INVESTIGATION OF OPTIMAL POWER FACTOR IN DISTRIBUTION NETWORKS BASED ON DG PLACEMENT



SYED NAQI HAIDER SHAH 01-244222-012

BAHRIA UNIVERSITY, ISLAMABAD.

APPROVAL FOR EXAMINATION

Scholar's name: Syed Naqi Haider ShahRegistration No: 01-244222-012Program of study: M.S.Thesis title: Investigation of Optimal Power Factor in Distribution Networks

Based on DG Placement.

It is to certify that the above scholar's thesis has been completed to my satisfaction and, to my belief, its standard is appropriate for submission for examination. I have also conducted plagiarism test of this thesis using HEC prescribed software and found similarity index at <u>12%</u> and from single source is <u>1%</u> that is within the permissible limit set by the HEC for the MS/MPhil/Equivalent degree thesis. I have also found the thesis in a format recognized by the BU for the MS thesis.

Principal Supervisor's
Signature: _____
Date: _____

Name: _____

AUTHOR'S DECLARATION

I, SYED NAQI HAIDER SHAH hereby state that my M.S. thesis titled "Investigation of Optimal Power Factor in Distribution Networks Based on Dg <u>Placement.</u>" is my own work and has not been submitted previously by me for taking any degree from this university <u>BAHRIA UNIVERSITY</u>, ISLAMABAD or anywhere else in the country/world.

At any time if my statement is found to be incorrect even after my graduation, the University has the right to withdraw/cancel my M.S. degree.

Name of scholar: SYED NAQI HAIDER SHAH .

Date: <u>30/08/2024</u>.

PLAGIARISM UNDERTAKING

I, <u>Syed Naqi Haider Shah</u>, solemnly declare that research work presented in the thesis titled "<u>Investigation of Optimal Power Factor in Distribution Networks Based on DG</u> <u>Placement</u>" is solely my research work with no significant contribution from any other person. Small contribution/help wherever taken has been duly acknowledged and that complete thesis has been written by me.

I understand the zero-tolerance policy of the HEC and Bahria University towards plagiarism. Therefore, I as an Author of the above-titled thesis declare that no portion of my thesis has been plagiarized and any material used as the reference is properly referred/cited.

I undertake that if I am found guilty of any formal plagiarism in the above-titled thesis even after award of M.S. degree, the university reserves the right to withdraw/revoke my M.S. degree and that HEC and the University has the right to publish my name on the HEC / University webSITE on which names of scholars are placed who submitted plagiarized thesis.

Scholar / Author's Sign: _____

Name of the Scholar: SYED NAQI HAIDER SHAH

DEDICATION

This research is dedicated to my loving parents who despite the various sacrifices they made for me to achieve my goals, have been very supportive throughout. To my father who had wished that his Son should strive to be a Bachelor and above all a Master of Electrical Engineering – I shall remain eternally indebted to him. His vision and faith in me have been my source of light right through the obstacles. DAD! You are so kind to me and I can never repay you for your support and trust. All the times that you have sacrificed for me, and encouraged me have been the very motivating factor for me. I will be forever grateful to my father because he encourages me to make it even when he didn't make it for himself.

The teachings of Ahlulbayt (A.S.) are another source of great inspiration for me; his insight on the value of moral integrity and the quest of knowledge has guided me throughout my life. "*Knowledge enlivens the soul*," as Imam Ali (A.S.) once stated. Imam Ja'far al-Sadiq (A.S.) went on to say, "*The one who seeks knowledge is like a warrior in the path of Allah*." My approach to learning has been affected by their teachings, which place a strong emphasis on using information ethically in addition to its acquisition. I hope that I will always be motivated and led by their example in all that I do.

Engr. Syed Naqi Haider Shah Enrolment No. 01-244222-012

ACKNOWLEDGEMENTS

I am very thankful to Almighty Allah for His infinite blessings and for providing me with the chance, power and capability to finish my master's degree with this thesis.

I wish to express my sincere thanks to my supervisor Prof. Dr. Asad Waqar. I would like to thank my supervisor for his constant encouragement, valuable suggestions and directions regarding this study. His commitment, love for learning, as well as efficiency in work has always been a motivation for me to work hard in this process.

I want especially to thank all the Professors whose knowledge and experience have helped in writing this thesis. I would like to appreciate my teachers for their support and contribution towards my research in form of one way or the other and my university fellows for the help they offered me in course of this research.

In addition, I want to express my deepest gratitude to my loving parents who have always been my support, my inspiration and my belief. Their support, hard work and belief in me have been the pillars that have helped me achieve my goals. I will always be grateful to them for teaching me the values which include hard work, perseverance and integrity which have been with me all through this academic exercise.

> Syed Naqi Haider Shah Enrolment No. 01-244222-012

ABSTRACT

This study optimizes Distributed Generator (DG) placement and sizing in regional distribution networks using the Sea Horse Optimization (SHO) algorithm and the Backward Forward Sweep Method (BFSM). Four distinct situations, each with a different type of DG, are used to systematically analyze the network using the Torrit software and the MATPOWER toolkit in MATLAB. The main goal is to find the ideal power factor for the deployment of distributed generation (DG), with an emphasis on active and reactive power injection and absorption. The results show notable improvements in voltage profiles and considerable reductions in power losses across a range of deployment scenarios. In addition, the study investigates how various DG types interact with one another to improve overall network efficiency. An extensive analysis comparing IESCO regional distribution networks to IEEE 33-bus and 69-bus systems confirms the efficacy of the suggested techniques. In addition, the study compares the outcomes of the SHO algorithm with those of the Artificial Bee Colony (ABC) algorithm to ascertain which method is more accurate. In summary, this research offers utility operators useful information by proposing methods to boost power system efficiency, boost dependability, and refine power distribution management decision-making procedures.

KEYWORDS: Distributed generation (DG) Backward Forward Sweep Method (BFSM) analysis Optimal sizing and placement of DGs Voltage Loss Power Factor Sea Horse Optimization (SHO).

TABLE OF CONTENTS

Approval For Examination	ii
Author's Declaration	iii
Plagiarism Undertaking	iv
Dedication	ii
Acknowledgements	iii
Abstract	iv
Table of Contents	v
List of Figures	X
Abbreviations	xvii
1 Introduction	
1.1 Overview	
1.2 Motivation and Problem Statement	
1.3 Objectives	
1.4 Limitations	
1.5 Organization of Thesis	24
2 Literature Review	
2.1 Discussion on Literature Review	
3 Methodology	
3.1 Objectives to be minimized	
3.1.1 Active Power Loss	

	3.1.2	Reactive Power Loss	35
	3.1.3	Voltage Deviation Index	36
	3.1.4	Network Power Factor Index	37
	3.1.5	Optimization Constraints	38
-	3.2 Loa	ad Flow Analysis	39
	3.2.1	Backward Forward Sweep Method Analysis	39
	3.3 Use	er Manual for Operating Torrit with MATLAB:	41
	3.3.1	Introduction	41
	3.3.2	System Requirements	41
	3.3.3	Installation	41
	3.3.4	Getting Started	44
	3.3.5	Detailed Steps for MATLAB Code Burning	46
	3.3.6	Performing Load Flow Analysis	49
	3.3.7	Optimization Algorithms for DG Placement	53
	3.3.8	Load Flow and Optimization Results	55
	3.3.9	Visualizing Results with Figures and Graphs	68
	3.3.10	Example Scenarios	70
4	Propose	ed Algorithm	73
2	4.1 Sea	a Horse Optimization (SHO) Algorithm	73
	4.1.1	Sea Horses Mobility Patterns	76
	4.1.2	Predation Behavior Phase	77

	4.1.3	Reproduction phase of Sea horses	17
5	Results	And Discussions	19
5	5.1 IEE	EE 33 Bus Network System:	30
	5.1.1 Algoritl	Backward Forward Sweep (BFS) Analysis of Network Using SH	
	5.1.2 Algoritl	Backward Forward Sweep (BFS) Analysis of Network Using AB	
	5.1.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 8	35
5	5.2 IEE	EE 69 Bus Network System:	35
	5.2.1 Algoritl	Backward Forward Sweep (BFS) Analysis of Network Using SH	
	5.2.2 Algorith	Backward Forward Sweep (BFS) Analysis of Network Using AB	
	5.2.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 9	90
5	5.3 IES	CO 15 Bus Network System:	<i>•</i> 0
	5.3.1 Algorith	Backward Forward Sweep (BFS) Analysis of Network Using SH	
	5.3.2 Algoritl	Backward Forward Sweep (BFS) Analysis of Network Using AB	
	5.3.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 9) 5
5	5.4 IES	CO 18 Bus Network System:) 5
	5.4.1 Algoritl	Backward Forward Sweep (BFS) Analysis of Network Using SH	

5.4.2	Backward Forward Sweep (BFS) Analysis of Network Using ABC
Algor	rithm:
5.4.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 100
5.5 II	ESCO 20 Bus Network System:
5.5.1 Algoi	Backward Forward Sweep (BFS) Analysis of Network Using SHO rithm:
5.5.2 Algor	Backward Forward Sweep (BFS) Analysis of Network Using ABC rithm:
5.5.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 106
5.6 2	2 Bus Network System: 106
5.6.1 Algor	Backward Forward Sweep (BFS) Analysis of Network Using SHO rithm:
5.6.2 Algoi	Backward Forward Sweep (BFS) Analysis of Network Using ABC
5.6.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 111
5.7 2	6 Bus Network System:112
5.7.1 Algor	Backward Forward Sweep (BFS) Analysis of Network Using SHO ithm:
5.7.2 Algoi	Backward Forward Sweep (BFS) Analysis of Network Using ABC rithm:
5.7.3	Comparison of Convergence Curve between SHO and ABC Algorithm: 117
5.8 C	Overall Result Comparison between Algorithms

	5.8	.1	Visualization of results comparison with SHO Algorithm:	122
	5.8	.2	Visualization of results comparison with ABC Algorithm:	123
	5.8	.3	Comparison between performance of Both Algorithms:	124
6	Co	nclus	sion And Future Work	126
(5.1	Cor	nclusion	126
(5.2	Fut	ure Work	126
Re	feren	ces		127
7	An	nexu	re	133

LIST OF FIGURES

Figure 1 Flowchart of Forward Backward Sweep (FBS) Method	
Figure 2 Download from MATPOWER website	
Figure 3 Extracted MATPOWER directory	
Figure 4 Run Install_MATPOWER	
Figure 5 MATPOWER Installation Options	
Figure 6 Open Network file	
Figure 7 Are branch impedances already in per unit	
Figure 8 Are loads already in MW	
Figure 9 Choose an Option for Analysis	
Figure 10 Choose Load Flow Algorithm	
Figure 11 Load Flow Analysis with DG or exit	50
Figure 12 Enter Population Size	51
Figure 13 Enter Maximum Iterations	51
Figure 14 Select DG type	
Figure 15 Enter Number of DGs	
Figure 16 Enter DG size range	
Figure 17 Average Voltage Comparison without and with DGs Graph	69
Figure 18 Voltage Comparison without and with DGs Graph	69
Figure 19 Comparison of Active Power Loss without and with DGs	
Figure 20 Reactive Power Loss Comparison without and with DGs	70

Figure 21 Flow Chart of SHO Algorithm
Figure 22 Torrit View of IEEE 33 Bus Network System
Figure 23 Active Power Loss (APL) in term of DG placement
Figure 24 Reactive Power Loss (RPL) in term of DG placement:
Figure 25 Average Voltage (Vavg) showing before and after placement of DGs
Figure 26 Minimum Voltage (Vmin) showing before and after placement of DGs 81
Figure 27 Active Power Loss (APL) in term of DG placement
Figure 28 Reactive Power Loss (RPL) showing before and after placement of DGs 83
Figure 29 Average Voltage (Vavg) showing before and after placement of DGs
Figure 30 of Minimum Voltage (Vmin) showing before and after placement of DGs 83
Figure 31 Comparison of Convergence Curve between SHO and ABC Algorithm 85
Figure 32 Torrit View of IEEE 69 bus network
Figure 33 Active Power Loss (APL) showing before and after placement of DGs 86
Figure 34 Reactive Power Loss (RPL) showing before and after placement of DGs 86
Figure 35 Average Voltage (Vavg) showing before and after placement of DGs
Figure 36 Minimum Voltage (Vmin) showing before and after placement of DG 86
Figure 37 Active Power Loss (APL) showing before and after placement of DGs 88
Figure 38 Reactive Power Loss (RPL) showing before and after placement of DGs 88
Figure 39 Average Voltage (Vavg) showing before and after placement of DGs
Figure 40 Minimum Voltage (Vmin) showing before and after placement of DGs 88
Figure 41 Comparison of Convergence Curve between SHO and ABC Algorithm 90

Figure 42 Torrit view of IESCO 15 Bus Network System
Figure 43 Active Power Loss (APL) showing before and after placement of DGs91
Figure 44 Reactive Power Loss (RPL) showing before and after placement of DGs 91
Figure 45 Average Voltage (Vavg) showing before and after placement of DGs91
Figure 46 Minimum Voltage (Vmin) showing before and after placement of DG91
Figure 47 Active Power Loss (APL) showing before and after placement of DGs 93
Figure 48 Reactive Power Loss (RPL) showing before and after placement of DGs 93
Figure 49 Average Voltage (Vavg) showing before and after placement of DGs93
Figure 50 Minimum Voltage (Vmin) showing before and after placement of DGs93
Figure 51 Comparison of Convergence Curve between SHO and ABC Algorithm95
Figure 52 Torrit view of IESCO 18 Bus Network System
Figure 53 Active Power Loss (APL) showing before and after placement of DGs 96
Figure 54 Reactive Power Loss (RPL) showing before and after placement of DGs 96
Figure 55 Average Voltage (Vavg) showing before and after placement of DGs96
Figure 56 Minimum Voltage (Vmin) showing before and after placement of DGs96
Figure 57 Active Power Loss (APL) showing before and after placement of DGs 98
Figure 58 Reactive Power Loss (RPL) showing before and after placement of DGs 98
Figure 59 Average Voltage (Vavg) showing before and after placement of DGs98
Figure 60 Minimum Voltage (Vmin) showing before and after placement of DG98
Figure 61 Comparison of Convergence Curve between SHO and ABC Algorithm 100
Figure 62 Torrit view of IESCO 20 Bus Network System 101

Figure 63 Active Power Loss (APL) showing before and after placement of DGs 102
Figure 64 Reactive Power Loss (RPL) showing before and after placement of DG 102
Figure 65 of Average Voltage (Vavg) showing before and after placement of DGs 102
Figure 66 Minimum Voltage (Vmin) showing before and after placement of DGs 102
Figure 67 Active Power Loss (APL) showing before and after placement of DGs 104
Figure 68 Reactive Power Loss (RPL) showing before and after placement of DG 104
Figure 69 of Average Voltage (Vavg) showing before and after placement of DGs 104
Figure 70 Minimum Voltage (Vmin) showing before and after placement of DGs 104
Figure 71 Convergence Curve showing comparison between SHO and ABC algorithm.
Figure 72 Torrit view of 22 Bus Network System
Figure 73 Active Power Loss (APL) showing before and after placement of DGs 108
Figure 74 Reactive Power Loss (RPL) showing before and after placement of DGs 108
Figure 75 Average Voltage (Vavg) showing before and after placement of DGs 108
Figure 76 Minimum Voltage (Vmin) showing before and after placement of DGs 108
Figure 77 Active Power Loss (APL) showing before and after placement of DGs 110
Figure 78 Reactive Power Loss (RPL) showing before and after placement of DGs 110
Figure 79 Average Voltage (Vavg) showing before and after placement of DGs 110
Figure 80 Minimum Voltage (Vmin) showing before and after placement of DG 110
Figure 81 Convergence Curve showing comparison between SHO and ABC algorithm
Figure 82 Torrit view of 26 Bus Network System112

Figure 83 Active Power Loss (APL) showing before and after placement of DGs 113
Figure 84 Reactive Power Loss (RPL) showing before and after placement of DGs 113
Figure 85 Average Voltage (Vavg) showing before and after placement of DGs 113
Figure 86 Minimum Voltage (Vmin) showing before and after placement of D 113
Figure 87 Active Power Loss (APL) showing before and after placement of DGs 116
Figure 88 Reactive Power Loss (RPL) showing before and after placement of DGs 116
Figure 89 Average Voltage (Vavg) showing before and after placement of DGs 116
Figure 90 Minimum Voltage (Vmin) showing before and after placement of DGs 116
Figure 91 Convergence Curve showing comparison between SHO and ABC algorithm
Figure 92 Visualization of results comparison with SHO Algorithm 122
Figure 93 Visualization of results comparison with ABC Algorithm 123

LIST OF TABLES

Table 1 Voltage at buses	
Table 2 Active Power Loss at branches	56
Table 3 Reactive Power Loss at branches	58
Table 4 MOI with Iterations	59
Table 5 Voltage at buses with DGs	61
Table 6 Voltage at busses with DG	62
Table 7 Total Active and Reactive Power Loss with respect to iterations	63
Table 8 Results Summary Using SHO Algorithm	
Table 9 Results Summary Using ABC Algorithm	
Table 10 Results Summary Using SHO Algorithm	
Table 11 Results Summary Using ABC Algorithm	
Table 12 Results Summary Using SHO Algorithm	
Table 13 Results Summary Using ABC Algorithm	94
Table 14 Results Summary Using SHO Algorithm	97
Table 15 Results Summary Using ABC Algorithm	99
Table 16 Results Summary Using SHO Algorithm	103
Table 17 Results Summary Using ABC Algorithm	105
Table 18 Results Summary Using SHO Algorithm	109
Table 19 Results Summary Using ABC Algorithm	111
Table 20 Results Summary Using SHO Algorithm	115

Table 21 Results Summary Using ABC Algorithm	117
Table 22 Overall Result Comparison between Algorithms	118
Table 23 Comparison between performance of Both Algorithms	

ABBREVIATIONS

ABC	Artificial Bee	IWD	Intelligent Water
Colony		Drop	
APL	Active Power Loss	IWO	Invasive Weed
BA	Bat Algorithm	Optimization	
BBO	Biogeography	LF	Load Flow
Based Optimization		MI	Maximum
BFA	Bacterial Foraging	Iterations	
Algorithm		NRLF	Newton Raphson
BFO	Bacterial Foraging	Load Flow	
Optimization		PSO	Particle Swarm
BSO	Backtracking	Optimization	
Search Algorithm		RDS	Radial Distribution
CBs	Capacitor Banks	System	
CSA	Cuckoo Search	RPL	Reactive Power
Algorithm		Loss	
DG	Distributed	SHO	Sea Horse
Generators		Optimization	
EA	Efficient	SKHA	Strud Kill Herd
Analytical		Algorithm	
BFSM	Backward Forward		
Sweep Method			
GA	Genetic Algorithm		
HGWO	Hybrid Grey Wolf		
Optimization			

Chapter 1

Introduction

1 INTRODUCTION

1.1 Overview

The modern electrical power system is a complex network consisting of three interconnected components: There are also three main stages involved in the process which includes generation, transmission and distribution. Generating plants generate electricity and this electricity is transmitted through the transmission networks across long distances. After that, a distribution network supplies electricity to consumer directly usually organized in radial system of distribution.

The distribution network is a vital part of the national power system in Pakistan whereas it is experiencing lot of challenges. Majority of power losses are normally experienced in both primary and secondary distribution networks due to such factors as long transmission distances, low voltage conditions, technical losses, and poor power factor control.

In view of the growing population that comes with urbanization and industrialization, electricity demand increases and hence the pressure on the distribution companies to deliver efficiently. Distribution networks are complex, and thus the operation of distribution networks is likely to cause system instability, increased power dissipation and voltage fluctuations due to high current loads during distribution.

As this thesis's overview demonstrates, distribution networks remain a key component of Pakistan's energy system, and that authoritative changes are required to improve the distribution infrastructure, robustness, and reliability in the power sector.

1.2 Motivation and Problem Statement

To illustrate this let us think of the electricity system in your neighbourhood. It is similar to the electrical power system where there are many wires that bring power to your home. Today, we are integrating small power producers such as PV panels, wind farm and others in this system. It is eco-friendly, however, it still has its challenges sometimes. The issue which I am attempting to address in this study is how to place these small power sources in the right positions and ensure that they are of the right size. This is something we would wish to achieve so that there is no wastage of electricity and all systems run as supposed to. Occasionally, the electricity can become a bit complex and we would like to improve it. Another way that one can know the performance of the electricity system is something called the "power factor" This should ideally be 1:1 but it is never exactly like that. We also need to reduce losses as electricity flows through the wires so that we do not use up all the energy that is being transmitted.

Distributed generations or DG systems have been integrated at the distribution networks, which have affected their operation and performance. There are several advantages that can be achieved through the use of DG resources and these include; Improvement of power quality within the network, and minimization of system loss. In this regard, one of the important issues that are important to address is the use of DG and its size for maximizing power factor in distribution networks.

The power factor is one of the most important parameters affecting the effectiveness of power distribution as well as its quality. This in turn causes high losses, voltage drop and a poor overall network performance due to low power factor. Thus, the primary purpose of this study is to address the following question: 'How can DG units be sized and located properly in distribution networks in order to achieve the least losses and the best power factor?

The progressive increase in the electrical load demand in the last few decades has added to the total line losses and voltage drops in distribution systems as compared to the transmission systems since their X/R ratio is lower [1]. In the past, the number of networks has been used as the conventional way of addressing these problems [2]. However, it has been realized that about 80% of the added infrastructure is underused as most of them are designed to meet peak load requirements that are only valid for 5% of the total working time [3].

Over the years, many methods have been introduced to mitigate the problems associated with power loss and voltage drop such as proper positioning of capacitors and injection of reactive power. Although capacitor bank placement seems to hold a lot of potential for the said methods, they do not usually offer voltage profiles that are above the nominal value of 1 p.u., primarily because of the passive nature of radial distribution systems. As a result of this inherent passivity and in order to achieve the goal of minimizing voltage drop, several approaches have been proposed among which is the incorporation of renewable energy sources known as Distributed Generation (DG).

Distributed Generation technologies have become popular due to the flexibility of installation and short time of response [4]. Sized and located appropriately, DGs have been proven to reduce tremendously line losses and voltage drops in electrical grids. Thus, both as generators of active power and as reactive power suppliers, DGs can be useful load that assists in reducing peak load demands; in turn, it makes it possible to dampen pulsing loads, minimize power losses and improve the stability of the power factor [5]. These are some of the complex issues that make the planning of the size and location of DGs in distribution networks very systematic, and for this reason, this thesis focuses on the following.

It is impossible to facilitate working electrical substation-connection distribution networks with or without Distributed Generators (DGs) which are vital in determining power factor optimization/size/placement. Thus it can be summarised that improvement of power factor is required for efficient operation of power transmission and distribution network. Distributed generators (DGs) playing a vital role in this regard since they can actively control reactive power which offsets the imbalance caused by inductive loads such as motors and transformers. DGs reduce the associated transmission losses that are usually experienced due to low power factor and also help in controlling and balancing of voltage by providing reactive power as a needed. DGs are categorized based on the power that is in the form of either reactive or active power that an islanded DG can consume or supply. The following are the four DG classifications:

Active Power Injection or Type 1 DG Just Solar photovoltaic (PV) systems are known as type 1 DGs these mainly inject active power (P) to the grid. Solar photovoltaic systems directly inject electrical power generated by sunshine to the distribution network. Reactive power is not generally produced by these systems in any isolated manner. For this reason, they are considered to be responsible for no contribution to the management of reactive power of the system but are only involved in the generation of active power. Reactive Power Injection, or Type 2 DG As for any capacitor bank type 2 DG inject reactive power (Q) to the system. They include the control of voltage by use of reactive power and ensuring that satisfactory power is available in the network. Capacitor banks to be connected in the system for controlling the amount of reactive power can be connected in the system according to the requirement. They assist in interlinking the PF and voltage regulation by providing for required short in the distribution network.

Active and Reactive Power Injection, type 3 DGs can inject power both in terms of real power (active power 'P') and Reactive power 'Q'. Examples of such equipments are Synchronous machines and certain types of wind power turbines. Due to the flexibility in the design of synchronous machines, they can be tailored to alter the amount of reactive and real power which is generated as one deems fit. In addition, this capability helps them to participate in creation of active power and to support stability of the grid by providing reactive power if required.

Active Power Injection with Reactive Power Absorption is Type 4 DG. The 4th type of DGs impose both active power (P) into the grid and absorb the reactive power (Q), for instance, wind power producers using Induction generators. For the purpose of running induction generators used in wind turbines, reactive power has to be taken from the grid. Therefore, they supply reactive energy which the distribution system has to obtain from other sources in order to maintain the voltage stability with the necessary reliability.

In this thesis, four types of DGs are introduced as follows: Each of which plays a specific role to improve and steady distribution networks with its respective advantages. With proper size and location of these DGs, it has been reported to reduce line loss up to 50%, produce optimum voltage profiles and enhance the efficiency and reliability of the grid. To fully harness such advantages whilst at the same time reducing the total energy costs and power losses, this thesis analyses the most suitable method for incorporating these DG kinds.

Achieving an ideal capacity for DG installation requires correlation with load demand in the various regions and voltage stability. This ensures that DGs to be employed are optimally used in meeting the requirements of the reactive power. Some types of inverters and some synchronous DGs may change their output voltage so as to maintain constant power factors which in turn enhance the systems performance in general. Intentionally sited DG systems are efficient since they minimize loss as they distribute energy and at the same time increase stability of the electricity grid.

In addition, DGs are self-inherently contributing to high power-factor qualities particularly when they derive from sources such as solar and wind. With the reduction of fossil base DGs, incorporation of these renewable distributed generation units (DGs) in the grid also enhances the environmental quality, while at the same time enhancing voltage control and grid stability.

1.3 Objectives

The objectives of this research is to improve the sustainability, dependability, and efficiency of power distribution networks. These objectives address significant aspects of network transmission in order to optimize distribution system performance. Research endeavors to enhance the resilience and efficiency of the electrical infrastructure by mitigating power losses, augmenting power factor, and maintaining constant voltage levels.

. The following are the main goals:

- Minimization of Active Power Loss (APL)
- Minimization of Reactive Power Loss (RPL)
- Improvement of Network Power Factor (PF)
- Minimization of Voltage Deviation Index (VDI)

These objectives collectively aim to optimize distribution network performance by addressing key challenges in power management and system efficiency. Achieving these objectives will enhance operational reliability, reduce environmental impact, and support the transition towards a more sustainable energy future.

1.4 Limitations

The limitations always occur with some extension:

- Limited Data Availability: Access to comprehensive and up-to-date data on network topology, load profiles, and operational parameters from IESCO's distribution networks may be restricted due to privacy and regulatory concerns.
- Data Consistency and Accuracy: Variations in data quality, such as incomplete datasets or measurement errors, can impact the reliability and precision of research findings and modelling outcomes.
- **Temporal and Spatial Resolution:** Data granularity, both in terms of temporal (time intervals) and spatial (geographical detail), may be insufficient for detailed analysis. Some IESCO grids still note and calculate data manually, which weakens accuracy and reliability.

1.5 Organization of Thesis

This thesis research consists of following chapters:

- Chapter 1 reviews introduction of the power system, motivation & problem description, objectives, limitations and organization of dissertation.
- Chapter 2 describes the literature review.
- Chapter 3 describes Backward Forward Sweep load flow analysis and mathematical mode of radial distribution.
- Chapter 4 explains the proposed SHO algorithm in comparison with ABC algorithm.
- Chapter 5 details the test system and the results obtained after the implementation of algorithm.
- Chapter 6 summarizes the conclusions and some future works.

Chapter 2

Literature Review

2 LITERATURE REVIEW

If distributed generation (DG) units are to reduce power losses and enhance voltage profiles, they must be scaled and placed in distribution networks in compliance with best practices. The connection between system performance and the DG power factor has been the subject of numerous studies.

There was a proposed flow placement, sizing, and power factor concerning several units of DG that utilized the Differential Evolution algorithm. The authors also pointed out that, compared to those DGs that have a unity power factor, the DGs which have a lagging power factor have proved to be more effective in terms of losses and voltage profile [1]. Another research presented an analytical index, which incorporates reliability-based index, voltage stability index and loss sensitivity factor to decide the precise location and size of the distributed generators. The application of the proposed index was tested and analyzed on IEEE 33 and 69 bus distribution systems to show the impact of DGs at leading power factor on the network reliability and voltage stability [1].

A study was therefore conducted with the objective of reducing power losses and enhancing the voltage profile by ascertaining with high accuracy the precise location and size of DG units by using three major indices, the IVM (Index Vector Method), VDI (Voltage Deviation Index), and VSI (Voltage Stability Index). The two types of DGs which the authors considered were the lagging power factor and unity power factor. They realized that it was the larger DGs that had the best overall outcome as far as minimum bus voltage and losses were concerned if operated at a power factor of 0. 9 and the VSI index the findings showed that the global gender gap report, the human capital index, and the VSI index were all correlated with each other. [2]

Another study investigated the places that are suitable for DG in various forms inside distribution networks. This was made possible because the authors managed to keep the total power distribution loss as small as possible to determine the most appropriate size and location of DGs. They also found that the quality of the supply to the DG also has an effect especially the ideal power factor of the system. [3]

There is another study which described how actual power losses can be minimized in primary distribution networks using distribution generators in an optimal way through employing reactive power and active power compensation particle swarm optimization method. The right power factor was also quantified in order to minimize energy loss. Thus we can state that the analytical technique can be applied for smaller systems to determine the optimal size and location within the system as for each system bus, the technique offers the best solution. Heuristic approaches are again better suited for the larger systems since the search of a solution is faster. [3]

The loss sensitivity factors method is employed to identify potential sites for Distributed generation (DG) which depends with the node's loss and their sensitivity after compensation. Therefore, the positions for distributed generators which should be connected in the system are identified as those which cause the highest bus real and reactive power losses. As mentioned before, when DG are integrated into the distribution system, the buses with high losses minimize the level of losses. [3]

How to place DG optimally; the placement of DG to minimize active power losses in the distribution networks has been the focus of many researchers. A new and versatile voltage stability index was proposed in one thesis to determine both the conductor stability limit as well as the optimal conductor size. Optimum switching configuration for a microgrid, according to the authors, were arrived at through reconfiguration with the help of BAT method thereby minimizing the loss of active power. Further, the proposed BAT algorithm was used to identify the ideal position and sizes of DG sources for better stability and minimized power loss. [4]

Another study provided a very effective method for reducing total power losses in distribution networks adding PV systems coupled to capacitors. From the obtained outcome it can be inferred that apart from the reduction of loss through the installation of capacitors, the installation of PV system may contribute significantly to the reduction of loss, as well as the number of capacitors installed [5].

The literature has also analyzed of how capacitors influence the minimization of power loss. One research has shown that the two factors that affect the reduction of the active power losses are load reactive power and capacitor capacity. To avoid increase of the overvoltage and the power losses, the size of the capacitor should be varied from zero to the reactive power of the load but not more. The contribution of the researchs also highlights the importance of the network power factor on the distribution network performance in the literature. One of the solutions after analyzing and comparing for the active power transmission loss allocation in power pools was that a method for improving the power factor at the network level [6].

Another factor that must be considered when placing the DG units at an optimal locations is the voltage fluctuations. To determine the conductor diameters needed for the best results and conductor stability limit, a study proposed a new voltage stability index. According to the authors, to minimize the voltage deviation index, the best switching arrangements together with the best DG locations were defined by using a BAT algorithm [6].

Here, load requirement is designated to increase significantly, which in turn makes the network more sensitive to voltage fluctuations. Apart from increas-ing the conductor size 'reconductoring', applying higher voltage levels, inducting capacitors, and network reconfiguration are some of the techniques that are applied to reduce or even eradicate such losses. Distributed generation (DG) systems can help in enhancing the voltage level and combating power outages at the same time and add efficiency to the distribution network [7]. DGs are located in the distribution network. The integration of distributed generation (DG) units in distribution networks has recently increased and this lead to an analysis of the effects of such units on the power grid from the environmental, technical and economical perspective.

The literature reveals that it is possible to have multiple techniques for allocating DGs into various distribution grids in an optimal manner. For the optimization of the size and the location of DG the authors employed the genetic algorithm (GA) inspired by nature in [8]. The goal and that is, reducing loses, has been achieved by the author. The authors have evaluated the performance of the presented method for IEEE 16, 37 and 75-bus distribution system. To find the optimal location of multiple distributed generators (DGs) in a microgrid, the authors in [9] have presented an improved re-initialized socially structured particle swarm optimisation technique known as IRS-PSO. The authors have applied the proposed method on IEEE 69-bus radial distribution system.

Most of the authors in [10] proposed the use of the sensitivity test technique to identify where and of what size to locate the DGs. The lone goal, that is to reduce power losses, has been achieved perfectly by the author. The authors have applied the given

approach on IEEE 33- and 69-bus systems. A new heuristic algorithm of the optimized sizing and positioning of a single DG using an artificial bee colony (ABC) has been developed to minimize power loss in the system. The authors have evaluated the proposed method in IEEE 33 and 69-bus systems and states that new method is much better than PSO and GA with varying parameters [11].

Several DGs' placement was optimized in [12] by utilizing hybrid grey wolf optimization (HGWO). The authors' single aim of minimizing the power loss was achieved when the authors applied the proposed method on the IEEE 33, 69 and Indian 85-bus radial networks. According to authors [13], BA is the best approach to provide the optimal PV power assignment. This is the real power loss and the authors endeavour to decrease its value. Implementation of the proposed approach was carried out on IEEE 33-bus distribution model. In [14], the authors applied particle swarm optimization (PSO) method in the allocation of capacitors and DGs. It was to make the highest level of profit as could be envisaged in a given period. IEEE 33 and 69 Bus distribution networks were used to assess the effectiveness of the presented technique.

Many DGs positioning was studied in [15] using the strud krill herd algorithm (SKHA). The authors have achieved their set objective of minimizing line loss, and they have tested there proposed method for test being the 33, 69, and the Portuguese 94 buses in radial distribution grid. The authors of [16] introduced the augmented PSO approach that was used to determine the location of the DGs that provides the best combinations of parameters as shown below. It was for the purpose of making the maximum amount of profit possible. The above-discussed method is proposed on an IEEE 34-bus electrical distribution network. In [17] the authors used PSO to assess the impact of DG location and size. The authors had the main goal of minimizing the cases of power interruptions. The effectiveness of the proposed technique was investigated for IEEE 33 and 69-bus systems.

The authors in [18] have used bat inspired algorithm (BA) to maximize the size and placement of a stand-alone DG. The methodology proposed in this research was simulated on test power networks namely the IEEE 33 and 69-bus networks. In [19], a PSO and gravitational search method-based population structure was proposed to address the issue of optimal sizing of DGs as well as their location in the distribution network. In [20], the authors have used bacterial foraging algorithm (BFA) to allocate DGs in different models such as IEEE 12, 34 and 69-bus distribution models effectively. The objectives that were set aimed at a higher bus voltage and a reduction of the power losses. Whereas, in [21], it was revealed by adopting an efficient analytical (EA) technique that the best distribution of all the various DGs. Such type of losses are called no load losses or losses that occur even when there is no current passing through the transformer.

Authors in presented a genetic algorithm (GA) for the optimal placement and sizing of diode generators (DGs) and capacitor banks (CBs) [22]. The two main objectives were power loss reduction and reliability. The proposed approach was verified in IEEE radial systems. One technique for placing PV arrays is biogeography-based optimization (BBO), as proposed by the authors in [23]. The sole objective is to minimize power loss. The proposed optimization method was evaluated in IEEE 33 and 69-bus radial systems. Harmony Search Algorithm (HSA) has been employed by authors to simultaneously find the DG optimal site and reconfigure the network by using power loss minimization as the only objective function [24].

The authors in [28] have proposed a hybrid method (HA) to appropriately size and place DGs. The DGs were placed by PSO, and their size was tuned through analytical techniques. The proposed method was assessed using standard IEEE networks. The objectives were to improve the voltage profile and decrease total power loss. As proposed by the authors in [29], one technique for determining the optimal location and capacity for DGs is backtracking search optimization (BSO). The proposed technique was tested on radial networks with Portuguese 94-bus and IEEE 33-bus traffic. The objective was to minimize power loss and voltage drop.

Bacterial foraging optimization (BFO) was used in [30] to propose the placement and dimensions of a single DG in the IEEE distribution system. Lowering active power loss, operating expenses, and voltage drop were the objectives. Natural intelligent water drop (IWD) optimization has been proposed in [31] to optimize the placement of DGs. The proposed optimization method is tested on IEEE 10, 33, and 69-bus radial grids. The objective was to lower total line loss.

None of the examined research have explicitly looked into determining the ideal power factor for placing distributed generation units, despite the fact that the published literature offers a variety of optimization strategies for this purpose. Most research has concentrated on goals such voltage profile enhancement and active power loss minimization, ignoring the effect of DG power factor on system performance. In order to close this gap, the current work suggests a novel way for deciding on the best location and size of distributed generation units (DG) units while taking their power factor into account.

This approach is based on the sea horse optimization (SHO) algorithm and the Backward-Forward Sweep (BFS) method. The BFS technique is used to efficiently assess the power flow in radial distribution networks, while the SHO algorithm is used to optimize the placement and sizing of distributed generators (DGs) with respect for power factor as a vital parameter. By integrating the power factor as a crucial element in the optimization process, the proposed method aims to provide a more comprehensive and workable solution for the optimal placement of DG units in distribution networks [32]. Better power loss results are expected when comparing the BFS-SHO method to other existing techniques that do not explicitly account for the impact of the DG power factor.

2.1 Discussion on Literature Review

It is evident from the literature review that most of the research focus on DG placement and sizing to minimize the power loss or/voltage profiles or both using various techniques. The goals are comprised in a broad range of objectives to enhance the performance of the distribution network as listed below; minimise active power loss (APL), minimise reactive power loss (RPL), improve the network power factor (PF), and maximise the voltage deviation index (VDI). Combined, these objectives aim at enhancing operational reliability, minimizing impact on the environment and towards the implementation of low carbon economy. Achievement of these goals will go a long way in improving the efficiency of the system and management of power. [32]

To achieve these goals, the thesis employs the advanced Sea Horse Optimization (SHO) algorithm, a newly developed technique, as a new approach for searching for useful solutions based on the actions of seahorses when foraging for food. Many variations of the SHO algorithm in the search patterns, which enables the addition of diverse solutions and reduce the parameter of local optimization. The conventional Artificial Bee Colony (ABC) algorithm, which is a swarm intelligence-based method

inspired by bees' foraging process, is also employed for comparison purposes. Thus, the limit cycle capacity of the ABC method increases solution variety by decreasing the possibility of local optimization [33].

Furthermore, the Sea Horse Optimization (SHO) algorithm was employed at first in two analyses. The first analysis concentrated on minimizing the four objectives in IEEE 33-bus and 69-bus radial networks, as well as some regional distribution networks, such as those of IESCO: Active power loss (APL), reactive power loss (RPL), correction of network power factor (PF) and voltage deviation index (VDI). These outcomes were then compared with the outcomes of other preceding used algorithms such as the Artificial Bee Colony (ABC) algorithm. The ABC method was also applied in the second analysis to respond to the same goals mentioned above. This was possible because both studies were conducted across all the four categories of DGs thus ensuring an effective assessment of the DGs.

Chapter 3

Methodology

3 Methodology

The approach combines the Backward-Forward Sweep (BFS) method and the Sea Horse Optimization (SHO) algorithm to achieve key objectives: Minimizing active power loss which is signified by APL, decreasing reactive power loss which is abbreviated as RPL, increasing the power factor of the network which abbreviated PF and lastly maintaining a Voltage deviation index abbreviated as VDI.

The performance of the proposed methodology is confirmed by comparison with the recognized IEEE 33-bus and IEEE 69-bus radial distribution networks. Further, the methodology will be employed on the regional distribution networks supplied by IESCO to have a practical implementation aspect in the study.

For load flow analysis in radial distribution networks, the BFS method will be used and for identification of DG placement and sizing the SHO, which is inspired from the foraging behavior of seahorses will be used. Another algorithm for comparison with the results is the Artificial Bee Colony (ABC) Algorithm as well.

The optimization process will involve using objective functions that will define the outcome metrics to be optimized such as the active power loss, reactive power loss, network power factor and the voltage deviation index.

To further prove the concept, the methodology will be conducted across a variety of DG units to demonstrate the versatility of the methodology to different power factor operating limits. Ideas from the SHO and ABC algorithms will be discussed in the context of other optimization methods for the clarity of the usefulness of the proposed method. The effectiveness of the presented methodology will be showcased in more detail by applying it to distribution networks of IESCO.

This work gives an overall method that offers a framework of analyzing and enhancing DG distribution into networks with custom designed power factor. Through the use of state-of-the-art optimization methodologies and highly detailed simulation models, the proposed research is expected to increase network performance, decreased running expenses and produce a more reliable system.

3.1 Objectives to be minimized

Reducing reactive power loss, cutting active power usage, and raising the network's power factor are the goals of the optimization task. **OBF** is the notation for the objective function. The values of the weights *W1,W2,W3* and *W4* are all equal at 0.25. The index for active power loss is called **APL**, the index for reactive power loss is called **RPL**, the index for Voltage Deviation Index is **VDI** and the network power factor index is called **PFI** [35].

$$OBF = w_1 \cdot APL + w_2 \cdot RPL + w_3 \cdot VDI + w_4 \cdot PFI$$
(3.1)

3.1.1 Active Power Loss

The first objective is minimization of total active power loss P_{loss} (kW), Where, P_{loss} is total active power loss in kW. APL is defined as the ratio of active power loss after the allocation of Distributed Generators (DGs) to the active power loss before the allocation of DGs. Active power loss refers to the energy that is lost as heat in electrical power systems, primarily due to the resistance of the electrical conductors (like wires and transformers) that carry the current. This type of power loss occurs when electric current flows through a conductor with resistance, causing energy to be dissipated as heat according to Joule's law. The formulation is given by [36] :

$$P_{loss}(i,j) = R_{ij} \cdot \frac{(P_i^2 + Q_i^2)}{|V_i|^2}$$
(3.2)

$$APL = \frac{APL_{after_DG}}{APL_{before_DG}}$$
(3.3)

3.1.2 Reactive Power Loss

Reactive power loss is prevalent in power systems because all systems contain inductive and capacitive components that generate both real and reactive power. While the active power is the power that actual performs useful work, the reactive power is that which needed to sustain the voltage levels required in the power system, but in itself does not deliver energy to the loads. Reactive power loss results into poor efficiencies in the power system as the current flow in the network increases leading to I²R losses or active power loss and increases the need for capacities equipment.

Several methods of power factor improvement can be applied, and one of the most important measures connected with reduction of the reactive power loss is increasing of the value of the power factor in the sense of the ratio of the active power to the apparent power. A low power factor means that a large proportion of the power is reactive and results in larger currents and, therefore, higher losses in the system. Reducing the real power and increasing the apparent power, power factor can be improved while use of equipments such as capacitors or synchronous condensers can help in removing most of the reactive power in the system. Not only does this, it also minimizes the overall current within the system and; therefore, the active power loss as well. Therefore, ensuring an improved power factor optimizes the utilization of power equipment to reduce energy consumption, enables power companies to limit infrastructure development by only having to upgrade instruments, which are overburdened by power factor related challenges. Reactive power loss after DG allocation is defined as the total amount of reactive power loss after the distribution of the integrated DGs. [37].

$$Q_{loss}(i,j) = X_{ij} \cdot \frac{(P_i^2 + Q_i^2)}{|V_i|^2}$$
(3.4)

$$RPL = \frac{RPL_{after_DG}}{RPL_{before_DG}}$$
(3.5)

3.1.3 Voltage Deviation Index

The Voltage Deviation Index (VDI), is an important integer in power systems which coincides with the extent of deviation of voltage levels from their intended nodes. Hence, if voltage deviation is allowed to occur and is allowed to persist, it will have adverse impacts on the operation and effectiveness of an electrical distribution network. when power factor improving measures are used like; proper placement of distributed generations or use of capacitor, the amount of reactive power in power system is regulated. Such negative reactive power flow is constrained when the flow of the reactive power is reduced and this results to controlled voltage in the electrical network. As a result a lower Voltage Deviation Index is likely to be achieved meaning that the Voltages

at various points in the power network is smaller and closer to the optimum. Another implication of a lower level of VDI is better PF implies more sinusoidal voltage and current waveforms – no under-voltage, over-voltage, which are some of the biggest challenges to the efficiency of electric energy and possible failure of electrical appliances. Therefore, it can be seen that measures that lead to an improvement in the measure of the power factor are advantageous in minimizing voltage fluctuations thereby enhancing the stability of the power distribution network. As stated, n is the total number of buses in the system, V_b is the bus voltage after placement of DG in the system and V_{ref} is any reference voltage ~ 1.03 p. u. The formula that can be used to show VDI is as indicated below. [37]

$$VDI = max_{\{b=1\}}^{\{n\}} \frac{\{|V_{\{ref\}}| - |V_b|\}}{|V_{\{ref\}}|}$$
(3.6)

3.1.4 Network Power Factor Index

Measures of the power factor at the network level Determine Network Power Factor Index, otherwise known as PFI are critical in the electrical distribution network. Power factor is the real power which may be used to do work or to deliver useful work to the product of real and reactive powers that flow through the circuit. Figure 8 reveals that higher power factor closer to unity is suggestive of efficient use of electrical power on the other hand a lower value indicates inefficiency due to utilization even of nonproductive or reactive power which does not do useful work but adds to the total apparent power and hence more losses.

Optimizing the power factor at the network level enable the enhancement of the efficiency of the power distribution network. From the table, when the value of the power factor is improved (closer to the value of 1) then it observes a higher PFI which is an added advantage in terms of efficiency and reduction in the use of reactive power. This improvement can be made by some techniques like inserting capacitors or synchronous condensers or Also, the location of Distributed Generation (DG) units can be effectively adjusted. Lower KVAR means that lesser current is required to supply the same amount of real power that in turn reduces the amount of I²R losses (active power loss), reduces the load on the transformers and conductors and hence decreases energy costs and increases the stability of the system. The PFI, therefore, functions as an indicator that utilities use in efforts to fix the power issue at the networks, hence enhancing effectiveness and reliability. PFI is calculated using the following relation:PFI is calculated using the following relation: [34]:

$$PFI = 1 - PF_{with DG}$$
(3.7)

The Network Power Factor with DGs.

$$PF_{with_DG} = \frac{P_g + P_{dg}}{S_{with_DG}}$$
(3.8)

 $S_{with_{DG}}$ is the apparent power of the network after the allocation of DGs and is defined as:

$$S_{\text{with}_DG} = \sqrt{(P_g + P_{dg})^2 + (Q_g + Q_{dg})^2}$$
 (3.9)

The Network Power Factor without DGs is formulated as:

$$PF_{without_DG} = \frac{P_g}{S_{without_DG}}$$
(3.10)

$$S_{\text{without}_\text{DG}} = \sqrt{P_g^2 + Q_g^2} \tag{3.11}$$

3.1.5 Optimization Constraints

This optimization problem is accompanied with a number of constraints such as load flow balance, voltage constraints on buses and DG's output constraints. Total real and reactive power supplied to the network together with that from DGs must be equal to the connected load and total power losses:

$$P_g + \sum P_{dg} = P_{\text{load}} + P_{\text{losses}}$$
(3.12)

$$Q_g + \sum Q_{dg} = Q_{\text{load}} + Q_{\text{losses}}$$
(3.13)

The power output of the DGs must not exceed the total connected load limit:

$$0 \le P_{dg} \le P_{\text{load}} \tag{3.14}$$

3.2 Load Flow Analysis

3.2.1 Backward Forward Sweep Method Analysis

The Backward-Forward Sweep (BFS) method is a cornerstone in the current work as we employ it to study the power flow characteristics of distribution feeders of a region. This conventional numerical method allows the studying of nodal voltage and power flows which are critical in analyzing the effects and benefits of introducing DGs and shunt capacitors into the grid. At its core, the BFS method entails two main steps: two main components which are the forward sweep and the backward sweep. When operating in the forward sweep phase, the algorithm systematically goes through each node in the network starting with the node where the electric power originates and ending at the load terminals. Finally, when convergence is reached the use of the BFS method allows for determination of the power flow in the branches and nodes of the network. This kind of analysis is critical in determining information such as the power losses and voltage profiles which is important when assessing the effectiveness of power factor improvement methods like placement of DGs and shunt capacitor [35].

For radial distribution networks, particularly characterized by a radial structure, low X / R ratio, and weakly meshed design, the BFS method is used in load flow analysis owing to its high efficiency [6]. Opposed to Gauss-Seidel and Newton-Raphson techniques more applicable to transmission networks, the BFS method is able to actually 'map' the

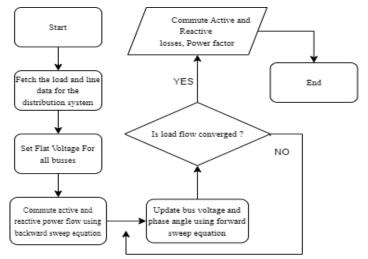


Figure 1 Flowchart of Forward Backward Sweep (FBS) Method

distribution system as a tree in essence with the slack bus as the root node and distant nodes as the leaves. The entire process of the methodology mainly involves the following steps:

- 1. Prepare the load and line data for the distribution system.
- 2. Assume initial voltages at all buses to be 1 per unit (p.u.).
- 3. During the backward sweep, calculate the real and reactive power flows using the following equations [39] :

$$P_i = P_j + P_{L_j} + R_{ij} \cdot \frac{(P_j'^2 + Q_j'^2)}{|V_j|^2}$$

(3.15)

$$Q_i = Q_j + Q_{L_j} + X_{ij} \cdot \frac{(P_j'^2 + Q_j'^2)}{|V_j|^2}$$

(3.16)

4. Update the voltages and phase angles using the forward sweep with these equations:

$$V_j = \left[V_i^2 - 2\left(P_i R_{ij} + Q_i X_{ij}\right) + \left(R_{ij}^2 + X_{ij}^2\right) \cdot \frac{\left(P_i^2 + Q_i^2\right)}{V_i^2}\right]^{1/2}$$

(3.17)

$$\theta_{j} = \theta_{i} + \tan^{-1} \left(\frac{Q_{i}R_{ij} - P_{i}X_{ij}}{V_{i}^{2} - (P_{i}R_{ij} + Q_{i}X_{ij})} \right)$$
(3.18)

5. Compute the power mismatch and check the termination criteria:

$$DeltaS_{i}^{(k)} = S_{i}^{sch} - V_{i}^{(k)} \left(I_{i}^{(k)} \right)^{*} \le \epsilon$$
(3.19)

- 6. If the termination criteria are not met, repeat the process.
- 7. Once the termination criteria are satisfied, compute the branch power losses and total system losses using these equations [39]:

$$P_{loss}(i,j) = R_{ij} \cdot \frac{P_i^2 + Q_i^2}{|V_i|^2}$$
(3.20)

$$Q_{loss}(i,j) = X_{ij} \cdot \frac{P_i^2 + Q_i^2}{|V_i|^2}$$
(3.21)

3.3 User Manual for Operating Torrit with MATLAB:

3.3.1 Introduction

Welcome to the Power System Simulation Tool User Manual. This manual provides a comprehensive guide to installing, configuring, and using the software for load flow analysis and optimization of distributed generator (DG) placement. Whether you're a beginner or an advanced user, this manual will walk you through each step to ensure you can effectively utilize the tool.

3.3.2 System Requirements

Before you start, ensure your system meets the following requirements:

- Operating System: Windows
- MATLAB: Version 2018b or later recommended
- MATPOWER: Version 8.0b1
- Hardware: Sufficient disk space and memory to run MATLAB
- **Processor**: Minimum of dual-core, quad-core or higher recommended for large-scale simulations
- **RAM**: Minimum 4GB, 8GB or higher recommended for large network simulations

3.3.3 Installation

3.3.3.1 Installing MATLAB

- 1. Download MATLAB:
 - Visit the MATLAB official website and download the installer.

• Ensure you have a valid license to activate the software.

2. Install MATLAB:

- Run the downloaded installer.
- Follow the on-screen instructions to complete the installation process.
- Select the components you need based on your usage.

3. Activate MATLAB:

- After installation, activate MATLAB using your MathWorks account.
- Follow the activation instructions provided.

3.3.3.2 Installing MATPOWER

1. Download MATPOWER:

Go to <u>MATPOWER's website</u> and download the latest version as a ZIP file.



Figure 2 Download from MATPOWER website

2. Extract MATPOWER:

• Extract the ZIP file to a directory of your choice on your computer.

3. Install MATPOWER:

• Open MATLAB or directory where you have extracted MATPOWER.

• Open install_matpower file from the extracted folder.

🧵 data	23-Dec-22 9:51 AM	File folder	
J docker	23-Dec-22 9:51 AM	File folder	
docs	23-Dec-22 9:51 AM	File folder	
extras	23-Dec-22 9:51 AM	File folder	
📕 lib	23-Dec-22 9:51 AM	File folder	
🧵 mips	23-Dec-22 9:51 AM	File folder	
🧵 most	23-Dec-22 9:51 AM	File folder	
mp-opt-model	23-Dec-22 9:51 AM	File folder	
Imptest	23-Dec-22 9:51 AM	File folder	
AUTHORS	23-Dec-22 9:51 AM	File	1 KB
CHANGES	23-Dec-22 9:51 AM	Markdown Source	130 KB
CITATION	23-Dec-22 9:51 AM	File	4 KB
CONTRIBUTING	23-Dec-22 9:51 AM	Markdown Source	13 KB
install_matpower	23-Dec-22 9:51 AM	MATLAB Code	15 KB
LICENSE	23-Dec-22 9:51 AM	File	3 KB
README	23-Dec-22 9:51 AM	Markdown Source	19 KB
1 startup	05-Feb-24 1:24 PM	MATLAB Code	2 KB

• Run install_matpower file in matlab.

Figure 3 Extracted MATPOWER directory

HOME PLOTS APPS	EDITOR PUBLISH VIEW	
Image: Compare with the second seco	Inset Comment & 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	
Þ → 💽 🏹 💭 💄 + F: + matpower8.0b1		
	Editor - F:\matpower8.0b1\matpower8.0b1\install_matpower.m	
Name	install_matpower.m × +	
	<pre>1</pre>	

Figure 4 Run Install_MATPOWER

4. Follow Installation Prompts:

- Choose option 3 to do modify the MATLAB path, and SAVE the updated path.
- Select Y to run the MATPOWER tests to ensure everything is installed correctly.

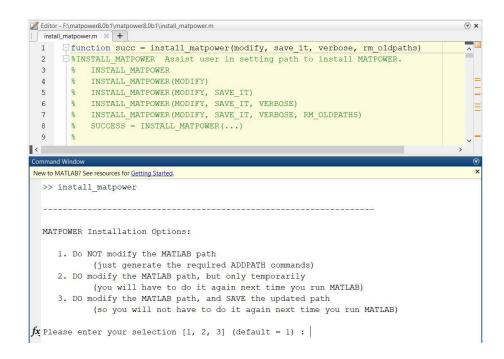


Figure 5 MATPOWER Installation Options

3.3.4 Getting Started

3.3.4.1 Launching the Torrit Software

- 1. Locate the Executable:
 - Find the MATPOWER Case Building with Torrit.exe file on your computer.

2. Open the Software:

• Double-click the executable file to launch the software.

3.3.4.2 Creating and Saving a Project

1. Create a New Project:

• Go to the "File" menu and select "New Project."

- Enter a name for your project and click "OK."
- A new project workspace will open.

2. Save Your Project:

- Regularly save your work by selecting "Save Project" or "Save Project As" from the "File" menu.
- Choose a location on your computer to save the project file.

3.3.4.3 Configuring System Parameters

1. Set System MVA:

- Locate the "System MVA" field at the top right corner of the interface.
- Enter the total power capacity (in MVA) for your system.
- This value is critical for scaling the power flow calculations.

2. Set Swing Bus:

- Choose the swing bus, which serves as the reference point for voltage and phase angle calculations.
- The swing bus is typically the bus connected to the largest generator.

3.3.4.4 Placing and Editing Components

1. Select and Place Components:

- In the "Components" tab, select from generators, loads, buses, and transmission lines.
- Click on the workspace to place each component.
- Use the grid layout to ensure components are placed accurately.

2. Connect Components:

- Use the Transmission Line/Connector tool to connect components by clicking on the starting and ending points.
- Ensure all buses are properly connected to avoid errors in load flow analysis.

3. Edit Component Properties:

- Double-click on any component to open its properties window.
- Adjust parameters such as impedance, power rating, and load values as needed.
- Click "OK" to save changes.

4. Rotate and Drag Components:

- Rotate components by selecting them and pressing the "R" key.
- Drag components by clicking and holding the mouse button, then moving them to the desired location.

3.3.5 Detailed Steps for MATLAB Code Burning

3.3.5.1 Open MATLAB Software

 Upon launching the software, it prompts you to open a network file designed in Torrit software.

the_N	fain.m 🗶 🕇
1	% Clear existing data and command window
2 -	delete *.mat
3 -	clear
4 -	clc
5 -	close all
6 -	commandwindow
7	
8	% Set MATLAB plot properties
9 -	<pre>set(0, 'DefaultLineLineWidth', 1.5);</pre>
10 -	<pre>set(0, 'defaultAxesFontSize', 10)</pre>
11 -	tic
12	
13 -	<pre>disp('Please open network file');</pre>
	Ad max (1) 13 -
Command	Window
New to M	ATLAB? See resources for Getting Started.
Plea	ase open network file

Figure 6 Open Network file

- Browse to the location of matpower file that is designed using torrit software and open it.
- The network data, including bus data, load data, branch data, and generator data, will be loaded into the workspace.

1. Input Prompts for Impedances and Loads

Impedances in Per Unit:

• The software asks if the branch impedances are in per unit (pu).

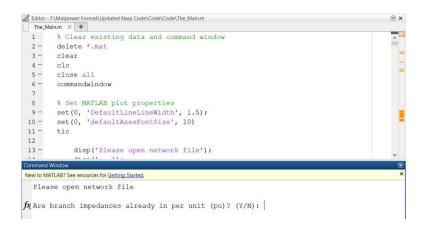


Figure 7 Are branch impedances already in per unit

- If the values are not in per unit, the software will automatically convert them.
- Ensure accurate entry to avoid calculation errors.

Loads in MW:

- The software asks if the load values are in MW.
- If the values are not in MW, the software will convert them accordingly.
- Accurate load data is essential for precise analysis.

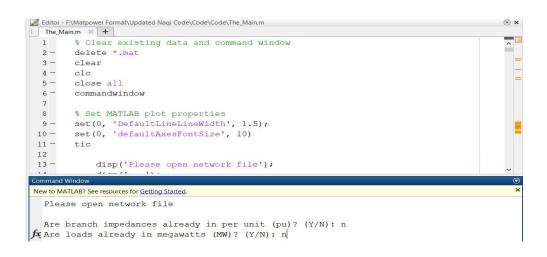


Figure 8 Are loads already in MW

Pro Tip: By default, the torrit software asks the user to enter the impedances and load data in per unit and MW respectively. However, if the user has entered the impedances actual values and loads in kW then the code is capable of making the conversions to per unit and MW.

3.3.5.2 User Choices

1. Choose Analysis Option:

- When prompted, choose whether to perform load flow analysis without DGs.
 - Enter 1 to perform Load Flow Analysis without DGs
 - Enter 2 to skip Load Flow Analysis without DGs

The S	- F3Madpower Format)Updated Nagi Code\Code\Code\The Mainm
	tenm × +
1	% Clear existing data and command window
2 -	delete *.mat
3 -	clear
4 -	clc
5-	close all
6-	commandwindow
7	
8	% Set MATLAB plot properties
9-	<pre>met(0, 'DefaultLineLineWidth', 1.5);</pre>
10 -	set(0, 'defaultAxesFontSize', 10)
11-	tic
12	
13 -	disp('Please open network file');
**	dd an 77 - 73 -
and the second se	
New to M	Window WTLABTSee resources for Getting Started.
New to M	ATLAB? See resources for Gernino Started.
New to M	MTLAB? See resources for Getting, Started.
Are	MTLAB? See resources for Getting, Started.
Are Nets	MINATion measurem by during forms. loads already in megawatts (MM)? (Y/N): n
Are Net	NUMAFile envoyees to Genergianed. loads already in megawatts (NW)? (Y/N): n work file file loaded successfully.
Are Network	NDAW is ensures to Gene Sand loads already in megawatts (MW)? (Y/N): n work file file loaded successfully. nsm an option: er 1 to perform Load Flow Analysis without DGs
Are Net Chose	NUMAFile envoyees to Genergianed. loads already in megawatts (NW)? (Y/N): n work file file loaded successfully.

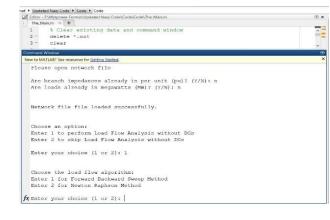


Figure 9 Choose an Option for Analysis

- Figure 10 Choose Load Flow Algorithm
- 2. Choose Load Flow Algorithm:

- If the user selects 1 in the previous prompt, then in next step the code asks the user to select the algorithm to use for load flow analysis.
 - Enter 1 for Backward Forward Sweep Method
 - Enter 2 for Newton Raphson Method

3.3.6 Performing Load Flow Analysis

3.3.6.1 Without Distributed Generators (DGs)

- 1. Run Analysis:
 - Perform load flow analysis using the selected algorithm.
 - The software calculates power losses, voltage at buses, power flows, and power factor.

2. Save Results:

- Results are saved in an Excel file and figures in .fig format.
- Review the data to understand the current system performance.
- 3. Review and Interpret Results:
 - Examine the power losses, voltage profiles, and power flows.
 - Identify areas with significant losses or voltage deviations.

Pro Tip: Regularly check for voltage drops or rises at each bus to ensure they remain within acceptable limits.

3.3.6.2 With Distributed Generators (DGs)

- 1. Choose Analysis Option:
- After the initial load flow analysis, decide whether to perform analysis with DGs.

- Enter 1 to perform Load Flow Analysis with DGs.
- \circ Enter 2 to exit.

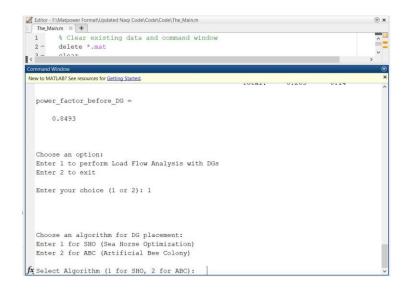


Figure 11 Load Flow Analysis with DG or exit

2. Select Optimization Algorithm:

- Choose the algorithm for DG placement.
 - Enter 1 for SHO (Sea Horse Optimization)
 - Enter 2 for ABC (Artificial Bee Colony)

3. Input Algorithm Parameters:

To input the algorithm parameters for Distributed Generator (DG) placement, follow these detailed steps:

I. Enter Population Size:

- A prompt will appear asking for the population size.
- This is the number of candidate solutions the algorithm will generate and evaluate in each iteration.
- **Example**: Enter 100 for a moderate-sized population.



Figure 12 Enter Population Size

II. Specify Maximum Iterations:

- Next, you will be asked to specify the maximum number of iterations.
- This determines how many times the algorithm will repeat its search process.
- **Example**: Enter 100 for a standard optimization run.

	>
mmand Window	(
ew to MATLAB? See resources for Getting Started.	
Choose an algorithm for DG placement:	
Enter 1 for SHO (Sea Horse Optimization)	
Enter 2 for ABC (Artificial Bee Colony)	
Select Algorithm (1 for SHO, 2 for ABC): 1	
Enter Population Size (integer): = 300	
Enter Maximum Iterations (integer): = 100	

Figure 13 Enter Maximum Iterations

III. Select DG Type:

- A prompt will ask you to choose the type of DG to place in the network.
- Enter 1 for DG type 1 injecting Active Power only, DG type 2 for injecting Reactive Power, 3 for injecting Active as well as Reactive Power, and 4 for injecting Active Power and absorbing Reactive Power.

- Typical types include renewable sources like solar and wind, capacitor banks or synchronous condenser type DG.
- **Example**: Enter 1 for Solar PV, 2 for Capacitor bank, 3 for synchronous condenser and 4 for Wind type DG.

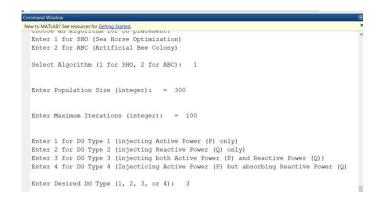


Figure 14 Select DG type

IV. Define Number of DGs:

- Specify the total number of DG units to be placed in the network.
- This helps the algorithm know how many units to optimize for.
- **Example**: Enter 3 if you plan to place three DG units.

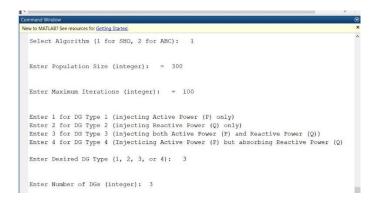


Figure 15 Enter Number of DGs

- V. Set DG Size Ranges:
 - Enter the size ranges for the DG units in terms of kW and kVAR.

- This includes the minimum and maximum capacities for active and reactive power.
- **Example**: For active power, enter 50 kW as the minimum and 200 kW as the maximum. For reactive power, enter 30 kVAR as the minimum and 100 kVAR as the maximum.

W LO WATE	AB? See resources for Getting Started.
Enter	Maximum Iterations (integer): = 100
Enter	1 for DG Type 1 (injecting Active Power (P) only)
Enter	2 for DG Type 2 (injecting Reactive Power (Ω) only)
Enter	3 for DG Type 3 (injecting both Active Power (P) and Reactive Power (Q))
Enter	4 for DG Type 4 (Injecticing Active Power (P) but absorbing Reactive Power (Q)
Encer	Desired DG Type (1, 2, 3, or 4): 3
Enter	Number of DGs (integer): 3
Enter	Minimum DG Size in kW: 1
Enter	Maximum DG Size in kW: 1000
Enter	Minimum DG Size in kVAR: 1
	Maximum DG Size in kVAR: 1000

Figure 16 Enter DG size range

VI. Run Optimization:

- The algorithm optimizes DG placement based on the specified objective function.
- The software displays results and visualizes them using graphs.

VII. Review and Save Results:

- Analyze the optimized DG placements and their impact on the system.
- Save the results for future reference and comparison.

Pro Tip: Use optimization results to adjust the network design for improved performance.

3.3.7 Optimization Algorithms for DG Placement

3.3.7.1 Sea Horse Optimization Algorithm (SHO)

1. Algorithm Overview:

- SHO mimics the mating behavior of sea horses.
- Balances exploration (diversity of solutions) and exploitation (convergence to optimal solution).
- 2. Algorithm Steps:
- **Initialization**: Generate initial population of solutions.
- **Evaluation**: Assess each solution using the objective function.
- Selection and Mating: Select top-performing solutions for mating.
- Update: Generate new solutions by combining parents' characteristics.
- **Termination**: Repeat until maximum iterations or convergence criteria are met.
- Parameters:
- Population Size: Number of candidate solutions.
- Maximum Iterations: Number of generations for optimization.
- Execution:
- The algorithm iterates through candidate solutions, refining them to find the optimal placement of DGs.
- Results include the best DG locations and their impact on the system.

Pro Tip: SHO is effective in balancing the search process, making it suitable for complex optimization problems.

3.3.7.2 Artificial Bee Colony Algorithm (ABC)

1. Algorithm Overview:

- ABC simulates the foraging behavior of honeybees.
- Known for efficient exploration of the solution space.

2. Algorithm Steps:

- **Initialization**: Generate initial population of solutions.
- Employed Bees Phase: Each bee modifies its solution based on local information.
- Onlooker Bees Phase: Bees choose solutions based on a probability related to their quality.
- Scout Bees Phase: Replace abandoned solutions with new random solutions.
- **Termination**: Repeat until maximum iterations or convergence criteria are met.

Parameters:

- Population Size: Number of bees (candidate solutions).
- Maximum Iterations: Number of generations for the algorithm.

Execution:

- The algorithm evaluates candidate solutions and refines them to identify optimal DG placements.
- Outputs include the best DG locations and their impact on system performance.

Pro Tip: ABC is excellent for avoiding local minima and exploring diverse solutions, making it ideal for DG placement.

3.3.8 Load Flow and Optimization Results

3.3.8.1 Load Flow Analysis Results

- 1. Voltage at Buses:
 - Displays voltage levels at each bus.
 - Identifies any buses with voltage deviations from nominal values.

Bus Number	Voltage (P.u)	
1	1	
2	0.99703226	
3	0.982937983	
4	0.975456413	
5	0.968059232	
6	0.949658177	
7	0.946172614	
8	0.941328437	
9	0.935059372	
10	0.929244423	

11	0.928384417
12	0.926884837
13	0.920771748
14	0.918504993
15	0.91709268
16	0.91572476
17	0.913697546
18	0.913090479
19	0.996503896
20	0.9929263
21	0.992221796
22	0.991584377

23	0.979352257
24	0.972681101
25	0.969356112
26	0.94772891
27	0.945165164
28	0.933725581
29	0.925507478
30	0.921950058
31	0.917788887
32	0.916873466
33	0.916589822

2. Average Voltage:

- Shows the average voltage across the system.
- Helps assess overall system voltage stability.
- 3. Active Power Loss:
 - Details power losses at each iteration and between branches.
 - Helps identify areas with significant power loss.

Table 2 Active Power Loss at branches

From Branch	To Branch	APL(kw)
1	2	6.239695649
2	3	26.37886194
3	4	10.13510604
4	5	9.523654326
5	6	33.01804265
6	7	6.32854428
7	8	1.598827083
8	9	3.003492598
9	10	2.524019638
10	11	0.183065222
11	12	0.291358035
12	13	2.09775361
13	14	0.959784732
14	15	0.317712893
15	16	0.205546455

16	17	0.335967627
17	18	0.04166665
18	19	0.153593488
19	20	0.749855262
20	21	0.117711008
21	22	0.057693062
22	23	2.173972168
23	24	4.061670613
24	25	1.007402689
25	26	1.324791602
26	27	1.694952169
27	28	9.963748379
28	29	6.824229017
29	30	1.984296279
30	31	1.574993316
31	32	0.248487415
32	33	0.020475079

4. Reactive Power Loss:

- Lists reactive power losses at each iteration and between branches.
- Important for understanding reactive power distribution.

From		
Branch	To Branch	RPL(kVAR)
1	2	12.24042423
2	3	51.79123431
3	4	19.90047645
4	5	18.69894211
5	6	38.24862367
6	7	1.914517597
7	8	4.837965066
8	9	4.180536994
9	10	3.560914193
10	11	0.553701887
11	12	0.881134477
12	13	2.666235757
13	14	0.729161749
14	15	0.356973992
15	16	0.28146664

Table 3 Reactive Power Loss at branches

Review Excel Files:

- Load flow results are saved in Excel files.
- Review power losses, voltage profiles, power flows, and power factors.

Review Figures:

• Load flow analysis figures are saved in .fig format.

• Visualize voltage profiles, power flows, and other key parameters.

Pro Tip: Use Excel's data analysis tools to gain deeper insights into the results.

3.3.8.2 Optimization Results

- 1. Convergence Curve:
 - Shows the MOI of the optimization algorithm over iterations.
 - Helps assess the efficiency and stability of the optimization process.

Iteration #	MOI	8	0.112861		16
1	0.115074	9	0.112861		17
2	0.115074	10	0.112861		18
3	0.115074	11	0.112861		19
4	0.115074	12	0.112861		20
5	0.115074	13	0.112861		21
6	0.112861	14	0.112861		22
7	0.112861	15	0.112861		23

Table 4 MOI with Iterations

	_	_			
0.112861		45	0.112861	66	
0.112861		46	0.112861	67	
0.112861		47	0.112861	68	
0.112861		48	0.112861	69	
0.112861		49	0.112861	70	
0.112861		50	0.112861	71	
0.112861		51	0.112861	72	
0.112861		52	0.112861	73	
0.112861		53	0.112861	74	
0.112861		54	0.112861	75	
0.112861		55	0.112861	76	
0.112861		56	0.112861	77	
0.112861		57	0.112861	78	
0.112861		58	0.112861	79	
0.112861		59	0.112861	80	
0.112861		60	0.112861	81	
0.112861		61	0.112861	82	
0.112861		62	0.112861	83	
0.112861		63	0.112861	84	
0.112861		64	0.112861	85	
0.112861		65	0.112861	86	

0.112861
0.112861
0.112861
0.112861
0.112861
0.112861
0.112861
0.112861
0.112861
0.112861
0.112861
0.099889
0.099889
0.09882
0.091721
0.091156
0.081343
0.080586
0.079696
-

0.112861

0.112861

		1		ı		
87	0.079109		92	0.079089	97	0.078855
88	0.079089		93	0.079089	98	0.078355
89	0.079089		94	0.079089	99	0.07092
90	0.079089		95	0.079056	100	0.070631
91	0.079089		96	0.079056		

2. Voltage Profiles:

- Displays voltage levels at buses before and after DG placement.
- Helps in evaluating the impact of DGs on voltage stability.

Bus Number	Voltage (P.u)	9	0.999759	18	0.992204
	(1.u)	10	1.003179	19	0.998907
1	1	11	1.003728	20	0.995339
2	0.999435	12	1.004902	21	0.994636
3	0.998179	13	0.999274	22	0.994
4	0.997312	14	0.997188	23	0.998665
5	0.996792	15	0.995888	24	1.000673
6	0.996111	16	0.994629	25	0.997441
7	0.996487	17	0.992763	26	0.996179
8	0.996842	L		L	

Table 5 Voltage at buses with DGs

27	0.996403	30	1.000199	33	0.995262
28	0.997569	31	0.996366		
29	0.998849	32	0.995523		

3. Power Losses with DGs:

- Details active and reactive power losses with DGs.
- Important for understanding the benefits of DG placement in reducing losses.
- Details power losses between branches.

Fro m Bran ch	To Bran ch	APL(kw)	RPL(kw)
1	2	0.671371 652	0.342239 346
2	3	1.591098 075	0.810394 983
3	4	0.275727 019	0.140424 908
4	5	0.112057 715	0.057072 691
5	6	0.129546 16	0.111830 446

Table 6 Voltage at busses with DG

6	7	0.080017 983	0.264503 887
7	8	0.301356 777	0.099590 917
8	9	1.080001 272	0.775923 244
9	10	1.367258 625	0.969129 677
10	11	0.318453 309	0.105287 208
11	12	0.735195 983	0.243101 663

12	13	2.260583 585	1.778592 671	23	24	1.812719 909	1.431402 77
13	14	0.618090 358	0.813583 117	24	25	1.215971 165	0.951470 295
14	15	0.302480 427	0.269212 698	25	26	0.031513 729	0.016051 821
15	16	0.238444 438	0.174128 659	26	27	0.065284 383	0.033239 445
16	17	0.213122 74	0.284549 446	27	28	0.378396 391	0.333624 844
17	18	0.045000 056	0.035286 929	28	29	0.434623 949	0.378634 094
18	19	0.160175 578	0.152850 476	29	30	0.538880 807	0.274484 116
19	20	0.828142 456	0.746220 107	30	31	1.351906 809	1.336090 166
20	21	0.100269 274	0.117139 977	31	32	0.180835 785	0.210771 242
21	22	0.043422 683	0.057413 007	32	33	0.011169 033	0.017366 044
22	23	0.826238 596	0.564559 75				

• Details power losses at each iteration.

Table 7 Total Active and Reactive Power Loss with respect to iterations

			I	1	1
)	APL per Iteration (kw)	RPL per Iteration (kVAR)	12	21.7530779 7	30.9002244
	24.3031105	33.4903421	13	21.7530779 7	30.9002244
	1 24.3031105	33.4903421	14	21.7530779 7	30.9002244
	1 24.3031105	9 33.4903421	15	21.7530779 7	30.9002244
	1 24.3031105	9 33.4903421	16	21.7530779 7	30.9002244
	1 24.3031105	9 33.4903421	17	21.7530779 7	30.9002244
	1 21.7530779	9	18	21.7530779	30.9002244
	7	30.9002244		21.7530779	
	21.7530779 7	30.9002244	19	7 21.7530779	30.9002244
	21.7530779 7	30.9002244	20	7 21.7530779	30.9002244
	21.7530779 7	30.9002244	21	7	30.9002244
	21.7530779 7	30.9002244	22	7	30.9002244
	21.7530779 7	30.9002244	23	21.7530779 7	30.9002244

	APL per	RPL per
Iteratio	Iteration	Iteration
n #	(kw)	(kVAR)
	24.3031105	33.4903421
1	1	9
	24.3031105	33.4903421
2	1	9
	24.3031105	33.4903421
3	1	9
	24.3031105	33.4903421
4	1	9
	24.3031105	33.4903421
5	1	9
	21.7530779	
6	7	30.9002244
	21.7530779	
7	7	30.9002244
	21.7530779	
8	7	30.9002244
	21.7530779	
9	7	30.9002244
	21.7530779	
10	7	30.9002244
	21.7530779	
11	7	30.9002244

1	I	I	I	ı
24	21.7530779 7	30.9002244		
25	21.7530779 7	30.9002244		
26	21.7530779 7	30.9002244		
27	21.7530779 7	30.9002244		-
28	21.7530779 7	30.9002244		
29	21.7530779 7	30.9002244		
30	21.7530779 7	30.9002244		
31	21.7530779 7	30.9002244		-
32	21.7530779 7	30.9002244		
33	21.7530779 7	30.9002244		
34	21.7530779 7	30.9002244		_
35	21.7530779 7	30.9002244		
			J	L

1	1	1
36	21.7530779 7	30.9002244
37	21.7530779 7	30.9002244
38	21.7530779 7	30.9002244
39	21.7530779 7	30.9002244
40	21.7530779 7	30.9002244
41	21.7530779 7	30.9002244
42	21.7530779 7	30.9002244
43	21.7530779 7	30.9002244
44	21.7530779 7	30.9002244
45	21.7530779 7	30.9002244
46	21.7530779 7	30.9002244
47	21.7530779 7	30.9002244

	I	I	1	I	I	l
	21.7530779				21.7530779	
48	7	30.9002244		60	7	30.9002244
	21.7530779				21.7530779	
49	7	30.9002244		61	7	30.9002244
	21.7530779				21.7530779	
50	7	30.9002244		62	7	30.9002244
	21.7530779				21.7530779	
51	7	30.9002244		63	7	30.9002244
	21.7530779				21.7530779	
52	7	30.9002244		64	7	30.9002244
	21.7530779				21.7530779	
53	7	30.9002244		65	7	30.9002244
	21.7530779				21.7530779	
54	7	30.9002244		66	7	30.9002244
	21.7530779				21.7530779	
55	7	30.9002244		67	7	30.9002244
	21.7530779				21.7530779	
56	7	30.9002244		68	7	30.9002244
	21.7530779				21.7530779	
57	7	30.9002244		69	7	30.9002244
	21.7530779				21.7530779	
58	7	30.9002244		70	7	30.9002244
	21.7530779				21.7530779	
59	7	30.9002244		71	7	30.9002244

1	I	I	I	i
72	21.7530779 7	30.9002244		
73	21.7530779 7	30.9002244		
74	21.7530779 7	30.9002244		
75	21.7530779 7	30.9002244		
76	21.7530779 7	30.9002244		
77	21.7530779 7	30.9002244		
78	21.7530779 7	30.9002244		(
79	22.5262577 9	32.4345657 2		(
80	22.5262577 9	32.4345657 2		Ç
81	21.490311	30.9935152 8		Ģ
82	18.7368869	26.8181777 1		Ģ
83	18.1211268 2	25.9724317 4		Ģ

	15.1643401	20.7776423
84	3	8
	14.8788164	20.3911800
85	6	9
	14.6743723	20.1692304
86	1	8
	14.2664543	
87	5	19.6368607
	14.1856315	19.5674187
88	8	6
	14.1856315	19.5674187
89	8	6
	14.1856315	19.5674187
90	8	6
	14.1856315	19.5674187
91	8	6
	14.1856315	19.5674187
92	8	6
	14.1856315	19.5674187
93	8	6
	14.1856315	19.5674187
94	8	6
	14.1847867	19.5764888
95	6	8

96	14.1847867 6	19.5764888 8
97	14.1570430 1	19.5577120 3
98	14.2057097 2	19.5300570 9

99	12.7287759 9	16.6647400 5
100	13.8961706 4	18.3193567 2

Review Excel Files:

- Optimization results are saved in Excel files.
- Review DG placements, power losses, and voltage profiles.

Review Figures:

- Optimization figures are saved in .fig format.
- Visualize the impact of DG placements on the system.

Pro Tip: Compare pre- and post-optimization results to assess the effectiveness of DG placements.

3.3.9 Visualizing Results with Figures and Graphs

3.3.9.1 Load Flow Analysis Parameters

1. Average Voltage:

- Shows the average voltage level across the system as shown in fig 22.
- Useful for assessing overall system health.

2. Active Power Loss:

 Plots active power losses at each iteration and between branches as shown in fig 24. Important for identifying areas with high losses.

3. Reactive Power Loss:

- Plots reactive power losses at each iteration and between branches as shown in fig 25.
- Helps in understanding reactive power distribution.

4. Voltage at Buses:

- Graphically represents voltage levels at each bus as shown in fig 23.
- Helps in visualizing voltage stability and identifying problematic areas.

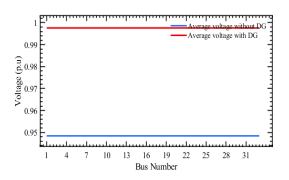
3.3.9.2 Optimization Graphs

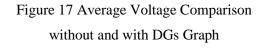
1. Average Voltage:

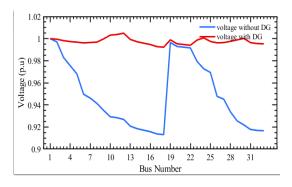
- Shows the average voltage level across the system as shown in fig 22.
- Useful for assessing overall system health.

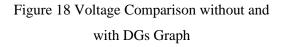
2. Voltage with DGs:

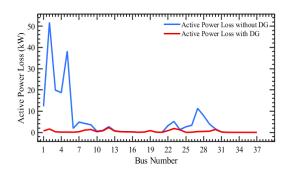
- Shows voltage levels at buses with DG placement as shown in fig 25.
- Helps in evaluating the impact of DGs on voltage stability.

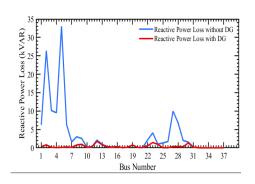


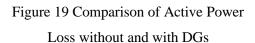


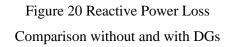












3. Active Power Loss with DGs:

- Plots active power losses with DGs as shown in fig 24.
- Important for understanding the benefits of DG placement in reducing losses.

4. Reactive Power Loss with DGs:

- Plots reactive power losses with DGs as shown in fig 25s.
- Helps in understanding the benefits of DG placement in managing reactive power.

3.3.9.3 Saving and Viewing Figures:

- Figures are saved in .fig format for later review in MATLAB.
- Use MATLAB's plotting tools to customize and analyze the figures.

Pro Tip: Regularly save graphical results to track performance changes over time and across different scenarios.

3.3.10 Example Scenarios

Scenario 1: Simple Load Flow Analysis

1. Load a Network File:

• Open the network file designed in Torrit software.

2. Perform Load Flow Analysis:

- Select the load flow algorithm (Backward Forward Sweep or Newton Raphson).
- Run the analysis and review the results.
- 3. Save Results:
 - The results will be saved in Excel file and figures in .fig format for later review.

Scenario 2: DG Placement Optimization

- 1. Load a Network File:
 - Open the network file designed in Torrit software.
- 2. Perform Initial Load Flow Analysis:
 - Conduct analysis without DGs to establish a baseline.
- 3. Select Optimization Algorithm:
 - Choose SHO or ABC for DG placement.
- 4. Input Parameters and Run Optimization:
 - Enter population size, maximum iterations, DG type, and size ranges.
 - Run the optimization and review the results.
- 5. Analyze and Save Results:
 - Examine voltage profiles, power losses, and DG placements.
 - Save the optimization results and graphs.

Chapter 4

Proposed Algorithm

4 PROPOSED ALGORITHM

4.1 Sea Horse Optimization (SHO) Algorithm

The Sea Horse Optimization (SHO) algorithm is a nature-inspired metaheuristic that draws inspiration from the foraging behaviour of sea horses, which exhibit unique swimming patterns, hunting strategies, and interactions with their environment. This behaviour has been abstracted into an optimization technique that effectively balances exploration (searching for new solutions) and exploitation (refining known solutions). It is especially beneficial when solving such combined optimization problems as placement of Distributed Generation units in power distribution networks, intending DG units distribution in the way that will maximize the power factor.

Furthermore, the location of DG units significantly defines the extent of improvement of the overall efficiency of the distribution network. Regarding specific performances, lower losses, higher power factor as well as appropriate voltage levels can be achieved with a well-sited DG, resulting in a better network response. To this end, the SHO algorithm is optimized to apply an objective function may include loss reduction, improvement of the power factor or both. They called the set of potential solutions 'Sea horses ' in SHO algorithm where each sea horse refers to a particular placement of DGs. In this case, the position of each sea horse in the search space corresponds to correspondent DG locations and sizes, which provides the basis for feedback phase to cover a wide variety of effective solutions. This exploration avoids the algorithm from being stuck in local optima, and indeed leads to optimality [41].

The proposed SHO algorithm uses the movement of sea horses as a base for its movements. For example, the hunting movement corresponds to sea horses moving toward areas with better solutions that could be achieved through refining the areas of DG placements featured in the improvement of the power factor. Also, the forming of swarms, so sea horses gather together, provides the directed motion in the direction of the areas with better solutions in order to turn into good ones. It also provides provisions for getting out of local optima which is similar to sea horse movement to escape predators

and thus is able to keep on preserving the variety of search space and avoid the convergence problem.

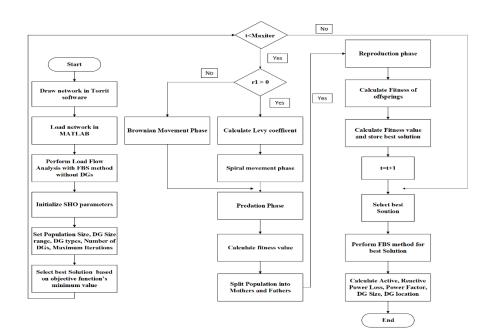


Figure 21 Flow Chart of SHO Algorithm

At the end of each iteration, the solutions are assessed against the objective function in order with regards to, for instance, the power factor improvement, loss reduction and voltage profile. Usually, the solutions that have had high performance ratings are maintained for the next iteration while others that had low ratings get eliminated or amended. This process is done until a certain stopping criterion is reached which may be certain number of iterations or a convergence criterion or a certain acceptable value of the power factor.

The major advantage of the SHO algorithm is in its use of exploration and exploitation, which makes SHO superior in searching for the global optimum not being stacked at the local one. Besides, its flexibility making it possible to fit the solution into different problem constraints and goals to solve, which makes it ideal in solving complicated, multiple objectives problems like determining the best placement of DGs. The algorithm is also lightweight, and this is desirable since power distribution networks can be extensive, and hence require efficient algorithms. However, many factors influence the run time and quality of the SHO Algorithm including the number of sea horses, maximum

iteration point and the specific staking movements. Furthermore, there is the need to successfully incorporate it with accurate models of the power distribution network for an efficient evaluation of the objective function. To learn more about the proposition of SHO for DG placement, heed has to be paid on comparing it with existing optimisation algorithms like PSO, GA, or ACO. Summing up, the Sea Horse Optimization can be viewed as a new and promising heuristic algorithm for designing the DG placement relating to the power factor improvement, as its tendency to combine the global and local search methods give us a hope to reach the optimal solution [43].

Step 1: Start setting requirements for a model that has to be optimized with the help of SHO such as number of sea horses (solutions), maximum number of iterations and regulating parameters such as exploration and exploitation.

Step 2: Produce the first set of sea horses where each one of them is a candidate solution of DG placement in the distribution network. Each position may be represented as a vector Xi where i=1,2,...,Ni = 1,2,...,N (with N being the total number of sea horses)

Step 3: Calculate the fitness of each sea horse based on the objective function, such as minimizing losses or improving the power factor:

$$Fitness = f(Xi) \tag{4.1}$$

Step 4: Update the position of each sea horse using movement strategies like hunting and swarming. A typical position update might be [42]:

$$Xinew = Xi + r \times (Xbest - Xi)$$
(4.2)

where r is a random factor, and *Xbest* is the best position found so far.

Step 5: Recalculate the fitness for the updated positions. If f(Xinew) is better than the previous fitness, update the sea horse's position.

Step 6: If diversity decreases or local optima are detected, introduce random perturbations to allow sea horses to escape local minima:

$$Xiescape = Xi + rand \times (Xmax - Xmin)$$
(4.3)

Step 7: Store the best solution found in the memory, then increment the iteration counter and repeat the process until the stopping criterion (e.g., maximum iterations) is met.

Step 8: Output the best solution for DG placement in the distribution network, which optimizes the power factor or other desired performance metrics.

4.1.1 Sea Horses Mobility Patterns

The movement of sea horses follows a normal distribution. It is divided into exploration and exploitation phases. In the exploitation phase, sea horses attempt to move towards the best sea horse represented by X_{elite} . The distance covered is represented by Levy flight [17], as described in (4.4).

$$X_{new}^{1}(t+1) = X_{i}(t) + Levy(\lambda)(X_{elite}(t) - X_{i}(t)) \times x \times y \times z + X_{elite}(t)$$
(4.4)

The coordinates of sea horses are denoted by x, y, z, and the equation for Levy flight is given by (4.5).

$$Levy(z) = s \times \frac{w \times \sigma}{|k|^{\frac{1}{\lambda}}}$$
 (4.5)

Parameters w and k have values in the range of 0 to 1, λ is in the range of [0,2], s has a value of 0.01, and σ is determined by (4.6).

$$\sigma = \left(\frac{\Gamma(1+\lambda) \times \sin\left(\frac{\pi\lambda}{2}\right)}{\Gamma\left(\frac{1+\lambda}{2}\right) \times \lambda \times 2^{\left(\frac{\lambda-1}{2}\right)}}\right)$$
(4.6)

Exploration is carried out using Brownian movement, where sea horses attempt to capture food by moving in a random direction, as given by (4.7).

$$X_{new}^{1}(t+1) = X_{i}(t) + rand \times l \times \beta_{t} \times (X_{i}(t) - \beta_{t} \times X_{elite})$$
(4.7)

The variable *l* is fixed, and β is a random walk variable determined by (4.8).

$$\beta = \frac{1}{\sqrt{2\pi}} \times e^{\left(\frac{-x^2}{2}\right)} \tag{4.8}$$

4.1.2 Predation Behavior Phase

In the predation behavior phase, the hunting phase is divided into two scenarios based on the probability of success and failure, denoted by 1 and 0, respectively. In the exploitation phase, it is assumed that sea horses catch their prey with a 0.90 probability, while the exploration phase has a 0.10 probability. If a sea horse captures its prey, it out-ranks its opponent, becoming the elite soldier; otherwise, it explores other ways to capture prey. This phase is represented by (4.9).

$$X_{new}^{2}(t+1) = \begin{cases} \alpha \times (X_{elite} - rand \times X_{new}^{1}(t)) + (1-\alpha) \times X_{elite} ifr_{2} > 0.1\\ (1-\alpha) \times (X_{new}^{1}(t) - rand \times X_{elite}) + \alpha \times X_{new}^{1}(t) ifr_{2} \le 0.1 \end{cases}$$

$$(4.9)$$

 X_{new}^1 represents the new location of sea horses. r2 ranges between 0 and 1, and alpha is determined by (4.10).

$$\alpha = \left(1 - \frac{t}{T}\right)^{\frac{2t}{T}} \tag{4.10}$$

T indicates the maximum iteration limit. Alpha can be changed to make the algorithm converge to its solution.

4.1.3 Reproduction phase of Sea horses

During the reproduction phase, the population is divided into two halves, with one half containing male sea horses and the other half containing female sea horses. This is given by (4.11).

$$X_i^{offspring} = r_3 X_i^{father} + (1 - r_3) X_i^{mother}$$
(4.11)

where r_3 is any number between 0 and 1.

Chapter 5

Results and Discussions

5 RESULTS AND DISCUSSIONS

Firstly, every network's required data was collected, including load profiles, voltage levels, line impedances, and the locations of any installed Distributed Generators (DGs). The ensuing analysis and optimization process need this baseline data. Following data collection, Torrit visualization was employed to offer a more complete image of each network's current state. By using this visualization technique, the inefficient or unbalanced areas of the network could be quickly identified, and determining the optimal locations for DG deployment was also made easier. Important components including load distribution, voltage stability, and power factor changes were all clearly depicted in each network's display.

After that, the experiment's findings are examined, paying close attention to how achieving the optimal power factor through strategic distributed generator (DG) placement impacts the network's overall performance. While several different DG types were employed, DG Type 3 was specifically used to obtain the results shown here. This choice was made because of its outstanding performance in power factor correction, voltage profile augmentation, and overall network efficiency. When it comes to achieving optimal performance, DG Type 3 performs better than other DG types due to its ability to effectively reduce power losses, boost voltage stability, and improve network efficiency [39]. In order to demonstrate the gains in power factor, voltage profiles, and overall network efficiency, this section presents a comparison between the initial state of the network and its optimized results.

The network characteristics that the IESCO Gujjar Khan Division uses to test and investigate the ideal power factor of distribution networks depending on the location of DG are described in detail. The efficiency of the used optimization techniques is demonstrated by the MATLAB results for each network, which also provide a quantitative analysis of the performance gains and highlight the noteworthy advantages of DG deployment strategy.

To find the ideal distribution of DGs, two optimization methods were applied in addition to the load flow analysis. The study's main focus, the Sea Horse Optimization (SHO) approach, was used to determine the network's ideal DG layout. The identical task was also finished with the Artificial Bee Colony (ABC) method for comparison's sake. To assess how well each load flow strategy enhanced the distribution network's overall performance, these two algorithms were combined with it.

Following are the detailed analysis of optimization implemented on each network:

5.1 IEEE 33 Bus Network System:

A benchmark radial distribution network, the IEEE 33-bus system has 33 buses, 32 branches, a 3.72 MW total load, and 2.3 MVAR. It is frequently employed to evaluate optimization strategies for distributed generation integration, voltage stability, and power flow.

Here is the Torrit network single line diagram:

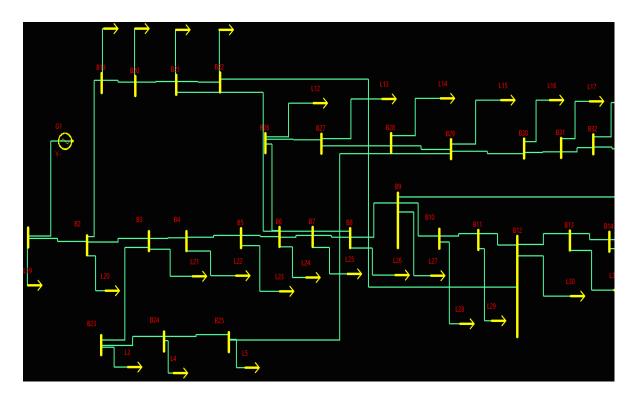


Figure 22 Torrit View of IEEE 33 Bus Network System

5.1.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.1.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

5.1.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

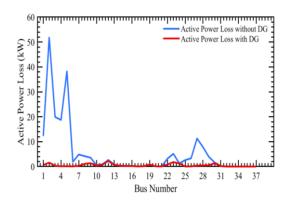


Figure 23 Active Power Loss (APL) in term

of DG placement

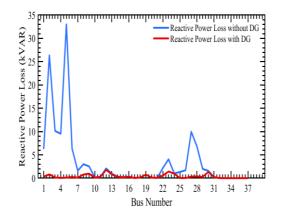


Figure 24 Reactive Power Loss (RPL) in term of DG placement:

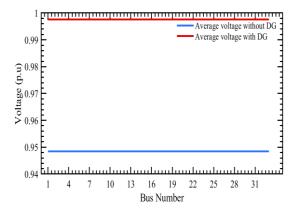


Figure 25 Average Voltage (Vavg)

showing before and after placement of DGs

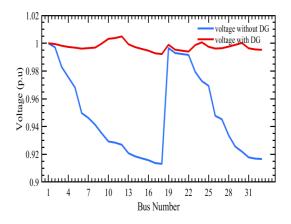


Figure 26 Minimum Voltage (Vmin) showing before and after placement of DGs

5.1.1.3 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.:

5.1.1.4 Results Summary Using SHO Algorithm:

The SHO algorithm's results summary emphasizes the Distributed Generators' (DGs) ideal placement and sizing within the network, showing improvements in voltage profiles and decreased power losses. An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable	Before	DG After	DG
'Min Voltage (p.u)'	0.9130	9' '0.992	2'
'Min Voltage at Bus Location'	'18'	'18'	
'Total Active Power Losses (kW)'	202.67	'18.31	94'
'Total Active Power Losses (%)'	'NA'	90.96	13'
'Total Reactive Power Losses (kV	AR)' '135.14	11' '13.89	62'
'Total Reactive Power Losses (%)	'NA'	189.71	73'
'Power Factor'	0.8493	'0.947	39'
dditional Information related to DG Variable	is:	Value	
	;s: 	Value	
Variable	57 Mail 5 7	Value	
Variable 'DG Type'	'3' '3'	Value	
Variable 'DG Type' 'No of DGs'	'3' '3'	Value 876.16707	756.45893'
'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '30 12 24'		756.45893' 963.7898'

Table 8 Results Summary Using SHO Algorithm

5.1.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.1.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) in fig 27 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.1.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) in fig 28 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

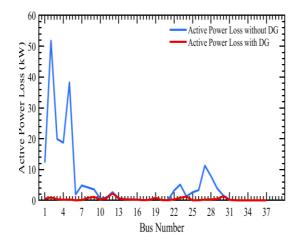
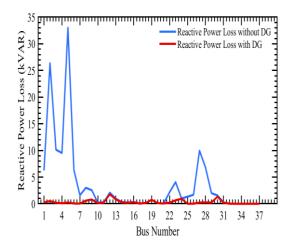
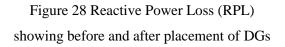


Figure 27 Active Power Loss (APL) in term of DG placement





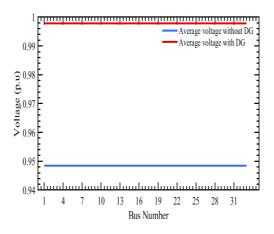


Figure 29 Average Voltage (Vavg) showing before and after placement of DGs

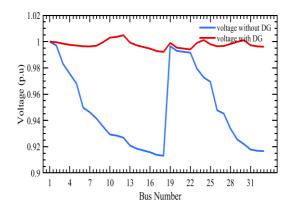


Figure 30 of Minimum Voltage (Vmin) showing before and after placement of DGs

5.1.2.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) in fig 28 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.1.2.4 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) in fig 30 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.1.2.5 Results Summary Using ABC Algorithm:

The impact of DG placement is shown in the results summary using the ABC algorithm, with a focus on how it affects voltage stability and power loss reduction. A quick comparison of the ABC algorithm's performance against alternative approaches is provided in this section.

Variable		Before_DG	After_D	G
'Min Voltage (p.u)'		'0.91309'	'0.9921	8'
'Min Voltage at Bus Location'		18'	'18'	
'Total Active Power Losses (kW)'	e	202.6771	' 15.037	8'
'Total Active Power Losses (%)'		'NA'	92.580	4 '
'Total Reactive Power Losses (kV	/AR) '	'135.141'	'11.585	
'Total Reactive Power Losses (%)		'NA'	91.427	5 '
'Power Factor'		'0.8493'	0.9322	21
	Gs:		Value	2
itional Information related to DG Variable				
itional Information related to DG Variable 'DG Type'	'3'			
itional Information related to DG Variable 'DG Type' 'No of DGs'	'3' '3'			
itional Information related to DG Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '12	24 30'	Value	
itional Information related to DG Variable 'DG Type' 'No of DGs'	'3' '3' '12 '968	24 30'		- 888.4894' 1073.2499

Table 9 Results Summary Using ABC Algorithm

5.1.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm:

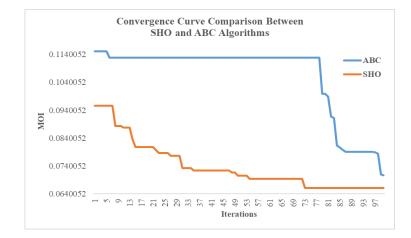


Figure 31 Comparison of Convergence Curve between SHO and ABC Algorithm

5.2 IEEE 69 Bus Network System:

The IEEE 69-bus system is a standard test case for distribution networks, featuring 69 buses and 68 branches with a total load of 3.8 MW and 2.69 MVAR. It's commonly used to study power flow, voltage control, and the impact of distributed generation in real-world scenarios.

Here is the Torrit network single line diagram

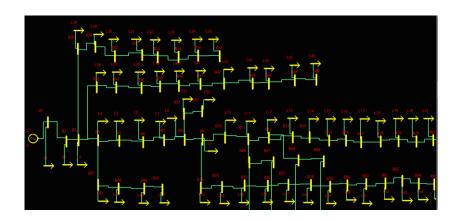


Figure 32 Torrit View of IEEE 69 bus network

5.2.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.2.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) in fig. 33 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm

5.2.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) in fig 34 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

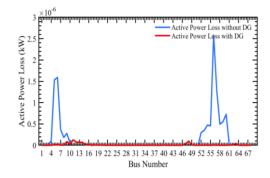


Figure 33 Active Power Loss (APL) showing before and after placement of DGs

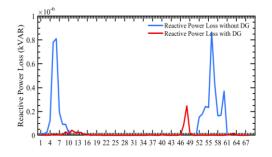


Figure 34 Reactive Power Loss (RPL) showing before and after placement of DGs

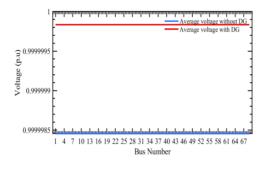


Figure 35 Average Voltage (Vavg) showing before and after placement of DGs

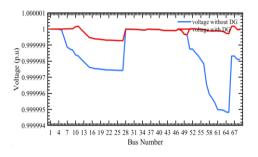


Figure 36 Minimum Voltage (Vmin) showing before and after placement of DG

5.2.1.3 Average Voltage (V_{avg}) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) in fig 35 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.2.1.4 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) in fig 36 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.2.1.5 Results Summary Using SHO Algorithm:

An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable		Befor	e_DG	After_DG
'Min Voltage (p.u)'		0.999	99'	·1·
'Min Voltage at Bus Location'		65'		27'
'Total Active Power Losses (kW)'		1.194	8e-05'	'7.5641e-07
'Total Active Power Losses (%)'		'NA'		'93.6691'
'Total Reactive Power Losses (kV	AR) '	5.477	6e-06'	'5.8448e-07
'Total Reactive Power Losses (%)		'NA'		'89.3296'
'Power Factor' itional Information related to DG		0.815	87'	'0.99982'
		'0.815	87' Value	'0.99982'
itional Information related to DG		'0.815		'0.99982'
itional Information related to DG Variable	;s: 	'0.815		'0.99982'
itional Information related to DG Variable 'DG Type' 'No of DGs'	;s: '3' '3'	2 11'		'0.99982'
itional Information related to DG Variable 'DG Type' 'No of DGs'	is: 	2 11'		'0.99982'
itional Information related to DG Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	;s: '3' '3' '61	2 11'	Value	

Table 10 Results Summary Using SHO Algorithm

5.2.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.2.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) in fig. 37 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm:

5.2.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) in fig. 38 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

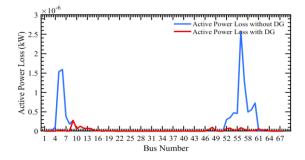
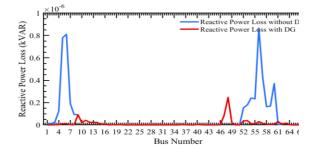
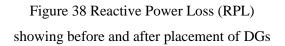


Figure 37 Active Power Loss (APL) showing before and after placement of DGs





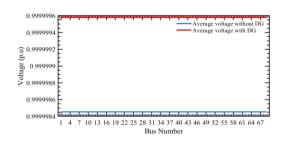


Figure 39 Average Voltage (Vavg) showing before and after placement of DGs

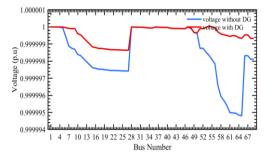


Figure 40 Minimum Voltage (Vmin) showing before and after placement of DGs

5.2.2.3 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.2.2.4 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) in fig 39 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.2.2.5 Results Summary Using ABC Algorithm:

The impact of DG placement is shown in the results summary using the ABC algorithm, with a focus on how it affects voltage stability and power loss reduction. A quick comparison of the ABC algorithm's performance against alternative approaches is provided in this section.

Table 11 Results Summary	Using ABC Algorithm
--------------------------	---------------------

Variable		Before_DG	After_DG
'Min Voltage (p.u)'		'0.99999'	'1'
'Min Voltage at Bus Location'		'65'	27'
'Total Active Power Losses (kW)'		'1.1948e-05'	'1.38 <mark>4</mark> 2e-06'
'Total Active Power Losses (%)'		'NA'	'88.4151'
'Total Reactive Power Losses (kV	AR) '	'5.4776e-06'	'8.3444e-07'
'Total Reactive Power Losses (%)	1	'NA'	'84.7665'
'Power Factor'		'0.81587'	'0.95267'
	is:	'0.81587' Value	'0.95267'
itional Information related to DG	:s: 		'0.95267'
itional Information related to DG Variable	<u></u>		'0.95267'
tional Information related to DG Variable 'DG Type'	'3' '3'	Value	'0.95267'
Itional Information related to DG Variable 'DG Type' 'No of DGs'	'3' '3'	Value	'0.95267'
'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '54	Value 47 62'	

5.2.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm:

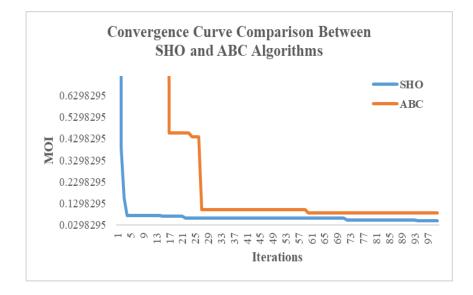


Figure 41 Comparison of Convergence Curve between SHO and ABC Algorithm

5.3 IESCO 15 Bus Network System:

The network in Figure 50 depicts the Torrit visualization of IESCO regional distributive network, comprising 15 buses and 14 branches. The total active and reactive load demand of this network is 1334.4 kW and 1439.5 kVAR respectively.

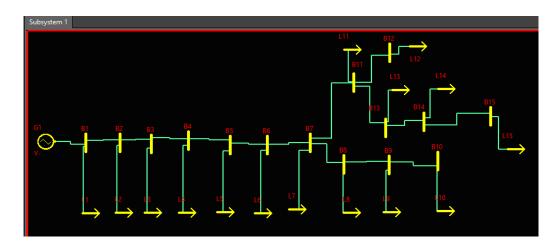


Figure 42 Torrit view of IESCO 15 Bus Network System

5.3.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.3.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) in fig 43showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.3.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) in fig 44 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

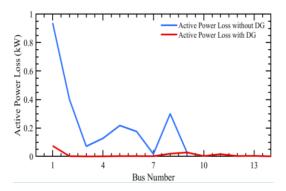


Figure 43 Active Power Loss (APL) showing before and after placement of DGs

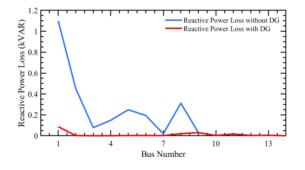


Figure 44 Reactive Power Loss (RPL) showing before and after placement of DGs

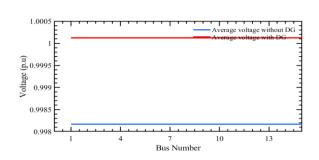


Figure 45 Average Voltage (Vavg) showing before and after placement of DGs

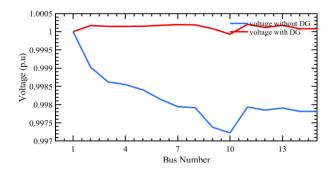


Figure 46 Minimum Voltage (Vmin) showing before and after placement of DG

5.3.1.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) in fig 45 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.:

5.3.1.4 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) in fig 46 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.3.1.5 Results Summary Using SHO Algorithm:

The SHO algorithm's results summary emphasizes the Distributed Generators' (DGs) ideal placement and sizing within the network, showing improvements in voltage profiles and decreased power losses. An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable		Before_	DG	After_D	G
'Min Voltage (p.u)'		0.9972	3'	'0.9999	31
'Min Voltage at Bus Location'		'10'		'10'	
'Total Active Power Losses (kW)'		2.2983	1	0.1525	7'
'Total Active Power Losses (%)'		'NA'		93.361	4'
'Total Reactive Power Losses (kV)	AR) '	12.6098		'0.1721	4'
'Total Reactive Power Losses (%)	10	'NA'		93.404	2'
'Power Factor' itional Information related to DG	s:	0.6648	2'	'0.969 <mark>8</mark>	1'
	s:	'0.6648		'0.9698 alue	1'
itional Information related to DG Variable	·	'0.6648			1.
itional Information related to DG. Variable 'DG Type'	'3'	'0.6648			1,
itional Information related to DG Variable 'DG Type' 'No of DGs'	·3'				1,
itional Information related to DG. Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '11	9 2'	V	alue	
itional Information related to DG. Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)' 'Sizes of DGs in kW'	'3' '3' '11 '386.	9 2' .171	V	alue .6767	1' 1' 449.9785

Table 12 Results Summary Using SHO Algorithm

5.3.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.3.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) in fig.47 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.3.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) in fig.48 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

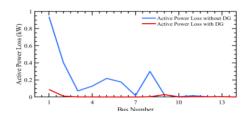


Figure 47 Active Power Loss (APL) showing before and after placement of DGs

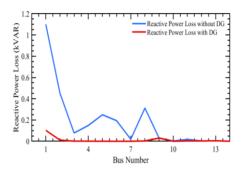


Figure 48 Reactive Power Loss (RPL) showing before and after placement of DGs

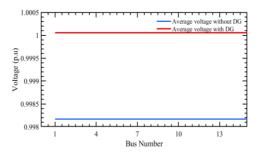


Figure 49 Average Voltage (Vavg) showing before and after placement of DGs

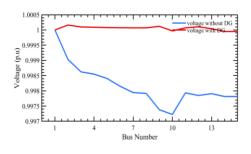


Figure 50 Minimum Voltage (Vmin) showing before and after placement of DGs

5.3.2.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) in fig.49 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.3.2.4 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) in fig.50 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.3.2.5 Results Summary Using ABC Algorithm:

The impact of DG placement is shown in the results summary using the ABC algorithm, with a focus on how it affects voltage stability and power loss reduction. A quick comparison of the ABC algorithm's performance against alternative approaches is provided in this section.

Variable	Before	_DG A1	ter_DG
'Min Voltage (p.u)'	'0.997	23' '0	.99995'
'Min Voltage at Bus Location'	'10'	• 1	.4'
'Total Active Power Losses (kW)'	12.298	3' '0	.13846'
'Total Active Power Losses (%)'	'NA'	1 9	93.9755'
'Total Reactive Power Losses (kVA	R)' '2.609	8' 'C	.15922'
'Total Reactive Power Losses (%)'	'NA'	19	93.8993'
'Power Factor'	'0.664	82' '0	.98211'
		Value	3
dditional Information related to DGs Variable 'DG Type'	'3'	Value	3
dditional Information related to DGs Variable 'DG Type' 'No of DGs'	'3' '3'	Value	3
dditional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '9 2 12'		
dditional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)' 'Sizes of DGs in kW'	'3' '3' '9 2 12' '590.2673	40.1	152 228.7612
dditional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '9 2 12'	40.1	152 228.7612

5.3.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm:

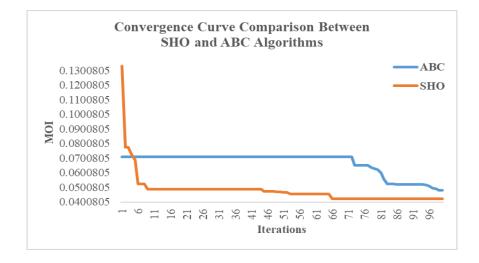


Figure 51 Comparison of Convergence Curve between SHO and ABC Algorithm

5.4 IESCO 18 Bus Network System:

The network in Figure 61 depicts the Torrit visualization of the IESCO regional distributive network, comprising 18 buses and 17 branches. The total active and reactive load demand of this network is 1390 kW and 1412 kVAR respectively.

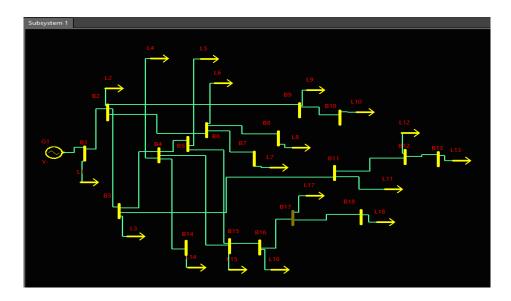


Figure 52 Torrit view of IESCO 18 Bus Network System

5.4.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.4.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

5.4.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

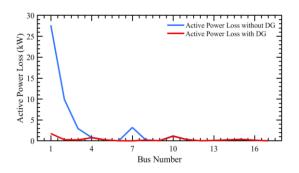
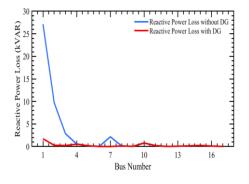
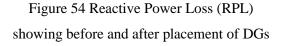


Figure 53 Active Power Loss (APL) showing before and after placement of DGs





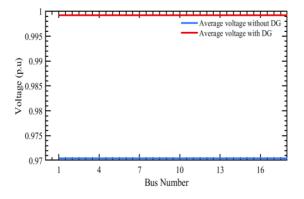
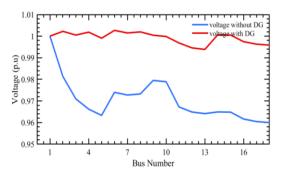
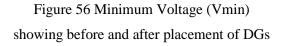


Figure 55 Average Voltage (Vavg) showing before and after placement of DGs





5.4.1.3 Average Voltage (V_{avg}) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.4.1.4 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.4.1.5 Results Summary Using SHO Algorithm:

An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable	Before	e_DG	After_DG	
	10.00		10,00000	•
'Min Voltage (p.u)'	0.96		0.99386	
'Min Voltage at Bus Location'	'18'		'13'	
'Total Active Power Losses (kW)'	47.6	148'	'5.9532'	
'Total Active Power Losses (%)'	'NA'		87.4971	1
'Total Reactive Power Losses (kV	'AR)' '44.4	792'	'4.71'	
'Total Reactive Power Losses (%)	'NA'		89.4107	1
'Power Factor' ditional Information related to DG		231'	'0.96926	
'Power Factor'			'0.96926 alue	
'Fower Factor' ditional Information related to DG Variable				
'Power Factor' ditional Information related to DG	s: 			
'Power Factor' ditional Information related to DG Variable 'DG Type'	's: 			
'Power Factor' ditional Information related to DG Variable 'DG Type' 'No of DGs'	'3' '3' '3'	v		
'Power Factor' ditional Information related to DG Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '4 2 6'	v	alue	, 370.5708' 381.3133'

5.4.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.4.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.4.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

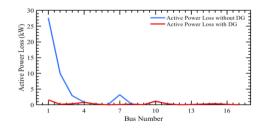


Figure 57 Active Power Loss (APL) showing before and after placement of DGs

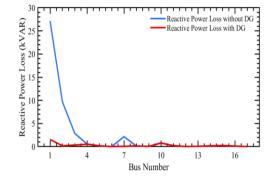


Figure 58 Reactive Power Loss (RPL) showing before and after placement of DGs

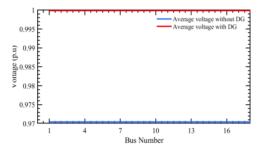


Figure 59 Average Voltage (Vavg) showing before and after placement of DGs

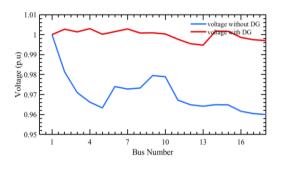


Figure 60 Minimum Voltage (Vmin) showing before and after placement of DG

5.4.2.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.:

5.4.2.4 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.4.2.5 Results Summary Using ABC Algorithm:

The impact of DG placement are shown in the results summary using the ABC algorithm, with a focus on how it affects voltage stability and power loss reduction. A quick comparison of the ABC algorithm's performance against alternative approaches is provided in this section.

Variable	Before_DO	G After_DG	2
'Min Voltage (p.u)'	'0.96'	0.99472	St. 1
'Min Voltage at Bus Location'	'18'	'13'	
'Total Active Power Losses (kW)'	47.6148	5.8931	
'Total Active Power Losses (%)'	'NA'	87.6234	·2
'Total Reactive Power Losses (kVA	AR)' '44.4792'	4.6144	
'Total Reactive Power Losses (%)'	'NA'	89.6257	
TOORT NERCOTTE LOWET DODDED (8)			
'Power Factor' itional Information related to DGs	'0.70231'	0. 95954	
'Power Factor'	'0.70231'	'0.95954 Value	n I
'Power Factor' itional Information related to DGs	'0.70231'		
'Fower Factor' itional Information related to DGs Variable	'0.70231' s:		. d
'Power Factor' itional Information related to DGs Variable 'DG Type'	'0.70231' s: '3' '3'		
'Power Factor' itional Information related to DGs Variable 'DG Type' 'No of DGs'	'0.70231' s: 		768.2108
'Power Factor' itional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'0.70231' s: '3' '2 7 4' '280.1845	Value	

5.4.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm:

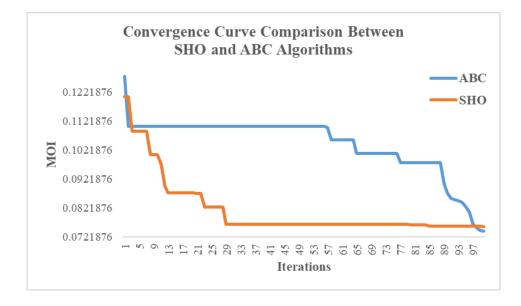


Figure 61 Comparison of Convergence Curve between SHO and ABC Algorithm

5.5 IESCO 20 Bus Network System:

The network in Figure 72 depicts the Torrit visualization of the IESCO regional distributive network, comprising 20 buses and 18 branches. The total active and reactive load demand of this network is 2234 kW and 2485 kVAR respectively.

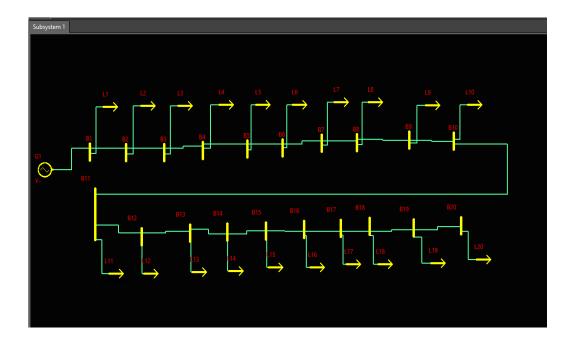


Figure 62 Torrit view of IESCO 20 Bus Network System

5.5.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.5.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) in fig.63 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm

5.5.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) in fig.64 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

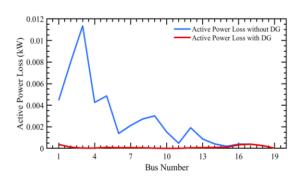


Figure 63 Active Power Loss (APL) showing before and after placement of DGs

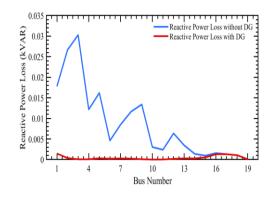


Figure 64 Reactive Power Loss (RPL) showing before and after placement of DG

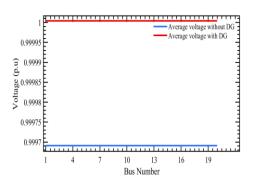


Figure 65 of Average Voltage (Vavg) showing before and after placement of DGs

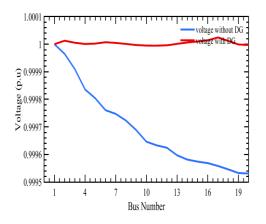


Figure 66 Minimum Voltage (Vmin) showing before and after placement of DGs

5.5.1.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) in fig.65 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.5.1.4 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) in fig.66 showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.5.1.5 Results Summary Using SHO Algorithm:

An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable	Before		After	DG
'Min Voltage (p.u)'	'0.999	53'	0.9999	99'
'Min Voltage at Bus Location'	'20'		'11'	
'Total Active Power Losses (kW)'	0.048	605'	'0.0021351'	
'Total Active Power Losses (%)'	'NA'		'95.60'	72'
'Total Reactive Power Losses (kVA	AR)' '0.162	88'	'0.008	1101'
'Total Reactive Power Losses (%)'	'NA'		95.020	09'
Davage Frankryl	10 950	65'	0.999	191
'Power Factor'		00	0.555	1.5
		Val		
tional Information related to DGs				
tional Information related to DGs Variable				
tional Information related to DGs Variable 'DG Type'	3: 			
tional Information related to DGs Variable 'DG Type' 'No of DGs'	3: 	Val		12.12384

Table 16 Results Summary Using SHO Algorithm

5.5.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.5.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.5.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

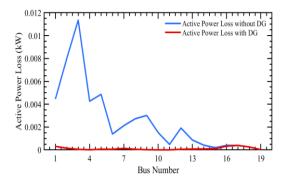


Figure 67 Active Power Loss (APL) showing before and after placement of DGs

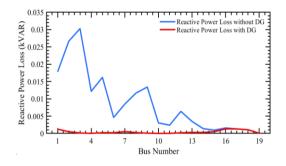


Figure 68 Reactive Power Loss (RPL) showing before and after placement of DG

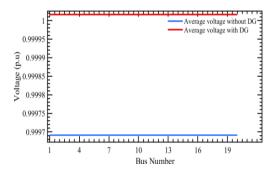


Figure 69 of Average Voltage (Vavg) showing before and after placement of DGs

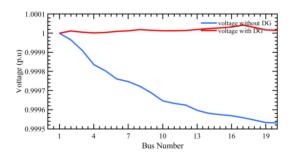


Figure 70 Minimum Voltage (Vmin) showing before and after placement of DGs

5.5.2.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.5.2.4 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.5.2.5 Results Summary Using ABC Algorithm:

The impact of DG placement are shown in the results summary using the ABC algorithm, with a focus on how it affects voltage stability and power loss reduction. A quick comparison of the ABC algorithm's performance against alternative approaches is provided in this section.

Variable		Before	DG	After	r_DG	
'Min Voltage (p.u)'		0.9995	3'	'1'		
'Min Voltage at Bus Location'		'20' '1		'1'	1'	
'Total Active Power Losses (kW)'		'0.048605'		'0.0021558'		
'Total Active Power Losses (%)'		'NA'		95.5	647'	
'Total Reactive Power Losses (kV	AR) '	0.1628	8'	10.008	81778'	
'Total Reactive Power Losses (%)	1	'NA'		'94.9794'		
'Power Factor' tional Information related to DG	s:	10.8506	5'	10.996	664'	
	;s: 	'0.8506		'0.990 lue	664'	
tional Information related to DG	is: 	'0.8506			664'	
tional Information related to DG Variable		'0.8506			564'	
tional Information related to DG Variable 'DG Type' 'No of DGs'	131 131	'0.8506 8 17'			664'	
tional Information related to DG Variable 'DG Type' 'No of DGs'	·31 ·31	8 17'	Va			
tional Information related to DG Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '2 '30.5	8 17'	Va 95.	lue		

Table 17 Results Summary Using ABC Algorithm

5.5.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm.

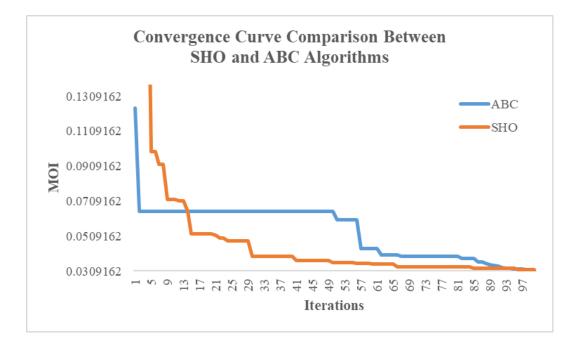


Figure 71 Convergence Curve showing comparison between SHO and ABC algorithm.

5.6 22 Bus Network System:

The network in Figure 83 depicts the Torrit visualization of the IESCO regional distributive network, comprising 20 buses and 28 branches. The total active and reactive load demand of this network is 2781 kW and 2973 kVAR respectively.

Here is the Torrit network single line diagram:

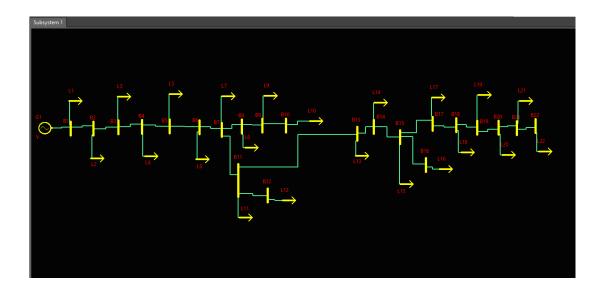


Figure 72 Torrit view of 22 Bus Network System

5.6.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.6.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm

5.6.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

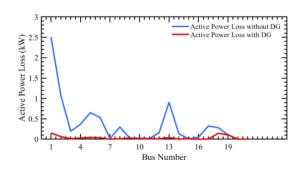


Figure 73 Active Power Loss (APL) showing before and after placement of DGs

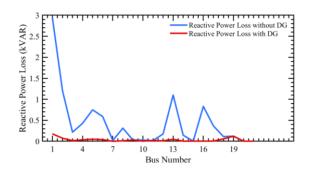


Figure 74 Reactive Power Loss (RPL) showing before and after placement of DGs

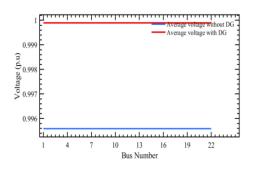
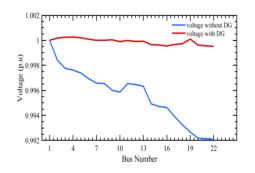
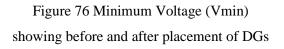


Figure 75 Average Voltage (Vavg) showing before and after placement of DGs





5.6.1.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.:

5.6.1.4 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.:

5.6.1.5 Results Summary Using SHO Algorithm:

An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable		Befo	ore_DO		After_DG	
'Min Voltage (p.u)'		'0.9	9209'		'0.99951	,
'Min Voltage at Bus Location'		'22'			'22'	
'Total Active Power Losses (kW)'		17.6	5956'		0.72232	•
'Total Active Power Losses (%)'		'NA			90.6139	
'Total Reactive Power Losses (kVA	AR) '	19.3	8969'		0.71291	•
'Total Reactive Power Losses (%)'		'NA			92.4133	•
			0.66467			
'Power Factor' itional Information related to DGs	з:	10.6	56467'		0.94627	
	s: 	'0.6	56467'	Val		2
itional Information related to DGs	s: '3'	'0.6	56467'			
itional Information related to DGs Variable		'0.6	56467'			
itional Information related to DGs Variable 'DG Type'		۰۵.۴ 9	4'			
Itional Information related to DGs Variable 'DG Type' 'No of DGs'	'3' '3'	9	4'	Val		414.5597
itional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '19	9	4'	V al 472.	ue	

Table 18 Results Summary Using SHO Algorithm

5.6.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.6.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm

5.6.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

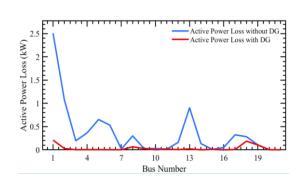


Figure 77 Active Power Loss (APL) showing before and after placement of DGs

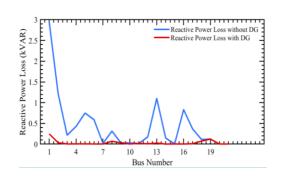


Figure 78 Reactive Power Loss (RPL) showing before and after placement of DGs

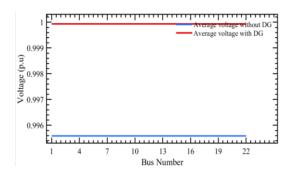


Figure 79 Average Voltage (Vavg) showing before and after placement of DGs

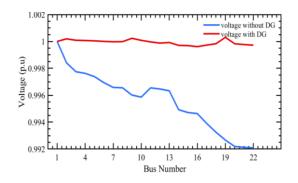


Figure 80 Minimum Voltage (Vmin) showing before and after placement of DG

5.6.2.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.6.2.4 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm

5.6.2.5 Results Summary Using ABC Algorithm:

The impact of DG placement are shown in the results summary using the ABC algorithm, with a focus on how it affects voltage stability and power loss reduction. A quick comparison of the ABC algorithm's performance against alternative approaches is provided in this section.

Variable	Before_DO	After_DG
'Min Voltage (p.u)'	'0.99209'	'0.99961'
'Min Voltage at Bus Location'	'22'	'16'
'Total Active Power Losses (kW)'	'7.6956'	'0.7155'
'Total Active Power Losses (%)'	'NA'	'90.7026'
'Total Reactive Power Losses (kVA	R)' '9.3969'	'0.66945'
'Total Reactive Power Losses (%)'	'NA'	'92.8759'
	Laboration and the second s	
'Power Factor' itional Information related to DGs	'0.66467'	'0.97214'
		Value
itional Information related to DGs		
itional Information related to DGs Variable	2.54 (1921) 2.5 2.5 2.5 2.5 2.5 2.5 2.5 2.5 2.5 2.5	
itional Information related to DGs Variable 'DG Type'	'3' '3'	
itional Information related to DGs Variable 'DG Type' 'No of DGs'	'3' '3' '3' '2 19 9'	
itional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'3' '3' '2 19 9' '1 576.307	Value

Table 19 Results Summary Using ABC Algorithm

5.6.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm

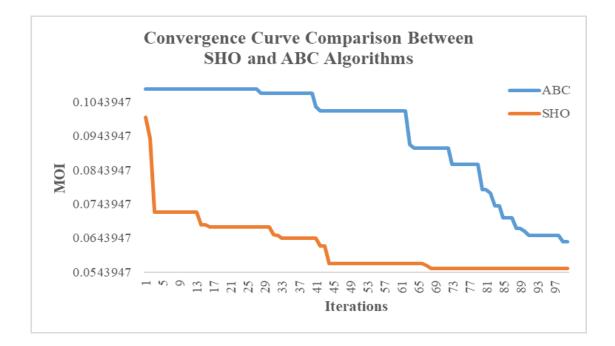


Figure 81 Convergence Curve showing comparison between SHO and ABC algorithm

5.7 26 Bus Network System:

The network in Figure 83 depicts the Torrit visualization of the IESCO regional distributive network, comprising 26 buses and 25branches. The total active and reactive load demand of this network is 3248 kW and 3692 kVAR respectively.

Here is the Torrit network single line diagram:

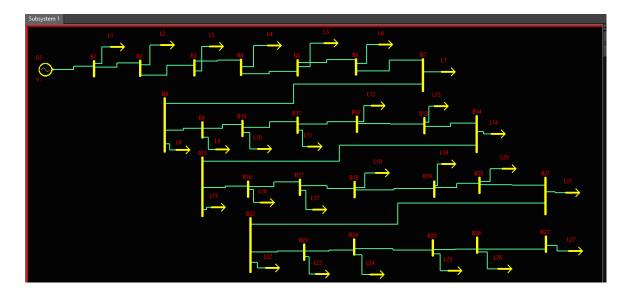


Figure 82 Torrit view of 26 Bus Network System

5.7.1 Backward Forward Sweep (BFS) Analysis of Network Using SHO Algorithm:

Below is the analysis of main parameters obtained using BFS method with SHO Algorithm:

5.7.1.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm

5.7.1.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm:

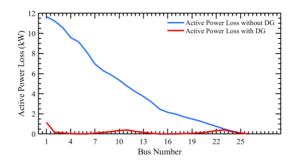
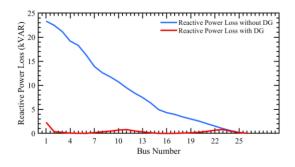
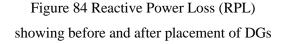


Figure 83 Active Power Loss (APL) showing before and after placement of DGs





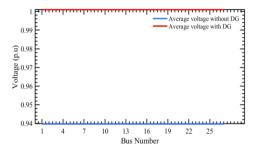


Figure 85 Average Voltage (Vavg) showing before and after placement of DGs

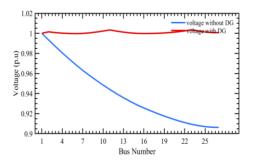


Figure 86 Minimum Voltage (Vmin) showing before and after placement of D

5.7.1.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.7.1.4 Minimum Voltage (V_{min}) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with SHO Algorithm.

5.7.1.5 Results Summary Using SHO Algorithm:

The SHO algorithm's results summary emphasizes the Distributed Generators' (DGs) ideal placement and sizing within the network, showing improvements in voltage profiles and decreased power losses. An outline of the efficiency benefits attained by the SHO algorithm is given in this section.

Variable	Before_D	G After_DG	
'Min Voltage (p.u)'	0.90625	'0.99971'	
'Min Voltage at Bus Location'	'27'	161	
'Total Active Power Losses (kW)'	114.419	5' '4.6809'	
'Total Active Power Losses (%)'	'NA'	'95.909'	
'Total Reactive Power Losses (kVA	AR)' '228.838	9' '9.3618'	
'Total Reactive Power Losses (%)'			
'Power Factor'	'0.69172	' '0.99649'	
'Power Factor'		' '0.99649' Value	
'Power Factor' Ltional Information related to DGs Variable			
'Power Factor' itional Information related to DGs Variable 'DG Type'	131		
'Power Factor' Lional Information related to DGs Variable 'DG Type' 'No of DGs'	131 131		
'Power Factor' tional Information related to DGs Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	*: '3' '3' '11 23 2'	Value	2.53118011
'Power Factor' tional Information related to DGs Variable 'DG Type' 'No of DGs'	131 131		2.5311801

Table 20 Results Summary Using SHO Algorithm

5.7.2 Backward Forward Sweep (BFS) Analysis of Network Using ABC Algorithm:

Below is the analysis of main parameters obtained using BFS method with ABC Algorithm:

5.7.2.1 Active Power Loss (APL) in term of DG placement:

Here is the graphical representation of Active Power Loss (APL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

5.7.2.2 Reactive Power Loss (RPL) in term of DG placement:

Here is the graphical representation of Reactive Power Loss (RPL) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.

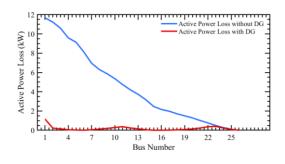
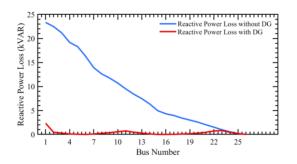
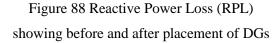


Figure 87 Active Power Loss (APL) showing before and after placement of DGs





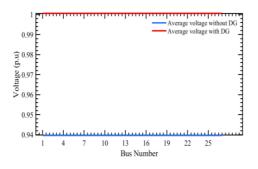


Figure 89 Average Voltage (Vavg) showing before and after placement of DGs

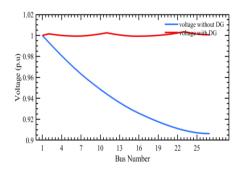


Figure 90 Minimum Voltage (Vmin) showing before and after placement of DGs

5.7.2.3 Average Voltage (Vavg) in term of DG placement:

Here is the graphical representation of Average Voltage (V_{avg}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.:

5.7.2.4 Minimum Voltage (Vmin) in term of DG placement:

Here is the graphical representation of Minimum Voltage (V_{min}) showing before and after placement of DGs optimally for Power factor improvement using BFS Method with ABC Algorithm.:

5.7.2.5 Results Summary Using ABC Algorithm:

The effects of DG location are illustrated in the findings summary of the ABC algorithm, with a focus on how they affect voltage stability and power loss reduction. A quick

comparison of the ABC algorithm's performance with alternative approaches is given in this section.

Variable	Before_DG	After_DG
'Min Voltage (p.u)'	'0.90625'	'0.99927'
'Min Voltage at Bus Location'	'27'	'16'
'Total Active Power Losses (kW)'	'114.4195'	'4.6554'
'Total Active Power Losses (%)'	'NA'	'95.9313'
'Total Reactive Power Losses (kVAR	()' '228.8389'	'9.3109'
		105 00101
'Total Reactive Power Losses (%)'	'NA '	95.9313
'Total Reactive Power Losses (%)' 'Power Factor' ditional Information related to DGs:	'0.69172'	'95.9313' '0.99705'
'Power Factor'	'0.69172'	
'Power Factor' ditional Information related to DGs: Variable	'0.69172'	'0.99705'
'Power Factor' ditional Information related to DGs: Variable 'DG Type'	'0.69172'	'0.99705'
'Power Factor' ditional Information related to DGs: Variable 'DG Type' 'No of DGs'	'0.69172' '3' '3'	'0.99705'
'Power Factor' ditional Information related to DGs: Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)'	'0.69172' '3' '2 11 23'	'0.99705' Value
'Power Factor' ditional Information related to DGs: Variable 'DG Type' 'No of DGs' 'Location of DGs (Bus Number)' 'Sizes of DGs in kW'	'0.69172' '3' '2 11 23' '74.403433	'0.99705'

Table 21 Results Summary Using ABC Algorithm

5.7.3 Comparison of Convergence Curve between SHO and ABC Algorithm:

Here is the graphical representation of Convergence Curve showing comparison between SHO and ABC algorithm

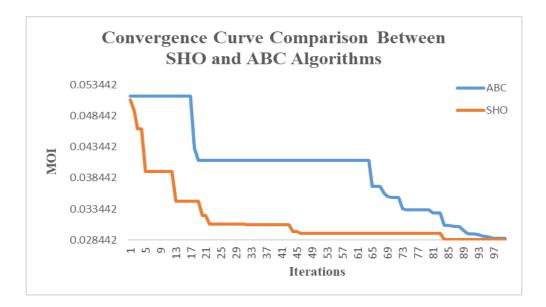


Figure 91 Convergence Curve showing comparison between SHO and ABC algorithm

5.8 Overall Result Comparison between Algorithms

The Sea Horse Optimization (SHO) and Artificial Bee Colony (ABC) algorithms' performances in various network architectures are contrasted in the table. It assesses important parameters before to and during the deployment of Distributed Generators (DGs), including minimum voltage, total active and reactive power losses, and power factor. The outcomes demonstrate how each tactic improved voltage stability, decreased power loss, and increased power factor. The comparison makes clear the relative benefits of the SHO and ABC algorithms for various distribution network optimization scenarios.

Network	Algorithm	gorithm Variable Before DG		After DG	% Improvement
IEEE 33 Bus	SHO	Min Voltage (p.u)	0.91309	0.9922	8.68%
		Total Active Power Losses (kW)	202.6771	18.3194	90.96%
		Total Reactive Power Losses (kVAR)	135.141	13.8962	89.72%
		Power Factor	0.8493	0.94739	11.56%
	ABC	Min Voltage (p.u)	0.91309	0.99218	8.68%
		Total Active Power Losses (kW)	202.6771	15.0378	92.58%
		Total Reactive Power Losses (kVAR)	135.141	11.585	91.43%
		Power Factor	0.8493	0.93222	9.77%

Table 22 Overall Result Comparison between Algorithms

					1
IEEE 69 Bus	SHO	Min Voltage (p.u)	0.99999	1	0.01%
		Total Active Power Losses (kW)	1.19E-05	1.76E-07	99.85%
		Total Reactive Power Losses (kVAR)	1.55E-06	5.84E-07	62.68%
		Power Factor	0.81587	0.99982	22.58%
	ABC	Min Voltage (p.u)	0.999999	1	0.01%
		Total Active Power Losses (kW)	1.19E-05	1.38E-06	88.42%
		Total Reactive Power Losses (kVAR)	1.55E-06	1.83E-07	88.17%
		Power Factor	0.81587	0.95267	16.73%
15 Bus	SHO	Min Voltage (p.u)	0.99723	0.99993	0.27%
		Total Active Power Losses (kW)	2.2983	0.15257	93.36%
		Total Reactive Power Losses (kVAR)	2.6098	0.17214	93.35%
		Power Factor	0.66482	0.96981	45.78%
	ABC	Min Voltage (p.u)	0.99723	0.99995	0.28%
		Total Active Power Losses (kW)	2.2983	0.13846	93.98%
		Total Reactive Power Losses (kVAR)	2.6098	0.15922	93.89%
		Power Factor	0.66482	0.98211	47.59%

[Ι	T	י ד	1
18 Bus	SHO	Min Voltage (p.u)	0.96	0.99386	3.53%
		Total Active Power Losses (kW)	47.6148	5.9532	87.50%
		Total Reactive Power Losses (kVAR)	44.4792	4.71	89.43%
		Power Factor	0.70231	0.96926	38.07%
	ABC	Min Voltage (p.u)	0.96	0.99472	3.61%
		Total Active Power Losses (kW)	47.6148	5.8931	87.62%
		Total Reactive Power Losses (kVAR)	44.4792	4.6144	89.65%
		Power Factor	0.70231	0.95954	36.58%
20 Bus	SHO	Min Voltage (p.u)	0.99953	0.999999	0.05%
		Total Active Power Losses (kW)	0.048605	0.0021351	95.63%
		Total Reactive Power Losses (kVAR)	0.16288	0.0081101	95.02%
		Power Factor	0.85065	0.99919	17.45%
	ABC	Min Voltage (p.u)	0.99953	1	0.05%
		Total Active Power Losses (kW)	0.048605	0.0021558	95.58%
		Total Reactive Power Losses (kVAR)	0.16288	0.0081778	95.02%
		Power Factor	0.85065	0.99664	17.23%

· · · · · ·				<u>г</u>	1
22 Bus	SHO	Min Voltage (p.u)	0.99209	0.99951	0.74%
		Total Active Power Losses (kW)	7.6956	0.72232	90.63%
		Total Reactive Power Losses (kVAR)	19.3969	0.71291	96.32%
		Power Factor	0.66467	0.94627	42.41%
	ABC	Min Voltage (p.u)	0.99209	0.99961	0.75%
		Total Active Power Losses (kW)	7.6956	0.7155	90.71%
		Total Reactive Power Losses (kVAR)	19.3969	0.66945	96.55%
		Power Factor	0.66467	0.97214	46.09%
26 Bus	SHO	Min Voltage (p.u)	0.90625	0.99971	10.30%
		Total Active Power Losses (kW)	27	6	77.78%
		Total Reactive Power Losses (kVAR)	NA	228.8389	NA
		Power Factor	0.69172	0.99649	44.24%
	ABC	Min Voltage (p.u)	0.90625	0.99927	10.28%
		Total Active Power Losses (kW)	27	16	40.74%
		Total Reactive Power Losses (kVAR)	NA	297.4579	NA
		Power Factor	0.69172	0.99752	44.24%

5.8.1 Visualization of results comparison with SHO Algorithm:

Figure 91 shows the impact of Distributed Generation (DG) deployment using the SHO (Sea Horse Optimization) algorithm on several network performance metrics. Among the many network types that are explored are many IESCO networks, IEEE 33-Bus, and IEEE 69-Bus.

Plots showing statistics on network performance before and after DG installation are included. Among these measurements are Power Factor, Total Active Power Losses (kW), Total Reactive Power Losses (VAR), and Min Voltage (p.u.). These measures are optimized using the SHO approach, and the results are displayed using two different colors: red for conditions after DG installation and dark blue for conditions before DG installation.

Solid lines represent measures after DG installation, while dotted lines represent metrics before DG installation.

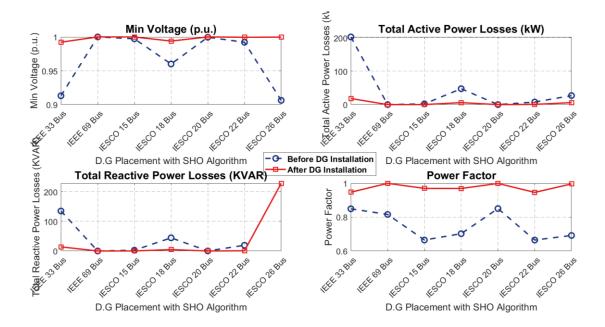


Figure 92 Visualization of results comparison with SHO Algorithm

This professional presentation highlights the effectiveness of the SHO algorithm in improving network performance through DG placement, offering a clear visual comparison of the enhancements achieved

5.8.2 Visualization of results comparison with ABC Algorithm:

The Graphs depicting the performance indicators of the ABC Algorithm before and after Distributed Generators (DGs) were integrated across various networks are shown in Figure 93. Power factor, total active and reactive power losses, and minimum voltage levels are among the important indicators that are shown. Red data is collected after DG deployment, and dark blue data is collected prior to DG placement. These visualizations allow one to assess the effectiveness of the ABC Algorithm in optimizing network performance with and without DGs. The graphs show how the integration of DGs has enhanced voltage stability, reduced power losses, and raised power factor through the examination of various metrics.

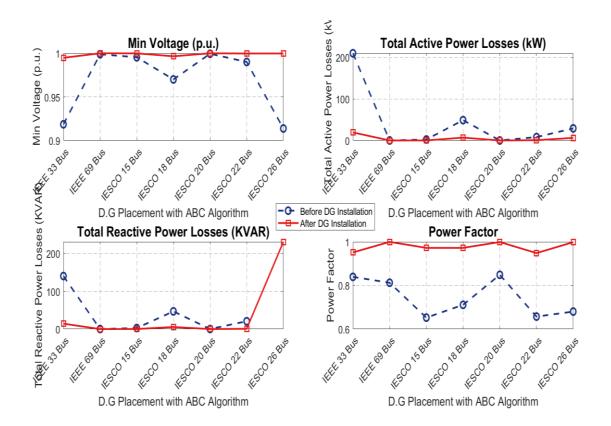


Figure 93 Visualization of results comparison with ABC Algorithm

5.8.3 Comparison between performance of Both Algorithms:

Based on the research, the following is a quick comparison of how well the SHO and ABC algorithms performed:

Metric	SHO Algorithm	ABC Algorithm
Voltage Levels	Shows a clear increase in voltage levels with DGs in every network, indicating improved stability.	Exhibits a noticeable increase in voltage levels with DGs, though slightly less than SHO.
Active Power Losses	Significant reduction in active power losses with DG placement, indicating better power distribution efficiency.	Reduction in active power losses is observed, but the improvement is less pronounced compared to SHO.
Reactive Power Losses	Major decrease in reactive power losses post-DG placement, showing effective power factor correction.	Decrease in reactive power losses is evident, but the reduction is not as significant as with SHO.
Power Factor	With the addition of DG, power factor considerably increased, increasing system efficiency	The power factor improvement is not as significant as the one obtained with SHO, however it is still noticeable

Table 23 Comparison between performance of Both Algorithms

Chapter 6

Conclusion and Future Work

6 CONCLUSION AND FUTURE WORK

6.1 Conclusion

This study concludes by offering a thorough analysis of various Distributed Generator (DG) types and their effects on distribution networks' power factor enhancement and reduction. The study successfully evaluated the performance of various distributed generator (DG) types and showed how, by utilizing the Sea Horse Optimization (SHO) algorithm and the Backward Forward Sweep Method (BFSM), they increase network efficiency.

An extensive investigation of IESCO regional networks and IEEE 33- and 69-bus systems validates the efficacy of the suggested solutions. Utility operators may optimize power system performance, maximize efficiency, and make informed decisions about power distribution management by utilizing the solid basis provided by this study. Utility operators can utilize these data to make informed decisions on the deployment of distributed generation (DGs). Depending on the particular requirements of the network, the type of distributed generation (DG) that is selected may have the impact of lowering power losses, increasing power factor, or striking a compromise between the two. Distribution network management techniques become more tailored and efficient as a result of this flexibility.

6.2 Future Work

Future work includes:

- Electric Vehicles (EVs) can be integrated in this framework.
- Utilization of Machine Learning techniques.
- The probabilistic Nature of Wind and Solar DGs can be incorporated to carry out probabilistic load flow.
- The objectives of economic analysis of DGs can also be taken into account.

REFERENCES

- [1] U. Sultana, Azhar B. Khairuddin, M.M. Aman, A.S. Mokhtar, N. Zareen, "A review of optimum DG placement based on minimization of power losses and voltage stability enhancement of distribution system," *Renewable and Sustainable Energy Reviews*, vol. 63, pp. 363-378, 2016.
- [2] Danish Ali, Ward Ul Hijaz Paul, Md Safdar Ali, Mubassir Ahmad, Haroon Ashfaq,
 "Optimal Placement of Distribution Generation Sources in Hybrid Generation Network," *Smart Grid and Renewable Energy*, vol. 5, no. 12, pp. 2151-4844, 2021.
- [3] Suresh, M.C.V., Belwin, E.J., "Optimal DG placement for benefit maximization in distribution networks by using Dragonfly algorithm," *Renewables*, vol. 4, no. 5, 2018.
- [4] P. Sundararaman, E. Mohan, S. V. Aswin Kumer, Sridhar Udayakumar, Abdissa Fekadu Moti, "Minimizing the Active Power Losses and Retaining the Voltage Profile of the Distribution System Using Soft Computing Techniques with DG Source," *Journal of Electrical Engineering*, 2022.
- [5] Nguyen TT, Dinh BH, Pham TD, Nguyen TT, "Active Power Loss Reduction for Radial Distribution Systems by Placing Capacitors and PV Systems with Geography Location Constraints," *Sustainability*, vol. 18, p. 12, 2020.
- [6] Salgado, Roberto & Moyano, Carlos & Medeiros, A.D.R., "Reviewing strategies for active power transmission loss allocation in power pools," *International Journal* of Electrical Power & Energy Systems, vol. 10, no. 26, pp. 81-90, 2004.
- [7] Ochoa, Luis F., and Gareth P. Harrison, "Minimizing energy losses: Optimal accommodation and smart operation of renewable distributed generation," *IEEE Transactions on Power Systems*, pp. 198-205, 2011.

- [8] D. Singh and K. Verma, "GA based energy loss minimization approach for optimal sizing & placement of distributed generation," *Int. J. Knowl. Intell. Eng. Syst*, vol. 12, pp. 147-156, 2008.
- [9] W.O. Prommee, W.Ongsakul, "Optimal multiple distributed generation placement in microgrid system by improved reinitialized social structures particle swarm optimization," *Eur. Trans. Electr. Power*, vol. 1, no. 21, pp. 489-504, 2011.
- [10] F. S. Abu-Mouti and M.E. El-Hawary, "Heuristic curve-fitted technique for distributed generation optimization in radial distribution feeder systems," *Proc. IET Gener. Trans. Distrib.*, vol. 2, no. 5, pp. 172-180, 2011.
- [11] F. S. Abu-Mouti and M. E. El-Hawary, ""Optimal Distributed Generation Allocation and Sizing in Distribution Systems via Artificial Bee Colony Algorithm," *In IEEE Transactions on Power Delivery*, vol. 26, no. 4, pp. 2090-2101, 2011.
- [12] R. Sanjay, T. Jayabarathi, T. Raghunathan, V. Ramesh and N. Mithulananthan,
 "Optimal Allocation of Distributed Generation Using Hybrid Grey Wolf Optimizer," *IEEE Access*, vol. 5, pp. 14807-14818, 2017.
- [13] Suresh Kumar Sudabattula, Kowsalya M, "Optimal allocation of solar based distributed generators in distribution system using Bat algorithm," *Perspectives in Science*, vol. 8, pp. 270-272, 2016.
- [14] Satish Kansal, Barjeev Tyagi and Vishal Kumar, "Cost-Benefit analysis for Optimal DG placement in distribution system," *Int. J. Ambient. Eng*, 2015.
- [15] S.A. ChithraDevi, L. Lakshminarasimman, R. Balamurugan, "Stud Krill herd Algorithm for multiple DG placement and sizing in a radial distribution system," *Engineering Science and Technology an International Journal*, vol. 20, no. 2, pp. 748-759, 2017.

- [16] K. Akbari, E. Rahmani, A. Abbasi, M. Askari, "Optimal placement of distributed generation in radial networks considering reliability and cost indices," *J. Intell. Fuzzy System*, vol. 2, no. 30, p. 1077–1086, 2016.
- [17] S. Kansal, V. Kumar, and B. Tyagi, "Optimal placement of different types of DGs in distribution networks," *Elect. Power Energy System*, no. 53, pp. 752–760, 2013.
- [18] Candelo-Becerra, John Edwin, & Hernández-Riaño, Helman Enrique, "Distributed generation placement in radial distribution networks using a bat-inspired algorithm," *DYNA*, vol. 82, no. 192, pp. 60-67, 2015.
- [19] Minnan Wang and JinZhong, "A Novel Method for Distributed Generation and Capacitor Optimal Placement considering Voltage Profiles," vol. 1, no. 5, p. 4577, 2011.
- [20] S. Devi, M. Geethanjal, "Application of Modified Bacterial Foraging Optimization algorithm for optimal placement and sizing of Distributed Generation," *Expert Syst Appl*, no. 41, p. 2772–81, 2014.
- [21] K. Mahmoud, N. Yorino and A. Ahmed, "Optimal Distributed Generation Allocation in Distribution Systems for Loss Minimization," *in IEEE Transactions on Power Systems*, vol. 31, no. 2, pp. 960-969, 2016.
- [22] T. R. Ayodele, A. S. O. Ogunjuyigbe, and O. O. Akinola, "Optimal location, sizing, and appropriate technology selection of distributed generators for minimizing power loss using genetic algorithm," *J Renew Energy*, 2015.
- [23] Duong, M.Q.; Pham, T.D.; Nguyen, T.T.; Doan, A.T.; Tran, H.V, "Determination of Optimal Location and Sizing of Solar Photovoltaic Distribution Generation Units in Radial Distribution Systems.," *Energies*, vol. 12, p. 174, 2019.
- [24] Acharya, Naresh, PukarMahat, and NadarajahMithulananthan, "An analytical approach for DG allocation in primary distribution network," *International Journal* of Electrical Power & Energy Systems, vol. 10, no. 28, pp. 669-678, 2006.

- [25] N. Mohd Zaid et al, "Multi-Objective Optimization for Sizing of Distributed Generation Using Cuckoo Search Algorithm," *Applied Mechanics and Materials*, vol. 785, pp. 34-37, 2015.
- [26] Thuan Thanh Nguyen, Anh Viet Truong, "Distribution network reconfiguration for power loss minimization and voltage profile improvement using cuckoo search algorithm," *Int. J. Electr. Power Energy System*, vol. 68, p. 233–242, 2015.
- [27] D. Rama Prabha, T. Jayabarathi, "Optimal placement and sizing of multiple distributed generating units in distribution networks by invasive weed optimization algorithm," *Ain Shams Engineering Journal*, vol. 7, pp. 683-694, 2016.
- [28] Satish Kansal, Vishal Kumar, Barjeev Tyagi, "Hybrid approach for optimal placement of multiple DGs of multiple types in distribution networks," *International Journal of Electrical Power & Energy Systems*, vol. 75, pp. 226-235, 2016.
- [29] Attia El-Fergany, "Optimal allocation of multi-type distributed generators using backtracking search optimization algorithm," *International Journal of Electrical Power & Energy Systems*, vol. 64, pp. 1197-1205, 2015.
- [30] Mohamed Imran A, Kowsalya M, "Optimal size and siting of multiple distributed generators in distribution system using bacterial foraging optimization," *Swarm and Evolutionary computation*, vol. 15, p. 2015, 58-65.
- [31] D. Rama Prabha, T. Jayabarathi, R. Umamageswari, S. Saranya, "Optimal location and sizing of distributed generation unit using intelligent water drop algorithm," *Sustainable Energy Technologies and Assessments*, vol. 11, pp. 106-113, 2015.
- [32] Zahid, M.; Chen, J.; Li, Y.; Duan, X.; Lei, Q.; Bo, W.; Mohy-ud-din, G.; Waqar,
 "A. New Approach for Optimal Location and Parameters Setting of UPFC for Enhancing Power Systems Stability under Contingency Analysis," *Energies 2017*, no. 1738, p. 10, 2017.

- [33] K. Rudion, A. Orths, Z. A. Styczynski, K. Strunz, "Design of Benchmark of Medium Voltage Distribution Network for Investigation of DG Integration" in IEEE, 2006
- [34] Z Srinivasa Rao Gampa, D. Das, Optimum placement of shunt capacitors in a radial distribution system for substation power factor improvement using fuzzy GA method, *International Journal of Electrical Power & Energy Systems*, Volume 77,2016
- [35] Samir M. Dawoud, Xiangning Lin, Merfat I. Okba, "Optimal placement of different types of RDGs based on maximization of microgrid," Journal of Cleaner Production, 2017
- [36] M.H. Moradi, M. Abedini, "A combination of genetic algorithm and particle swarm optimization for optimal DG location and sizing in distribution systems," International Journal of Electrical Power & Energy Systems, Vols. 34, no. 1, pp. 66-74, 2012.
- [37] R. S. Rao, K. Ravindra, K. Satish and S. V. L. Narasimham, "Power Loss Minimization in Distribution System Using Network Reconfiguration in the Presence of Distributed Generation," in *IEEE Transactions on Power Systems*, vol. 28, no. 1, pp. 317-325, Feb. 2013.
- [38] M. Kefayat, A. Lashkar Ara, S.A. Nabavi Niaki, "A hybrid of ant colony optimization and artificial bee colony algorithm for probabilistic optimal placement and sizing of distributed energy resources," Energy Conversion and Management, vol. 92, pp. 149-161, 2016.
- [39] Suresh, M.C.V., Belwin, E.J. Optimal DG placement for benefit maximization in distribution networks by using Dragonfly algorithm. *Renewables* 5, 4 (2018). https://doi.org/10.1186/s40807-018-0050-7
- [40] Dervis Karaboga, Bahriye Basturk, "A powerful and efficient algorithm for numerical function optimization:Artificial bee colony (abc) algorithm," Journal of Global Optimization, vol. 39, pp. 459-471, 2007

- [41] Dervis Karaboga, Bahriye Basturk, "An artificial bee colony (abc) algorithm for numeric function optimization," IEEE Swarm Intelligence Symposium, Indianapolis, Indiana, USA, 2006
- [42] Dervis Karaboga, Bahriye Basturk, "On the performance of artificial bee colony (ABC) algorithm," Applied Soft Computing, vol. 8, pp. 687-697, 2008

7 ANNEXURE

Attested Report for Data validation from IESCO Operation Department:

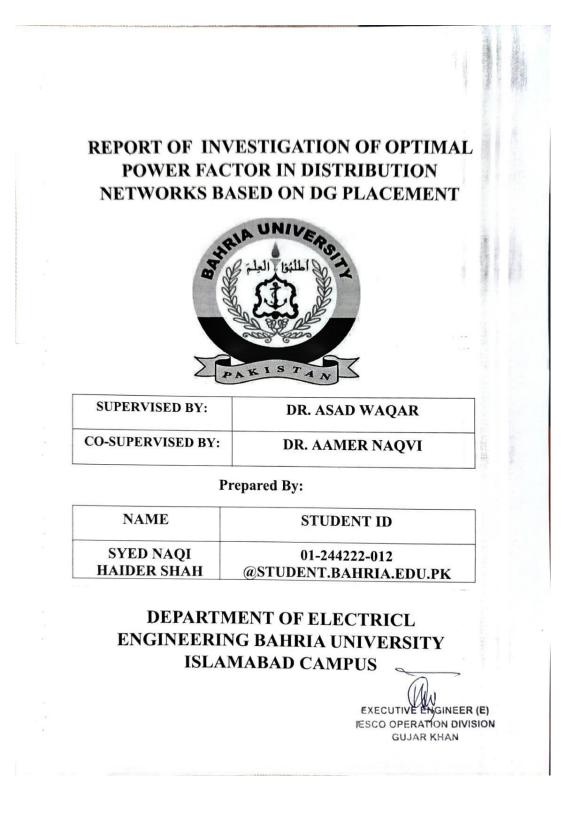


Table	of Contents
IESCO (ujjar Khan Division Networks:
1.	15 Bus 15 Branch Network System:
Si	gle Line Diagram Of Network:
R	sults Summary Comparison Before & After D.G Placement:
2.	18 Bus 17 Branch Network System:5
S	ngle Line Diagram Of Network:
R	sults Summary Comparison Before & After D.G Placement:
3.	20 Bus 18 Branch Network System:
S	ngle Line Diagram Of Network:10
R	esults Summary Comparison Before & After D.G Placement:10
4.	22 Bus 28 Branch Network System:11
S	ngle Line Diagram Of Network:
R	esults Summary Comparison Before & After D.G Placement:
5.	27 Bus 27 Branch Network System:
S	ngle Line Diagram Of Network:
R	sults Summary Comparison Before & After D.G Placement:
IEEE St	andard Networks:
1.	IEEE 33 Bus Network System:
S	ngle Line Diagram Of Network:
R	sults Summary Comparison Before & After D.G Placement:
2.	IEEE 69 Bus Network System:
S	ngle Line Diagram Of Network:
R	sults Summary Comparison Before & After D.G Placement:
Param	eters Description:
1.	Bus Data Units and Format:
2.	Generator Data Units and Format:
3.	Branch Data Units and Format:

EXECUTIVE ENGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

IESCO Gujjar Khan Division Networks:

Following are the details about parameters of networks which are retrieved from IESCO Gujjar khan Division for testing and INVESTIGATION OF OPTIMAL POWER FACTOR IN DISTRIBUTION NETWORKS BASED ON DG PLACEMENT.

1. 15 Bus 15 Branch Network System:

Input Parameters being used in network are shown below (Parameters are converted into P.U):

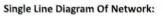
										-	_	Contraction of the
have 1		Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	Zone	Vmax	Vmin
bus_i	type	Design and the second second	Complete per de la la la	Annalytic State	Anothers in	1	1	0	11	1	1.06	0.94;
1	3	0	0	0	0		Contraction of the second		11	1	1.06	0.94;
2	1	0	0	0	0	1	1	0			1.06	0.94;
3	1	0.0445	0.05	0	0	1	1	0	11	1	CONTRACTOR OF CONTRACTOR OF CONTRACTOR	
4	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
5	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
6	1	0.0089	0.01	0	0	1	1	0	11	1	1.06	0.94;
Contraction of the local division of the	The second second		unbolicele of state	C STATUTE	AND PARTY	1	1	0	11	1	1.06	0.94;
7	1	0	0	0	0	and only of the local	Lard Dave Large	BOCOVERSE.	Concession of the second of the	1	1.06	0.94;
8	1	0.0445	0.05	0	0	1	1	0	11	Contraction Contraction	and the stand of the stand	COMPANY INCOME.
9	1	0.356	0.4	0	0	1	1	0	11	1	1.06	0.94;
10	1	0.178	0.2	0	0	1	1	0	11	1	1.06	0.94;
11	1	0	0	0	0	1	1	0	11	1	1.06	0.94;
12	1	0.178	0.2	0	0	1	1	0	11	1	1.06	0.94;
safesting of the second	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
13	and the second second	and the second se	Par start Constraints	V HOLES	-	-	The second		the state of the state of the		1.06	0.94;
14	1	0.0445	0.05	0	0	1	1	0	11	1	CONTRACTOR OF STREET	Matter Science Load made
15	1	0	0	0	0	1	1	0	11	1	1.06	0.94;

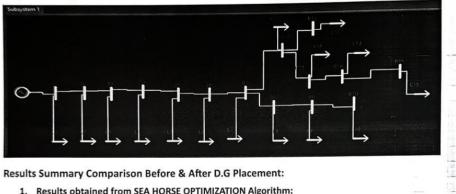
	La L			in the second second	ienerator Dr	ta		Storage Party			
Bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1	
1	0	0 10			10 -10 1		100	100 1		0	0
Pc2	Qc1min	Qc1max	Qc2min	Qc2max	ramp_agc	ramp_10	ramp_30		ramp_q	Apf	
0	0	0	0	0	0	0	0		0	0	

e de la complete	Branch data												
fbus	tbus	r	x	b	rateA	rateB	rateC	ratio	angle	status	angmin	angmax	
1	2	0.011	0.013	0	0	0	0	0	0	1	-360	360.0;	
2	3	0.0047	0.0053	0	0	0	0	0	0	1	-360	360.0;	
3	4	0.0009	0.001	0	0	0	0	0	0	1	-360	360.0;	
4	5	0.0018	0.0021	0	0	0	0	0	0	1	-360	360.0;	
5	6	0.0034	0.0039	0	0	0	0	0	0	1	-360	360.0;	

EXECUTIVE SINGINEER (E) ESCO OPERATION DIVISION GUJAR KHAN

6	7	0.0028	0.0031		0	0	0	0	0	1	-360	360.0;
7	8	0.00028	0.0031	0	0	0	1000	0	0	1	-360	360.0;
8	9		0.0006	0	0	0	0	0	0	1	-360	360.0;
9		0.012	0.0125	0	0	0	0	0	0	1	-360	360.0;
3	10	0.01	0.011	0	0	0	0	0	0	1	-360	360.0;
/	11	0.0004	0.0004	0	0	0	0	0	0	1	-360	360.0;
11	12	0.0057	0.0062	0	0	0	0		0	1	-360	360.0;
11	13	0.0036	0.0039	0	0	0	0	0		1	-360	360.0;
13	14	0.023	0.028	0	0	0	0	0	0	1	-360	360.0;
14	15	0.0039	0.0042	0	0	0	0	0	0	-	-300	550107





Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

Variable		Bef	ore	DG	After_D	G
'Min Voltage (p.u)'		·o.	9972	3'	'0.9999	3'
'Min Voltage at Bus Location'		10			.10.	
'Total Active Power Losses (kW)'		12.	2983		10.1525	7'
'Total Active Power Losses (%)'		'NA			193.361	4'
'Total Reactive Power Losses (kV)	AR) '	•2.	6098		10.1721	.4 '
'Total Reactive Power Losses (%)	•	'NA			193.404	2'
'Power Factor'		·o.	6648	2'	10.9698	1'
'DG Type'	.3.					
'No of DGs'	.3.	10.000				
'Location of DGs (Bus Number)'	•11	9	2'			
'Sizes of DGs in kW'	'386.				. 6767	1,
'Sizes of DGs in kVAR'	1673.			39	7.0128	449.9785
'MOI'	10.04	17933				

EXECUTIVE ENGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm: Sum

ary of Results Comparison: Variable	Before_DC	After_DG
'Min Voltage (p.u)' 'Nin Voltage at Bus Location' 'Total Active Power Losses (KW)' 'Total Active Power Losses (%)' 'Total Reactive Power Losses (%)' 'Total Reactive Power Losses (%)'	'0.99723' '10' '2.2983' 'NA' '2.6098' 'NA! '0.66482'	'0.99995' '14' '0.13846' '93.9755' '0.15922' '93.8993' '0.98211'

Additional Information related to DGs:

And the second se

Variable	Value	
'DG Type'	·3· ·3·	
'No of DGs' 'Location of DGs (Bus Number)'	'9 2 12'	
'Sizes of DGs in kW'	-390.2073 10.200	7612'
'Sizes of DGs in kVAR' 'MOI'	'650.629 596.8152 298.3 '0.04208'	533.

Value

2. 18 Bus 17 Branch Network System:

Parameters are entered in their own units as Pd and Qd are specified in kW & kVAr. Also r and x specified in ohms here. Code will Automatically Convert them into P.U before burning

							ata					
bus_i	type	Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	zone	Vmax	Vmin
1	3	0	0	0	0	1	1	0	11	1	1	1;
2	1	43.5	45.5	0	0	1	1	0	11	1	1.1	0.9;
3	1	71.5	72.5	0	0	1	1	0	11	1	1.1	0.9;
4	1	140.5	142	0	0	1	1	0	11	1	1.1	0.9;
5	1	45	46	0	0	1	1	0	11	1	1.1	0.9;
6	1	140.5	142	0	0	1	1	0	11	1	1.1	0.9;
7	1	139.5	141.5	0	0	1	1	0	11	1	1.1	0.9;
8	1	71	72	0	0	1	1	0	11	1	1.1	0.9;
9	1	69.5	71	0	0	1	1	0	11	1	1.1	0.9;
10	1	43	44	0	0	1	1	0	11	1	1.1	0.9;
11	1	139	141	0	0	1	1	0	11	1	1.1	0.9;

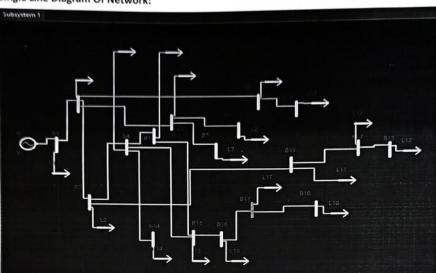
EXECUTIVE ENGINEER (E) ESCO OPERATION DIVISION GUJAR KHAN

12	1	71	72	0	0	1	1	0	11	1	1.1	0.9;
13	1	44	45	0	0	1	1	0	11	1	1.1	0.9;
14	1	69	71	0	0	1	1	0	11	1	1.1	0.9;
15	1	141	142.5	0	0	1	1	0	11	1	1.1	0.9;
16	1	70	71	0	0	1	1	0	11	1	1.1	0.9;
17	1	70.5	71.5	0	0	1	1	0	11	1	1.1	0.9;
18	1	43.5	44.5	0	0	1	1	0	11	1	1.1	0.9;

bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1
	1	0	0	10	-10	1	100	1	10	0
Pc2	Qc1min	Qc1max	Qc2min	Qc2max	ramp_agc	ramp_10	ramp_30	ramp_q	ap	of
0	0	0	0	0	0	0	0	0 -	0	

		in the second				Bit	anch dat	а				
fbus	tbus	r	X	ь	rateA	rateB	rateC	ratio	angle	status	angmin	angmax
1	2	0.775	0.76	0	0	0	0	0	0	1	-360	360;
2	3	0.67	0.657	0	0	0	0	0	0	1	-360	360;
3	4	0.485	0.47	0	0	0	0	0	0	1	-360	360;
4	5	0.875	0.59	0	0	0	0	0	0	1	-360	360;
2	9	1.155	0.78	0	0	0	0	0	0	1	-360	360;
9	10	0.967	0.653	0	0	0	0	0	0	1	-360	360;
2	6	1.47	0.99	0	0	0	0	0	0	1	-360	360;
6	7	0.625	0.42	0	0	0	0	0	0	1	-360	360;
6	8	0.72	0.485	0	0	0	0	0	0	1	-360	360;
3	11	1.03	0.696	0	0	0	0	0	0	1	-360	360;
11	12	1.405	0.95	0	0	0	0	0	0	1	-360	360;
12	13	1.155	0.78	0	0	0	0	0	0	1	-360	360;
4	14	1.28	0.865	0	0	0	0	0	0	1	-360	360;
4	15	0.685	0.465	0	0	0	0	0	0	1	-360	360;
5	16	0.625	0.42	0	0	0	0	0	0	1	-360	360;
16	17	0.72	0.485	0	0	0	0	0	0	1	-360	360;
17	18	0.72	0.485	0	0	0	0	0	0	1	-360	360;

EXECUTIVE ANGINEER (E) ESCO OPERATION DIVISION GUJAR KHAN



Single Line Diagram Of Network:

Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

Summary of Results Comparison: Variable Before_DG After_DG _____ -'Min Voltage (p.u)' 0.96' '0.99386' 'Min Voltage at Bus Location' 18' 13' 'Total Active Power Losses (kW)' 47.6148' '5.9532' 'Total Active Power Losses (%)' 'NA' '87.4971' 'Total Reactive Power Losses (kVAR)' 44.4792' '4.71' 'Total Reactive Power Losses (%)' 'NA' '89.4107' 'Power Factor' 0.70231 '0.96926' Additional Information related to DGs: Variable Value

'DG Type'	'3'		
'No of DGs'	131		
'Location of DGs (Bus Number)'	'4 2 6'		
'Sizes of DGs in kW'	'768.1084	134.6035	370.5708'
'Sizes of DGs in kVAR'	'744.3397	817.2148	381.3133'
'MOI'	'0.074188'		
		-	0

EXECUTIVE FIGINEER (E) JESCO OPERATION DIVISION GUJAR KHAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm:

Variable	Before	_ba	After_D	
'Min Voltage (p.u)' 'Min Voltage at Bus Location' 'Total Active Power Losses (kW)' 'Total Active Power Losses (%)' 'Total Reactive Power Losses (%)	'NA' 'AR)'''44.47 'NA'	48' 92'	0.9947 13 5.8931 87.623 4.6144 89.625	1 * 1 * 5 7 *
'Power Factor'	10.702	31	0.5550	
'Power Factor' Additional Information related to DG Variable			lue	
Additional Information related to DG		Va		768.2108'

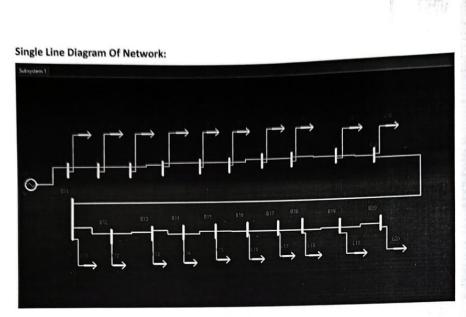
3. 20 Bus 18 Branch Network System:

Parameters are entered in their own units as Pd and Qd are specified in kW & kVAr. Also r and x specified in ohms here. Code will Automatically Convert them into P.U before burning

bus_i	type	Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	zone	Vmax	Vmin
1	3	20	10	0	0	1	1	0	12.66	1	1.05	0.95;
2	1	0	0	0	0	1	1	0	12.66	1	1.05	0.95;
3	1	25	15	0	0	1	1	0	12.66	1	1.05	0.95;
4	1	30	20	0	0	1	1	0	12.66	1	1.05	0.95;
5	1	15	10	0	0	1	1	0	12.66	1	1.05	0.95;
6	1	20	10	0	0	1	1	0	12.66	1	1.05	0.95;
7	1	10	5	0	0	1	1	0	12.66	1	1.05	0.95;
8	1	10	5	0	0	1	1	0	12.66	1	1.05	0.95;
9	1	15	10	0	0	1	1	0	12.66	1	1.05	0.95;
10	1	10	5	0	0	1	1	0	12.66	1	1.05	0.95;
11	1	20	15	0	0	1	1	0	12.66	1	1.05	0.95;
12	1	10	5	0	0	1	1	0	12.66	1	1.05	0.95;

EXECUTIVE GINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

\$h.)



Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

mary of Results Comparison: Variable	Before_DG	After_DG
'Min Voltage (p.u)'	0.99953	0.99999'
'Min Voltage at Bus Location'	'20'	'11'
'Total Active Power Losses (kW)'	'0.048605'	0.0021351
'Total Active Power Losses (%)'	'NA'	'95.6072'
'Total Reactive Power Losses (kVAR)'	'0.16288'	10.0081101
'Total Reactive Power Losses (%)'	'NA'	'95.0209'
'Power Factor'	·0.85065 ·	·0.99919·

Additional Information related to DGs: Variable

'DG Type'	131		
	and a second second		
'No of DGs'	'3'		
'Location of DGs (Bus Number)'	'6 17 2'		
'Sizes of DGs in kW'	111.1325	177.0516	12.12384
'Sizes of DGs in kVAR'	48.2828	121.2475	137.0119'
'MOI'	'0.030916'		

EXECUT SINGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

Value

13	1	15	10	0	0	1	1	0	12.66	1	1.05	0.95;
14	1	15	10	0	0	1	1	0	12.66	1	1.05	0.95;
15	1	20	15	0	0	1	1	0	12.66	1	1.05	0.95;
16	1	10	5	0	0	1	1	0	12.66	1	1.05	0.95;
17	1	25	15	0	0	1	1	0	12.66	1	1.05	0.95;
18	1	20	15	0	0	1	1	0	12.66	1	1.05	0.95;
19	1	30	20	0	0	1	1	0	12.66	1	1.05	0.95;
20	1	20	10	0	0	1	1	0	12.66	1	1.05	0.95;

				G	emanation Dat	ta .				
bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1
1	0	0	30	-30	1	100	1	30	0	0
Pc2	Qc1min	Qc1max	Qc2min	Qc2max	ramp_agc	ramp_10	ramp_30	ramp_q	ар	of
0	0	0	0	0	0	0	0	0	0	; 11

						Br	anch da	ta				
fbus	tbus	r	x	b	rateA	rateB	rateC	ratio	angle	status	angmin	angmax
1	2	0.005	0.02	0	0	0	0	0	0	1	-360	360;
2	3	0.009	0.03	0	0	0	0	0	0	1	-360	360;
3	4	0.015	0.04	0	0	0	0	0	0	1	-360	360;
4	5	0.007	0.02	0	0	0	0	0	0	1	-360	360;
5	6	0.009	0.03	0	0	0	0	0	0	1	-360	360;
6	7	0.003	0.01	0	0	0	0	0	0	1	-360	360;
7	8	0.005	0.02	0	0	0	0	0	0	1	-360	360;
8	9	0.007	0.03	0	0	0	0	0	0	1	-360	360;
9	10	0.009	0.04	0	0	0	0	0	0	1	-360	360;
10	11	0.005	0.01	0	0	0	0	0	0	1	-360	360;
11	12	0.002	0.01	0	0	0	0	0	0	1	-360	360;
12	13	0.009	0.03	0	0	0	0	0	0	1	-360	360;
13	14	0.005	0.02	0	0	0	0	0	0	1	-360	360;
14	15	0.003	0.01	0	0	0	0	0	0	1	-360	360;
15	16	0.002	0.01	0	0	0	0	0	0	1	-360	360;
16	17	0.005	0.02	0	0	0	0	0	0	1	-360	360;
17	18	0.009	0.03	0	0	0	0	0	0	1	-360	360;
18	19	0.012	0.05	0	0	0	0	0	0	1	-360	360;
19	20	0.006	0.02	0	0	0	0	0	0	1	-360	360;

EXECUTIVE UNGINEER (E) JESCO OPERATION DIVISION GUJAR KHAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm:

Variable		Before_DO	1	After_DG	
'Min Voltage (p.u)'		0.99953		11	
'Min Voltage at Bus Location'		'20'		.1.	
'Total Active Power Losses (kW) '		10.048605		0.0021558	
'Total Active Power Losses (%)'		'NA'		95.5647'	
'Total Reactive Power Losses (kVA	R) '	'0.16288'		0.0081778	0
'Total Reactive Power Losses (%) '		'NA'		94.9794	
'Power Factor'		10.85065		'0.99664'	
Additional Information related to DGs	:				
Variable			Valu	8	
'DG Type'	131				
'No of DGs'	131				
'Location of DGs (Bus Number)'	•2	8 17'			
'Sizes of DGs in kW'	130.5	5908	95.69	79 177	. 439
1 million and the second second		1	The States	1994	동안 가신하겠

4. 22 Bus 28 Branch Network System:

'Sizes of DGs in kVAR'

'MOI'

Input Parameters being used in network are shown below (Parameters are converted into P.U):

119.1476

10.031761

57.7806

121.1417'

						Busic	latte				interes not	
bus_i	type	Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	zone	Vmax	Vmin
1	3	0	0	0	0	1	1	0	11	1	1.06	0.94;
2	1	0	0	0	0	1	1	0	11	1	1.06	0.94;
3	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
4	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
5	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
6	1	0.0089	0.01	0	0	1	1	0	11	1	1.06	0.94;
7	1	0	0	0	0	1	1	0	11	1	1.06	0.94;
8	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
9	1	0.356	0.4	0	0	1	1	0	11	1	1.06	0.94;
10	1	0.178	0.2	0	0	1	1	0	11	1	1.06	0.94;
11	1	0	0	0	0	1	1	0	11	1	1.06	0.94;
12	1	0.178	0.2	0	0	1	1	0	11	1	1.06	0.94;
13	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;
14	1	0.0445	0.05	0	0	1	1	0	11	1	1.06	0.94;

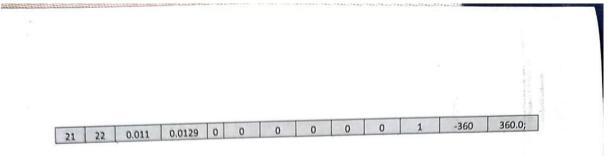


	Const Property and	In the second second					1	0	11	1	1.06	0.94;
15	1	0	0	0	0	1	1			1	1.06	0.94;
16	1	0.089	0.1	0	0	1	1	0	11	1	1.06	0.94;
17	1	0.0445	0.05	0	0	1	1	0	11	1	Concernation of the	and the second second
18	1	0.178	0.2	0	0	1	1	0	11	1	1.06	0.94;
	-		1 1 1 1 1 1 1	0	0	1	1	0	11	1	1.06	0.94;
19	1	0.089	0.1			-	1	0	11	1	1.06	0.94;
20	1	0.178	0.2	0	0	1	-		11	1	1.06	0.94;
21	1	0	0	0	0	1	1	0		1	1.06	0.94;
22	1	0.0445	0.05	0	0	1	1	0	11	1	1.00	0.54)

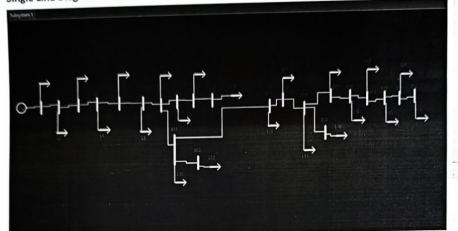
						1			-	
bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1
1	0	0	10	-10			1	10	0	0
PcZ	Qc1min	Qc1max Qc2min Qc2ma	Qc1max Qc2min Qc2max ramp_agc ram	Qc2min Qc2max	ramp_10	ramp_30	ramp_q	ар	f	
0	0	0	0	0	0	0	0	0	0.0);

ma	and the	- Internet				Branc	n data			and the second		
fbus	tbus	r	x	b	rateA	rateB	rateC	ratio	angle	status	angmin	angmax
1	2	0.011	0.0129	0	0	0	0	0	0	1	-360	360.0;
2	3	0.0047	0.0053	0	0	0	0	0	0	1	-360	360.0;
3	4	0.00091	0.001	0	0	0	0	0	0	1	-360	360.0;
4	5	0.0018	0.0021	0	0	0	0	0	0	1	-360	360.0;
5	6	0.0034	0.0039	0	0	0	0	0	0	1	-360	360.0;
6	7	0.0028	0.0031	0	0	0	0	0	0	1	-360	360.0;
7	8	0.00057	0.00063	0	0	0	0	0	0	1	-360	360.0;
8	9	0.012	0.0125	0	0	0	0	0	0	1	-360	360.0;
9	10	0.01	0.011	0	0	0	0	0	0	1	-360	360.0;
7	11	0.00042	0.0004	0	0	0	0	0	0	1	-360	360.0;
11	12	0.0057	0.0062	0	0	0	0	0	0	1	-360	360.0;
11	13	0.0036	0.0039	0	0	0	0	0	0	1	-360	360.0;
13	14	0.023	0.028	0	0	0	0	0	0	1	-360	360.0;
14	15	0.0039	0.0042	0	0	0	0	0	0	1	-360	360.0;
15	16	0.0159	0.006	0	0	0	0	0	0	1	-360	360.0;
15	17	0.002	0.033	0	0	0	0	0	0	1	-360	360.0;
17	18	0.0153	0.0173	0	0	0	0	0	0	1	-360	360.0;
18	19	0.033	0.013	0	0	0	0	0	0	1	-360	360.0;
19	20	0.0245	0.028	0	0	0	0	0	0	1	-360	360.0;
20	21	0.0153	0.0173	0	0	0	0	0	0	1	-360	360.0;





Single Line Diagram Of Network:



Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

mary of Results Comparison: Su

ummary of Results Comparison: Variable	Before_D	G After_DG	_
'Min Voltage (p.u)' 'Min Voltage at Bus Location' 'Total Active Power Losses (kW)' 'Total Active Power Losses (%)' 'Total Reactive Power Losses (kVAS 'Total Reactive Power Losses (%)' 'Power Factor'	'0.99209 '22' '7.6956' 'NA' '9.3969' 'NA' '0.66467	'22' '0.72232 '90.6139 '0.71291 '92.4133	
dditional Information related to DGs: Variable		Value	
'DG Type' 'No of DGs' 'Location of DGs (Bus Number)' 'Sizes of DGs in kW' 'Sizes of DGs in kVAR' 'MOI'	'3' '19 9 4' '490.7568 '636.5237 '0.063264'	472,1768 746.2824	414.5597

EXECUTIVE ENGINEER (E) GUJAR KHAN

÷.,				
~	÷			
		``	×	

	1	75	75	0	0	1	1	0	11	1	1.1	0.9;
16	1		- CONTRACTOR	0	0	1	1	0	11	1	1.1	0.9;
17	1	45	45			-	1	0	11	1	1.1	0.9;
18	1	70	70	0	0	1	-	and the second second	A DECEMBER OF THE PARTY OF THE	1	1.1	0.9;
19	1	60	60	0	0	1	1	0	11	- Comment		
20	1	65	65	0	0	1	1	0	11	1	1.1	0.9;
21	1	90	90	0	0	1	1	0	11	1	1.1	0.9;
22	1	95	95	0	0	1	1	0	11	1	1.1	0.9;
and the second second	1	120	120	0	0	1	1	0	11	1	1.1	0.9;
23		125	125	0	0	1	1	0	11	1	1.1	0.9;
24	1			0	0	1	1	0	11	1	1.1	0.9;
25	1	145	145					0	11	1	1.1	0.9;
26	1	160	160	0	0	1	1			1	1.1	0.9;
27	1	65	65	0	0	1	1	0	11	1	1.1	0.5,

					enereto: Det	ŧ1				
bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1
1	0	0	10	-10	1	100	1	10	0	0
Pc2	Qc1min	Qc1max	Qc2min	Qc2max	ramp_agc	ramp_10	ramp_30	ramp_q	ар	f
0	0	0	0	0	0	0	0	0	0	1-5

and the							Branch d	lata		and the second second		
fbus	tbus	r	x	b	rateA	rateB	rateC	ratio	angle	status	angmin	angmax
1	2	0.1	0.2	0	0	0	0	0	0	1	-360	360;
2	3	0.1	0.2	0	0	0	0	0	0	1	-360	360;
3	4	0.1	0.2	0	0	0	0	0	0	1	-360	360;
4	5	0.1	0.2	0	0	0	0	0	0	1	-360	360;
5	6	0.1	0.2	0	0	0	0	0	0	1	-360	360;
6	7	0.1	0.2	0	0	0	0	0	0	1	-360	360;
7	8	0.1	0.2	0	0	0	0	0	0	1	-360	360;
8	9	0.1	0.2	0	0	0	0	0	0	1	-360	360;
9	10	0.1	0.2	0	0	0	0	0	0	1	-360	360;
10	11	0.1	0.2	0	0	0	Ó	0	0	1	-360	360;
11	12	0.1	0.2	0	0	0	0	0	0	1	-360	360;
12	13	0.1	0.2	0	0	0	0	0	0	1	-360	360;
13	14	0.1	0.2	0	0	0	0	0	0	1	-360	360;
14	15	0.1	0.2	0	0	0	0	0	0	1	-360	360;
15	16	0.1	0.2	0	0	0	0	0	0	1	-360	360;
16	17	0.1	0.2	0	0	0	0	0	0	1	-360	360;
17	18	0.1	0.2	0	0	0	0	0	0	1	-360	360;
18	19	0.1	0.2	0	0	0	0	0	0	1	-360	360;
19	20	0.1	0.2	0	0	0	0	0	0	1	-360	360;
20	21	0.1	0.2	0	0	0	0	0	0	1	-360	360;

EXECUTIVE ENGINEER (E) ESCO OPERATION DIVISION GUJAR KHAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm:

()	1
'0.99209' '22' '7.6956' 'NA' '9.3969' 'NA' '0.66467'	'0.99961' '16' '0.7155' '90.7026' '0.66945' '92.8759' '0.97214'
	'22' '7.6956' 'NA' '9.3969' 'NA'

Variable

AND ADDED TO ADDED TO ADDR.

'DG Type'	131				
'No of DGs'	131				
'Location of DGs (Bus Number)'	'2 19	9'			
'Sizes of DGs in kW'	•1	576.30	071	741.2	049'
'Sizes of DGs in kVAR'	905.84	29	628.15	32	917.0452'
'MOI'	0.0553	95'			

5. 27 Bus 27 Branch Network System:

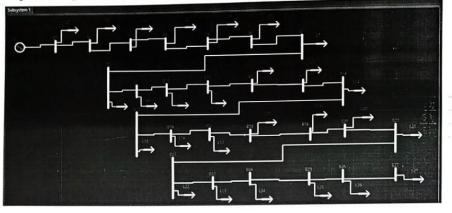
Parameters are entered in their own units as Pd and Qd are specified in kW & kVAr. Also r and x specified in ohms here. Code will Automatically Convert them into P.U before burning

bus_i	type	Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	zone	Vmax	Vmin
1	3	0	0	0	0	1	1	0	11	1	1	1;
2	1	50	50	0	0	1	1	0	11	1	1.1	0.9;
3	1	75	75	0	0	1	1	0	11	1	1.1	0.9;
4	1	120	120	0	0	1	1	0	11	1	1.1	0.9;
5	1	55	55	0	0	1	1	0	11	1	1.1	0.9;
6	1	130	130	0	0	1	1	0	11	1	1.1	0.9;
7	1	160	160	0	0	1	1	0	11	1	1.1	0.9;
8	1	95	95	0	0	1	1	0	11	1	1.1	0.9;
9	1	65	65	0	0	1	1	0	11	1	1.1	0.9;
10	1	80	80	0	0	1	1	0	11	1	1.1	0.9;
11	1	100	100	0	0	1	1	0	11	1	1.1	0.9;
12	1	95	95	0	0	1	1	0	11	1	1.1	0.9;
13	1	85	85	0	0	1	1	0	11	1	1.1	0.9;
14	1	110	110	0	0	1	1	0	11	1	1.1	0.9;
15	1	150	150	0	0	1	1	0	11	1	1.1	0.9;



									0	1	-360	360;
21	22	0.1	0.2	0	0	0	0	0	0	-	-360	360;
22	23	0.1	0.2	0	0	0	0	0	0	1		360;
23	24	0.1	0.2	0	0	0	0	0	0	1	-360	
-		-	and the second second				0	0	0	1	-360	360;
24	25	0.1	0.2	0	0	0	0	0	0	1	-360	360;
25	26	0.1	0.2	0	0	0	0	U	0	1	-360	360;
26	27	0.1	0.2	0	0	0	0	0	0	1	-500	0001

Single Line Diagram Of Network:



Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

Variable	Before_	DG After_DG	_
'Min Voltage (p.u)'	0.9062	5' '0.99971	
'Min Voltage at Bus Location'	'27'	. 6.	
'Total Active Power Losses (kW)'	'114.41	95' '4.6809'	
'Total Active Power Losses (%)'	'NA'	'95.909'	
'Total Reactive Power Losses (kV	AR)' '228.83	9.3618'	
'Total Reactive Power Losses (%)	' 'NA'	'95.909'	
'Power Factor'	10.6917	'0.99649	•
'DG Type'	'3' '3'		
'No of DGs'	'3'		
'Location of DGs (Bus Number)'	'11 23 2'		
'Sizes of DGs in kW'	1002.8568	1073.3716	2.5311801
'Sizes of DGs in kVAR'	1094.7992	1062.6175	1445.0493
'NOI'	'0.028685'		
		(A	
	E	XECUTIVE	GINEER (E)
		SCO OPERATIO	
	10		
		GUJAR K	LIAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm:

Summary of Results Comparison: Variable	1	Sefore_DG	After	DG
'Nin Voltage (p.u)' 'Nin Voltage at Bus Location' 'Total Active Power Losses (kW)' 'Total Active Power Losses (%)' 'Total Reactive Power Losses (kV 'Total Reactive Power Losses (%) 'Power Factor'	AR) '	0.90625' 27' 114.4195 NA' 228.8389 NA' 0.69172'	'16' '4.65 '95.9 '9.31 '95.9	54' 313' 09' 313'
Additional Information related to DG Variable	a:		Value	-
'DG Type' 'No of DGs' 'Location of DGs (Bus Number)' 'Sizes of DGs in KW' 'Sizes of DGs in kVAR' 'MOI'	'3' '2 11 '74.40; '1500 '0.028	999.	978.62644 19051	105 1116.876

IEEE Standard Networks:

1111

State State States

Following are the detailed parameters of IEEE Standard networks which are retrieved from Google for testing and INVESTIGATION OF OPTIMAL POWER FACTOR IN DISTRIBUTION NETWORKS BASED ON DG PLACEMENT.

1. IEEE 33 Bus Network System:

Parameters are entered in their own units as Pd and Qd are specified in kW & kVAr. Also r and x specified in ohms here. Code will Automatically Convert them into P.U before burning

		11- 11-								a second second		
bus i	type	Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	zone	Vmax	Vmin
1	3	0	0	0	0	1	1	0	12.66	1	1	1
2	1	100	60	0	0	1	1	0	12.66	1	1.1	0.9
3	1	90	40	0	0	1	1	0	12.66	1	1.1	0.9
4	1	120	80	0	0	1	1	0	12.66	1	1.1	0.9
5	1	60	30	0	0	1	1	0	12.66	1	1.1	0.9
6	1	60	20	0	0	1	1	0	12.66	1	1.1	0.9
2	1	200	100	0	0	1	1	0	12.66	1	1.1	0.9
7	1	200	100	0	0	1	1	0	12.66	1	1.1	0.9
8	1	60	20	0	0	1	1	0	12.66	1	1.1	0.9
	1	60	20	0	0	1	1	0	12.66	1	1.1	0.9
10	-	45	30	0	0	1	1	0	12.66	1	1.1	0.9
11	1	and the second	35	0	0	1	1	0	12.66	1	1.1	0.9
12	1	60	1.1.1	-		1	1	0	12.66	1	1.1	0.9
13	1	60	35	0	0	-		0	12.66	1	1.1	0.9
14	1	120	80	0	0	1	1	0	12.00	1	1.1	0.9

EXECUTIVE GINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

1053.8896'

1116.8764'

10	1	60	10	0	0	1	1	0	12.66	1	1.1	0.9
15		60	20	0	0	1	1	0	12.66	1	1.1	0.9
16	1	and the local division of the	20	0	0	1	1	0	12.66	1	1.1	0.9
17	1	60	40	0	0	1	1	0	12.66	1	1.1	0.9
18	1	90	40	0	0	1	1	0	12.66	1	1.1	0.9
19	1	90	and the second second	0	0	1	1	0	12.66	1	1.1	0.9
20	1	90	40	and the second	0		1	0	12.66	1	1.1	0.9
21	1	90	40	0	and the special designed in		1	0	12.66	1	1.1	0.9
22	1	90	40	0	0	1	1	0	12.66	1	1.1	0.9
23	1	90	50	0	0	1	and the second second	the barrene war	and the second se	1	1.1	0.9
24	1	420	200	0	0	1	1	0	12.66		and the second sec	0.9
25	1	420	200	0	0	1	1	0	12.66	1	1.1	
26	1	60	25	0	0	1	1	0	12.66	1	1.1	0.9
27	1	60	25	0	0	1	1	0	12.66	1	1.1	0.9
28	1	60	20	0	0	1	1	0	12.66	1	1.1	0.9
29	1	120	70	0	0	1	1	0	12.66	1	1.1	0.9
30	1	200	600	0	0	1	1	0	12.66	1	1.1	0.9
31	1	150	70	0	0	1	1	0	12.66	1	1.1	-0.9
32	1	210	100	0	0	1	1	0	12.66	1	1.1	0.9
33	1	60	40	0	0	1	1	0	12.66	1	1.1	0.9

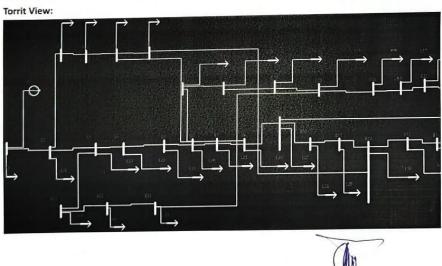
						tta -			in a subset	
Bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1
1	0	0	10	-10	1	100	1	10	0	0
Pc2	Qc1min	Qc1max	Qc2min	Qc2max	ramp_agc	ramp_10	ram	o_30	ramp_q	Apf
0	0	0	0	0	0	0	()	0	0

						Bran	ch data					
fbus	tbus	r	x	b	rateA	rateB	rateC	ratio	angle	status	angmin	angmax
1	2	0.0922	0.047	0	0	0	0	0	0	1	-360	360;
2	3	0.493	0.2511	0	0	0	0	0	0	1	-360	360;
3	4	0.366	0.1864	0	0	0	0	0	0	1	-360	360;
4	5	0.3811	0.1941	0	0	0	0	0	0	1	-360	360;
5	6	0.819	0.707	0	0	0	0	0	0	1	-360	360;
6	7	0.1872	0.6188	0	0	0	0	0	0	1	-360	360;
7	8	0.7114	0.2351	0	0	0	0	0	0	1	-360	360;
8	9	1.03	0.74	0	0	0	0	0	0	1	-360	360;
9	10	1.044	0.74	0	0	0	0	0	0	1	-360	360;
10	11	0.1966	0.065	0	0	0	0	0	0	1	-360	360;
11	12	0.3744	0.1238	0	0	0	0	0	0	1	-360	360;
12	13	1.468	1.155	0	Ó	0	0	0	Ó	1	-360	360;
13	14	0.5416	0.7129	0	0	0	0	0	0	1	-360	360;
14	15	0.591	0.526	0	0	0	0	0	0	1	-360	360;



			OFAF	0	0	0	0	0	0	1	-360	360;
15	16	0.7463	0.545	0		0	0	0	0	1	-360	360;
16	17	1.289	1.721	0	0	0	0	0	0	1	-360	360;
17	18	0.732	0.574	0	0		0	0	0	1	-360	360;
2	19	0.164	0.1565	0	0	0	0	0	0	1	-360	360;
19	20	1.5042	1.3554	0	0	0	0	0	0	1	-360	360;
20	21	0.4095	0.4784	0	0	0	0	0	0	1	-360	360;
21	22	0.7089	0.9373	0	0	0		0	0	1	-360	360;
3	23	0.4512	0.3083	0	0	0	0	0	0	1	-360	360;
23	24	0.898	0.7091	0	0	0	0	0	0	1	-360	360;
24	25	0.896	0.7011	0	0	0	0		0	1	-360	360;
6	26	0.203	0.1034	0	0	0	0	0	0	1	-360	360;
26	27	0.2842	0.1447	0	0	0	0	0		1	-360	360;
27	28	1.059	0.9337	0	0	0	0	0	0	1	-360	360;
28	29	0.8042	0.7006	0	0	0	0	0	0	1	-360	360;
29	30	0.5075	0.2585	0	0	0	0	0	0	1	-360	360;
30	31	0.9744	0.963	0	0	0	0	0	0	1	-360	360;
31	32	0.3105	0.3619	0	0	0	0	0	0		-360	360;
32	33	0.341	0.5302	0	0	0	0	0	0	1	-360	360;
21	8	2	2	0	0	0	0	0	0	0		360;
9	15	2	2	0	0	0	0	0	0	0	-360	
12	22	2	2	0	0	0	0	0	0	0	-360	360;
18	33	0.5	0.5	0	0	0	0	0	0	0	-360	360;
25	29	0.5	0.5	0	0	0	0	0	0	0	-360	360;

Single Line Diagram Of Network:



EXECUTIVE PNGINEER (E) IESCO OPERATION DIVISION GUJAK KROM

16	1	45.5	30	0	0	1	1	0	12.66	1	1.1	0.0.
17	1	60	35	0	0	1	1	0	12.66	1	1.1	0.9;
18	1	60	35	0	0	1	1	0	12.66	1	1.1	0.9;
19	1	0	0	0	0	1	1	0	12.66	1		0.9;
20	1	1	0.6	0	0	1	1	0	12.66	and the state of the	1.1	0.9;
21	1	114	81	0	0	1	1	0	12.66	1	1.1	0.9;
22	1	5.3	3.5	0	0	1	1	0	and the second se	1	1.1	0.9;
23	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
24	1	28	20	0	0	1	1	0	12.66	1	1.1	0.9;
25	1	0	0	0	0	1	1		12.66	1	1.1	0.9;
26	1	14	10	0	0	1	The section of the	0	12.66	1	1.1	0.9;
27	1	14	10	0		Part Internet	1	0	12.66	1	1.1	0.9;
28	1	26	18.6	1	0	1	1	0	12.66	1	1.1	0.9;
29	1			0	0	1	1	0	12.66	1	1.1	0.9;
30	1	26	18.6 0	0	0	1	1	0	12.66	1	1.1	0.9;
	1	-		0	0	1	1	0	12.66	1	1.1	0.9;
31	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
		-	0	0	0	1	1	0	12.66	1	1.1	0.9;
33	1	14	10	0	0	1	1	0	12.66	1	1.1	0.9;
34	1	19.5	14	0	0	1	1	0	12.66	1	1.1	0.9;
35	1	6	4	0	0	1	1	0	12.66	1	1.1	0.9;
36	1	26	18.6	0	0	1	1	0	12.66	1	1.1	0.9;
37	1	26	18.6	0	0	1	1	0	12.66	1	1.1	0.9;
38	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
39	1	24	17	0	0	1	1	0	12.66	1	1.1	0.9;
40	1	24	17	0	0	1	1	0	12.66	1	1.1	0.9;
41	1	1.2	1	0	0	1	1	0	12.66	1	1.1	0.9;
42	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
43	1	6	4.3	0	0	1	1	0	12.66	1	1.1	0.9;
44	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
45	1	39.2	26.3	0	0	1	1	0	12.66	1	1.1	0.9;
46	1	39.2	26.3	0	0	1	1	0	12.66	1	1.1	0.9;
47	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
48	1	79	56.4	0	0	1	1	0	12.66	1	1.1	0.9;
49	1	384.7	274.5	0	0	1	1	0	12.66	1	1.1	0.9;
50	1	384.7	274.5	0	0	1	1	0	12.66	1	1.1	0.9;
51	1	40.5	28.3	0	0	1	1	0	12.66	1	1.1	0.9;
52	1	3.6	2.7	0	0	1	1	0	12.66	1	1.1	0.9;
53	1	4.3	3.5	0	0	1	1	0	12.66	1	1.1	0.9;
54	1	26.4	19	0	0	1	1	0	12.66	1	1.1	0.9;
55	1	24	17.2	0	0	1	1	0	12.66	1	1.1	0.9;



56	1	0	0	0	0	1	1	0	12.66	1		
57	1	0	0	0	0	1	1	0	and the second se		1.1	0.9;
58	1	0	0	0	0	1			12.66	1	1.1	0.9;
59	1	100	72	0	0		1	0	12.66	1	1.1	0.9;
60	1	0	0	-		1	1	0	12.66	1	1.1	0.9;
61				0	0	1	1	0	12.66	1	1.1	0.9;
	1	1244	888	0	0	1	1	0	12.66	1	1.1	0.9;
62	1	32	23	0	0	1	1	0	12.66	1	1.1	0.9;
63	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
64	1	227	162	0	0	1	1	0	12.66	1	1.1	0.9;
65	1	59	42	0	0	1	1	0	12.66	1	1.1	0.9;
66	1	18	13	0	0	1	1	0	12.66	1	1.1	0.9;
67	1	18	13	0	0	1	1	0	12.66	1	1.1	0.9;
68	1	28	20	0	0	1	1	0	12.66	1	1.1	0.9;
69	1	28	20	0	0	1	1	0	12.66	1	1.1	0.9;

										-
_				G	enerator dat					
bus	Pg	Qg	Qmax	Qmin	Vg	mBase	status	Pmax	Pmin	Pc1
-	1	0	0	10	-10	1	100	1	10	0
PcZ	Qc1min	Qc1max	Qc2min	Qc2max	ramp_agc	ramp_10	ramp_30	ramp_q	ар	COMPANY OF THE OWNER
0	0	0	0	0	0	0	0	0	0	the second

						Bran	ch data					
fbus	tbus	r	x	b	rateA	rateB	rateC	ratio	angle	status	angmin	angmax
1	2	0.0005	0.0012	0	0	0	0	0	0	1	-360	360;
2	3	0.0005	0.0012	0	0	0	0	0	0	1	-360	360;
3	4	0.0015	0.0036	0	0	0	0	0	0	1	-360	360;
4	5	0.0251	0.0294	0	0	0	0	0	0	1	-360	360;
5	6	0.366	0.1864	0	0	0	0	0	0	1	-360	360;
6	7	0.381	0.1941	0	0	0	0	0	0	1	-360	360;
7	8	0.0922	0.047	0	0	0	0	0	0	1	-360	360;
8	9	0.0493	0.0251	0	0	0	0	0	0	1	-360	360;
9	10	0.819	0.2707	0	0	0	0	0	0	1	-360	360;
10	11	0.1872	0.0619	0	0	0	0	0	0	1	-360	360;
11	12	0.7114	0.2351	0	0	0	0	0	0	1	-360	360;
12	13	1.03	0.34	0	0	0	0	0	0	1	-360	360;
13	14	1.044	0.34	0	0	0	0	0	0	1	-360	360;
14	15	1.058	0.3496	0	0	0	0	0	0	1	-360	360;
15	16	0.1966	0.065	0	0	0	0	0	0	1	-360	360;
16	17	0.3744	0.1238	0	0	0	0	0	0	1	-360	360;

EXECUTIVE GINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm:

Variable	Before_DO	After_DG
'Min Voltage (p.u)'	'0.91309'	·0.99218·
'Min Voltage at Bus Location'	18'	.18.
'Total Active Power Losses (kW)'	202.6771'	15.0378
'Total Active Power Losses (%)'	'NA'	92.5804
'Total Reactive Power Losses (KVAR)'	'135.141'	'11.585'
'Total Reactive Power Losses (%)'	'NA'	91.4275
'Power Factor'	'0.8493'	'0.93222'
dditional Information related to DGs:		
Variable		Value

'DG Type'	131		
'No of DGs'	131		
'Location of DGs (Bus Number)'	'12 24 30'		
'Sizes of DGs in kW'	968.0763	925.546	888.4894'
'Sizes of DGs in kVAR'	1524.21426	777.92735	1073.2499'
'MOI'	'0.066105'		

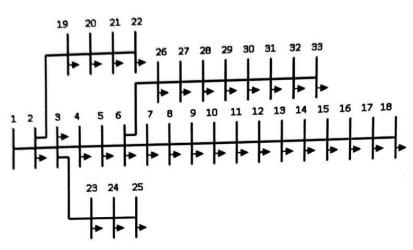
2. IEEE 69 Bus Network System:

Parameters are entered in their own units as Pd and Qd are specified in kW & kVAr. Also r and x specified in ohms here. Code will Automatically Convert them into P.U before burning

	HUS GENERAL											
bus_i	type	Pd	Qd	Gs	Bs	area	Vm	Va	baseKV	zone	Vmax	Vmin
1	3	0	0	0	0	1	1	0	12.66	1	1	1;
2	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
3	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
4	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
5	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;
6	1	2.6	2.2	0	0	1	1	0	12.66	1	1.1	0.9;
7	1	40.4	30	0	0	1	1	0	12.66	1	1.1	0.9;
8	1	75	54	0	0	1	1	0	12.66	1	1.1	0.9;
9	1	30	22	0	0	1	1	0	12.66	1	1.1	0.9;
10	1	28	19	0	0	1	1	0	12.66	1	1.1	0.9;
11	1	145	104	0	0	1	1	0	12.66	1	1.1	0.9;
12	1	145	104	0	0	1	1	0	12.66	1	1.1	0.9;
13	1	8	5.5	0	0	1	1	0	12.66	1	1.1	0.9;-
14	1	8	5.5	0	0	1	1	0	12.66	1	1.1	0.9;
15	1	0	0	0	0	1	1	0	12.66	1	1.1	0.9;



Standard View:



Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

Summary of Results Comparison: Variable	Before_DG	After_DG
<pre>'Min Voltage (p.u)' 'Min Voltage at Bus Location' 'Total Active Power Losses (kW)' 'Total Active Power Losses (%)' 'Total Reactive Power Losses (kVAR)' 'Total Reactive Power Losses (%)' 'Power Factor'</pre>	'0.91309' '18' '202.6771' 'NA' '135.141' 'NA' '0.8493'	'0.9922' '18' '18.3194' '90.9613' '13.8962' '89.7173' '0.94739'

Additional Information related to DGs: Variable

'DG Type'	131
'No of DGs'	131
'Location of DGs (Bus Number)'	30 12 24
'Sizes of DGs in kW'	1033.1426
'Sizes of DGs in kVAR'	'840.551
'MOI'	'0.070631'

12 24' .1426 876.

876.16707 756.45893' 675.122 963,7898'

Value

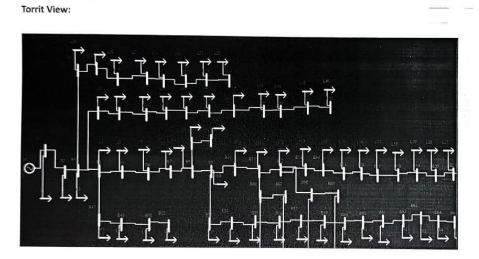
EXECUTIVE ANGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

											250	260
17	18	0.0047	0.0016	0	0	0	0	0	0	1	-360	360;
18	19	0.3276	0.1083	0	0	0	0	0	0	1	-360	360;
19	20	0.2106	0.069	0	0	0	0	0	0	1	-360	360;
20	21	0.3416	0.1129	0	0	0	0	0	0	1	-360	360;
21	22	0.014	0.0046	0	0	0	0	0	0	1	-360	360;
22	23	0.1591	0.0526	0	0	0	0	0	0	1	-360	360;
23	24	0.3463	0.1145	0	0	0	0	0	0	1	-360	360;
24	25	0.7488	0.2475	0	0	0	0	0	0	1	-360	360;
25	26	0.3089	0.1021	0	0	0	0	0	0	1	-360	360;
26	27	0.1732	0.0572	0	0	0	0	0	0	1	-360	360;
3	28	0.0044	0.0108	0	0	0	0	0	0	1	-360	360;
28	29	0.064	0.1565	0	0	0	0	0	0	1	-360	360;
1010	-	0.3978	0.1305	0	0	0	0	0	0	1	-360	360;
29	30	0.3978	0.0232	0	0	0	0	0	0	1	-360	360;
30	31	0.351	0.116	0	0	0	0	0	0	1	-360	360;
31		Y- Charles	0.2816	0	0	0	0	0	0	1	-360	360;
32	33	0.839	0.2816	0	0	0	0	0	0	1	-360	360;
33	34	1.708		-	0	0	0	0	0	1	-360	360;
34	35	1.474	0.4873	0		0	0	0	0	1	-360	360;
3	36	0.0044	0.0108	0	0		0	0	0	1	-360	360;
36	37	0.064	0.1565	0	0	0	0	0	0	1	-360	360;
37	38	0.1053	0.123	0	-	0	0	0	0	1	-360	360;
38	39	0.0304	0.0355	0	0	0	0	0	0	1	-360	360;
39	40	0.0018	0.0021	0	0	0	0	0	0	1	-360	360;
40	41	0.7283	0.8509	0	0	0	0	0	0	1	-360	360;
41	42	0.31	0.3623	0	0	0	0	0	0	1	-360	360;
42	43	0.0092	0.0116	0	0	0	0	0	0	1	-360	360;
43 44	44	0.1089	0.1373	0	0	0	0	0	0	1	-360	360;
44	45	0.0009	0.0012	0	0	0	0	0	0	1	-360	360;
45	40	0.0034	0.0012	0	0	0	0	0	0	1	-360	360;
47	48	0.0851	0.2083	0	0	0	0	0	0	1	-360	360;
48	49	0.2898	0.7091	0	0	0	0	0	0	1	-360	360;
49	50	0.0822	0.2011	0	0	0	0	0	0	1	-360	360;
8	51	0.0928	0.0473	0	0	0	0	0	0	1	-360	360;
51	52	0.3319	0.114	0	0	0	0	0	0	1	-360	360;
9	53	0.174	0.0886	0	0	0	0	0	0	1	-360	360;
53	54	0.203	0.1034	0	0	0	0	0	0	1	-360	360;
54	55	0.2842	0.1447	0	0	0	0	0	0	1	-360	360;
55	56	0.2813	0.1433	0	0	0	0	0	0	1	-360	360;
56	57	1.59	0.5337	0	0	0	0	0	0	1	-360	360;



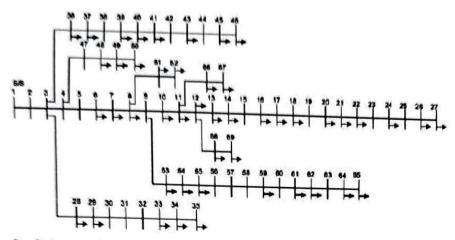
57	58	0.7837	0.263	0	0	0	0	0	0	1	200	260
58	59	0.3042	0,1006	0	0	0	0	0		-	-360	360;
59	60	0.3861	0.1172	0	0	0	0	0	0	1	-360	360;
60	61	0.5075	0.2585	0	0	0	0	0	0	1	-360	360;
61	62	0.0974	0.0496	0	0	0	0	-	0	1	-360	360;
62	63	0.145	0.0738	0	0	0	0	0	0	1	-360	360;
63	64	0.7105	0.3619	0	0	0		0	0	1	-360	360;
64	65	1.041	Contract Contractory of the second	0	and the second second	the second s	0	0	0	1	-360	360;
			0.5302		0	0	0	0	0	1	-360	360;
11	66	0.2012	0.0611	0	0	0	0	0	0	1	-360	360;
66	67	0.0047	0.0014	0	0	0	0	0	0	1	-360	360;
12	68	0.7394	0.2444	0	0	0	0	0	0	1	-360	360;
68	69	0.0047	0.0016	0	0	0	0	0	0	1	-360	360:

Single Line Diagram Of Network:



Standard View:

EXECUTIVE ENGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN



Results Summary Comparison Before & After D.G Placement:

1. Results obtained from SEA HORSE OPTIMIZATION Algorithm:

Summary of Results Comparison:

Variable	Before_DG	After_DG
'Min Voltage (p.u)' 'Min Voltage at Bus Location'	'0.99999' '65'	·1·
'Total Active Power Losses (kW)'	'1.1948e-05'	'27' '7.5641e-07'
'Total Active Power Losses (%)'	'NA'	'93.6691'
'Total Reactive Power Losses (kVAR)'	'5.4776e-06'	'5.8448e-07'
'Total Reactive Power Losses (%)'	'NA'	'89.3296'
'Power Factor'	'0.81587'	'0.99982'

Value

Additional Information related to DGs: Variable

	4
'DG Type'	131
'No of DGs'	'3'
'Location of DGs (Bus Number)'	'61 2 11'
'Sizes of DGs in kW'	'1.5721 1 1'
'Sizes of DGs in kVAR'	'1.1781 1.8237 1'
'MOI'	'0.04983'

EXECUTIVE ENGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

2. Results obtained from ARTIFICIAL BEE COLONY Algorithm:

After_DG	Before_DO	Variable
'1' '27'	'0.99999' '65'	'Min Voltage (p.u)' 'Min Voltage at Bus Location'
contraction of the second s	'1.1948e-05' 'NA'	'Total Active Power Losses (KW)' 'Total Active Power Losses (%)'
877 H-100 R-72 - 000 R-74	'5.4776e-06'	'Total Reactive Power Losses (kVAR)' 'Total Reactive Power Losses (%)'
'0.95267'	'0.81587'	'Power Factor'
		Additional Information related to DGs:
e	Valu	Variable

·	2
'DG Type'	'3'
'No of DGs'	131
'Location of DGs (Bus Number)'	'54 47 62'
'Sizes of DGs in kW'	'1 1.0002 1.3301'
'Sizes of DGs in kVAR'	'1 1.2848 1'
'MOI'	'0.086161'

Parameters Description:

1. Bus Data Units and Format:

1	bus number (positive integer)
2	bus type
	PQ bus = 1
	PV bus = 2
	reference bus = 3
	isolated bus = 4
3	Pd, real power demand (MW)
4	Qd, reactive power demand (MVAr)
5	Gs, shunt conductance (MW demanded at V = 1.0 p.u.)
6	Bs, shunt susceptance (MVAr injected at V = 1.0 p.u.)
7	area number, (positive integer)
8	Vm, voltage magnitude (p.u.)
9	Va, voltage angle (degrees)

EXECUTIVE TO INEER (E) GUJAR KHAN

(-) (bus name)	and the second second second
10 baseKV, base voltage (kV)	Mark Mary Mark
11 zone, loss zone (positive integer)	Course and the second
(+) 12 maxVm, maximum voltage magnitude (p.u.)	and the second second
(+) 13 minVm, minimum voltage magnitude (p.u.	

2. Generator Data Units and Format:

1	bus number
(-)	(machine identifier, 0-9, A-Z)
2	Pg, real power output (MW)
3	Qg, reactive power output (MVAr)
4	Qmax, maximum reactive power output (MVAr)
5	Qmin, minimum reactive power output (MVAr)
6	Vg, voltage magnitude setpoint (p.u.)
(-)	(remote controlled bus index)
7	mBase, total MVA base of this machine, defaults to baseMVA
(-)	(machine impedance, p.u. on mBase)
(-)	(step up transformer impedance, p.u. on mBase)
(-)	(step up transformer off nominal turns ratio)
8	status, > 0 - machine in service
	<= 0 - machine out of service
(-)	(% of total VAr's to come from this gen in order to hold V at
	remote bus controlled by several generators)
9	Pmax, maximum real power output (MW)
1	0 Pmin, minimum real power output (MW)
(2)	11 Pc1, lower real power output of PQ capability curve (MW)
(2)	12 Pc2, upper real power output of PQ capability curve (MW)
(2)	13 Qc1min, minimum reactive power output at Pc1 (MVAr)
(2)	14 Qc1max, maximum reactive power output at Pc1 (MVAr)
(2)	15 Qc2min, minimum reactive power output at Pc2 (MVAr)
(2)	16 Qc2max, maximum reactive power output at Pc2 (MVAr)
(2)	17 ramp rate for load following/AGC (MW/min)
(2)	18 ramp rate for 10 minute reserves (MW)
(2)	19 ramp rate for 30 minute reserves (MW)
(2)	20 ramp rate for reactive power (2 sec timescale) (MVAr/min)
(2)	21 APF, area participation factor

3. Branch Data Units and Format:

1	f, from bus number	
2	t, to bus number	

EXECUTIVE MIGINEER (E) IESCO OPERATION DIVISION GUJAR KHAN

(-)	(circuit identifier)
3	r, resistance (p.u.)
4	X, reactance (p.u.)
5	b, total line charging susceptance (p.u.)
6	rateA, MVA rating A (long term rating)
7	rateB, MVA rating B (short term rating)
8	rateC, MVA rating C (emergency rating)
9	ratio, transformer off nominal turns ratio (= 0 for lines)
	(taps at 'from' bus, impedance at 'to' bus,
	i.e. if r = x = 0, then ratio = Vf / Vt)
1	0 angle, transformer phase shift angle (degrees), positive => delay
(-)	(Gf, shunt conductance at from bus p.u.)
(-)	(Bf, shunt susceptance at from bus p.u.)
(-)	(Gt, shunt conductance at to bus p.u.)
(-)	(Bt, shunt susceptance at to bus p.u.)
1	1 initial branch status, 1 - in service, 0 - out of service
(2)	12 minimum angle difference, angle(Vf) - angle(Vt) (degrees)
(2)	13 maximum angle difference, angle(Vf) - angle(Vt) (degrees)
	(The voltage angle difference is taken to be unbounded below
	if ANGMIN < -360 and unbounded above if ANGMAX > 360.
	If both parameters are zero, it is unconstrained.)

EXECUTIVE CHIGINEER (E) ESCO OPERATION DIVISION GUJAR KHAN

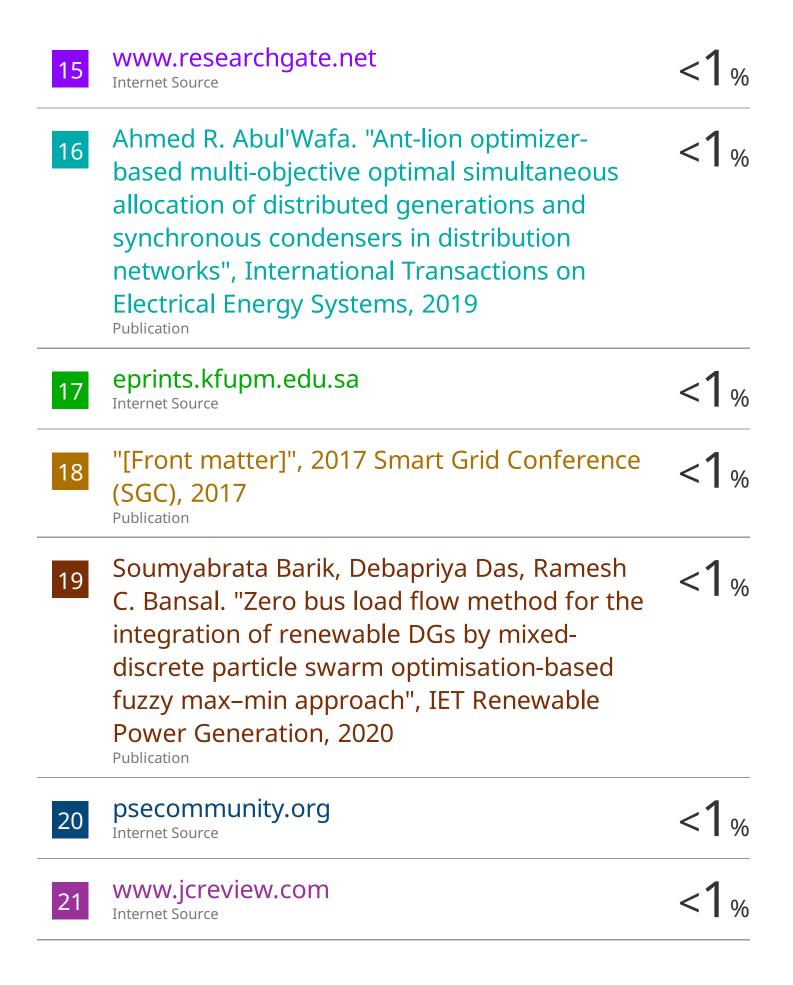
and a second

01-2	244222-01	2			
ORIGIN	ALITY REPORT				
1 SIMIL	2% ARITY INDEX	7% INTERNET SOURCES	10% PUBLICATIONS	3% STUDENT PAR	PERS
PRIMAR	RY SOURCES				
1	Placeme	bjective Optimi ent of Multiple D e Loads", Energ	Gs for Voltag		1%
2	Submitte Pakistan Student Paper		ucation Com	mission	1%
3	"Intellige Wiley, 20 Publication	ent Renewable I 022	Energy Syster	ms",	1%
4	prr.hec.g				1%
5	Waqar, N Haq, Mu "ABC alg placeme consider	A. Al-Ammar, Kira Juhammad Aar Jorithm based o ent of DGs in dis ring multiple ob ring Journal, 202	nir, Saifullah, , Memoona E ptimal sizing tribution netv jectives", Ain	Azhar Ul Batool. and works	<1%



6

7	Omid Khoubseresht, Heidarali Shayanfar. "The role of demand response in optimal sizing and siting of distribution energy resources in distribution network with time- varying load: An analytical approach", Electric Power Systems Research, 2020 Publication	< 1 %
8	"Optimization of Power System Problems", Springer Science and Business Media LLC, 2020 Publication	<1%
9	downloads.hindawi.com Internet Source	<1%
10	www.mdpi.com Internet Source	<1%
11	hdl.handle.net Internet Source	<1%
12	mobt3ath.com Internet Source	<1%
13	www.hindawi.com Internet Source	<1%
14	elibrary.tucl.edu.np Internet Source	<1%



22 Snigdha R. Behera, B. K. Panigrahi. "A multi objective approach for placement of multiple DGs in the radial distribution system", International Journal of Machine Learning and Cybernetics, 2018 Publication

<1%

23	Crpase.com Internet Source	<1 %
24	D. Sudha Rani, N. Subrahmanyam, M. Sydulu. "Self Adaptive Harmony Search algorithm for Optimal Network Reconfiguration", 2014 Power and Energy Conference at Illinois (PECI), 2014 Publication	< 1 %
25	Ramesh Bansal. "Power System Protection in Smart Grid Environment", CRC Press, 2019 Publication	<1 %
26	Ravi Shankar Pandey, S. R. Awasthi. "A Multi- objective Hybrid Algorithm for Optimal Planning of Distributed Generation", Arabian Journal for Science and Engineering, 2020 Publication	<1%
27	"Handbook of Distributed Generation", Springer Science and Business Media LLC, 2017	<1%

Publication

28	Fathabadi, Hassan. "Power distribution network reconfiguration for power loss minimization using novel dynamic fuzzy c- means (dFCM) clustering based ANN approach", International Journal of Electrical Power & Energy Systems, 2016. Publication	<1%
29	Kayode E. Adetunji, Ivan Hofsajer, Adnan M. Abu-Mahfouz, Ling Cheng. "A Review of Metaheuristic Techniques for Optimal Integration of Electrical Units in Distribution Networks", IEEE Access, 2020 Publication	<1%
30	Submitted to Universiti Putra Malaysia Student Paper	<1%
31	journals.tubitak.gov.tr Internet Source	<1%
32	"Applied Computer Sciences in Engineering", Springer Science and Business Media LLC, 2019 Publication	<1%
33	coek.info Internet Source	<1%
34	link.springer.com	<1%
35	mars.gmu.edu Internet Source	

		<1%
36	Submitted to University of Wales Institute, Cardiff Student Paper	<1%
37	www.tandfonline.com	<1%
38	Almoataz Y. Abdelaziz, Yasser G. Hegazy, Walid El-Khattam, Mahmoud M. Othman. "Optimal Planning of Distributed Generators in Distribution Networks Using Modified Firefly Method", Electric Power Components and Systems, 2015 Publication	<1%
39	Submitted to Singapore Institute of Technology Student Paper	<1%
40	escholarship.mcgill.ca Internet Source	<1%
41	WWW.NOVINT.COM Internet Source	<1%
42	Muhammad Usama, Hazlie Mokhlis, Nurulafiqah Nadzirah Mansor, Mahmoud Moghavvemi et al. "A multi-objective optimization of FCL and DOCR settings to mitigate distributed generations impacts on	<1 %

distribution networks", International Journal of Electrical Power & Energy Systems, 2023 Publication

43 Sachin Kumar, Kumari Sarita, Akanksha Singh S Vardhan, Rajvikram Madurai Elavarasan, R. K. Saket, Narottam Das. "Reliability Assessment of Wind-Solar Pv Integrated Distribution System Using Electrical Loss Minimization Technique", Energies, 2020

Publication

Internet Source

44

psasir.upm.edu.my

<1% <1%

- T. Yuvaraj, K. Ravi, K.R. Devabalaji.
 "DSTATCOM allocation in distribution networks considering load variations using bat algorithm", Ain Shams Engineering Journal, 2015 Publication
- 46 Submitted to VIT University Student Paper <1 %
- Bader N. Alajmi, M. F. AlHajri, Nabil A. Ahmed, Ibrahim Abdelsalam, Mostafa I. Marei. "Multiobjective Optimization of Optimal Placement and Sizing of Distributed Generators in Distribution Networks", IEEJ Transactions on Electrical and Electronic Engineering, 2023 Publication

 Bazilah Ismail, Noor Izzri Abd Wahab, Mohammad Lutfi Othman, Mohd Amran Mohd Radzi et al. "A Comprehensive Review on Optimal Location and Sizing of Reactive Power Compensation Using Hybrid-Based Approaches for Power Loss Reduction, Voltage Stability Improvement, Voltage Profile Enhancement and Loadability Enhancement", IEEE Access, 2020 Publication

Isaac Ortega-Romero, Xavier Serrano-Guerrero, Antonio Barragán-Escandón, Chistopher Ochoa-Malhaber. "Optimal Integration of Distributed Generation in Long Medium-Voltage Electrical Networks", Energy Reports, 2023 Publication

50 Sundar, S.. "A swarm intelligence approach to the quadratic minimum spanning tree problem", Information Sciences, 20100901 Publication

51 Submitted to Symbiosis International <1% University Student Paper

52 Yuvaraj, T., K.R. Devabalaji, and K. Ravi. "Optimal Placement and Sizing of DSTATCOM Using Harmony Search Algorithm", Energy Procedia, 2015. <1%

<1 %

53	pp.bme.hu Internet Source	<1%
54	rkala.in Internet Source	<1%
55	www.slideshare.net	<1%
56	"Advances in Electrical Control and Signal Systems", Springer Science and Business Media LLC, 2020 Publication	<1%
57	A. Bayat, A. Bagheri, R. Noroozian. "Optimal siting and sizing of distributed generation accompanied by reconfiguration of distribution networks for maximum loss reduction by using a new UVDA-based heuristic method", International Journal of Electrical Power & Energy Systems, 2016 Publication	<1 %
58	Mohamed Saad Suliman, Hashim Hizam, Mohammad Lutfi Othman. "Determining penetration limit of central PVDG topology considering the stochastic behaviour of PV generation and loads to reduce power losses	< 1 %

generation and loads to reduce power losses and improve voltage profiles", IET Renewable Power Generation, 2020

59	Mohammed Kdair Abd, S. J. Cheng, H. S. Sun. "Optimal DG placement and sizing for power loss reduction in a radial distribution system using MPGSA and sensitivity index method", 2016 IEEE 11th Conference on Industrial Electronics and Applications (ICIEA), 2016 Publication	<1%
60	Om Prakash Mahela, Baseem Khan, Puneet Kumar Jain. "Emerging Electrical and Computer Technologies for Smart Cities - Modelling, Solution Techniques and Applications", CRC Press, 2024 Publication	<1%
61	Shrikaant Kulkarni, Jaiprakash Narain Dwivedi, Dinda Pramanta, Yuichiro Tanaka. "Edge Computational Intelligence for AI-Enabled IoT Systems", CRC Press, 2024 Publication	< 1 %
62	c.coek.info Internet Source	<1%
63	ijeecs.iaescore.com Internet Source	<1%
64	www.coursehero.com Internet Source	<1%
65	www.networktechinc.com	<1%



71 Devabalaji, K.R., A. Mohamed Imran, T. Yuvaraj, and K. Ravi. "Power Loss

Minimization in Radial Distribution System", Energy Procedia, 2015.

Publication

Florina Rotaru, Gianfranco Chicco, Gheorghe Grigoras, Gheorghe Cartina. "Two-stage distributed generation optimal sizing with clustering-based node selection", International Journal of Electrical Power & Energy Systems, 2012 Publication

<1%

- Hanning Chen, Yunlong Zhu, Kunyuan Hu,
 Xiaoxian He. "Hierarchical Swarm Model: A
 New Approach to Optimization", Discrete
 Dynamics in Nature and Society, 2010
 Publication
- 74 Hsieh, T.J.. "Forecasting stock markets using wavelet transforms and recurrent neural networks: An integrated system based on artificial bee colony algorithm", Applied Soft Computing Journal, 201103 Publication
- N. D. Hatziargyriou, A. G. Anastasiadis, A. G. Tsikalakis, J. Vasiljevska. "Quantification of economic, environmental and operational benefits due to significant penetration of Microgrids in a typical LV and MV Greek network", European Transactions on Electrical Power, 2011 Publication

76	Nursyarizal Mohd Nor, Abid Ali, Taib Ibrahim,	<1
70	Nursyarizal Mohd Nor, Abid Ali, Taib Ibrahim, Mohd Fakhizan Romlie. "Chapter 4 Planning	
	of Distributed Renewable Energy Resources	
	Using Genetic Algorithm", Springer Science	
	and Business Media LLC, 2018	
	Publication	

%

<1 %

- Prabhat Kumar Vidyarthi, Ashiwani Kumar, Ravi Shankar. "Improvement of AGC for Multi-Area Deregulated Power Systems with Integrated RESs using a modified TID.", 2023 IEEE 3rd International Conference on Smart Technologies for Power, Energy and Control (STPEC), 2023 Publication
- Preetham Goli, Suresh Makkena, Srinivasa
 Rao Gampa, Debapriya Das. "Fuzzy Ant
 Colony Optimization Technique for Predefined
 Performance of Distribution Systems
 Considering DGs and Shunt Capacitors", 2019
 North American Power Symposium (NAPS),
 2019
 Publication
- TianNing Zhu, Yan Liu, JianHui Li, WanRu
 Zhao. "Linear Array Synthesis Using Sea-horse
 Optimization Algorithm", 2023 IEEE 13th
 International Conference on Electronics
 Information and Emergency Communication
 (ICEIEC), 2023

80	idr.mnit.ac.in Internet Source	<1%
81	ijeei.org Internet Source	<1%
82	mospace.umsystem.edu Internet Source	<1%
83	ujcontent.uj.ac.za Internet Source	<1%
84	www.diva-portal.org	<1%
85	www.engineeringletters.com	<1%
86	"Optimization in the Energy Industry", Springer Science and Business Media LLC, 2009 Publication	<1%
87	Adel A. Abou El-Ela, Sohir M. Allam, Nermine K. Shehata. "Optimal Allocation of a Hybrid Wind Energy-Fuel Cell System Using Different Optimization Techniques in the Equation	< 1 %

Optimization Techniques in the Egyptian Distribution Network", Energy and Power Engineering, 2021 Publication

<1 %

Angela Amphawan. "Dragonfly Algorithm and
Its Hybrids: A Survey on Performance,
Objectives and Applications", Sensors, 2021
Publication

- 89 Konrad Diwold, Madeleine Beekman, Martin Middendorf. "Chapter 13 Honeybee Optimisation – An Overview and a New Bee Inspired Optimisation Scheme", Springer Science and Business Media LLC, 2011 Publication
- 90 Mariesa L. Crow. "Computational Methods for Electric Power Systems", CRC Press, 2019 Publication <1%
- 91 Md. Shadman Abid, Hasan Jamil Apon, Khandaker Adil Morshed, Ashik Ahmed. "Optimal Planning of Multiple Renewable Energy-Integrated Distribution System With Uncertainties Using Artificial Hummingbird Algorithm", IEEE Access, 2022 Publication
- Mohamad Khairuzzaman Mohamad Zamani, Ismail Musirin, Saiful Izwan Suliman.
 "Symbiotic Organisms Search Technique for SVC Installation in Voltage Control", Indonesian Journal of Electrical Engineering and Computer Science, 2017 Publication
- <1%

<1 %

- Mohammadi, M., and Mehdi Nafar. "Optimal placement of multitypes DG as independent private sector under pool/hybrid power market using GA-based Tabu Search method", International Journal of Electrical Power & Energy Systems, 2013.
- Nurul Idayu Yusoff, Abdullah Asuhaimi Mohd Zin, Azhar Bin Khairuddin. "Congestion management in power system: A review", 2017 3rd International Conference on Power Generation Systems and Renewable Energy Technologies (PGSRET), 2017 Publication

<1%

<1 %

Oktoviano Gandhi. "Reactive Power Support Using Photovoltaic Systems", Springer Science and Business Media LLC, 2021 Publication

95

- Priyanka Maurya, Prabhakar Tiwari, Arvind Pratap. "Electric eel foraging optimization algorithm for distribution network reconfiguration with distributed generation for power system performance enhancement considerations different load models", Computers and Electrical Engineering, 2024 Publication
- 97 S. Gopiya Naik, D.K. Khatod, M.P. Sharma. "Optimal allocation of combined DG and

capacitor for real	power	loss minimiz	ation in
distribution netwo	orks", Ir	nternational _	Journal
of Electrical Powe	r & Ene	rgy Systems	, 2013
Publication			

98

Sangeeta Das, Debapriya Das, Amit Patra. "Distribution network reconfiguration using distributed generation unit considering variations of load", 2016 IEEE 1st International Conference on Power Electronics, Intelligent Control and Energy Systems (ICPEICES), 2016 Publication

Srinivasa Rao Gampa, D. Das. "Optimum 21%
placement of shunt capacitors in a radial distribution system for substation power factor improvement using fuzzy GA method", International Journal of Electrical Power & Energy Systems, 2016
Publication

100	article.nadiapub.com Internet Source	<1%
101	espace.library.uq.edu.au	<1%
102	i-scholar.in Internet Source	<1%
103	openscholar.dut.ac.za Internet Source	<1%



105 "High Performance Computing in Biomimetics", Springer Science and Business Media LLC, 2024 Publication

- Dharageshwari, K, and C Nayanatara. "Multiobjective optimal placement of multiple distributed generations in IEEE 33 bus radial system using simulated annealing", 2015 International Conference on Circuits Power and Computing Technologies [ICCPCT-2015], 2015. Publication
- Jaser A. Sa'ed, Mohammad Amer, Ahmed Bodair, Ahmad Baransi, Salvatore Favuzza, Gaetano Zizzo. "A Simplified Analytical Approach for Optimal Planning of Distributed Generation in Electrical Distribution Networks", Applied Sciences, 2019 Publication
- Mlungisi Ntombela, Kabeya Musasa, Moketjema Clarence Leoaneka. "Power Loss Minimization and Voltage Profile Improvement by System Reconfiguration, DG Sizing, and Placement", Computation, 2022

<1%

<1 %

Sara Ashfaq, Daming Zhang, Zhao Yang Dong. "Multi-objective Power Optimization of Microgrid with Distributed Generation", TENCON 2018 - 2018 IEEE Region 10 Conference, 2018 Publication

110	Shenxi Zhang, Haozhong Cheng, Ke Li, Nengling Tai, Dan Wang, Furong Li. "Multi- objective distributed generation planning in distribution network considering correlations among uncertainties", Applied Energy, 2018 Publication	<1%

111 V V S N Murty, Ashwani Kumar. "Optimal DG integration and network reconfiguration in microgrid systemwith realistic time varying load model using hybrid optimization", IET Smart Grid, 2019 Publication

<1 %

 "Applications of Artificial Intelligence Techniques in Engineering", Springer Science and Business Media LLC, 2019 Publication

113	"Electric Distribution Network Planning", Springer Science and Business Media LLC,	<1%
	2018 Publication	

- A. Forooghi Nematollahi, A. Dadkhah, O. Asgari Gashteroodkhani, B. Vahidi. "Optimal sizing and siting of DGs for loss reduction using an iterative-analytical method", Journal of Renewable and Sustainable Energy, 2016 Publication
- Agnihotri, Ganga, Manisha Dubey, and AASHISH BOHRE. "Optimal Sizing and Sitting of DG with Load Models using Soft Computing Techniques in Practical Distribution System", IET Generation Transmission & Distribution, 2016. Publication
- Ashwani Kumar, P. Vijay Babu, V. V. S. N. Murty. "Distributed Generators Allocation in Radial Distribution Systems with Load Growth using Loss Sensitivity Approach", Journal of The Institution of Engineers (India): Series B, 2016 Publication
- 117 Ashwin Raiyani, Sheetal Pandya. "chapter 15 Development Environment, Tools, and SDKs for Serverless Computing", IGI Global, 2024 Publication
- G. Srinivasan, S. Visalakshi. "Application of AGPSO for Power loss minimization in Radial Distribution Network via DG units, Capacitors and NR", Energy Procedia, 2017

<1%

<1%

119	Hossein Shayeghi, Masoud Alilou. "Distributed generation and microgrids", Elsevier BV, 2021 Publication	<1%
120	Lecture Notes in Electrical Engineering, 2015. Publication	<1%
121	Moshood Akanni Alao, Olawale Mohammed Popoola, Temitope Raphael Ayodele. "An improved particle swarm optimisation algorithm for optimum placement and sizing of biogas-fuelled distributed generators for seasonal loads in a radial distribution network", Energy Reports, 2024 Publication	<1%
122	Natarajan, . "Power Flow Analysis", Power Engineering (Willis), 2002. Publication	< 1 %
<mark>123</mark>	Shahzad, Mohsin, Ikram Ullah, Peter Palensky, and Wolfgang Gawlik. "Analytical approach for simultaneous optimal sizing and placement of multiple Distributed Generators in primary distribution networks", 2014 IEEE 23rd International Symposium on Industrial Electronics (ISIE), 2014. Publication	<1%

124	Sultana, U., Azhar B. Khairuddin, A.S. Mokhtar, N. Zareen, and Beenish Sultana. "Grey wolf optimizer based placement and sizing of multiple distributed generation in the distribution system", Energy, 2016. Publication	< 1 %
125	Thai Dinh Pham, Thang Trung Nguyen, Bach Hoang Dinh. "Find optimal capacity and location of distributed generation units in radial distribution networks by using enhanced coyote optimization algorithm", Neural Computing and Applications, 2020 Publication	<1%

Exclude quotes	On	Exclude matches	Off
Exclude bibliography	On		