in the evolution processes.
- 2) Numbers of particles (10–50) are usually sufficient.
- 3) Usually  $C1+C2 = 4$ , with no good reason other than empiricism (the value taken is solely based on experience).

### Repair algorithm

After updating the position of each particle, a position check is carried out to make sure that none of the particles have flown out of the search space bounds. That is all the generated solutions are feasible. If a violation is detected then apply repair algorithm is used to force the violated particle to return to the feasible region as follows: If position of particle ( $pos(j)$ ) is greater than the maximum position ( $mp$ ), then  $pos(j)=mp$  else if  $pos(j)$  is less than one, then  $pos(j)=1$  else end.

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## V. SOLUTION MECHANISM

The network reconfiguration problem is equivalent to the problem of finding an optimal radial configuration such that the loss is minimized. In this section, the DPSO algorithm is adapted to solve the network reconfiguration problem. Detailed discussions of each step in implementing the DPSO are as in the following sections.

- Step1:- For each particle, the position and velocity vectors will be randomly initialized with same size.
- Step2:- check the network radiality using radiality algorithm this is explained in section III.
- Step3:- Measure the fitness (power loss) of each particle(pbest), the position and store the particle with the best fitness (gbest) value by running the load flow program based on forward sweep method.
- Step4:- Update velocity and position of each particle according to Eqs. (7) and (8)
- Step5:- perform violation check. If violation is detected then apply repair algorithm explained in section IV.
- Step6:- Repeat steps 2-5 until a stopping a criterion is satisfied.

The test system consisting of the standard IEEE 33-bus radial distribution system is shown in Fig.4. The system consists of one feeder, 32 normally closed sectionalize switch and five normally open tie switch (dotted line). The tie switches are located on branch No. 33, 34, 35, 36 and 37. The system load is assumed to be constant and  $S_{base} = 100\text{MVA}$  and  $V_{base} = 12.66\text{KV}$ . The line and load data details can be referred in [4]. The total load on the system is 3715kW and 2300kVAr.

The minimum and maximum voltages are set at 0.95 and 1.05p.u.respectively. All calculations for this method are carried out in the per-unit. A computer program is developed to implement the proposed DPSO algorithm using MATLAB6.5. The DPSO parameter used during the simulation for the reconfiguration problem is summarized in table 1.

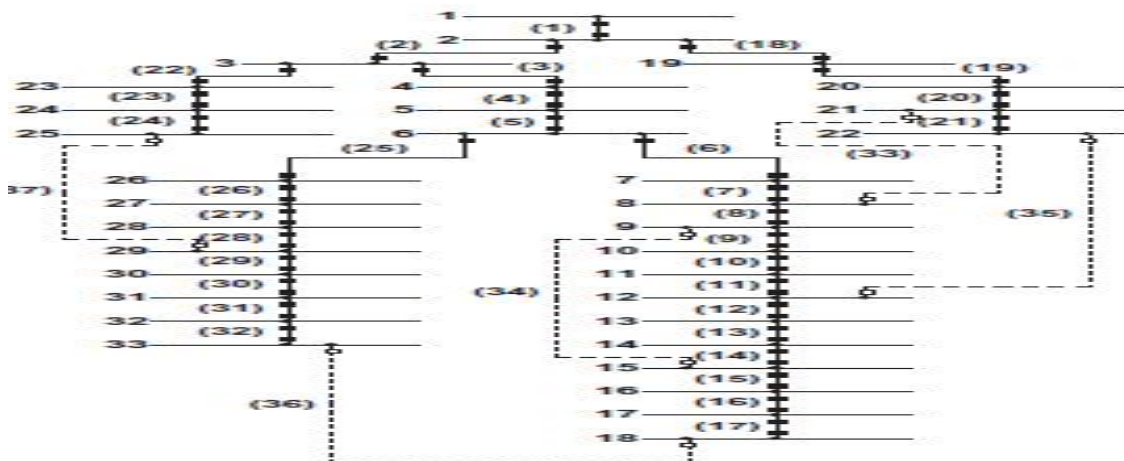


Fig 4 Initial configuration of the 33 bus distribution system



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Parameter	Value
Number of particles	15
Initial velocity of the agent	0
Maximum inertia weight	0.9
Minimum inertia weight	0.2
C1 and C2	2

Table:- DPSO parameters used during the simulation

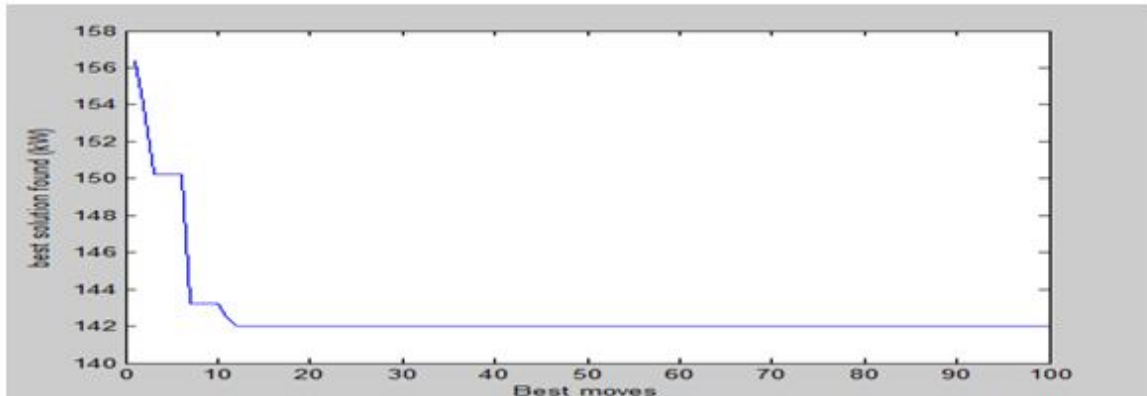


Fig 5 The best moves recorded during the search process for IEEE 33 bus system using DPSO algorithm

## VI. RESULTS AND DISCUSSION

The proposed method is applied to the IEEE 33 bus system to solve the reconfiguration problem. Under base case condition the real power loss is 202.86kW. The lower and upper bounds of the nodal voltage magnitude is  $V_{max}=1p.u$  and  $V_{min}=0.9097p.u$  from the table 2. Simulation results show that the power loss after reconfiguration is 139.68kW which is reduced by 31.15% of its initial value. Fig 5. Shows the best moves recorded during the search process, after 10 iteration system leads to power loss reduction (out of 100). The obtained results using the proposed DPSO algorithm have been reached after 50 trails.

From Fig 6. The voltage profile has been improved by DPSO algorithm. The minimum bus voltage is 0.9097 under base case and it is raised to 0.9362 after reconfiguration.

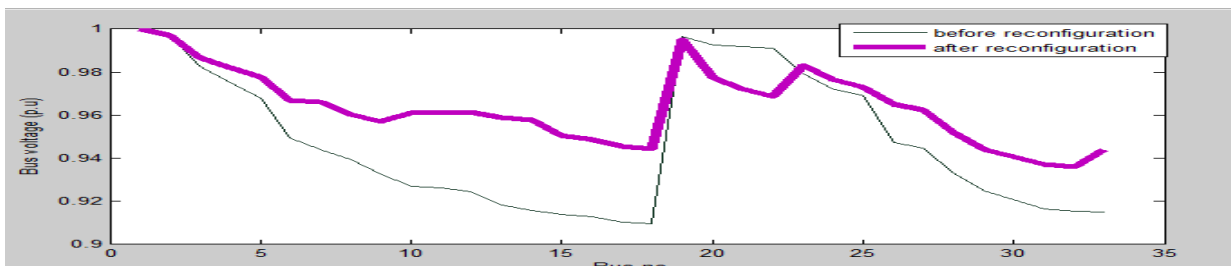


Fig 6. Voltage profile for the 33 bus system after using DPSO algorithm

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System	Power loss	Maximum and Minimum Voltage limits (p.u)	Optimal configuration
Before reconfiguration	202.86 kW	$V_{max}=1$ $V_{min}=0.9097$	33,34,35,36,37
Ant colony algorithm [7]	142.68kW	$V_{max}=1$ $V_{min}=0.9320$	7,10,14,36,37
Hammony search algorithm [8]	139.98kW	$V_{max}=1$ $V_{min}=0.9404$	7,9,14,28,32
<b>Proposed method (DPSO)</b>	<b>139.68kW</b>	$V_{max}=1$ $V_{min}=0.9362$	<b>7,9,14,32,37</b>

Table 2. Result of 33 bus system using the DPSO algorithm

Best objective function	Average objective function	Worst objective function
139.68kW	142.18kW	148.36kW

Table 3. Objective function statistics for IEEE -33 bus System using the DPSO algorithm

## VII. CONCLUSION

This paper proposed the DPSO algorithm to solve the optimal reconfiguration problem. The advantage of DPSO over other method is simplicity. The results obtained during the simulation shows that the DPSO algorithm is capable of finding the optimal solution. The main objective of this paper is to reduce the real power loss and also improve the voltage profile the bus. A 33-bus distribution system is used to demonstrate the effectiveness of the proposed technique. DPSO showed the tremendous improvement in term of processing time, number of iterations to reach the optimal value of power losses. The simulation result indicated that the optimal on/off patterns of the tie line can be identified which give the minimum power loss while keeping bus voltage magnitudes within the acceptable limits. Based on these reasons, it is strongly expected that DPSO is capable of solving large-scale problems arose in network reconfiguration as compared to the existing methods.

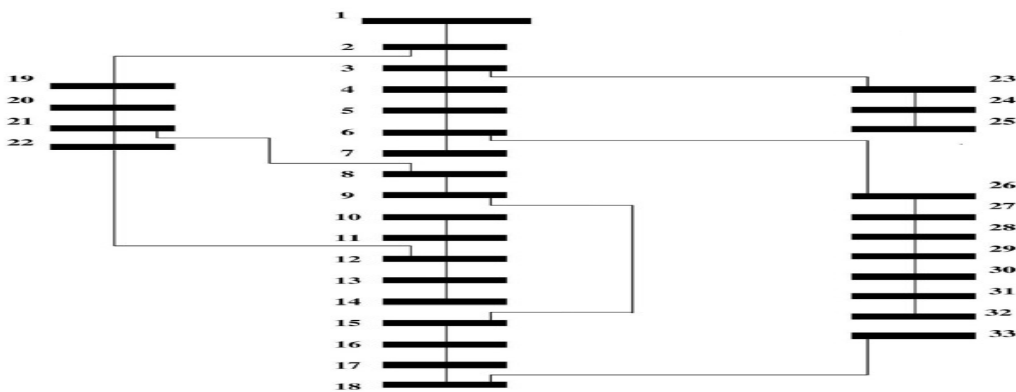


Fig 7. Optimal configuration of 33-bus system using DPSO algorithm with Total Power Loss of 139.7 kW



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