

Distribution Feeder Reconfiguration for Reliability Evaluation by Using Binary Particle Swarm Optimization

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ABSTRACT— This paper presents the probabilistic reliability methodology in the radial distribution system. Reliability is a key aspect of power system design and planning. Reliability indices are calculated at each load points by analytical method. The reconfiguration is aimed at maximizing the reliability of power supplied to the customers. Binary particle swarm optimization (BPSO) is used to determine optimal configuration of switches in network in order to maximize reliability at load points. The proposed methodology applied on IEEE-9 bus system.

KEYWORDS— distribution system, probabilistic reliability model, reconfiguration, reliability indices, BPSO

I. INTRODUCTION

The economic and social effects of loss of electric service have significant impacts on both the utility supplying electric energy and the end users of electric service. The power system is vulnerable to system abnormalities such as control failures, communication system failures, protection and disturbances is due to lightning error, and human operational errors. Hence, maintaining a reliable power supply is a very important issue for power systems design and operation. The objective of a reliability-driven design of a distribution system is to reduce the frequency and duration of power interruptions to the customers. This implies the reduction in the number of customers affected by individual faults, then reduction in time needed to locate and isolate a fault, this reduces the time required to restore power to affected customers, and strengthening of the network by improving

the existing power lines and installing new power lines and equipment, if required. The function of an electric

power system is to satisfy the system load requirement with a reasonable assurance of continuity and quality and the ability of the system to provide an adequate supply of electrical energy is usually designated by the term of reliability. The concept of power-system reliability is extremely broad and covers all aspects of the ability of the system to satisfy the customer requirements.

There is a reasonable subdivision of the concern designated as "system reliability". This represents two basic aspects of a power system: system adequacy and security. This relates Adequacy to the existence of sufficient facilities within the system to satisfy the consumer load demand. These include facilities necessary to generate sufficient energy and the associated transmission and distribution facilities required to transport the energy to the actual consumer load points. This relates security to the ability of the system to respond to disturbances arising within that system. Therefore, Security is associated with the response of the system to whatever perturbations it is subject to Most of the probabilistic techniques presently available for power-system reliability evaluations are in the domain of adequacy assessment. Intuitively, adequacy is tied to forecasting (load and capacity). However, power system reliability evaluations are usually conducted separately in the functional zones of generation, transmission and distribution. The Assessments are made in the combined generation and transmission systems are known as bulk system evaluation. In generation system reliability evaluation, the generation capacity of the total system is examined to determine its adequacy to meet the total system load requirement. Generation adequacy analysis is not intended to provide the reliability indices for individual load points. On the other hand, the distribution system is usually analyzed as a separate entity. Distribution network reconfiguration is the process of

altering the topology of the system by changing the open/closed status of the sectionalizing devices such that no feeder overloads or abnormal voltages are created. This research focuses on reconfiguration for service restoration as an intermediate process in reliability analysis. Reconfiguration for restoration improves system reliability by reducing customer down times following outages power. In this research, a reconfiguration for restoration algorithm is developed that takes into account multiple concurrent failures and also load priorities. Reliability evaluation of an electric power system can be divided into two main categories, system adequacy evaluation and system security evaluation. System adequacy is concerned with the evaluation of adequacy of resources to supply the demanded load, whereas system security is concerned with the response of system to disturbances. Electric power systems are complex and large systems. It is difficult to analyze the entire system at once. For this reason the power system is generally divided into three main hierarchy levels for reliability evaluation.

HLI (Hierarchy level 1) are concerned with the reliability evaluation of generation systems. The main indices analyzed in this study are loss of load expectation (LOLE), loss of energy expectation (LOEE), failure frequency and failure duration. In this study only the generation systems are considered and the effect of transmission and distribution systems on reliability are neglected.

HLII studies constitute of both the generation and the transmission systems. This study can be used to assess the reliability of an existing system or a proposed system. These studies give two main outputs, one related to each bus and the other for overall system. The main indices considered here are failure frequency and failure duration.

HLIII constitutes of the entire power system consisting of generation, transmission and distribution. Due to the complexity of such a system, evaluation of the entire system is not done. The reliability analysis of distribution system is done separately, wherein the indices from HLII evaluation are inputs to the reliability evaluation of the distribution system. The main indices to be considered in this evaluation are SAIFI, SAIDI and CAIDI.

Section II gives an overview of probabilistic reliability assessment to calculate the reliability indices are used in this paper. Section III briefly describes the BPSO algorithm. Section IV gives the Formulation of optimization problem. Simulated results are presented in Section V and Section VI concludes the paper.

II. RELIABILITY ASSESSMENT USING PROBABILISTIC TECHNIQUE

Reconfiguration methodology followed in this paper is to maximize the reliability at various load points. At each and every load point reliability indices are calculated. By using the algorithm optimal placement of switches are found, in order to reduce the frequency as well as the duration of interruptions. First step is to evaluate the reliability at various load points by using probabilistic technique.

A. Component Statistics

A power system consists of various components, such as lines, breakers, cables, switches, reactors, transformers, and capacitors. Any single component outage may cause a partial or even entire system outage. The availability of component functionally is characterized by failure rates and repair or replacement times.

- average failure rate X_s ,
- average outage time r_s ,
- average annual outage time U_s .

$$X_s = \sum X_i \tag{1}$$

$$U_s = \sum X_i \cdot r_i \tag{2}$$

$$r_s = \frac{\sum X_i \cdot r_i}{\sum X_i} \tag{3}$$

B. Reliability Indices

They classify into two categories: System-wide indices and customer indices. And the former is used to give an insight of the overall system performance while the latter is more specific and reflects what the individual customer experiences[2]. They are classified based on the interruptions dealt with sustained or momentary.

(i) System average interruption frequency index (SAIFI):

$$SAIFI = \frac{\text{Total number of customer interruptions}}{\text{Total number of customers served}} \tag{4}$$

(ii) Customer average interruption frequency index (CAIFI):

$$CAIFI = \frac{\text{Total number of customer interruptions}}{\text{Total number of customers affected}} \tag{5}$$

(iii) System average interruption duration index (SAIDI):

$$SAIDI = \frac{\text{Sum of the customer interruption durations}}{\text{Total number of customers}} \tag{6}$$

(iv) Customer average interruption duration index (CAIDI):

$$CAIDI = \frac{\text{Sum of the customer interruption durations}}{\text{Total number of customer interruptions}} \tag{7}$$

(v) Average service availability (unavailability) index ASAI (ASUI):

$$ASAI = \frac{\text{Customer hours of available service}}{\text{Customer hours demanded}} \tag{8}$$

$$ASUI = \frac{\text{Customer hours of unavailable service}}{\text{Customer hours demanded}} \tag{9}$$

C. Component Reliability Models

Probabilistic reliability models of various components appearing in the minimal cut sets are developed based on their outage history and details about commissioning. The availability of a component may be defined as

$$P = \frac{MTF}{MTTF + MTTR} = \frac{\sum_i \frac{1}{X_i}}{\sum_i \frac{1}{X_i} + \sum_i \frac{1}{U_i}} \tag{10}$$

Distribution Feeder Reconfiguration for Reliability Evaluation by using Binary Particle Swarm Optimization

Where MTTF is the mean time to failure, and MTTR is the mean time to repair for a component; and are different types of outage and repair rates considered. The types of models for various components are considered in this paper are defined as follows:

1). Generator, Transformer, and Circuit-Breaker (CB) Model:

The common model for the generator, transformer, and CB is shown in Fig. 1. It includes failure modes, active and passive. Passive event is a failure component mode that does not cause the operation of protection breakers and they does not have an impact on the remaining healthy components. Service is restored by repairing or replacing the failed component the examples are open circuits and inadvertent opening of breakers and the active event is a component failure mode that causes the operation of the primary protection zone around the failed component and can therefore cause the removal of other healthy components and branches from service. Outage maintenance and end-of-life probabilities are also included in the model.

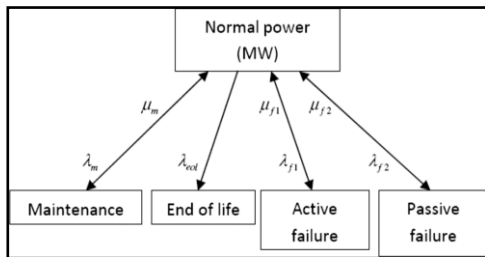


Fig. 1 Generator, Transformer and CB Model

2). Bus Model:

The reliability model of a bus bar is shown in Fig. 2 this includes maintenance and failure modes.

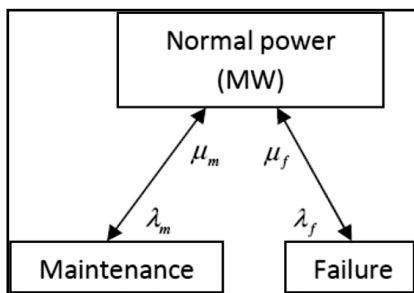


Fig. 2 Bus Model

3). Distribution Line and Switch Model:

The reliability model of the transmission line and switch is shown in Fig. 3, which includes active and passive failure modes and outage maintenance state.

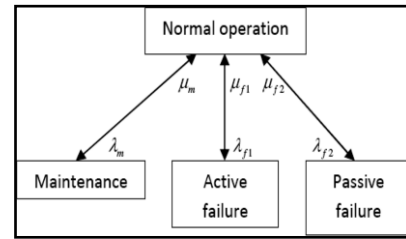


Fig. 3 Distribution Line and Switch model

D. Evaluating the Reliability at the Load Centers

The computational time can be prohibitively large even for an offline stage of planning and one of the major contribution of this paper is the application of the probabilistic method with very less computational time even for large systems. In this paper has been identified that the number of components appearing between the feeder and a load point in a distribution system is limited. Hence, the computation of the joint probability of outage of the components is possible within a short time, as explained in the following text. The minimal cut sets between the feeder and the load points are determined by using the algorithm[5]. For the power failure at a load point of all the components of a cut set must fail. In addition, the load point fails if failure of any one of the cut sets occurs, and each cut set is effectively connected in series with all other cut sets and the details of evaluating the reliability at the load points for cut sets of different orders are described in the following text. Unreliability at the load point is given by

$$Q = P(U_i C_i) \quad (11)$$

Where C_i is the failure event of the i th minimal cut set between the feeder and the load point, $P(\cdot)$ denotes the probability of occurrence of an event.

The reliability at the load point is given by

$$R = 1 - Q \quad (12)$$

1). First-Order Cut Sets:

Fig.4 shows a set of first-order cutsets between the feeder and the load point. The failure event of the i th first-order cut set is denoted by C_i^l ; and N_l is the total number of first-order cut sets.

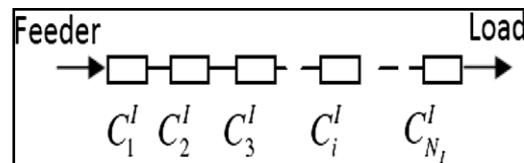


Fig. 4 First Order Cut Sets

The unreliability at the load point due to elements of outage belonging to one or more first-order cut sets is given by

$$Q^l = p(U_{i=1}^{N_l} C_i^l) \quad (13)$$

2). Second-Order Cut Sets:

Fig.5 shows a number of second-order cut sets between the feeder and the load point and the power failure at a load point due to the failure of a second-order cut set occurs when both components in the cut set fails, reason is they are connected in parallel.

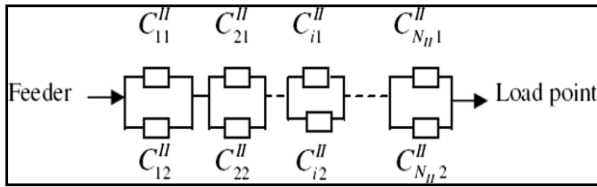


Fig .5 Second Order Cut Sets

The unreliability of the load point due to components outage belonging to one or more second-order cut sets is given by

$$Q^{II} = P(C^{II}) \tag{14}$$

Where the Failure event C^{II} is given by

$$C^{II} = \bigcup_{i=1}^{N_{II}} C_i^{II} \tag{15}$$

Here, N_{II} is the total number of second-order cut sets. This cut set can cause a failure only when both of its components fail.

$$C_i^{II} = C_{i1}^{II} \cap C_{i2}^{II} \tag{16}$$

Once the unreliability is computed for various orders of cut sets in equation (2) can be used to compute the total unreliability of points under load condition. The corresponding measure of the reliability is obtained by using (3). For a distribution system having multiple load points, then the average value of the unreliability is taken as

$$Q_{SA} = \frac{1}{L} \sum_{i=1}^L Q_i(\text{loadpoint}) \tag{17}$$

Where Q_{SA} is the average unreliability for the system, and L is the total number of load points considered in the distribution system.

III. BINARY PARTICLE SWARM OPTIMIZATION

The reconfiguration of the distribution system discussed in this paper is aimed at maximizing the reliability of the power at the load points and minimizing the system power loss. And the BPSO based algorithm is used to find the optimal configuration of the switches in the network. Reliability at the load points is evaluated by using the probabilistic method described in Section II. It has been successfully applied to optimize various continuous nonlinear functions. The basic principles of PSO are taken from the collective movement of a flock of bird. Number of agents or particles is employed in finding the optimal solution for the problem under consideration. PSO is based on the exchange of information between individuals, so called particles, of the population, so called swarm. Each particle adjusts its own position towards its previous experience and towards the best previous position obtained in the swarm. The movement of particles toward finding the optimal solution is guided by the knowledge of individual and other particles. The position of particle at any instant is determined by its velocity at that instant and the position at the previous instant, as shown in equation (18)

$$X_i(t) = X_i(t-1) + V_i(t) \tag{18}$$

Where $X_i(t)$ and $X_i(t-1)$ are the position vectors of the i th particle at the instant t and $t-1$ respectively, and $V_i(t)$ is the velocity vector of the particle. The velocity vector is updated by using the experience of the individual particles and the knowledge of the performance of the other particles in its neighborhood and the velocity update rule for a basic PSO is

$$V_i(t) = V_i(t-1) + \phi_1 r_1 (Pbest_i - X_i(t-1)) + \phi_2 r_2 (gbest - X_i(t-1)) \tag{19}$$

where ϕ_1 and ϕ_2 are adjustable parameters called individual and social acceleration constant respectively; r_1 and r_2 are random numbers in the range $[0, 1]$; $pbest_i$ is the best position vector found by the i th particle; $gbest$ is the best among the position vectors found by all the particles. The vectors $pbest$ and $gbest$ are evaluated by using a suitably defined fitness function. ϕ_1 and ϕ_2 are usually defined such that $\phi_1 + \phi_2 = 4$, with $\phi_1 = \phi_2 = 2$. The maximum and minimum values of the components of velocity are limited by the following constraints to avoid large oscillations around the solution.

$$v_{ij} = \begin{cases} -v_{max}, & \text{if } v_{ij} < -v_{max} \\ v_{max}, & \text{if } v_{ij} \geq v_{max} \end{cases} \tag{20}$$

In BPSO, each element of position vector can take only binary values i.e. 0 or 1. At each stage of iteration, the element of position vector x_i are updated according to the following rule.

$$x_{ij}(t) = \begin{cases} 1, & \text{if } \rho_{ij} < S(v_{ij}) \\ 0, & \text{otherwise} \end{cases} \tag{21}$$

Where ρ_{ij} is a random number in the range $[0, 1]$. $S(v_{ij})$ is a sigmoidal function defined as,

$$S(v_{ij}) = 1 / (1 + \exp(-v_{ij})) \tag{22}$$

In a BPSO algorithm, population is initiated randomly with particles and evaluated to compute fitnesses together with finding the particle best (best value of each individual so far) and global best (best particle in the whole swarm). Initially, each individual with its dimensions and fitness value is assigned to its particle best. The best individual among particle best population, with its dimension and fitness value, assigned to the global best. Then a loop starts to converge to an optimum solution. In the loop, particle and global bests are determined to update the velocity first. Then the current position of each particle is updated with the current velocity. Evaluation is again performed to compute the fitness of the particles in the swarm. This loop is terminated with a stopping criterion predetermined in advance.

```

Initialize Parameters
Initialize Population
Evaluate
Do{
    Find particle best
    Find global best
    Update velocity
    Update position

```

Evaluate
 } While (termination)

IV.OPTIMAL RECONFIGURATION USING BPSO

The distribution reconfiguration problem is essentially the determination of status (open or close) of the switches in the system, depending on desired performance criterion, which, in this case, is the reliability of the power supplied to the customers. The reconfiguration can be an operational or planning problem. In the operating state of a distribution system, the existing sectionalizing and tie switches constitute the set of switches for which the optimal statuses are to be determined. In the planning stage, along with the existing switches in the system, potential locations where new switches may be installed need to be identified. For the present work, it is assumed that the locations of the existing switches and the potential new switches are known. The BPSO algorithm searches through various possible set of switch configurations. At each stage of the search process, reliability of the reconfigured system is evaluated in terms of the chosen reliability index. Constraint imposed on the feasibility of a set of switch configuration is that the system should be radial, and that the electrical connection from the feeder to the end load should be maintained.

In the proposed implementation of the BPSO, the position vectors of the particles represent the potential solutions for the distribution system reconfiguration problem. A fitness function needs to be defined to evaluate the suitability of the solutions found by the particles at each stage. The individual position vector of a particle, $pbest_i$, and the global best position vector $gbest$ are evaluated based on this fitness function. The objective of the distribution system reconfiguration problem in this paper is to maximize the reliability of the power supplied to the customers. The fitness function therefore should evaluate, for the position vector of each particle, (1) whether the reconfigured system is feasible in terms of network constraints, and (2) in case it is feasible, what is the reliability of the power supplied to the customers.

1. *LOLE*: Loss of load expectation (LOLE) is defined as the average number of hours or days in a given period in which the hourly or daily peak load exceeds the available generating capacity. This is a measure of the adequacy of the power supplied by one or more feeders to the distribution system.

2. *LOEE*: Loss of energy expectation (LOEE) is the expected energy (in MWhr) not supplied due to the loads exceeding the available power supplied by the feeders.

The fitness function $J(x)$ for using the BPSO is formulated as follows:

Line no	Failure rate (failure per year)	Repair rate (hrs)	Load points	No of customers	Demand (kw)
1	0.10	2.0	-	-	-
2	0.15	2.5	2	100	1000
3	0.20	3.0	3	200	2000
4	0.20	3.0	4	300	3000
5	0.15	2.25	5	200	4000
6	0.10	2.0	6	300	5000
7	0.10	2.0	7	100	2000
8	0.15	2.35	8	400	3000
9	0.20	3.25	9	100	1000

$$J(x) = \begin{cases} K, & \text{if the configuration is not feasible} \\ w_1J_1 + w_2J_2, & \text{if the configuration is feasible} \end{cases} \quad (23)$$

Where K is a large number assigned to the fitness function if the position vector representing the set of switch configurations is not feasible (i.e., the reconfigured system is not radially connected); w_1 and w_2 are two weights with values such that w_1J_1 and w_2J_2 are comparable in magnitude. J_1 and J_2 are the reliability indices as shown below

$$J_1 = LOLE$$

$$J_2 = LOEE$$

The number of elements in the binary position vector x is equal to the number of switches in the system. The elements of x are defined as follows:

$$x = \begin{cases} 1, & \text{if the switch is closed} \\ 0, & \text{if the switch is opened} \end{cases} \quad (24)$$

The search process starts with a randomly selected binary position vector x , i.e., each element of x is randomly assigned a value of either 0 or 1. Using this switch configuration, it is examined whether the network constraints are being satisfied or not. If the switch configuration fails to render a connected and radial network, it is considered an infeasible solution, and a large numerical value, K is assigned to the fitness function. When the switch configuration satisfies the network constraints, a Monte Carlo simulation is performed, and the reliability indices *LOLE* and *LOEE* are evaluated based on the reliability models of the components that are involved in the network. The fitness function is the weighted sum of *LOLE* and *LOEE*.

V. SIMULATION RESULTS

The effectiveness of the proposed methodology is tested on a 9 bus radial distribution system. The possible location of switches is shown in Fig. 6. Each switch can either open or close. The optimal status (open or close) of the switches is determined by using computing techniques of BPSO. For simplified analysis, it is assumed that the reconfigured system should be radial, since the radiality of the system is taken as a necessary constraint.

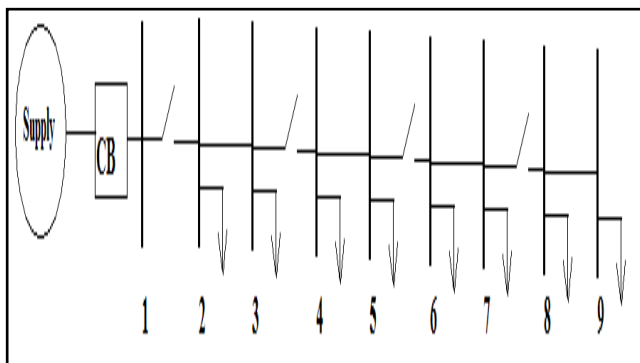


Fig.6 Single line diagram of IEEE 9 bus system

The table I shows the line and load data of IEEE 9 Bus system. Bus 1 is taken as a slack bus.

TABLE I

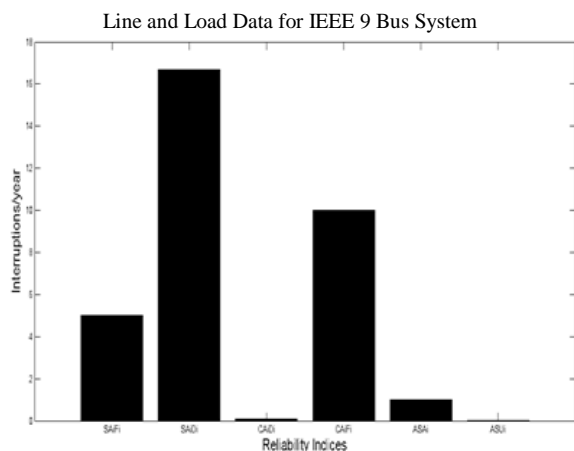


Fig. 7 Before optimal placement of switches in IEEE 9 bus system

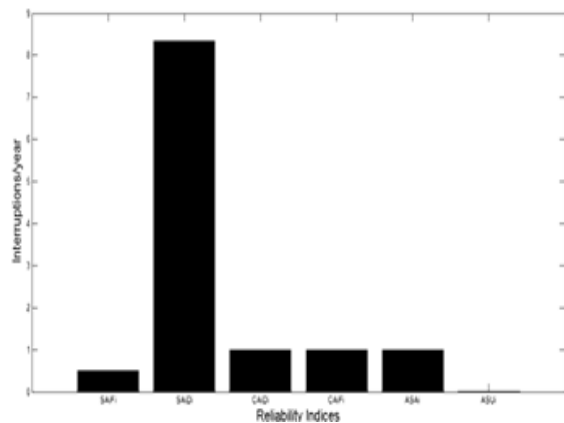


Fig. 8 After optimal placement of switches in IEEE 9 bus system

VI. CONCLUSIONS

In this project, reliability of IEEE 9 bus radial distribution system has been improved by Network Reconfiguration using Binary Particle Swarm Optimization (BPSO). Distribution system reconfiguration methodology aimed at maximizing the reliability by reducing the frequency and duration of interruption power supplied to the customers. In the present work, probabilistic reliability models of the components are used to estimate the future behavior of the distribution system in terms of the reliability of supplying

power to its customers. Binary particle swarm (BPSO) algorithm is used to determine the optimal statuses of the switches in the distribution system. By using algorithm BPSO fault rate of the equipment is reduced and the reliability is maximized. Thus frequency and duration of interruptions are reduced after the optimal placement of switches in IEEE 9 bus system and the Reliability is improved.

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