Check for updates

OPEN ACCESS

EDITED BY Ch. Rami Reddy, Chonnam National University, Republic of Korea

REVIEWED BY Nagi Reddy B., Vignana Bharathi Institute of Technology, India Balasubbareddy Mallala, Chaitanya Bharathi Institute of Technology, India

*CORRESPONDENCE Abdullah Altamimi, a.altmimi@mu.edu.sa

RECEIVED 09 December 2023 ACCEPTED 03 January 2024 PUBLISHED 18 January 2024

CITATION

Altamimi A (2024), Techno-economic evaluation of meshed distribution network planning with load growth and expansion with multiple assets across multiple planning horizons. Front. Energy Res. 12:1353071. doi: [10.3389/fenrg.2024.1353071](https://doi.org/10.3389/fenrg.2024.1353071)

COPYRIGHT

© 2024 Altamimi. This is an open-access article distributed under the terms of the [Creative](https://creativecommons.org/licenses/by/4.0/) [Commons Attribution License \(CC BY\).](https://creativecommons.org/licenses/by/4.0/) The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

[Techno-economic evaluation of](https://www.frontiersin.org/articles/10.3389/fenrg.2024.1353071/full) [meshed distribution network](https://www.frontiersin.org/articles/10.3389/fenrg.2024.1353071/full) [planning with load growth and](https://www.frontiersin.org/articles/10.3389/fenrg.2024.1353071/full) [expansion with multiple assets](https://www.frontiersin.org/articles/10.3389/fenrg.2024.1353071/full) [across multiple planning horizons](https://www.frontiersin.org/articles/10.3389/fenrg.2024.1353071/full)

Abdullah Altamimi^{1,2*}

1 Department of Electrical Engineering, College of Engineering, Majmaah University, Al-Majma'ah, Saudi Arabia, ²Engineering and Applied Science Research Center, Majmaah University, Al-Majma'ah, Saudi Arabia

The smart grid paradigm has ushered in an era where modern distribution systems are expected to be both robust and interconnected in topology. This paper presents a techno-economic-based sustainable planning (TESP) strategy, which can be used as a planning framework for linked distribution systems, seeking to discover a realistic solution among competing criteria of diverse genres. In this comparative analysis-based study, three voltage stability assessment indices—VSA_A, VSA_B, and VSA_W—and a loss minimization condition (LMC)-based framework are used in the initial stage to achieve optimal distributed generation (DG)-based asset optimization for siting, followed by sizing. The respective techniques are evaluated across two variants of multiple load growth horizons spread across 10 years. The suggested TESP technique is tested on two variants of a mesh-configured microgrid (MCMG) with varied load growth scenarios. One variant considers a 65-bus MG with a fixed load growth of 2.7% across two load growth horizons. The other variant considers a 75-bus MG with varied load growth across four load growth horizons, encapsulating an expansion-based planning perspective. The numerical results of the suggested TESP approach in a comparative study demonstrate its effectiveness, and it can be used by researchers and planning engineers as a planning framework for interconnected distribution tools across multiple planning horizons. The proposed study would contribute to enhancing the robustness and interconnectivity of smart grid distribution systems. This dual focus could lead to more cost-effective and reliable power distribution systems.

KEYWORDS

power system, renewable energy, distributed generation, energy, energy consumption

1 Introduction

To keep up with the standards of modern societies, the global demand for power has skyrocketed. Distribution networks (DNs) are at the forefront of working at or near operational limits, which causes a variety of techno-economic issues ([Evangelopoulos et al.,](#page-13-0) [2016](#page-13-0)). Furthermore, in competitive deregulated markets, smartly addressed increasing demands must be subjected to acceptable voltage gradients and system losses. DNs were deterministically designed to retain unidirectional power flow with radial structure, allowing for ease of control and minimal protection requirements [\(Kazmi et al., 2017a\)](#page-14-0). Furthermore, distributed generation (DG) was not considered during the planning phase, and any future changes to the DN topology will be subjected to new planning tools and considerations. The limitations of traditional grid are being overcome with the advent of smart grids, which are expected to be reliable and provide a variety of feasible techno-economic solutions ([Kazmi et al., 2017b](#page-14-1)). In addition, unlike radialstructured DN, the smart distribution network has given way to loop and meshed topology by closing certain normally open tieswitches (RDN). The incorporation of DG assets into interconnected topology provides a reliable and consistent alternative, helping transform RDN into an active DN [\(Mallala](#page-14-2) [et al., 2023](#page-14-2)).

The optimal DG placements (ODGPs) in future smart distribution mechanisms have paved way for interconnected configured DNs and microgrids (MGs), which provides an appropriate opportunity for tightly inhabited urban centers subjected to existing infrastructure modernization ([Mahmoud](#page-14-3) [et al., 2017](#page-14-3)). The techno-economic objective attributed to ODGPs includes acceptable voltage gradients, low system losses and affordable power along with profitability, and overall saving incurred during operations across certain planning horizons ([Alvarez-Herault et al., 2015](#page-13-1)). In various reported research works, researchers have made efforts to address concerned limitations in ODGP-based planning across various distribution mechanisms. Primarily, interconnected DNs have reviewed in terms of index-based optimization methods aimed at the attainment of techno-economic objectives. The prime assets considered are renewable and traditional DGs and reactive power compensation devices like capacitors and distribution static compensator (D-STATCOM). The loop or meshed DN infrastructure-based power system with ODGP-based optimization has considered tieswitches, and the respective impact on various load levels has been evaluated across various load growth horizons besides normal loading conditions. The index-based methodologies have been applied for ODGP in LDN aiming at achieving various technical objectives [\(Ali et al., 2019;](#page-13-2) [Mallala and Dwivedi, 2022\)](#page-14-4) under normal and load growth conditions for the attainment of techno-economic objectives [\(Arshad et al., 2018](#page-13-3); [Javaid et al., 2019\)](#page-13-4).

The DN modernization under SG encompasses several aspects that necessitate the use of decision-making (DM) tools and techniques to reach Pareto optima by considering a multitude of constraints and objective functions. Furthermore, in addition to technical and economic standards, the geographical spread of a new tool necessitates an assessment of environmentally friendly and socially acceptable solutions. As a result, many potential alternative solutions must be assessed across various dimensions (objectives) in order to achieve Pareto optimal solution ([Javaid](#page-13-4) [et al., 2019](#page-13-4)).

Traditional asset optimization methods used in DN were aimed to observe the lowest cost solution. However, they may lack solutions that can be applied to all required rubrics. Furthermore, one of the distinguishing factors of traditional RDNs has been the radiality constraint ([Das et al., 2017](#page-13-5)). According to the literature, DN-centric asset optimization within system restriction primarily intends to siting and sizing of individual assets ([Al-Shara](#page-13-6)fi et al., [2017;](#page-13-6) [Das et al., 2022](#page-13-7)). On the one hand, the technical side intends to reduce system active/reactive power losses, improve voltage profiles, and maximize DG-based renewable energy penetration, short circuit levels, system stability, acceptable bidirectional power flows, and overall power quality (Al-Sharafi [et al., 2017;](#page-13-6) [Das et al., 2022\)](#page-13-7). On the other hand, it aims to improve monetary benefits tend to reduce system loses and costs through optimal allocation of resources and optimal plaining to minimize the maintenance cost.

According to the literature, there are many approaches to reach optimal solutions for the objectives discussed above. These approaches include traditional techniques, i.e., numerical methods, analytical methods, and deterministic methods, in addition to nature-inspired heuristic, meta-heuristic, and AI-inspired neural networks. However, in various scenarios and cases, such algorithms are subjected to achieve solutions which might lead to local optima [\(Kazmi et al., 2017a;](#page-14-0) [Al-](#page-13-6)Sharafi [et al., 2017;](#page-13-6) [Kazmi et al., 2017b;](#page-14-1) [Das et al., 2017;](#page-13-5) [Javaid et al.,](#page-13-4) [2019](#page-13-4); [Das et al., 2022;](#page-13-7) [Mallala et al., 2023](#page-14-2)). However, the addition of further objectives or constraints can potentially increase the computational cost, and the results might not be the optimal result. This limitation is commonly associated with algorithms created by the hybridization of many algorithms which aim to achieve the global optima. Furthermore, multi-criteria optimization techniques are used to achieve Pareto optima amongst conflicting criteria/objectives [\(Kazmi](#page-13-8) [et al., 2019;](#page-13-8) [Mallala et al., 2023](#page-14-2)). While the literature acknowledges significant progress in the techno-economic optimization of DN under smart grid paradigms, ongoing research is required to address the complexities of such systems, especially considering the dynamic nature of load growth and system expansion.

The reviewed work from the perspective of voltage stability indices (VSIs) aiming at ODGPs has mostly focused on the RDN and fairly less for the loop of meshed configured DNs across various planning horizons ([Kazmi et al., 2019](#page-13-8); [Paliwal, 2021\)](#page-14-5). The limitation in all of them includes the fact that the load growth is usually considered constant across a certain large-scale horizon of 5 years, and expansion-based planning with an increased number of nodes are usually not catered, despite the reviewed works addressing the concerned issues partially ([Modarresi et al., 2016](#page-14-6); [Kazmi et al.,](#page-13-9) [2021\)](#page-13-9). However, based on the search results, it can be inferred that techno-economic assessments are a common approach for evaluating distribution network planning ([Gholami et al., 2022\)](#page-13-10). Additionally, the use of power electronic transformers (PETs) has been proposed for economic dispatch in mixed AC/DC systems ([Chen et al., 2021](#page-13-11)). Furthermore, optimal asset placement in interconnected and reliable modern distribution networks has been considered for smart grid modernization ([Khan et al., 2022\)](#page-14-7).

This paper aims to address the limitations in the existing literature by proposing a sustainable planning approach for meshed configured distribution networks (MDNs) with load growth and expansion across multiple planning horizons. The proposed approach, based on optimal distributed generation placement (ODGP) and voltage stability index (VSI), considers two constant load growth horizons and four variable load growth horizons. The approach evaluates techno-economic factors, rather than evaluating only technical or economic factors, and utilizes three VSIs, namely, VSI_A, VSI_B, and VSI_W, for ODGP, followed by loss minimization for the optimal sizing of DGs. The proposed approach is evaluated on an actual MDN-based campus microgrid and offers planning engineers and researchers an efficient and realistic solution for addressing load growth. The main contributions of this paper are as follows:

- (i) Evaluating multiple DGs across various VSIs for siting and loss minimization condition (LMC) for sizing.
- (ii) Assessing solutions based on techno-economic indices.
- (iii) Evaluating solutions across multiple load growth horizons.
- (iv) Including planning for both load growth and node expansion.
- (v) Conducting numerical assessments on an actual meshconfigured microgrid (MCMG).

The remainder of the paper is organized as follows: [Section 2](#page-2-0) presents the proposed approach with mathematical terms and computation processes. The simulation arrangement and performance valuation-based indices are shown in [Section 3.](#page-2-1) [Section 4](#page-5-0) illustrates the effectiveness of the approach across load growth and expansion of node-based planning horizons. The conclusions are reported in [Section 5](#page-13-12).

2 Proposed approach

The proposed method for optimal DG siting uses three voltage stability assessment indices, namely, VSA_A, VSA_B, and VSA_D, which are derived from literature sources [\(Kazmi et al., 2019;](#page-13-8) [Paliwal, 2021;](#page-14-5) [Das et al., 2022](#page-13-7)), aiming at achieving an optimal DG siting using load flow calculation. Load flow calculation is a numerical method used to analyze and calculate the steady-state behavior of an electrical power system. It determines the voltages, currents, and power flows in a power system under different operating conditions. VSA_A is calculated using Eq. [1](#page-2-2), which measures the critical value of the sum of the fourth power of the voltage sensitivity index (V_seb) divided by the square of the total number of buses (k) minus a term involving the sum of squares of V_{seb} and a combination of coefficients $(A_{AA}, B_{AA}, C_{AA},$ and D_{AA}) related to the power flow solution. The critical value of VSA_A ranges from 0, indicating instability, to 1, indicating stability.

Similarly, VSA_B is calculated using Eq. [2](#page-2-3), which measures the maximum value of the ratio of a term involving the sum of squares of V_{seb} and a combination of coefficients (E_{BB} and F_{BB}) related to the power flow solution divided by the sum of the fourth power of Vseb. The critical value of VSA_B ranges from 1, indicating instability, to 0, indicating stability. VSA_D in Eq. [3](#page-2-4) shows the deviation, which is positive, pointing toward critical loading conditions of a bus in a distribution network that is close to voltage collapse.

$$
VSA.A = \sum_{i=1}^{k_1} \left(\frac{V_{seb}}{k}\right)^4 - \frac{4}{k} \sum_{i=1}^{k_1} \left(\frac{V_{seb}}{k}\right)^2 \left[\left(\frac{A_{AA}}{C_{AA}}\right) + \left(\frac{B_{AA}}{D_{AA}}\right)\right] - \frac{4}{k^2} \left[\left(\frac{A_{AA}}{C_{AA}}\right) - \left(\frac{B_{AA}}{D_{AA}}\right)\right]^2 \ge 0,
$$
 (1)

where

$$
A_{AA} = P_{2A} \left[\prod_{i=1}^{n} R_{nr} \right] \left[1 - \left(\frac{X_{1x} X_{2x}}{R_{1r} R_{2r}} + \frac{X_{1x} X_{3x}}{R_{1r} R_{3r}} + \frac{X_{2x} X_{3x}}{R_{2r} R_{3r}} \right) \right] + Q_{2A} \left[\prod_{i=1}^{n} nx \right] \left[\left(\frac{R_{1r} R_{2r}}{X_{1x} X_{2x}} + \frac{R_{1r} R_{3r}}{X_{1x} X_{3x}} + \frac{R_{2r} R_{3r}}{X_{2x} X_{3x}} \right) - 1 \right];
$$

$$
B_{AA} = P_{2A} \left[\prod_{i=1}^{n} X_{nx} \right] \left[\left(\frac{R_{1r} R_{2r}}{X_{1x} X_{2x}} + \frac{R_{1r} R_{3r}}{X_{1x} X_{3x}} + \frac{R_{2r} R_{3r}}{X_{2x} X_{3x}} \right) - 1 \right]
$$

\n
$$
- Q_{2A} \left[\prod_{i=1}^{n} R_{nr} \right] \left[1 - \left(\frac{X_{1x} X_{2x}}{R_{1r} R_{2r}} + \frac{X_{1x} X_{3x}}{R_{1r} R_{3r}} + \frac{X_{2x} X_{3x}}{R_{2r} R_{3r}} \right) \right];
$$

\n
$$
C_{AA} = \left[abs \left\{ \sum_{k \neq l}^{n} R_{k} R_{l} - \sum_{k \neq l}^{n} X_{k} X_{l} \right\} + 0.001 \right];
$$

\n
$$
D_{AA} = abs \left(R_{1} X_{2} + R_{2} X_{1} + R_{1} X_{3} + R_{3} X_{1} + R_{2} X_{3} + R_{3} X_{2} \right)
$$

\n
$$
= abs \left[\sum_{k \neq l}^{n} R_{k} X_{l} \right].
$$

\n
$$
VSA_{-}B = 4k^{2} \frac{\left[E_{BB} \sum_{i=1}^{k_{l}} \left(\frac{V_{sb}}{k} \right)^{2} + \left(\frac{F_{BB}}{k} \right)^{2} \right]}{\sum_{i=1}^{k_{l}} \left(\frac{V_{sb}}{k} \right)^{4}} \leq 1,
$$

\n(2)

where

$$
E_{BB} = [abs\{ (P_{2b}R_{1r} + P_{4b}R_{2r} + P_{6b}R_{3r})
$$

+ (Q_{2b}X_{1x} + Q_{4b}X_{2x} + Q_{6b}X_{3x})\} + 0.001];

$$
F_{BB} = [abs\{ (P_{2b}X_{1x} + P_{4b}X_{2x} + P_{6b}X_{3x})
$$

- (Q_{2b}R_{1r} + Q_{4b}R_{2r} + Q_{6b}R_{3r})\} + 0.001].

$$
VSA_D = \sum_{i=1}^{k_1} (V_{seb} - V_{reb})^2 \ge 0,
$$
 (3)

where V_{seb} is the voltage value as a reference of substation voltage (sending end bus). V_{reb} represents the voltage value of the receiving end node/bus throughout the distribution network.

The weighted VSI factor is delegated as VSI_W and is based on weighted normalized values of VSA_Aw, VSA_Bw, and VSA_Dw, as shown in Eq. [4](#page-2-5):

$$
VSA_W = [(\omega_{Aw} \times VSI_{Aw}) + (\omega_{Bw} \times VSI_{Bw}) + (\omega_{Dw} \times VSI_{Dw})],
$$
\n(4)

where $\omega_{Aw}, \omega_{Bw},$ and ω_{Dw} are the weight factors of each individual normalized values of VSI, and their addition should be 1.

The technique of loss minimization condition (LMC) remains the same as mentioned in [Mahmoud et al. \(2017\)](#page-14-3). The expressions for LMC for P_{Loss} and Q_{Loss} subjected to zero loop currents are illustrated in Eqs [5,](#page-2-6) [6](#page-2-7) as P_LMC and Q_LMC, respectively, as follows:

$$
P_{LMC} = [(I_{2B})^{2}R_{2r} + (I_{1B})^{2}R_{1r} + (I_{3B})^{2}R_{3r}] \ge 0; \qquad (5)
$$

$$
Q_LMC = [(I_{2B})^2 X_{2x} + (I_{1B})^2 X_{1x} + (I_{3B})^2 X_{3x}] \ge 0,
$$
 (6)

where I_{1B} , I_{2B} , and I_{3B} are the individual line current across different feeders. R_{1r} , R_{2r} , and R_{3r} are the individual line resistance across different feeders. X_{1x} , X_{2x} , and X_{3x} are the individual line reactance across different feeders.

3 The simulation arrangement and performance valuation-based indices

[Figure 1](#page-3-0) shows the flow chart of the techno-economic-based sustainable planning (TESP) strategy for evaluating several load increase scenarios throughout DG placement planning periods. The

TABLE 1 Technical performance evaluation parameters [\(Kazmi et al., 2021](#page-13-9)).

research is carried out in two scenarios: one with two horizons of load expansion through a fixed node distribution network and the other with four horizons of expandable load development. A test system setup of a meshed configured microgrid is considered with two variants in this work: the first variant with a 65-bus system for fixed node evaluation, as shown in [Figure 2,](#page-3-1) and the second variant with a 75-bus expanded system, as shown in [Figure 3.](#page-4-0) The entire load horizon is 10 years, and both distribution network types are set up in a meshed topology. Without any DG in the meshed design, the normal active and reactive power loads in both networks are 3,950.505 KW and 1,913.317 KVAR, respectively, with losses of 60.51 KW and 34.63 KVAR. The load growth for 65-node MG is 2.7% in horizon 1 of 5 years (until 2025) and 2.7% in horizon 2 of 5 years (until 2030). The load growth for 75-node MG is 14.9% in horizon 1 of 3 years (until year 2023), 2.7% in horizon 2 of 2 years (until year 2025), 17.75% in horizon 3 of 3 years (until year 2028),

and 2.7% in horizon 4 of 2 years (until year 2030). [Table 1](#page-4-1) and [Table 2](#page-5-1) show the technical and economic performance evaluation factors, respectively ([Kazmi et al., 2021](#page-13-9)). [Table 3](#page-5-2) presents the load growth values for 65- and 75-bus meshed MGs in various planning horizons.

The base case mathematical model is created in MATLAB, and outcomes from the m-file are used to identify the weakest nodes according to the VSI. The outcome-based numericals were collected from the Simulink model setup and were used to perform the simulation The loop currents across TSs will be simulated until became nearly zero and voltages across the various buses became identical, which corresponded to the optimal sizing of assets based on 1% termination criteria. Finally, the obtained values from m-files are implemented in a MATLAB 2018a program, where the suggested multi-criteria sustainable planning technique is assessed using various matrices.

S. no.	Technical parameter	Designation	Relationship	Objective
	Cost of $PLoss$ in millions USD (M\$)	CP _{Loss}	$[P_L \times E_U \times T_Y (8760 hrs)]$	Minimize
	Savings of P_{Loss} in millions USD (M\$)	SP _{Loss}	$\frac{PLC_{No-DG}-PLC_{M-DG}}{PLC_{No-DG}} \times 100$	Maximize
	Cost of active power for DG (\$/MWh)	CDGP	$a \times P_{DG}^2 + b \times P_{DG} + c$, where $a = 0$, $b = 20$, and $c = 0.25$	Minimize
4	Cost of reactive power for DG (\$/MVAR)	CDGQ	$[C(S_{DG_M}) - C(\sqrt{(S_{DG_M}^2 - P_{DG}^2)})] \times k$, where $S_{DG_M} = \frac{P_{DG_M}}{\cos \theta} = \frac{1.1 \times P_{DG}}{\cos \theta}$; $k = 0.5 - 1$	Minimize

TABLE 2 Cost-economic performance assessment parameters [\(Kazmi et al., 2021\)](#page-13-9).

TABLE 3 Load growth values across 65- and 75-bus meshed MGs.

4 Results and discussion

4.1 Case 1 evaluation results

The cases with respective designations, in terms of nomenclature, are illustrated with case (C#) that have been evaluated across the following scenarios (C#/S#):

Case 1 (C1): DG placement in the 65-bus meshed configured microgrid.

Scenario 1 (C1/S1): evaluation across the radial network across two planning horizons of 5 years each of same load growth of 2.7%.

Scenario 2 (C1/S2): evaluation across the mesh network with VSA_A across two planning horizons of 5 years each of same load growth of 2.7%.

Scenario 3 (C1/S3): evaluation across the mesh network with VSA_B across two planning horizons of 5 years each of same load growth of 2.7%.

Scenario 4 (C1/S3): evaluation across the mesh network with VSA_W across two planning horizons of 5 years each of same load growth of 2.7%.

The two load growth horizons for C1 are shown with the following nomenclature:

Normal load:

• 2020

• 2020 optimal reinforcement (2020_O)

Load growth 1 across 5 years with a 2.7% growth rate:

• 2025

• 2025 optimal reinforcement (2025_O)

Load growth 1 across 5 years with a 2.7% growth rate:

- 2030
- 2030 optimal reinforcement (2030_O)

The DG placement and sizing at a 0.9 lagging power factor (LPF) in the 65-bus meshed configured MG evaluated across two planning horizons of 5 years each with a load growth of 2.7% increase per annum are illustrated in [Tables 4](#page-6-0)–[6.](#page-7-0) The 2.7% increment of power demand was chosen to simulate a realistic load growth scenario in the distribution system. This value was based on historical trends and future projections of population growth, urbanization, and industrialization in the area served by the distribution system. The DG siting and sizing based on VSA_A–LMC, VSA_B–LMC and VSA_W–LMC approaches across various parameters are shown in [Tables 4](#page-6-0)–[6,](#page-7-0) respectively. In all those tables, DG units with their capacities and an operating lagging power factor of 0.9 have been illustrated across abovementioned C1 scenarios. The 0.9 power factor was chosen as a typical value for the loads in the distribution system. This value is not necessarily regulated, but it is often used as a benchmark for power factor correction and improvement efforts in distribution systems. It is also a reasonable assumption for modeling and simulation purposes as it represents a moderate level of reactive power demand in the system.

TABLE 4 DG siting and sizing for a 65-bus meshed MG at 0.9 lagging power factor based on VSA_A and LMC approaches.

TABLE 5 DG siting and sizing for a 65-bus meshed MG at 0.9 lagging power factor based on VSA_B and LMC approaches.

DG siting and sizing considering VSA_B- and LMC-based parameters	Normal case 2020	Optimal case 2020_O	5-year non- optimal case 2025	5-year optimal case 2025_O	10-year non- optimal case 2030	10 -year optimal case 2030_O
Active/reactive load (KW + jKVAR)	$3949.77 +$ j1912.97	$3949.77 +$ j1912.97	$4512.64 + j2185.82$	$4512.64 + j2185.82$	$5155.65 + j2497.17$	$5155.65 + j2497.17$
Grid active/reactive power (KW + jKVAR)	$4010.28 +$ i1947.6	$922.45 + j440.09$	$1488.62 + j716.39$	$1200.61 + j576.59$	$1847.99 + j892.89$	$1442.97 + j696.25$
DG1 bus and size $(KW + jKVAR)$		DG 5 @ 270 + i130.77	DG_5 @ 270 + i130.77	DG 5 @ 270 + i130.77	DG 5 @ 270 + i130.77	DG 5 @ 315 + i152.56
DG2 ω bus and size (KW + jKVAR)		DG 14 @ 630 + 1305.12	DG 14 @ 630 + 1305.12	DG 14 @ 720 + i348.71	DG 14 @ 720 + i348.71	DG 14 @ 810 + 1392.3
DG3 $@$ bus and size (KW + jKVAR)		DG 17 @ $477 +$ i231.02	DG 17 @ 477 + i231.02	DG 17 @ 630 + 1305.12	DG 17 @ 630 + 1305.12	DG 17 @ 720 + j348.71
DG4 @ bus and size $(KW + jKVAR)$		DG 38 @ 855 + 1414.1	DG 38 @ 855 + 1414.1	DG 38 @ 900 + i435.89	DG 38 @ 900 + i435.89	DG 38 @ 1080 + i523.07
DG5 ω bus and size (KW + jKVAR)		DG 43 @ 855 + 1414.1	DG 43 @ 855 + 1414.1	DG 43 @ 855 + 1414.1	DG 43 @ 855 + j414.1	DG 43 @ 855 + i414.1
Active power loss (KW)	60.51	59.68	62.98	62.97	67.34	67.32
Reactive power loss (KVAR)	34.63	22.23	25.68	25.36	30.31	29.82

For the VSA_A–LMC-based approach, the normal scenario in both cases is the same, such as the year 2020 scenario. Load has increased linearly at 2.7 percent per year in the actual model after 3 years, as well as due to the addition of new load programs. Because the load has changed and the DG capacity has remained the same as in scenario 2020, DG values for scenario 2023 will need to be adjusted for this model after 3 years. The DG capacity has been re-optimized to meet the model's current requirement. According to scenario 2020, weak nodes would remain the same.

Similarly, for VSA_B–LMC- and VSA_W–LMC-based approach variants, load has increased linearly at 2.7 percent per year in the actual model after 3 years, as well as due to the addition of new load programs. Because the load has changed and the DG capacity has remained the same as in scenario 2020, DG values for scenario 2023 will need to be optimized for this model after 3 years. The DG capacity has also been reoptimized to meet the model's current requirement. All three VSIs were performed according to their respective criteria in order to increase the voltage profile and maximize the model's technical characteristics.

TABLE 6 DG siting and sizing for a 65-bus meshed MG at 0.9 lagging power factor based on VSA_W and LMC approaches.

In [Figure 4](#page-7-1), case 1 shows an active power loss trend across two planning horizons (accumulatively 10 years) in the 65-bus meshed configured MG-based distribution network. It can be observed that the active power losses have reduced in meshed configured approaches in C1, scenarios 2–4 compared to the radial counterpart in scenario 1.

[Figure 5](#page-7-2) and [Figure 6](#page-8-0) show that a substantial improvement is active and prominent in reactive power loss minimization across all scenarios of C1. The reason being that reactive power loss is high because the distribution network is laid underground rather than overhead cables. [Figure 7](#page-8-1) shows that each VSI enhanced the voltage profile of the network according to its own specifications, and that all VSIs are significantly superior to a simple interconnected network. After maximizing the value of DG penetration in accordance with the policy, the voltage of each node is brought to unity. For this scenario, a new optimized model has been

4.2 Case 2 evaluation results

power losses and an increase in savings.

The cases with respective designations, in terms of nomenclature, are illustrated with case (C#) that have been evaluated across the following scenarios (C#/S#):

acquired. [Figure 8](#page-8-2) shows a phenomenal reduction in the cost of active

Case 2 (C2): DG placement in the 75-bus meshed configured microgrid.

Scenario 1 (C2/S1): evaluation across the radial network across four different planning horizons of 5 years each of various load growth levels.

FIGURE 6

Active and reactive power loss minimization trends across two planning horizons in the 65-bus meshed configured MG-based distribution network.

Voltage profile trend across two planning horizons in the 65-bus meshed configured MG-based distribution network.

Scenario 2 (C2/S2): evaluation across the mesh network with VSA_A across four different planning horizons of various load growth levels.

Scenario 3 (C2/S3): evaluation across the mesh network with VSA_B across four different planning horizons of various load growth levels.

Scenario 4 (C2/S3): evaluation across the mesh network with VSA_W across four different planning horizons of various load growth levels.

The four load growth horizons for C2 are shown with the following nomenclature:

Normal load:

- 2020
- 2020 optimal reinforcement (2020_O)

Load growth 1 across 3 years with a 14.9% growth rate:

- 2023
- 2023 optimal reinforcement (2023_O)

Load growth 2 across 2 years with a 2.7% growth rate:

- -2025
- 2025 optimal reinforcement (2025_O)

Load growth 1 across 3 years with a 17.75% growth rate:

- 2028
- 2028 optimal reinforcement (2028_O)

Load growth 4 across 2 years with a 2.7% growth rate:

- 2030
- 2030 optimal reinforcement (2030_O)

The DG placement and sizing at 0.9 lagging power factor in the 75-bus meshed configured MG evaluated across four planning horizons of 2–3 years each with a variable expansion-based load growth increase per annum are illustrated in [Tables 7](#page-9-0)–[9.](#page-11-0) The DG siting and sizing based on VSA_A–LMC-, VSA_B–LMC-, and VSA_W–LMC approaches are illustrated across various parameters in [Tables 7](#page-9-0)–[9,](#page-11-0) respectively. In all those tables, DG units with their capacities and an operating lagging power factor (LPF) of approximately 0.9 have been illustrated across abovementioned C2 scenarios. All the data are provided in a selfexplanatory manner.

In [Figure 9](#page-12-0), case 2 shows an active power loss trend across four planning horizons (accumulatively 10 years) in the 75-bus meshed configured MG-based distribution network. It can be observed that the active power losses have reduced in meshed configured approaches in C2, scenarios 2–4 compared to the radial counterpart in scenario 2.

[Figure 10](#page-12-1) and [Figure 11](#page-12-2) show that a substantial improvement is active and prominent in reactive power loss minimization across all scenarios of C2 and is quite greater than that in C1 and respective scenarios. The reason being that reactive power loss is high because the distribution network is laid underground rather than overhead

DG place/size considering VSA_A –LMC-based parameters

Active/reactive load (KW ⁺ jKVAR)

Grid active/reactive power (KW ⁺ jKVAR)

DG1 bus and size (KW ⁺ jKVAR)

DG2 @ bus and size (KW ⁺ jKVAR)

DG3 @ bus and size (KW ⁺ jKVAR)

DG4 @ bus and size (KW ⁺ jKVAR)

DG5 @ bus and size (KW ⁺ jKVAR)

Normalcase 2020

3949.77 + j1912.97

4010.28 + j1947.6

—

 \overline{a}

—

 \overline{a}

 \overline{a}

10-year nonoptimal case 2030

7882.44 KW ⁺ j 4085.16 KVAR

10870 KW ⁺ j 5264.24 KVAR 10-year optimal case 2030_O

2541.47 KW ⁺ j 1255.03 KVAR

10870 KW ⁺ j 5264.24 KVAR

DG_5 @ 630 KW ⁺ j305.12 KVAR

DG_22 @ 1350 KW $+$ j653.84 KVAR

DG_44 @ 1620 KW + j784.6 KVAR

DG_57 @ 4293 KW + j2079.2 KVAR

DG_70 @ 630 KW + j305.12 KVAR

3-year nonoptimal case 2023

3151.17 KW + j1529.96 KVAR

5988.14 KW + j2899.97 KVAR

 $DG_5 = 558$ KW + j270.25 KVAR

DG_22 ⁼ 441 KW ⁺ j213.59 KVAR

DG_44 ⁼ 783 KW ⁺ j379.22 KVAR

DG_57 ⁼ 729 KW⁺ j353.07 KVAR

 DG 70 = 414 KW ⁺ j.51 KVAR

3-year optimal case 2023_O

2196.99 KW + j1065.18 KVAR

5988.14 KW + j2899.97 KVAR

DG_5 ⁼ 540 KW ⁺ j261.53 KVAR

DG_22 ⁼ 990 KW ⁺ j479.48 KVAR

DG_44 ⁼ 855 KW ⁺ j414.1 KVAR

DG $57 =$ 1080 KW + j523.07 KVAR

 $DG_{70} = 414$ KW ⁺ j200.51 KVAR

5-year nonoptimal case 2025

3221.19 KW + j1564.47 KVAR

6315.96 KW + j3058.84 KVAR

 $DG_5 = 558$ KW + j270.25 KVAR

DG_22 ⁼ 630 KW ⁺ j305.12 KVAR

DG_44 ⁼ 855 KW ⁺ j414.1 KVAR

DG_57 ⁼ 729 KW⁺ j353.07 KVAR

 DG 70 = 414 KW ⁺ j200.51 KVAR

Reactive power loss (KVAR) 34.63 43.9 46.63 KVAR 43.9 KVAR 48.68 KVAR 48.68 KVAR 325.77 KVAR 87.98 KVAR 363.97 KVAR 118.67 KVAR

5-year optimal case 2025_O

2240.01 KW ⁺ j 1086.63 KVAR

6315.96 KW ⁺ j 3058.84 KVAR

DG_5 ⁼ 540 KW ⁺ j261.53 KVAR

DG_22 ⁼ 990 KW ⁺ j479.48 KVAR

DG $44 =$ 1008 KW + j488.2 KVAR

DG $57 =$ 1215 KW + j588.45 KVAR

 DG 70 = 414 KW ⁺ j200.51 KVAR

8-year nonoptimal case 2028

7303.55 KW + j3773.83 KVAR

10305.94 KW + j4991.11 KVAR

DG_5 @ 558 KW ⁺ j 270.25 KVAR

DG_22 @ 630 KW ⁺ j305.12 KVAR

DG_44 @ 855 KW ⁺ j414.1 KVAR

DG_57 @ 729 KW⁺ j353.07 KVAR

DG_70 @ 414 KW⁺ j200.51 KVAR

8-year optimal case 2028_O

2521.24 KW + j1221.49 KVAR

10305.94 KW ⁺ j 4991.11 KVAR

DG_5 @ 585 KW ⁺ j283.33 KVAR

DG_22 @ 1350 KW +j653.84 KVAR

DG_44 @ 1620 KW + j784.6 KVAR

DG 57 @ 3870 KW + j1874.3 KVAR

DG_70 @ 540 KW⁺ j261.53 KVAR

Optimal case 2020_O

> 2196.99 + j1065.2

5988.14 +j2899.97

 DG_5 @540 ⁺ j261.53

 DG_22 @ 990 ⁺ j479.48

 DG_44 @ 855 $+$ j414.1

 DG_57@ 1080 +j523.07

 DG_70 @ 414 ⁺ j200.51

TABLE 9 DG siting and sizing for ^a 75-bus meshed MG at 0.9 lagging power factor based on VSA_W and LMC approaches.

Active and reactive power loss minimization trends across four planning horizons in the 75-bus meshed configured MG-based distribution network.

cables. [Figure 12](#page-12-3) shows that each VSI enhanced the voltage profile of the network according to its own specifications and all VSIs are significantly superior to a simple interconnected network. After maximizing the value of DG penetration in accordance with the policy, the voltage of each node is brought to unity. For scenario, a new optimized model has been acquired. [Figure 13](#page-12-4) shows a phenomenal reduction in the cost of active power losses and an increase in savings.

Further examination in [Figure 10](#page-12-1) and [Figure 11](#page-12-2) reveals that both active and reactive power losses are considerably minimized in C2 across all scenarios, surpassing the results of case 1 (C1). The higher reactive power loss in the baseline scenario is attributed to the network's underground cabling system. In [Figure 12,](#page-12-3) the implementation of various voltage stability indices (VSIs) has been shown to significantly improve the voltage profile over a simple interconnected network, bringing each node's voltage closer to unity after the integration of maximum distributed generation (DG) penetration according to the set policy.

The obtained results in all situations were compared with the existing literature, and a close approximation was observed in the results, especially in evaluating contradictory criteria with cases of load increase over numerous planning horizons. It is also observed that fixed load growth by percentages does not capture the realistic pictures across medium horizons of 5 years, as evident from evaluated

meshed configured MG-based distribution network.

case 1. It is observed from evaluated case 2 that small planning horizons with variable load growth levels capture the requirements of expansion-based planning efficiently with the fixed counterpart.

5 Conclusion

Within the smart grid paradigm, current distribution networks aim to enhance reliability and interconnected topology while meeting diverse performance requirements. This study introduces a TESP (transmission expansion planning) strategy to address grid planning challenges. The proposed multi-stage comprehensive strategy employs voltage stability assessment indices (VSA_A, VSA_B, and VSA_W) and a load margin constraint (LMC) to assess and optimize the positioning and sizing of distributed generation (DG) assets. The evaluation of potential solutions considers technical and economic performance indicators across two load growth scenarios spanning 10 years. The TESP technique is tested on two variants of distribution grids (65-bus and 75-bus) with different load growth patterns. The results are compared with the existing literature, demonstrating close alignment, especially in assessing performance amid conflicting standards across various planning horizons. Notably, fixed load growth percentages may not accurately depict realistic scenarios over five-year planning horizons, as evident from case 1. Conversely, smaller planning horizons with variable load growth levels effectively address expansion-based planning needs, reducing assessment time and sensitivity analysis. Additionally, the TESP method offers a wide range of trade-off options for performance indicators, making it a valuable planning tool for academics and distribution system planners in interconnected distribution networks.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material; further inquiries can be directed to the corresponding author.

References

Ali, F., Nazir, Z., Inayat, U., and Ali, S. M. (2019). "Infrastructure of south Korean electric power system and potential barriers for the implementation of smart grid: a review," in International Conference on Innovative Computing (ICIC), Japan, August 29-31, 2023 (IEEE), 1–7.

Al-Sharafi, A., Sahin, A. Z., Ayar, T., and Yilbas, B. S. (2017). Techno-economic analysis and optimization of solar and wind energy systems for power generation and hydrogen production in Saudi Arabia. Int. J. Renew. Sust. Ener Rev. 69, 33–49. doi[:10.](https://doi.org/10.1016/j.rser.2016.11.157) [1016/j.rser.2016.11.157](https://doi.org/10.1016/j.rser.2016.11.157)

Alvarez-Herault, M.-C., N'Doye, N., Gandioli, C., Hadjsaid, N., and Tixador, P. (2015). Meshed distribution network vs reinforcement to increase the distributed eneration connection. Sustain. Energy, Grids Netw. 1, 20-27. doi[:10.1016/j.segan.](https://doi.org/10.1016/j.segan.2014.11.001) generation

Arshad, M. A., Ahmad, S., Afzal, M. J., and Kazmi, S. A. A. (2018). "Scenario based performance evaluation of loop configured microgrid under load growth using multi-criteria decision analysis," in Proceedings of the 14th International Conference on Emerging Technologies (ICET), Islamabad, 21–22 Nov, 2018 (IEEE), 1–6.

Chen, L., Deng, X., Xia, F., Chen, H., Liu, C., Chen, Q., et al. (2021). A technoeconomic sizing approach for medium-low voltage DC distribution system. IEEE Trans. Appl. Supercond. 31 (8), 1–6. doi:[10.1109/tasc.2021.3101773](https://doi.org/10.1109/tasc.2021.3101773)

Das, B. K., Hoque, N., Mandal, S., Pal, T. K., and Raihan, M. A. (2017). A technoeconomic feasibility of a stand-alone hybrid power generation for remote area application in Bangladesh. Energy 134 (1), 775–788. doi[:10.1016/j.energy.2017.](https://doi.org/10.1016/j.energy.2017.06.024) [06.024](https://doi.org/10.1016/j.energy.2017.06.024)

Author contributions

AA: conceptualization, data curation, funding acquisition, resources, software, validation, writing–original draft, and writing–review and editing.

Funding

The author(s) declare financial support was received for the research, authorship, and/or publication of this article. The authors extend their appreciation to the Deanship of Scientific Research, Majmaah University, Saudi Arabia, for funding this research work through project number R-2023-192.

Conflict of interest

The author declares that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors, and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Das, S. K., Sarkar, S., and Das, D. (2022). Performance enhancement of gridconnected distribution networks with maximum penetration of optimally allocated distributed generation under annual load variation. Arabian J. Sci. Eng. 47 (11), 14809–14839. doi[:10.1007/s13369-022-06951-x](https://doi.org/10.1007/s13369-022-06951-x)

Evangelopoulos, V. A., Georgilakis, P., and Hatziargyriou, N. D. (2016). Optimal operation of smart distribution networks: a review of models, methods and future research. Electr. Power Syst. Res. 140, 95–106. doi:[10.1016/j.epsr.2016.06.035](https://doi.org/10.1016/j.epsr.2016.06.035)

Gholami, K., Islam, M. R., Rahman, M. M., Azizivahed, A., and Fekih, A. (2022). State-of-the-art technologies for volt-var control to support the penetration of renewable energy into the smart distribution grids. Energy Rep. 8, 8630–8651. doi[:10.1016/j.egyr.2022.06.080](https://doi.org/10.1016/j.egyr.2022.06.080)

Javaid, B., Arshad, M. A., Ahmad, S., and Kazmi, S. A. A. (2019). "Comparison of different multi criteria decision analysis techniques for performance evaluation of loop configured Micro grid," in Proceedings of the 2019 2nd International Conference on Computing, Mathematics and Engineering Technologies (iCoMET), Sukkur, Pakistan, 30–31 January 2019 (IEEE), 1–7.

Kazmi, S. A. A., Ahmad, H. W., and Shin, D. R. (2019). A new improved voltage stability assessment index-centered integrated planning approach for multiple asset placement in mesh distribution systems. Energies 12, 3163. doi[:10.3390/](https://doi.org/10.3390/en12163163) [en12163163](https://doi.org/10.3390/en12163163)

Kazmi, S. A. A., Ameer Khan, U., Ahmad, W., Hassan, M., Ibupoto, F. A., Bukhari, S. B. A., et al. (2021). Multiple (TEES)-Criteria-Based sustainable planning approach for mesh-configured distribution mechanisms across multiple load growth horizons. Energies 14, 3128. doi[:10.3390/en14113128](https://doi.org/10.3390/en14113128)

Kazmi, S. A. A., Shahzaad, M. K., and Shin, D. R. (2017b). Voltage stability index for distribution network connected in loop configuration. IETE J. Res. 63, 281–293. doi[:10.](https://doi.org/10.1080/03772063.2016.1257376) [1080/03772063.2016.1257376](https://doi.org/10.1080/03772063.2016.1257376)

Kazmi, S. A. A., Shahzad, M. K., and Shin, D. R. (2017a). Multi-objective planning techniques in distribution networks: a composite review. Energies 10, 208. doi:[10.3390/](https://doi.org/10.3390/en10020208) [en10020208](https://doi.org/10.3390/en10020208)

Khan, S. N., Kazmi, S. A. A., Altamimi, A., Khan, Z. A., and Alghassab, M. A. (2022). Smart distribution mechanisms—Part I: from the perspectives of planning. Sustainability 14 (23), 16308. doi[:10.3390/su142316308](https://doi.org/10.3390/su142316308)

Mahmoud, P. H. A., Phung, D. H., and Vigna, K. R. (2017). A review of the optimal allocation of distributed generation: objectives, constraints, methods, and
algorithms. *Renew. Sustain. Energy Rev.* 75, 293–312. doi:10.10 [10.071](https://doi.org/10.1016/j.rser.2016.10.071)

Mallala, B., and Dwivedi, D. (2022). Salp swarm algorithm for solving optimal power flow problem with thyristor-controlled series capacitor. J. Electron. Sci. Technol. 20 (2), 100156. doi[:10.1016/j.jnlest.2022.100156](https://doi.org/10.1016/j.jnlest.2022.100156)

Mallala, B., Venkata Prasad, P., and Palle, K. (2023). "Multi-objective optimization with practical constraints using AALOA," in International Conference on Information and Communication Technology for Intelligent Systems, Singapore, APRIL 4 – 5, 2024 (Springer Nature Singapore), 165–177.

Modarresi, J., Gholipour, E., and Khodabakhshian, A. (2016). A comprehensive review of the voltage stability indices. Renew. Sustain Energy Rev. 63, 1–12. doi:[10.1016/](https://doi.org/10.1016/j.rser.2016.05.010) [j.rser.2016.05.010](https://doi.org/10.1016/j.rser.2016.05.010)

Paliwal, P. (2021). Comprehensive analysis of distributed energy resource penetration and placement using probabilistic framework. IET Renew. Power Gener. 15 (4), 794–808. doi:[10.1049/rpg2.12069](https://doi.org/10.1049/rpg2.12069)