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# Critical Review of Different Methods for Siting and Sizing Distributed-generators

**Shomefun TS\*, Ademola A, Awosope COA, Adekitan AI**

Department of Electrical and Information Engineering, Covenant University, Canaan land, KM 10, Idiroko, Road, P.M.B. 1023, Ota, Ogun State, Nigeria

\*Corresponding author, e-mail: tobi.shomefun@covenantuniversity.edu.ng

## Abstract

*Due to several benefits attached to distributed generators such as reduction in line losses, improved voltage profile, reliable system etc., the study on how to optimally site and size distributed generators has been on the increase for more than two decades. This has propelled several researchers to explore various scientific and engineering powerful simulation tools, valid and reliable scientific methods like analytical, meta-heuristic and hybrid methods to optimally place and size distributed generator(s) for optimal benefits. This study gives a critical review of different methods used in siting and sizing distributed generators alongside their results, test systems and gaps in literature.*

**Keywords:** distributed generator; analytical; meta-heuristic; hybrid; site & size

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## 1. Introduction

Electric power system (EPS) is one of the most complex conceptions by mankind, and it is a non-linear system. Aside this, its construction and operation are very complex and complicated because of several factors and constraints that must be considered in terms of location, type, available resources, etc. The purpose of Electrical Power System (EPS) is to generate and supply electrical energy to users [1]. It comprises generation station, transmission network, distribution network and load centres. The load centres receive and consume generated power by the generation stations via the link of transmission and distribution networks. However, in a deregulated electricity market, congestion on transmission lines maybe unavoidable because of insufficient capacity of lines [2]. Moreso, under voltages and over voltages in the lines lead to poor power quality and lack of stable power system [3]. Not with standing, power engineers during planning stage, give a margin or forecast to accommodate future load demand on the network, however, development brings about increase in the load demand which will outgrow the specified margin at some points. Hence, there will be need for expansion when load demand equals or greater than the supply power from generation stations.

On the contrary, construction of a new generation station requires a huge capital. This has propelled several researchers to investigate alternative means to offset overshoot in load demand against the supplied power from the generation stations. One major solution discovered was the installation of distributed generator (DG) close to the load centres. Though there are some other solutions, DG gives the best option to overcome load demand, economic and environmental challenges [4] among other methods such as FACTS devices for power system improvement, network reconfiguration, capacitor compensation, static VAR etc. [5-16]. Review of work done with distributed generators, will be the area of focus in this work.

## 2. What then is Distributed Generator?

Several researchers have put forward diverse definitions of distributed generator as given in [4], [6], [17-26]. According to CIRED (1999), there is no consensus yet on the generally acceptable definition of DG. However, the definition given by T. Ackermann et al.

in [17] will suffice for this work. It is defined as “the electric power generation source linked directly to distribution network or the meter side of customer”. Other criteria for which DG can be classified are given in Figure 1 [27]. Unlike conventional Central Generation (CG), distributed generation is not location bound (as the name implies). Table 1 gives comparison between CG and DG.

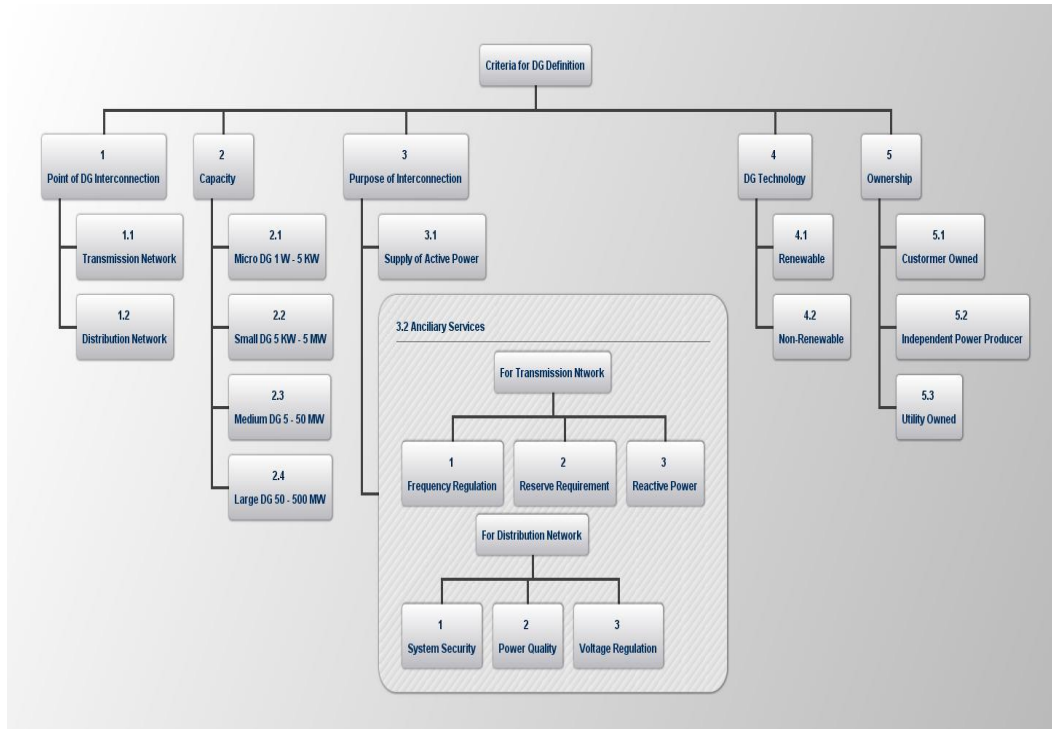


Figure 1. Criteria for DG classification

Table 1. Comparison between Central Generation and Distributed Generation

S/N	Central Generation	Distributed Generation
1	Centrally located	It is not location bound. It is distributed
2	Specific site of installation	It can be installed anywhere in which energy source is present
3	Excellent economies of scale	Small-scale power generation technologies (in the range of 1KW to 500MW)
4	Transmits electricity over a long distance	Transmits electricity over a short distance
5	Negatively affects the environment	Environmental friendly
6	It is part of the grid	It can be isolated or integrated into the grid
7	Basically, gas and hydro turbine	The technologies adopted in DG comprise small gas turbines, micro-turbines, fuel cells, wind and solar energy, biomass, small hydro-power etc.

### 3. Significance of Distributed Generator

According to the IEA (2002), there are five major factors that contribute to the advancement in distributed generation namely; developments in distributed generation technologies, constraints on the construction of new transmission lines, increased customer demand for highly reliable electricity, the electricity market liberalization and concerns about climate change. However, these factors can be summarized under these two key issues: cost effectiveness of distributed generation and friendly environmental impact. This is because DG technology is being developed (not explored) continuously because it is cost effective. Also, DG saves a huge amount of money that would have been budgeted for transmission lines. Meanwhile, consumers can only demand for a

reliable system that is affordable. However, no investors will like to venture into any business that is not profit oriented. And lastly, advent of renewable energy sources (RES) technology, which is free from greenhouse gas emission, mitigates concerns about climate change since they are environment friendly and readily available by nature. Hence, it can be said that Installation of distributed generation permits the utilization of freely available fuel opportunities [6].

Installation of DG is a short-time project and it is a less expensive alternative for electric power system expansion compared to construction of a new generation station [24, 27]. Employing this method will not only help to meet load demand but also, improve voltage profile, increase the system reliability level [27], minimize Total Harmonic Distortion (THD) [28], minimize cost of electricity [29], lower short-circuit level [30], relieve transmission and distribution congestion and minimize line losses. This is because it is located closer to the point of consumption than the main source for the distribution network [31].

However, with so much positive impacts which DG adds to electric power system, it must be strategically and optimally located to achieve the intended results [24]. It must also be properly and optimally sized to avoid excessive generation cost, increase in the power loss, and bus voltage fluctuating in and out of the statutory limit [18, 32, 33].

Several researchers have used various methods to site and size distributed generator(s) ranging from analytical methods to hybrid-based optimization methods. Some of these methods are Gradient and second-order method, Hereford Ranch algorithm, Heuristic iterative method, Analytical based on 2/3 rule, Tabu search, Hybrid fuzzy nonlinear goal programming, Heuristic iterative search method, Linear programming, Sensitivity analysis, Hybrid e-constraint-based multi-objective programming, Optimal power flow, GA, Mixed integer non-linear programming, Iterative search technique with load flow [34]. The details of the various methodologies that have been deployed to date are as presented in Table 2. Techniques for DG placement differ, and they are dependent on the objectives to be achieved. These techniques have their strengths and drawbacks. Table 3 gives detailed comparison of these techniques.

Table 2. Different Methods Used in Siting and Sizing DG with Test System, Result and Observed Gaps

S/ N	Methodology	Test System	Result	Gaps	Reference
A 1	ANALYTIC Analytical method based on exact loss formula	Nigerian 33-kV network	6.2% reduction in active power losses on the 33-kV Nigerian network (i.e. from 92.7MW to 87.0MW). The results showed an improvement in the voltage profile of the six load buses whose voltages were outside the statutory limit of between 0.95 pu and 1.05pu	The work required that DG should be installed at each bus. The methods cannot optimally place DGs.	J. N. Nweke, A. O. Ekwue and E. C. Ejiogu [1]
2	Loss sensitivity factor based on current injection method	12-, 34- and 69-bus distribution test systems	Proposed method (i.e. Current injection) alongside with Acharya's method and the Classical grid search algorithm were compared. The results for the proposed method and Acharya's method are almost the same for optimum sizes and estimated power losses. Based on the test systems, the optimal locations for DG are busses 9, 21 and 61 respectively. The proposed method is 1.5 times faster than the Acharya's method. However, Classical grid search algorithm gives a worse result.	It did not consider different types of load models in the analysis.	Tuba Gözel, M. Hakan Hocaoglu [17]

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S/ N	Methodology	Test System	Result	Gaps	Reference
3	N-R method for load flow study	IEEE 6-bus, IEEE 14-bus and IEEE 30-bus systems	For 6-, 14-, and 30-bus systems considered, a minimum value of Optimal Flow (OF) is obtained when DG is placed at bus nos. 3(6MW), 8(16MW) and 11(35MW) respectively.	It did not specify the type of DG technology. Likewise, it did not consider different types of load models in the analysis.	Sudipta Ghosh, S.P. Ghoshal, Saradindu Ghosh [18]
4	Power flow algorithm	13-bus radial system	Optimal size and placement of the theoretical analysis are valid for constant power, current and impedance load models. It is found that the optimum location does not change with the chosen model.	DG technology used was not specified, method is limited to non-complex network, hence single DG placement.	T. Gozel, M. H. Hocaoglu, U. Eminoglu, and A. Balikci [19]
5	Second-order power flow sensitivities,	70-bus distribution system	The results show that the total power losses are dependent on the DG location	The method cannot randomly find optimal location for DG	Hugo M. Ayres and Walmir Freitas [20]
6	Exact loss formula	30-bus, 32-branch loop system, 33-bus, 32-branch and 69-bus, 68-branch radial systems	The result gives 59.6%, 47.3% and 62.8% reductions in the active power loss of the three test systems respectively.	It did not specify the type of DG technology used and types of load models considered in the analysis.	Naresh Acharya, Pukar Mahat, N. Mithulana nthan [21]
7	Method based on load flow	IEEE 13- and 37-bus distribution test systems	In the 13 & 37-bus distribution test systems, the optimum sizes range from 1 to 1.7 MW & 0.7 to 10 respectively.	It did not specify the type of DG technology used and types of load models considered in the analysis.	P. Alemi and G.B. Gharehpetian [22]
8	Second-order algorithm	Six-bus 25-kV distribution network with lines ranging in length from 16 to 32 km	The total injection is maximally distributed to nodes 2, 3,4 and 6 for loss minimization.	It did not specify the type of DG technology used and types of load models considered in the analysis.	Narayan S. Rau, SM Yih-heui Wan, M [23]

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S/ N	Methodology	Test System	Result	Gaps	Reference
B 9	META-HEURISTIC Differential Evolution	IEEE 33-bus radial system consisting of 32 sections.	System losses reduced by 47% for the installation of one DG. The nodes violating the voltage limits reduced to 3 from 18 and the sum of squares of voltage error dropped from 0.1369 p.u. to 0.02968 p.u.	It did not consider different types of loads in the analysis. Single DG was considered	M. Abbagana, G. A. Bakare, and I. Mustapha [4]
10	Discrete particle swarm optimization (DPSO) technique	69-bus radial distribution test network	Proposed DPSO yields better result compared to analytical methods, GA and ABC.	It did not consider different types of loads in the analysis.	Idris Musa Shady Gadoue, and Bashar Zahawi [24]
11	Improved Genetic Algorithm (IGA), DG integration	Model distribution network	IGA helps to improve the network reconfiguration by reducing total non-restored loads from 488A (with 118.8kW losses) to 151A (with 311.1kW losses). Upon integration of DG, the total non-restored loads further reduced to 68A.	It did not use standard test system. However, the method employed did not reduce losses on the line as the lost loads were restored, but rather increased it.	M. Shahrin A. H. et al. [26]
12	Ant colony optimization (ACO) algorithm implemented in the hyper cube (HC) framework and random search musician-behavior-inspired evolutionary algorithm, harmony search (HS)	32-bus and 69-bus distribution systems	Both methods are viable in the sense that they give a configuration with minimal losses.	The simulation did not consider optimal siting or sizing of DG	ALMOATA Z Y. ABDELAZI Z, et al. [6]
13	Particle swarm algorithm based on two-dimensional depth-coded	IEEE69-bus	Proposed method has fast convergence with restoration of all outages	It did not consider optimal location of DGs for minimal loss and voltage improvement	Chen Dan et al. [27]
14	Teaching and Learning-Based Optimization (TLBO)	IEEE 33-bus and IEEE 69-bus radial distribution systems	TLBO was compared with GA and PSO. TLBO gave results with minimal loss, higher voltage profile and higher DG size.	It did not specify the type of DG technology used and types of load models considered in the analysis.	Phanindra k.G. and Chintham V. [28]

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S/ N	Methodology	Test System	Result	Gaps	Reference
15	Backtracking Search Optimization Algorithm (B.S.A)	IEEE 69-bus test system	B.S.A was compared with Harmony Search Algorithm (H.S.A) and Artificial Bee Colony Algorithm (A.B.C) and the results showed that B.S.A had the least loss and the most improved voltage	It did not consider network restoration and reliability indices of the network	Vivek Gupta, Sudha Rani Donepudi, N. Subrahmanyam [29]
16	Goal attainment method	IEEE 34-bus test system	The proposed algorithm calculates the reactive power injections by the DG and the reactive power compensation devices, such that not only the system bus voltages are regulated, but also reduces the total power loss.	It did not consider optimal location and size of DGs. It did also consider harmonic effect of the DGs	Vahid Asgharian, V.M. Istemihan Genc [30]
17	Particle Swarm Optimization (PSO)	38-bus radial system and an IEEE 30-bus meshed system	For 38-bus system, the reduction in the active power loss was in the range of 54–67%. The reduction in the reactive power loss was in the range of 58–67%. The reduction in the total MVA intake was about 30%. For the 30-bus system, the reduction in the active power loss was in the range of 30–37%. The reduction in the reactive power loss was in the range of 26–31%. The reduction in the total MVA intake was about 62%.	The author limits the number of DGs to three before simulation was carried out	A.M. El-Zonkoly [31]
18	Particle Swarm Optimization	12-bus, 30-bus and 69-bus test systems	The proposed method converged for all the cases observed in each test system. The optimal number of DGs for the test systems is two.	When two DGs were considered, the optimal location remained the same as the case of single DG. It did not specify the type of DG technology used.	Naveen Jain, S.N. Singh, and S.C. Srivastava [32]
C 19	HYBRID Voltage stability index-based method and Particle Swarm Optimization	A 30-bus test system and a 41-bus Indian distribution system	It found optimal point between benefit from DG placement and DG sizing for minimal loss and improved voltage profile.	It did not provide justification for the factor used.	Naveen Jain [33]
20	Fuzzy adaptive hybrid particle swarm optimization (FAHPSO) method	Modified version of the IEEE 33-node distribution system.	FAHPSO searched better schedule for the studied distribution network using fewer evolution cycles, compared with HPSO method.	Thermal effect was ignored	Shuheng Chen et al. [34]

Table 2. Different Methods Used in Siting and Sizing DG with Test System, Result and Observed Gaps

S/ N	Methodology	Test System	Result	Gaps	Reference
21	GA, Power flow and Analytical methods	Brazilian actual distribution system	In terms of power flow, the maximum voltage drop was reduced to 0.022404 pu (3%) and the losses became 85.4 kW (83.74%) after DG allocation.	It did not consider optimal location and sizing of DG. It is only limited to a location and some specified values of DG capacity.	Carmen L.T. Borges, Djalma M. Falcao [35]
22	Sensitivity-based approach and Modified Primal-Dual Interior Point Algorithm (MPDIPA)	IEEE 123-node test feeder	The proposed approach i.e. MPDIPA searches for the optimal solutions quickly with improvement in the voltage profile and obvious reduction in network loss.	It did not specify the type of DG technology likewise, it did not consider different types of loads model in the analysis.	Zhipeng Liu, Fushuan Wen, Gerard Ledwich and Xingquan Ji [36]
23	Rank Evolutionary Particle Swarm Optimization (REPSO), Evolutionary Particle Swarm Optimization (EPSO), and Traditional Particle Swarm Optimization (PSO) method	69-bus radial distribution system	REPSO converges faster than EPSO and PSO and gives a better standard deviation. There is 47% loss reduction	The method is limited to optimisation of already placed or installed DG.	J.J. Jamian et al. [25]
24	Mixed integer quadratic constraint programming (MIQCP) model and information gap decision theory (IGDT) using Robust Restoration Optimisation(RRO) and Determined Restoration Optimisation(DRO)	Modified Pacific Gas and Electric Company (PG&E) 69-node distribution network	RRO offers stable automatic strategies which meet the essential of self-healing that is absent in DRO	The method is limited to optimisation of already sized and placed or installed DG. This means that if the DG(s) were wrongly placed, this method cannot identify or correct the error.	Kening Chen et al. [27]
25	Cat-Swarm-Optimization (CSO) and composite reliability index, AWPSO, PSO-CF	34-bus radial test distribution system and IEEE 69 bus radial test distribution system	CSO yields better performance as compared to AWPSO and PSO-CF in terms of active power loss reduction, power transfer capacity and computational time	It did not consider optimal sizing of DGs and bus voltage limit	Deepak Kumar and S. R. Samantaray [37]



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S/N	Methodology	Test System	Result	Gaps	Reference
26	Non-dominated sorting genetic algorithm (NSGA-II) and fuzzy decision-making analysis	IEEE 69-bus test system	The application of network reconfiguration, shunt capacitors and distributed generators altogether on distribution system enhancement yields better system performance when compared to the application of only one or two options.	The number, size, and set of locations of capacitors and DGs are limited to pre-defined value	Russel John C. Gallano and Allan C. Nerves [8]
27	Ranked Evolutionary particle swarm optimization (REPSO)	IEEE 69-bus test system	The result indicates that as the number of DG units installed increases, the power loss reduction also increases and voltage profile increases.	The method used to locate DG placement can be localised. The analysis did not consider load variation	Haruna Musa, Sanusi Sani Adamu [38]
28	Fuzzy adaptive hybrid particle swarm optimization (FAHPSO) method	EEE 33-node distribution system with two newly installed distributed generators and eight newly installed capacitors banks.	FAHPSO gives better convergence and search schedule for the studied distribution network using fewer evolution cycles, compared with HPSO method.	It did not consider optimal location of DGs for minimal loss	Shuheng Chen et al. [34]
29	Fuzzy-GA method	=12-bus sample systems	Aside improvement in voltage profile, the result also shows that there will be \$7,554.5 saving costs in 10 years	It did not specify the type of DG technology used and types of loads model considered in the analysis.	Kyu-Ho Kim, Yu-Jeong Lee and Sang-Bong Rhee, Sang-Kuen Lee and Seok-Ku You [39]

Table 3. Comparison of Different Methods Used

S/N	Optimization method	Benefits	Drawback
1	Analytical	<ul style="list-style-type: none"> <li>• Computing time is short</li> <li>• Easy to implement</li> <li>• Non-iterative in nature</li> <li>• Unlike other techniques, does not pose convergence problems</li> </ul>	<ul style="list-style-type: none"> <li>• When problem becomes complex, assumptions used in order to simplify problem may override accuracy of solution</li> <li>• Lacks robustness</li> </ul>
2	Meta-Heuristic	<ul style="list-style-type: none"> <li>• Can rapidly locate solutions, even for large search space</li> <li>• Works with discrete and continuous parameters</li> <li>• Bad proposals do not affect end solution negatively</li> </ul>	<ul style="list-style-type: none"> <li>• Repeated fitness function evaluation for large and complex problems may be time consuming</li> <li>• May not suggest best solution always, possibility of trapping into local optima</li> <li>• Lack of accuracy, not suitable when a high-quality solution is required</li> </ul>
3	Hybrid	<ul style="list-style-type: none"> <li>• Very useful for complex problems</li> <li>• Higher efficiency</li> <li>• Higher possibility of global optima</li> <li>• Less computational time</li> </ul>	<ul style="list-style-type: none"> <li>• Increased complexity</li> </ul>

#### 4. Discussion

Although several researchers have considered integration of distributed generation into EPS as an alternative to construction of centralised generation station, some prevailing research problems, which require more investigations, are still open and they are listed as follows:

- a. One major prevailing problem in the planning of power system to incorporate DGs is to take into account various factors such as nature of DG technology, impact of DG on operating characteristics of power system and economic considerations [40].
- b. Another problem of integrating DG into the grid is islanding issue for which IEEE 1547 standard [41] was established: a criterion for interconnection of DG sources. The present standards do not allow islanded operation of DG [42]
- c. The possibility of reliability enhancement with increased penetration of RES-based DGs is another prevailing problem and it has also not been investigated. Likewise, the reliability assessment studies during islanded mode, incorporating RES-based DGs and storage has not been reported in literature. [40]
- d. DGs integration impact on system reliability, line losses, emissions, voltage profile and cost for an optimum system planning [40].

#### 5. Conclusion

This study gave a critical, comprehensive and systematic survey of the existing methods for integrating DG(s) into EPS in order to mitigate continuous increase in load demand. Three categories of optimisation techniques i.e. analytical, meta-heuristic and hybrid optimisation methods were considered. This categorization, as well as the representative techniques described under each category, will benefit optimisation techniques' researchers for choosing from proper state-of-the-art population-initialization-based techniques for their research. The volume of the surveyed techniques revealed that optimisation techniques have become an active research topic in electrical power system domain. However, some questions are yet to be resolved. Some of these questions were highlighted for future investigation.

Based on the reviewed literature, this study also gave a review of different optimisation methodologies for siting and sizing distributed generators in a distribution network. The test systems/networks as well as results obtained from these methods were also recorded. The observed gaps in the reviewed literature were also provided and finally, the strengths and weaknesses of the available methods were also included. However, most of the previous works were carried out on conventional DGs. Though many researchers did not specify the DG technology employed, their analyses prove that RES were not considered. Therefore, recommendations for further studies in this area of research will include integration of RES into the grid, consideration for islanding in integrating DG into the grid and protection coordination of a network with DG(s).

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