American Journal of Engineering Research (AJER)2020American Journal of Engineering Research (AJER)e-ISSN: 2320-0847 p-ISSN: 2320-0936Volume-9, Issue-1, pp-247-263www.ajer.orgResearch PaperOpen Access

Optimal Capacitor Placement in Radial Distribution Systems Using Grasshopper Optimization for Power Loss and Energy Cost Minimization

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Abstract: This paper applies the Grasshopper Optimization Algorithm (GOA), as a new approach to determine the optimal capacitor placement in radial distribution systems. The objective function is adapted to minimize both of power loss and energy cost. GOA is inspired from grasshoppers swarm motion. The grasshoppers can fly individually (which emulate the local searching in optimization techniques) or in swarm (which emulate the exploration of optimization techniques). The developed algorithm is validated based on different standard distribution systems such as 33-bus, 69-bus, and 85-bus test systems. However, the obtained results are compared with other algorithms to highlight the advantages of the developed approach. Numerical results stated that the GOA can generate optimal solutions for losses reduction and capacitor locations with quality better than many existing algorithms.

Keywords: Radial distribution system; Optimal capacitor placement; Power loss and energy cost minimization; Grasshopper optimization algorithm.

Nomenclatures

Nomer	iciatures		
K _e	Energy cost per each KWh	g	gravitational constant
T_j	Duration for which jth load level operates	$\overrightarrow{e_g}$	Unity vector towards the center of earth
P_j	Active power loss during jth load level	D_{ij}	Distance between <i>i</i> and <i>j</i> grasshoppers
K _c	Purchase cost of capacitor per KVAR	$s(D_{ij})$	Represents the social force
Q_{ci}	Size of capacitor placed at the ith bus	List of abb	reviation
псар	Number of capacitor locations.	GOA	Grasshopper Optimization Algorithm
R_k	Resistance of line	PGS	Plant Growth Simulation
X_k	Reactance of line	GA	Genetic Algorithm
Ploss	Active power loss of line	DSA	Direct search algorithm
Q_{loss}	Reactive power loss of line	ACO	Ant colony optimization
P_{Tloss}	total network active loss	IHA	Improved harmony algorithm
$P_{L,k}$	Real power load fed through bus j	MOHEA	Multiobjective hybrid evolutionary algorithm
$Q_{L,k}$	Reactive power load fed through bus j	IP	Interior Point
V _{max}	Maximum value of bus k voltage	FRCGA	Fuzzy-Real Coded GA
V _{min}	Minimum value of bus k voltage	DEA	Differential Evolution Algorithm
Q_{Ln}	Total load reactive power	GWO	Gray Wolf Optimizer
V_k	Bus voltage magnitude at bus k.	CSA	Cuckoo Search Algorithm

k+1

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P_l	Real power flow through the branch between bus k and $k+1$	TLBO	Teaching Learning Based Optimization
Q_l	Reactive power flow through the branch between bus k and $k+1$	PSO	Particle Swarm Optimization
R _l	Resistance of branch between bus k and $k+1$	MINLP	Mixed integer nonlinear programming
X_l	Reactance of branch between bus k and	BFOA	Bacterial foraging Optimization Algorithm

I. INTRODUCTION

Electrical power systems become more complicated and sophisticated systems. They contain of three primary components: the generation system, the transmission system, and the distribution system. Each component is essential to deliver the electrical energy from the generation site to the customer site. The major inductive loads are connected to the network through the distribution systems. This type of loads causes low voltage levels, high currents and power losses. However, studies have indicated that as much as 13% of the total generated power is lost as line losses [1]. Therefore, these losses must be diminished to improve the power system stability, power factor and voltage profile [2-4]. Connecting shunt capacitors is considered one of the basic methods which can be used to achieve that target [5], [6]. In general, the inappropriate locating of capacitors may be led to more voltage drop and higher losses. Moreover, the capacitor allocation problem has a combinatorial nature because capacitor locations and sizes are discrete variables [7]. Therefore several optimization algorithms have been proposed in recent years to solve that problem.

Recently many methods and optimization algorithm have been proposed in order to find the optimal capacitor placement problem [8]. Plant growth simulation algorithm PGS was presented for optimal allocation of capacitor with the objective of improving voltage profile and reduction of power loss. [1]. Genetic Algorithm GA was used to determine the optimal sizing of fixed and switched capacitor at different load levels [9-12]. Fuzzy based GA was used to find the optimal size with the multi objective of minimizing the energy cost and to enhance voltage profile of the system [13]. Direct search algorithm (DSA) was presented to find the optimal location and size of fixed and switched capacitor and it was tested on IEEE 22, 69, 85 bus radial distribution system with the objective of maximizing net savings and minimizing the power loss [14]. Ant colony optimization ACO algorithm was proposed to solve capacitor placement in radial distribution system [15]. Taher and Bagherpour proposed the hybrid honey bee colony optimization algorithm to place the shunt capacitor in IEEE 25, 37 bus radial distribution system to minimize power loss and maintains total harmonic distortion [16]. Baran and Wu introduced mixed integer programming for the capacitor placement [17]. Prakash and Sydulu introduced the particle swarm optimization to determine the optimal size of the capacitor bank to minimize the power loss [18]. Sayyad Nojavan et al. proposed mixed integer nonlinear programming approach to determine the optimal location and size of the capacitor to minimize the power loss and increased the net benefits [19]. Improved harmony algorithm IHA [20] is used to solve this problem depend on power loss index. Multiobjective hybrid evolutionary algorithm MOHEA [21] and improved harmony algorithm IHA [22] are introduced to solve the problem of the placement and size of the capacitor.

Grasshopper Optimization Algorithm (GOA) is a very new optimization technique GOA that has been introduced by Seyedali [23]. GOA is conceptualized from grasshoppers swarm motion where grasshoppers can move individually which emulate the local searching in optimization techniques or fly in swarm which emulate the exploration of optimization techniques. The mathematical model of grasshopper motion is described in [24], [25]. The GOA is a recent optimization technique and it has been under study. However, it should point out that GOA is similar to all new optimization technique where it has been tested with standard benchmark functions and also applied for finding the optimal shape for a 52-bar truss, 3-bar truss, and cantilever beam.

However, the application of GOA in solving the optimal capacitor placement problem for distribution systems has not been studied before. Hence, this paper develops the GOA in order to determine the optimal locations and sizes of capacitors and minimize the power losses and total cost of radial distribution systems considering different load levels. Three standard test systems; 33-bus, 69-bus and 85-bus systems are used to validate the effectiveness of the developed approach. In addition, the obtained results are compared with different well-known optimization algorithms to confirm the superiority of developed approach

The rest of paper is organized as follows: Section 2 presents the problem formulation. The GOA is described in Section 3. Section 4 presents the solution process of optimal capacitor placement problem using GOA. Section 5 presents the numerical results of developed GOA approach based on three standard radial distribution systems. Finally, the conclusions of paper are presented in Section 6.

II. PROBLEM FORMULATION

2.1 Forward/ Backward sweep three phase load flow

Forward/backward sweep power flow method have been proven to be effective in analyzing radial distribution systems. However, this load flow method based on forward/backward sweep approach using Kirchhoff's voltage and current laws [26]. In general, the forward/backward sweep load flow algorithm includes the following steps:

First step: Identify different layers in the radial distribution network as described in Fig. 1.



Fig. 1 layers in radial distribution network (Step1)

Second step: Calculate the injected currents of each phase with the initial voltages for all nodes using (1).

$$I_i^s = \left(\frac{S_i^s}{V_i^s}\right)^* \tag{1}$$

Third step:

This step based on backward sweep approach. The total branch currents beginning from the lower to upper layers can be calculated. The current flowing through the line segment l can be calculated as given in (2).

$$J_l^s = -I_j^s + \sum_{m \in \mathcal{M}} J_m^s \tag{2}$$

Final step: The forward sweep step (also known as the voltage update step). The node voltages are corrected beginning from the first layer towards the last layer as given in (3).

$$V_j^s = V_i^s - Z_{ij}^s J_l^s \tag{3}$$

The above steps are repeated until the load flow convergence is done.

1.1 Power loss calculation

The power flow equations of distribution system can be simply obtained from Fig. 2 which represents the radial distribution system as follows:



Fig. 2 Single line diagram of a radial distribution system.

$$P_{l+1} = P_l - P_{L,k+1} - R_l \left(\frac{P_l^2 + Q_l^2}{|V_k|^2}\right)$$
(4)

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$$Q_{l+1} = Q_l - Q_{L,k+1} - X_l \left(\frac{P_l^2 + j Q_l^2}{|V_k|^2}\right)$$

$$P_l, Q_l, R_l, X_l$$
(5)

Voltages of transmission line can be calculated from (6) as follows:

$$V_{k+1}^{2} = V_{k}^{2} - 2(R_{l}P_{l} + X_{l}Q_{l}) + (R_{l}^{2} + X_{l}^{2}) \left(\frac{P_{l}^{2} + Q_{l}^{2}}{|V_{k}|^{2}}\right)$$
(6)

The active and reactive power losses of l^{th} line between buses k and k+1 are given as:

$$P_{loss(k,k+1)} = R_l \left(\frac{P_l^2 + Q_l^2}{|V_k|^2} \right)$$

$$Q_{loss(k,k+1)} = X_l \left(\frac{P_l^2 + Q_l^2}{|V_k|^2} \right)$$
(8)

The total system loss can be calculated by summing all line losses as:

$$P_{T \ loss} = \sum_{k=1}^{n-1} P_{loss(k,k+1)}$$
(9)

The capacitor banks can be installed in distribution systems for enhancing the power quality and minimizing the total cost by injecting reactive power to the systems. Fig. 3 shows the single line diagram of radial distribution system including a shunt capacitor at bus k+1. However, the new value of the reactive power thought the transmission line can be calculated as:

$$Q_{k+1} = Q_k - Q_{L,k+1} - X_k \left(\frac{P_k^2 + Q_k^2}{|V_k|^2}\right) + Q_{C,k+1}$$
(10)



Fig. 3 Radial distribution system with shunt capacitor.

2.2 Objective Function

The objective of optimal capacitor placement in distribution system is to optimize a certain objective functions such as; real power losses and total cost per year (including the energy power losses and the cost of installing capacitors), voltage profile, voltage stability index, subject to voltage and reactive power limits. In the current study, two objective functions; real power losses and total cost per year are investigated. The following mathematical statement can be performed to achieve these objective function:

$$f_1 = (P_{T \ loss})$$
,
 $f_2 = (Cost)$. (11)
(12)

In general, installation of capacitors in distribution network can improve the voltage profile and reduce current flow through the lines. Consequentially, the power loss, energy loss cost and energy efficiency of the network will be better. However, the installation of capacitors increases the investment cost. Therefore, the objective function in this case is to minimize the total cost which can be defined as [14]:

$$Cost = K_e \sum_{j=1}^{L} T_j P_j + \sum_{i=1}^{ncap} K_c Q_{ci}$$
⁽¹³⁾

2.3 Operational constraints

The above objective functions are subjected to the following constrains:

Voltage limits

The bus voltage magnitude of each bus must be limited at its allowable ranges as:

 $V_{min} \leq |V_k| \leq V_{max}$

The lower and upper values are taken as 0.90 and 1.05 p.u, respectively.

Total reactive power limits

The total injected reactive power by capacitor banks must be less than or equal the total load reactive power as:

$$Q_{ci} \le Q_{Ln} \tag{15}$$

Compensation limits

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(14)

III. GRASSHOPPER OPTIMIZATION ALGORITHM

GOA is a new efficient optimization technique that is inspired form the life style (movement, migration) of grasshopper in natural. The adult insects of grasshopper traveling together over long distance which mimics exploration of optimization technique. The nymphs have no wing so it move in small area which mimics the exploitation of optimization technique [23].

Grasshoppers are harmful insects that can destroy a wide area of the agriculture and crops. The grasshoppers swarm consists of million members which can cover wide area up to 1000 KM. The life cycle of grasshopper consists of three stages as depicted in Fig. 4. The grasshopper can be found in two phases. In the first phase the individual of grasshoppers avoid interaction together (solitary phase) while in another phase (gregarious phase), grasshoppers became sociable and form a swarm. The flying swarm of grasshoppers depend upon environmental consideration such as air temperature, sunshine and wind speed [27].



Fig. 4 The life cycle of grasshopper.

The swarm of grasshopper moves in rolling motion. The groups are formed firstly by collection of insects which move in ground or locally and short flight. Then these groups became coordinated together and the insects share a common spatial orientation. However, the behavior of grasshopper swarm can be summarized as:

- The swarm flying with downwind.
- The grasshoppers in front of swarm settle on the ground.
- The settled insects start eating and resting.
- The swarm starts taking of gain to altitude.
- The grasshopper swarm navigation behavior aligned wind is depicted in Fig. 5.

Xi



Fig. 5 Motion of grasshopper swarm aligned with wind.

The grasshopper swarm behavior depends upon social interaction between grasshopper, the gravity force and the downwind advection. Hence mathematical behavior can be represented as [28]:

$$= r_1 S_i + r_2 B_i + r_3 C_i$$

A social force between two grasshoppers is established biologically, where the repulsion forces is existed in order to prevent collisions over a short length scale and attraction force is existed for aggregation. The social interaction between grasshoppers can define as:

$$S_i = \sum_{j=1}^N \sum_{i \neq j} s(D_{ij}) \left(\frac{x_i - x_j}{D_{ij}} \right)$$

Where

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(16)

(17)

$$D_{ij} = |x_i - x_j| s(D_{ij}) = F e^{\frac{D_{ij}}{l}} - e^{D_{ij}}$$
(18)

where F is the intensity of attractive force and l is the attractive length scale. The swarm motion is directly affected by the gravity force which can be found as:

$$B_i = -g\vec{e_g} \tag{19}$$

The wind advection effect on the motion swarm: $C_{i} = u\vec{e}^{i}$

By substituting the value of
$$S_i$$
, B_i and C_i from (18), (19) and (20) in (16):

$$X_{i} = \sum_{j=1}^{N} \sum_{i \neq j} s(D_{ij}) \left(\frac{x_{i} - x_{j}}{D_{ij}}\right) - g\overline{e_{g}} + u\overline{e_{w}}$$

$$(21)$$

where, N is number of grasshoppers. The previous equation is modified to be implemented for optimization problems and for enhancing the capability global searching of the algorithm. However, it can be modified as follows:

$$X_i^m = C\left(\sum_{j=1}^N \sum_{i\neq j} C\left(\frac{Up, m-Lp, m}{2}\right) s(D_{ij})\left(\frac{x_i - x_j}{D_{ij}}\right)\right) + X_{best}^m$$
(22)

where, U_p and L_p are the upper and lower limits of the control variable, respectively. x_{best}^m is the best position (the target position). *C* is an adaptive coefficient decreased linearly for enhancing the search capability of GOA. It can be represented as follows:

$$C = C_{max} - T \frac{C_{max} - C_{min}}{T_{max}}$$
(23)

where, C_{max} , C_{min} are the maximum and the minimum values of *C*, respectively. *T* and T_{max} are the current iterations and the maximum number of iterations, respectively. However, GOA algorithm can be summarized as follows:

Step 1 : Determine the input data of GOA including number of the search agents (N), maximum number of iterations, C_{min} , C_{max} , F, l and the upper and lower boundaries of control variables.

Step 2 : Initialize the population of GOA as follows:

$$P_i^m = Lp(i,m) + rand * (Up(i,m) - Lp(i,m))$$
⁽²⁴⁾

Step 3 : Calculate the fitness functions for each search agent.

Step 4 : Determine the best position (target position) in term of the best fitness function.

Step 5 : Update the position of search agent according to (21).

Step 6 : Check the boundaries of the updated agents and bring the violated variable to accepted limit.

Step 7 : Calculate the fitness function for the updated positions and determine the target position.

Step 8 : Repeat steps form (5) to (7) until the stopping criterion is achieved (current iteration equals to maximum iteration).

Step 9 : Find the best solution (target position) and the related fitness function.

IV. SOLUTION PROCESS OF OPTIMAL CAPACITOR PLACEMENT PROBLEM USING GOA

The sizes and location of capacitors are considered as decision variables and used to form different objective functions. The implementing procedure of the developed GOA in solving the optimal capacitor placement problem can be summarized as follows:

Step 1 : Read the line and bus data.

Step 2 Initialize the maximum number of iterations, the number of capacitors, the number of the search agents (N), C_{min} , C_{max} , F, l and the upper and lower boundaries of control variables.

Step 3 : Select positions of capacitors randomly depending upon the capacitor numbers.

Step 4 : Initialize randomly the size of the capacitors within the operating constraints as follows:

$$O_i = O_i^{\min} + rand * (O_i^{\max} - O_i^{\min})$$

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(20)

Step 5 Run the load flow to find the objective function (power losses or the total cost) then determine the best position (best locations and sizes of capacitors) in term of the best objective function.

Step 6 : Normalize the distances between grasshoppers

Step 7 : Update the position of search agent according to (21). In other words the locations and sizes of

capacitor are updated with respected to the best solution.

Step 8 If the stopping criterion is satisfied, stop and print the best solution, else go to Step 5.

However, the above steps can be summarized in the following flowchart.

Start Set Grasshopper algorithm Parameters Set System Data Generate initial populations Calculate the fitness for each search agent and determine the best solution Set $T = \theta$ <mark>00 = 00,000 - 00,000 - 00,000</mark> Update C value The Normalize the distances between search agents Ш Update the position of search agent m T = T + IRun load flow & Calculate the fitness for each search agent, then determine the best solution No Is $T > Tmax^2$ Yes Find the best so lutio n and its current position End

Fig. 6 Flowchart of developed Grasshopper optimization algorithm for optimal capacitor placement in radial distribution systems

V. NUMERICAL RESULTS

The developed GOA approach has been applied on different distribution radial test systems to observe its effectiveness and performance. Power loss and total cost are minimized for three standard test systems (33-bus, 69-bus and 85-bus radial distribution systems). The developed approach has been modelled and implemented using the MATLAB 7.14. All case studies are achieved using a personal computer having 2.5 GHz core i5 processor with 4 GB RAM. The number of compensated buses selected based on the size of test system.

The selected parameters of GOA algorithm are listed in Table 1. The available sizes of capacitors are given in Table 2. In all the mentioned test systems, GOA algorithm is performed 50 times and the best obtained results are reported and compared with other meta- heuristics optimization techniques.

Parameter			T_{max} C_{max}		C_{min}	F	L	
Value			100	1	0.00004	0.5	1.5	
		Тя	hle 2 The av	ailable canaci	tor sizes (KVAR)			
150	200			1	tor sizes (KVAR).	1200	1250	
150	300	450 Ta	ble 2 The av 600	vailable capaci 750	tor sizes (KVAR). 900 1050	1200	1350	
150 1500	300 1650			1		1200 2550	1350 2700	

 Table 1 The selected parameters of GOA.

5.1 Power loss minimization.

In first case study, the developed GOA technique is tested to minimize the power loss of different standard test systems as 33-bus, 69-bus and 85-bus radial distribution systems. The results obtained based on GOA are compared with that obtained from different optimization methods.

5.1.1 33-bus test system

To demonstrate the impact of developed GOA on medium scale of radial distribution system, 33-bus test system used in the first case study. The single line diagram of this test system given Fig. 7. The system voltage is 12.66 KV. However, the details of this system are given in [29]. In case of the system is working without capacitors, the minimum voltage magnitude is 0.9036 p.u at bus no. 18 and the total active power loss is 210.97 KW with annual energy losses cost about 35442.96 \$. Based on the developed GOA, only three capacitors are allocated at optimal locations. Table 3 shows the power loss, minimum bus voltage, optimal locations and optimal size of capacitors obtained by different optimization techniques. From this table, it can be observed that installation of capacitors reduces the power loss significantly. However, the active power loss using GOA is reduced to be 138.772 KW. Consequentially, the total losses are reduced by 33.008 KW compared to Analytical interior point IP [30], 5.268 KW compared to GA [13], 2.468 KW compared to Fuzzy-Real Coded GA FRCGA [31], and 798 W compared to Differential Evolution Algorithm DEA [32], 1.648 KW compared to Gray Wolf Optimizer GWO. In addition, the system voltage profile is improved significantly in case on including the optimal capacitor placement compared with uncompensated case as shown in Fig. 8. This shows that the developed GOA is more effective than the other optimization techniques. Referring to Fig.9 it can obvious that the best objective function is obtained rapidly at the first 10th iterations. On other hand, the objective function with GWO reach to the best solution after 50th iterations



Fig. 7 Single line diagram of 33-bus radial distribution system.

	W/O	With Capacitor											
	Capacitor	IP	[30]	G	A [13]		CGA 31]	G	WO	DEA[32]		GOA	
Optimal location & size of capacitors		9	450	7	850	6 8	475 175	12	450	5	NA	12	450
(KVAR)	-	29	800	29	25	9 28	350 25	4	300	27	NA	24	600
		30	900	30	900	29 30	300 400	30	1050	28	NA	30	900
Total KVAR	-	2	150		1775	1	725	1	800	Ν	IA	1	950

Table 3 Optimal re	results of 33-bus network	for power loss minimization
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Power loss kw	210.97	171.78	144.04	141.24	140.42	139.57	138.772
% reduction in power loss	0	18.576	31.72	33.05	33.44	33.84	34.22
Minimum voltage bus	18	18	NA	NA	18	18	18
Minimum voltage p.u.	0.9038	0.9501	NA	NA	0.9307	0.93	0.9295



Fig. 8 Voltage profile of 33-bus radial distribution system.



Fig. 9 Convergence characteristics of the objective function for 33-bus radial distribution system

5.1.2 69-bus test system

The developed GOA also validated using standard 69-bus radial distribution system, in case of taking the power loss minimization as an objective function. This test system is consisting of 69 buses and 68 branches as shown in Fig. 10. Its rated voltage is 12.66 kV and total system load is (1.896MW+j1.347MVAR). The details bus and line data are reported in [17]. The obtained KW loss without incorporating any capacitor equals to 224.96 KW and the lowest voltage is 0.9092 p.u. at bus no. 65. The optimal locations and sizes of capacitors with applying GOA are listed in Table 4. However, when applying the developed technique, the best active power loss is reduced to 145.404 KW. Meanwhile the percentage of power loss reduction is enhanced to be 35.37% which is the best value compared with other optimization algorithms as reported in Table 4. Moreover, the voltage profile is enhanced significantly with determining the optimal location and size of capacitors using GOA as shown in Fig. 11. The minimum voltage improved and became 0.9324 p.u which is compatible with the voltage constrains. However, the convergence characteristics for this case compared to GWO is shown in Fig.12.



Fig. 10 Single line diagram of 69-bus radial distribution system.

	W/O							With (Capacitor						
	Capacito r		[30]		two [33]	CS	A[34]		10EA 21]	TLB	O[35]	G	WO	C	δΟΑ
Optimal		11	900	19	225	62	1200	61	1150	12	600	16	450	12	450
location & size (KVAR) of	-	29	1050	62	900	21	250	18	250	61	1050	61	120 0	61	1200
capacitors		30	450	63	225					64	150	9	300	21	150
Total KVAR	-	2	400	1.	350	1450		1	400	1800		1950		1800	
Power loss KW	224.96	16	53.28	148.91		148		147.74		146.36		146.118		145.4047	
% reduction in power loss	0	2	7.42	3	3.8	3	34.21 34.33		4.33	34.945		35.05		35.37	
Minimum voltage bus	65		65		65		65		65		65	(65		65
Minimum voltage p.u. 0.9092		0.	9532	0.9	9288	().93	0.	9288	0.9	9312	0.9	313	0.	9308

Table 4 Optimal results of 69-bus network for power loss minimization







Fig. 12 Convergence characteristics of the objective function for 69-bus radial distribution system

5.1.3 85-bus test system

To investigate the effectiveness of the developed GOA on large scale radial distribution systems, 85bus test system is used. This system is operated with 100 MVA base and 11 KV rated voltage. Its single line diagram is shown in Fig. 13. All data of lines and loads are given in [36]. The total active power loss without incorporating capacitors in the system is 315.714 KW. The optimal results that obtained by GOA and other algorithms including optimal locations and sizes of capacitors are listed in Table 5. The active power loss is reduced to 148.9274 KW with percentage reduction of 52.83% with applying GOA. It should point out that the lowest bus voltage is increased from 0.8713 p.u. to 0.92182 p.u. Moreover, the voltage profile is significantly improved as shown in Fig. 14. Referring to Fig. 15, the GOA is performing well in stable and smooth convergence characteristics.



Fig. 13 Single line diagram of 85-bus radial distribution system.

With Capacitor W/O Capacitor PGS [1] PSO [37] MINLP BOFA [39] GWO GOA [38] Optimal location and 200 324 300 9 840 48 450 67 600 7 7 7 8 size (KVAR) 8 1200 8 796 700 34 660 9 900 34 600 of capacitors 26 58 908 27 901 29 900 60 650 68 450 600 58 453 58 500 29 450 12 450 Total KVAR 2308 2474 2400 2150 2250 2550 Power loss KW 315.714 174.0048 163.32 159.87 152.25 149.2728 148.9274 % reduction in power 44.82 52.72 52.8288 _ 48.21 49.03 51.71 loss Minimum voltage bus 54 54 54 54 54 54 54 Minimum voltage p.u. 0.92182



Fig. 14 Voltage profile of 85-bus radial distribution system



Fig. 15 Convergence characteristics of the objective function for 85-bus radial distribution system.

 Table 5 Optimal results of 85-bus network for power loss minimization

Energy cost optimization.

To clarify the effectiveness of the developed GOA for minimizing the annual cost of real power loss and total capacitors, simulation studies carried out on standard 69-bus and 85-bus bus test systems. In this section, three different load demand patterns of light load 50%, nominal load (100%) and peak load (160%) are considered to show the performance of the developed technique. Duration of time for light load, nominal load and peak load are 2000 h, 5260 h and 1500 h, respectively [14]. Here, energy cost is assumed as US \$ 0.06 per KWh and purchase cost of capacitor is taken as US \$ 3.0 per KVAR [14].

5.2.1 69-bus system.

5.2

The developed GOA is applied on 69-bus system to determine the optimal location and size of capacitors to minimize the total energy cost. In case of the test system does not consist of compensation units, the active power loss for peak, nominal and light load are 652.40 KW; 224.96 KW, and 51.60 KW, respectively, while the minimum voltage level are 0.84449 p.u., 0.90919 p.u., and 0.95668 p.u, respectively. The total cost equals to 135,905\$ [35]. The effect of optimal location and size of capacitors on the energy loss cost, capacitor cost and the total cost of the system are given in Table 6. Referring to the simulation results of Table 6, it is clear that the obtained results using GOA are better than the other reported algorithms such as the fuzzy GA [40], DSA [14], GWO and TLBO [35]. Due to the installing the capacitors in three locations the objective function decreased with respect to GWO as displayed in Fig.16.

Table 6 Simulation results using fuzzy GA, DSA, TLBO and GOA for cost analysis (69-bus system)

	W/O	-				W	ith Capaci	tor			
	Capacitor		zy GA 40]	DS.	A [14]	G	WO	TLBO	D[35]	G	юA
Light load											
Optimal location & size (KVAR) of capacitors	-	59 61 64	0 0 300	15 60 61	300 300 450	12 61 18	150 750 150	22 61 62	150 450 150	18 61 64	150 450 150
Minimum voltage bus	65		65		65 65		6	5	65		
Minimum voltage p.u. Power loss KW	0.95668 51.6		0.9622 40.48		0.9683 35.52		9688 4.87	0.9662 34.43		0.9666 34.36	
Nominal load											
Optimal allocation and size (KVAR)	-	59 61 64	100 700 800	15 60 61	450 450 900	12 61 18	150 1200 300	22 61 62	300 1050 300	18 61 64	300 1200 150
Minimum voltage bus	65		65		65		65	6	5		65
Minimum voltage p.u.	0.90919	0.9	3693	0.	9318	0.9303 145.776		0.9321 146.8		0.9325 146.45	
Power loss KW	224.96	15	6.52		147						
Peak load											
Optimal allocation and size (KVAR)	-	59 61 64	1100 800 1200	15 60 61	900 900 1800	12 61 18	150 1500 300	22 61 62	300 1050 750	18 61 64	300 1500 300
Minimum voltage bus	65		65		65		65	6	5		65
Minimum voltage p.u.	0.84449		0014		8936		8748	0.87			8803
Power loss KW	652.40		50.5		27.3		26.15	417			6.32
Energy loss cost (\$)	135,905	95727	.00	8911	2.6	8854	4.37	88016.3	37	8781	3.82
Capacitor cost (\$)	0	9.	300	10	0800	5	850	63	00	6	300
Total cost (\$)	135,905	105,02	27	99,91	2.6	9439	4.74	94316.	37	94113	.82



Fig. 16 Convergence characteristics of the cost analysis for 69-bus radial distribution system.

5.2.2 85-bus system.

Finally, the developed GOA is applied on 85-bus test system to find the optimal location and size of capacitors for minimizing the total cost. Without incorporating of compensation units, the minimum voltage level and the power loss of the system for peak, nominal and light load are 0.7722 p.u., 975.93 KW; 0.8713 p.u., 316.11 KW; and0.9397 p.u., 70.11 KW, respectively, and The total cost equals to 196,011\$ [41]. The effect of the optimal location and size of capacitors on the energy loss cost, capacitor cost and the total cost of the system are listed in Table 7. The obtained cost using GOA technique is better than DSA [14], GWO and TLBO [35]. The total cost is decreased to optimal value as shown in Fig.17

Table 7 Simulation results using DSA, TLBO and GOA for cost analysis (85-bus system)

	W/O				With C	Capacitor			
	Capacitor	DSA[14]	GW	0	TLB	O[35]	G	OA
Light load									
Optimal location and size		6	0	35	0	15	0	21	150
(KVAR) of capacitors		8	0	70	0	23	150	26	0
		14	150	52	150	26	150	27	150
		17	150	32	0	32	150	37	0
		18	0	4	0	36	150	39	0
		20	0	11	150	38	0	46	150
		26	0	73	150	45	0	53	150
	-	30	150	63	0	52	0	56	150
		36	300	15	150	57	150	62	0
		57	150	8	150	61	0	65	150
		61	0	31	150	64	150	67	0
		66	150	27	150	73	150	70	150
		69	150	49	0	82	0	80	150
		80	0	19	150	84	150	81	0
Minimum voltage bus	54	54		54	Ļ	4	54	-	54
Minimum voltage p.u.	0.9397	0.96	29	0.96	26	0.9	616	0.9	663
Power loss kw	70.11	34.7	6	34.48		34.11		35.7186	
Normal load									
Optimal allocation and size		6	150	35	0	15	150	21	300
(KVAR)		8	150	70	150	23	300	26	150
		14	150	52	150	26	300	27	150
		17	150	32	150	32	150	37	150
		18	150	4	150	36	150	39	150
		20	150	11	450	38	150	46	150
		26	150	73	150	45	150	53	300
		30	300	63	450	52	150	56	0
		36	450	15	0	57	300	62	150
		57	150	8	0	61	150	65	150
		61	150	31	150	64	300	67	150
		66	150	27	300	73	150	70	150
		69	300	49	150	82	150	80	300

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		80	150	19	150	84	0	81	150		
Minimum voltage bus	54	54		54	Ļ	5	54	5	4		
Minimum voltage p.u.	0.8713	0.9224		0.9213			241	0.9229			
Power loss kw	316.11	144.01		144.464		143.	.2493	145.	0935		
Peak load											
Optimal allocation and size	-	6	150	35	300	15	150	21	300		
(KVAR)		8	300	70	300	23	0	26	150		
		14	150	52	150	26	300	27	150		
		17	150	32	300	32	450	37	150		
		18	300	4	150	36	150	39	300		
		20	300	11	300	38	150	46	150		
		26	150	73	150	45	150	53	300		
		30	450	63	450	52	300	56	300		
		36	900	15	300	57	300	62	300		
		57	300	8	300	61	150	65	300		
		61	300	31	300	64	450	67	150		
		66	300	27	300	73	300	70	300		
		69	600	49	150	82	450	80	300		
		80	450	19	0	84	0	81	150		
Minimum voltage bus	54	54		54			54		4		
Minimum voltage p.u.	0.7722	0.87		0.8			647		362		
Power loss KW	975.93	410.6		411.			1.6		9425		
Energy loss cost (\$)	196,011	8658		872			346		513		
Capacitor cost (\$)	0	1395		103			250		00		
Total cost (\$)	196,011	10053	36	976	44	97	590	974	413		



Fig. 17 Convergence characteristics of the cost analysis for 85-bus radial distribution system.

VI. CONCLUSION

This paper has applied one of the recently developed optimization techniques (grasshopper optimization algorithm (GOA)), in radial distribution systems to solve the problem of capacitors placement. Minimizing the total real power losses and annual energy cost are taken as objective functions which are considered an attractive economic issue. The developed GOA has been successfully applied on several standard distribution systems to prove its superiority and effectively compared with different well-known optimization techniques. IEEE 33-node, IEEE 69-node and IEEE 85-node test systems have been selected as small, medium and large scale radial distribution network, respectively. In all the mentioned test systems, the locations and sizes of capacitors are optimized to achieve the determined objective functions using the developed GOA algorithm. Furthermore, the nodes voltage has been improved. However, all numerical results obtained have been compared with other well-known optimization algorithms. GOA presents a favorable and promising performance with stable convergence over the other algorithms.

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Abdel-Raheem Youssef, etal. "Optimal Capacitor Placement in Radial Distribution Systems Using Grasshopper Optimization for Power Loss and Energy Cost Minimization". *American Journal of Engineering Research (AJER)*, vol. 9(01), 2020, pp. 247-263.

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