

Received February 20, 2020, accepted March 10, 2020, date of publication March 26, 2020, date of current version April 17, 2020. *Digital Object Identifier* 10.1109/ACCESS.2020.2982988

Optimal Reactive Power Dispatch Using Chaotic Bat Algorithm

SYLVERE MUGEMANYI^{®1}, ZHAOYANG QU^{®1,2}, FRANÇOIS XAVIER RUGEMA³, YUNCHANG DONG^{®1}, CHRISTOPHE BANANEZA⁴, AND LEI WANG^{®1}

¹College of Information Engineering, Northeast Electric Power University, Jilin City 132012, China
²Jilin Engineering Technology Research Center of Intelligent Electric Power Big Data Processing, Jilin City 132012, China
³School of Electrical Engineering, Northeast Electric Power University, Jilin City 132012, China

⁴College of Energy and Electrical Engineering, Hohai University, Nanjing 210098, China

Corresponding author: Zhaoyang Qu (qzywww@neepu.edu.cn)

This work was supported in part by the Key Projects of National Natural Science Foundation of China under Grant 51437003, and in part by the Jilin Science and Technology Development Plan Project of China under Grant 2016062004T and Grant 20180201092GX.

ABSTRACT In this paper, the chaotic bat algorithm (CBA) is applied to solve the optimal reactive power dispatch (ORPD) problem taking into account small-scale, medium-scale and large-scale power systems. ORPD plays a key role in the power system operation and control. The ORPD problem is formulated as a mixed integer nonlinear programming problem, comprising both continuous and discrete control variables. The most outstanding benefit of the bat algorithm (BA) is its good convergence for optimal solutions. The BA, however, together with other metaheuristics, often gets stuck into local optima and in order to cope with this shortcoming, the use of the CBA is proposed in this paper. The CBA results from introducing the chaotic sequences into the standard BA to enhance its global search ability. The CBA is utilized to find the optimal settings of generator bus voltages, tap setting transformers and shunt reactive power sources. Three objective functions such as minimization of active power loss, total voltage deviations and voltage stability index are considered in this study. The effectiveness of the CBA technique is demonstrated for standard IEEE 14-bus, IEEE 39 New England bus, IEEE 57-bus, IEEE 118-bus and IEEE 300-bus test systems. The results yielded by the CBA are compared with other algorithms available in the literature. Simulation results reveal the effectiveness of the CBA for solving the ORPD problem.

INDEX TERMS Chaotic bat algorithm, optimal reactive power dispatch, chaotic sequences.

I. INTRODUCTION

In 1962, Carpentier introduced for the first time the optimal power flow (OPF) problem [1] and later on it was developed by Dommel and Tinney [2]. Henceforth, the OPF has aroused widespread interests among researchers in power system operation and planning [3]. The ORPD problem is a special case of the OPF problem and has an ever-increasing impact on the reliability, security as well as economic operation of power system [4].

From the mathematical optimization point of view, the ORPD problem is regarded as a mixed integer nonlinear programming problem which combines discrete control variables such as tap setting transformers and reactive compensators; and continuous variables such as generator bus voltages [5], [6].

The associate editor coordinating the review of this manuscript and approving it for publication was Ludovico Minati^D.

It is worth pointing out that the ORPD problem is addressed in an attempt to minimize either total active power transmission losses (P_{Loss}), or total voltage deviation (TVD), or enhance voltage stability index (VSI) while satisfying specific operating and physical constraints [3].

Over the last decades, extensive efforts have been undertaken to solve the ORPD problem using many classical optimization techniques including quadratic programming [7]–[10], mixed integer programming [11], interior point method [12]–[14], Newton-based method [15], linear programming [16], [17], non-linear programming [18], dynamic programming [19], gradient-based algorithm [20], decomposition approach [21] etc. Despite their excellent convergence characteristics, these classical optimization techniques fail to find the global solution due to their non-convexity characteristics. Furthermore, they are unable to handle the problems with discontinuous and non-differentiable objective functions, as well as discrete variables. Hence they are inappropriate to address the ORPD problem [3], [22].

The metaheuristic algorithms are typically inspired from physical phenomena such as gravitational force, electromagnetic force, inertia force, ray casting; animals' behavior such as schools of fish, flocks of birds, herds of land animals, ants, bees; or evolutionary concepts. The metaheuristics have become outstandingly popular owing to their simplicity, flexibility, derivation-free mechanism, and local optima avoidance [23]. Therefore, the metaheuristic algorithms overcome the shortcomings associated with the classical optimization algorithms [22], [24], [25]. Renowned metaheuristics include, among others, genetic algorithm (GA) [26], differential evolution (DE) [27], particle swarm optimization (PSO) [28], bat algorithm (BA) [29], firefly algorithm (FA) [30], flower pollination algorithm (FFA) [31], harmony search algorithm (HAS) [32], simulated anneal (SA) [33], gravitational search algorithm (GSA) [34], ant colony optimization (ACO) [35], wolf search algorithm (WSA) [36], grey wolf optimizer (GWO) [23], whale optimization algorithm (WOA) [37], elephant search algorithm (ESA) [38], ant lion optimizer (ALO) [39], salp swarm algorithm (SSA) [40], dragonfly algorithm (DA) [41], moth-flame optimization (MFO) [42], harris hawks optimization (HHO) [43], grasshopper optimization algorithm (GOA) [44], multi-verse optimization (MVO) [45], sine cosine Algorithm (SCA) [46], etc.

In recent times, some of metaheuristics including genetic algorithm (GA) [47]–[49], particle swarm optimization (PSO) [50], [51], differential evolution (DE) [52], seeker optimization algorithm (SOA) [53], gravitational search algorithm (GSA) [24], [54], [55], harmony search algorithm (HSA) [56], biogeography-based optimization (BBO) [57], teaching-learning-based optimization (TLBO) [58], krill herd algorithm (KHI) [59], bacteria foraging algorithm (BFA) [60], gray wolf optimizer (GWO) [61], ant-lion optimizer (ALO) [62], exchange market algorithm (EMA) [3], moth-flame algorithm (MFA) [4], imperialist competitive algorithm (ICA) [6], invasive weed optimization (IWO) [63], bat algorithm (BA) [64], cuckoo search algorithm (CSA) [65], etc, have been successfully employed to address the ORPD problem.

The BA is a novel metaheuristic optimization algorithm, introduced by Xin-She Yang in 2010, inspired by the echolocation of micro-bats [66]. The BA has recently been applied in numerous areas of research, including engineering optimization problems [67]–[69], economic load dispatch problems [70], microelectronic applications [71], scheduling problems [72], biological systems [73], image processing [74], classification, clustering and data mining [75]–[82], etc.

In an attempt to ameliorate its performance, various variants of BA have been proposed in the literature such as K-Means bat algorithm (KMBA) [75], fuzzy logic bat algorithm (FLBA) [77], multi-objective bat algorithm (MOBA) [69], binary bat algorithm (BBA) [82], bat algorithm with differential operator and Lévy flights (DLBA) [83], improved bat algorithm (IBA) [84], quantumbehaved bat algorithm [85]–[87], etc.

Thus far, chaos theory has been widely utilized to a broadspectrum of applications and to numerous metaheuristics in an effort to further improve their performance such as providing better convergence speed and avoiding entrapment into local optima [88]. In addition, the use of chaotic sequences has proved to outperform the use random numbers in terms of enhancing the performance of the metaheuristics (by tuning certain parameters) [89], [90]. In literature, examples of meta-heuristics which employ chaos include genetic algorithm (GA) [91], harmony search algorithm (HSA) [92], differential evolution (DE) [93], particle swarm optimization (PSO) [94], imperialist competitive algorithm (ICA) [95], krill herd algorithm (KHA) [96], gravitational search algorithm (GSA) [97], bat algorithm (BA) [98], firefly algorithm (FA) [99], cuckoo search algorithm (CSA) [100], butterfly optimization algorithm (BOA) [101], whale optimization algorithm (WOA) [88], grey wolf optimization (GWO) [102], moth-flame optimization (MFO) [103], salp swarm algorithm (SSA) [104]-[106], etc.

Some chaos-based metaheuristics have been efficiently applied to solve the ORPD problem [59], [107], [108] as well as economic load dispatch (ELD) problem [90], [109], [110], [111].

In the present work, the chaotic bat algorithm is implemented on the standard IEEE 14-bus, IEEE 39 New England, IEEE 57-bus, IEEE 118-bus and IEEE 300-bus test power systems with the aim to demonstrate its performance in solving the ORPD problem. The objective functions considered in this paper are the minimization of the active power loss, minimization of the total voltage deviation and minimization of the voltage stability index. The simulation results approve that the CBA is more potential and effective than the compared metaheuristic algorithms reported in the literature.

The rest of the paper is organized as follows: Section II presents the mathematical formulation of the ORPD problem, the CBA algorithm is described in Section III, the implementation of the CBA algorithm on the ORPD problem is described in Section IV, Section V presents the simulation results along with the discussions, and Section VI draws the conclusion of this paper.

II. PROBLEM FORMULATION

In this paper, three different objective functions are considered and described as follows:

A. OBJECTIVE FUNCTIONS

The purpose of the ORPD problem is to minimize either the active power loss (P_{Loss}) or total voltage deviation (TVD) or voltage stability index (VSI) while simultaneously fulfilling various equality and inequality constraints.

The mathematical formulation of the ORPD problem in general form is expressed as follows [3], [112], [113]:

$$Minimize J(x, u) \tag{1}$$

$$g\left(x,\,u\right) = 0\tag{2}$$

$$h\left(x,\,u\right) \le 0\tag{3}$$

where J(x, u) is the objective function to be minimized, g(x, u) = 0 represents the equality constraints, $h(x, u) \le 0$ represents the inequality constraints.

x is the vector of dependent variables (state vector) including:

- Load bus voltage V_L .
- Generator reactive power output Q_G .
- Transmission line loading S_l.

Hence, the vector *x* can be expressed as:

$$\boldsymbol{x}^{T} = \begin{bmatrix} V_{L1} \dots V_{LNPQ}, Q_{G1} \dots Q_{GNG}, S_{l1} \dots S_{lNTL} \end{bmatrix}$$
(4)

where NG depicts the number of generators; NPQ is the number of PQ buses and NTL is the number of transmission lines.

u is the vector of independent variables (control variables) including:

- Generator bus voltages V_G (continuous control variable).
- Transformer tap settings *T*(discrete control variable).
- Shunt VAR compensation Q_C (discrete control variable).

Therefore, *u* can be illustrated as follows:

$$u^{T} = [V_{G1} \dots, V_{GNG}, Q_{C1} \dots Q_{CNC}, T_1 \dots T_{NT}]$$
 (5)

where *NT* and *NC* define the number of tap regulating transformers and number of shunt VAR compensators, respectively.

1) MINIMIZATION OF ACTIVE POWER LOSS

The mathematical formulation of the ORPD problem for the minimization of the active power loss (P_{Loss}) is expressed as follows [113]:

Minimize $J_1(x_1, u_1) = \text{minimize } P_{Loss}$

$$=\sum_{k=0}^{NTL} g_k \left(V_i^2 + V_j^2 - 2V_i V_j \cos \delta_{ij} \right) \quad (6)$$

where $J_1(x_1, u_1)$ denotes the active power loss minimization function of the transmission network, g_k is the conductance of the branch k, V_i and V_j are the voltages of i th and jth bus, respectively, *NTL* is the number of transmission lines, δ_{ij} is the phase difference of voltages between bus i and j.

2) MINIMIZATION OF TOTAL VOLTAGE DEVIATIONS

The voltage deviations at load buses are minimized in an effort to enhance the voltage profile as well as the security of

the electric power network. Mathematically, the total voltage deviations (TVD) at load buses are expressed as [3]:

Minimize
$$J_2(x_2, u_2) = \text{minimize } TVD$$

$$= \sum_{i=1}^{NPQ} \left| V_i - V_i^{ref} \right|$$
(7)

where $J_2(x_2, u_2)$ is the total voltage deviation minimization function, *i* is the element of *NPQ*, V_i^{ref} is the reference voltage magnitude at *i* th load bus which is taken as 1p.u.

3) IMPROVEMENT OF VOLTAGE STABILITY INDEX

The L- index/stability index is an index which plays a significant role in voltage stability of the power system. Its values range from 0 to 1. In fact, the closer the L- index is to 0, the more stable is the voltage stability; the closer the L- index is to 1, the more unstable is the voltage stability [3].

The voltage stability index (VSI) or L_{max} can be mathematically expressed as [3]:

Minimize
$$J_3(x_3, u_3) = \text{minimize } L_{\text{max}}$$

= min[max (L_j)],
 $i=1, 2, 3, \dots, NPO$ (8)

where L_j is the voltage stability indicator (*L*- index) of *j* th node.

The value of L_i is given by

$$L_j = \left| 1 - \sum_{i=1}^{NPV} F_{ij} \frac{V_i}{V_j} \right| \tag{9}$$

$$F_{ij} = -[Y_1]^{-1}[Y_2]$$
(10)

where *i* and *j* represent the elements of PV(Generator) and PQ (Load) buses, respectively. $[Y_1]$ and $[Y_2]$ denote the sub-matrices of the system *Y* bus which result from the separation of PV and PQ buses as given in the following equation:

$$\begin{bmatrix} I_{PQ} \\ I_{PV} \end{bmatrix} = \begin{bmatrix} Y_1 & Y_2 \\ Y_3 & Y_4 \end{bmatrix} \begin{bmatrix} V_{PQ} \\ V_{PV} \end{bmatrix}$$
(11)

B. CONSTRAINTS

1) EQUALITY CONSTRAINTS

In (2), g is the set of equality constraints illustrating the load flow equations as follows:

$$P_{Gi} - P_{Di} - V_i \sum_{j=1}^{NB} V_j \left(G_{ij} \cos \delta_{ij} + B_{ij} \sin \delta_{ij} \right) = 0 \qquad (12)$$

$$Q_{Gi} - Q_{Di} - V_i \sum_{j=1}^{NB} V_j \left(G_{ij} \cos \delta_{ij} + B_{ij} \sin \delta_{ij} \right) = 0 \qquad (13)$$

where *NB* is the number of buses, P_{Gi} and Q_{Gi} are active and reactive power generation at the *i* th bus, P_{Di} and Q_{Di} are active and reactive power demands at the *i* th bus, G_{ij} and B_{ij} are the transfer conductance and susceptance between the *i* th and the *j*th buses respectively.

In (3), h is the set of inequality constraints that consists of:

1) GENERATOR CONSTRAINTS

The generation bus voltages including slack bus and reactive power outputs including slack bus must be restricted within their lower and upper limits as follows:

$$V_{Gi}^{\min} \le V_{Gi} \le V_{Gi}^{\max}, \quad i = 1, 2, \dots, NG$$
 (14)

$$Q_{Gi}^{\min} \le Q_{Gi} \le Q_{Gi}^{\max}, \quad i = 1, 2, \dots, NG$$
 (15)

where V_{Gi}^{\min} and V_{Gi}^{\max} are the minimum and maximum generator voltage of *i* th generating unit, Q_{Gi}^{\min} and Q_{Gi}^{\max} are the minimum and maximum reactive power output of *i* th generating unit.

2) TRANSFORMER CONSTRAINTS

Transformer tap settings are constrained within their lower and upper limits as below:

$$T_i^{\min} \le T_i \le T_i^{\max}, \quad i = 1, 2, \dots, NT$$
(16)

where T_i^{\min} and T_i^{\max} depict the minimum and maximum tap setting limits of *i* th transformer.

3) SHUNT VAR COMPENSATOR CONSTRAINTS

The lower and upper limits of the shunt VAR compensators are given as:

$$Q_{Ci}^{\min} \le Q_{Ci} \le Q_{Ci}^{\max}, \quad i = 1, 2, \dots, NC$$
 (17)

where Q_{Ci}^{\min} and Q_{Ci}^{\max} are the minimum and maximum VAR injection limits of *i* th shunt compensator.

4) SECURITY CONSTRAINTS

These includes the constraints on voltages at load buses and transmission line loadings as

$$V_{Li}^{\min} \le V_{Li} \le V_{Li}^{\max}, \quad i = 1, 2, \dots, NPQ$$
 (18)

$$S_{li} \le S_{li}^{\max}, \quad i = 1, 2, \dots, NTL$$
 (19)

where V_{Li}^{\min} and V_{Li}^{\max} are the minimum and maximum load voltages of *i* th unit. S_{li} denotes apparent power flow of *i* th branch and S_{li}^{\max} denotes maximum apparent power flow limit of *i* th branch.

D. CONSTRAINT HANDLING

The independent variables of the ORPD problem are selfconstrained whereas the dependent variables are constrained by means of the penalty functions. Hence, the equation (1) becomes [3]:

Minimize

$$F = F_{obj} + \lambda_V \sum_{i=1}^{NPQ} \left(V_{Li} - V_{Li}^{\lim} \right)^2 + \lambda_Q \sum_{i=1}^{NG} \left(Q_{Gi} - Q_{Gi}^{\lim} \right)^2 + \lambda_S \sum_{i=1}^{NTL} \left(S_{li} - S_{li}^{\lim} \right)^2$$
(20)

generation reactive power, and line flow, respectively. They

$$V_{Li}^{\lim} = \begin{cases} V_{Li}^{\max} & if \quad V_{Li} > V_{Li}^{\max} \\ V_{Li}^{\min} & if \quad V_{Li} < V_{Li}^{\min} \\ V_{Li} & if \quad V_{Li}^{\min} \le V_{Li} \le V_{Li}^{\max} \end{cases}$$
(21)

$$Q_{Gi}^{\lim} = \begin{cases} Q_{Gi}^{\max} & if \quad Q_{Gi} > Q_{Gi}^{\max} \\ Q_{Gi}^{\min} & if \quad Q_{Gi} < Q_{Gi}^{\min} \\ Q_{Gi} & if \quad Q_{Gi} < Q_{Gi}^{\min} \end{cases}$$
(22)

$$S_{li}^{\lim} = \begin{cases} Q_{Gi} & if \quad Q_{Gi}^{\min} \le Q_{Gi} \le Q_{Gi}^{\max} \\ S_{li}^{\min} & if \quad S_{li} > S_{li}^{\max} \\ S_{li}^{\min} & if \quad S_{li} < S_{li}^{\min} \\ S_{li} & if \quad S_{li} \le S_{li} \le S_{li}^{\max} \end{cases}$$
(23)

III. CHAOTIC BAT ALGORITHM

are expressed as follows

A. BRIEF REVIEW OF THE BASIC BAT ALGORITHM

The BA mimics the echolocation behavior of the microbats while searching for prey and avoiding obstacles [66].

Three rules have been suggested in [66] in order to idealize the echolocation characteristics of microbats:

- 1) Each microbat utilizes echolocation to estimate the distance between prey and surroundings.
- 2) Bats fly randomly with velocity V^i at position X^i with a fixed frequency f^{\min} , varying wavelength λ and loudness A^0 to search for prey. They can automatically adjust the wavelength (or frequency) of their emitted pulses and adjust the rate of pulse emission $r_1 \in$ [0, 1],depending on the proximity of their target.
- 3) It is assumed that the loudness varies from a large (positive) A^0 to a minimum constant value A^{\min} .

Each bat *i* has a position X^i , a velocity V^i , a frequency f^i in a d- dimensional space, and they should be updated iteratively towards the current best position as follows

$$f^{i} = f^{\min} + r_1 \left(f^{\max} - f^{\min} \right) \tag{24}$$

$$V^{i}(t+1) = V^{i}(t) + f^{i}\left(X^{i}(t) - X^{best}(t)\right)$$
(25)

$$X^{i}(t+1) = X^{i}(t) + V^{i}(t+1)$$
(26)

where r_1 is a uniformly distributed random number in the range [0, 1]; f^{\min} and f^{\max} are the minimum and maximum allowable frequencies while f^i is the frequency for the *i* th bat. For the present ORPD problem, the values of f^{\min} and f^{\max} are set to 0 and 100, respectively as given in [66] and in [90], *t* is the current iteration number, X^{best} is the location (solution), that has the best fitness in the current population. At initialization, V^i is assumed to be 0.

A new solution for each bat can be generated locally through random walk as follows

$$X^{i,new}(t) = \begin{cases} X^{best}(t) + r_3 A^i(t) & if \quad r_2 > R^i(t) \\ X^r(t) + r_3 A^i(t) & else \end{cases}$$
(27)

where $R^i(t)$ is the pulse emission rate, $A^i(t)$ is the loudness, $r_2 \in [0, 1], r_3 \in [-1, 1], r \in [1, 2, ..., N_b], r \neq i$ is a randomly chosen integer, N_b is the number of bats or solutions, $X^r(t)$ is a randomly chosen solution in the current iteration, and different from the *i* th iteration.

The fitter solution is given by

$$X^{i}(t) = X^{i,new}(t), \quad \text{if } f\left(X^{i,new}(t)\right) < f\left(X^{i}(t)\right), \text{ and}$$

$$r_{4} < A^{i}(t) \forall i, \quad i \in [1, 2, \dots, N_{b}]$$
(28)

where $r_4 \in [0, 1]$ is a uniformly distributed random number.

Once the bat has found its prey, the loudness keeps on decreasing whereas the pulse rate emission keeps on increasing. The loudness A^i and the pulse emission rate R^i are iteratively updated as follows

$$A^{i}(t+1) = \alpha A^{i}(t) \tag{29}$$

$$R^{i}(t+1) = R^{i}(0) \left[1 - \exp(-\gamma t)\right]$$
(30)

where $A^i(0) \in [1, 2]$ and $R^i(0) \in [0, 1]$ are randomly generated within their respective limits. For the sake of simplicity, we set $\alpha = \gamma = 0.9$, as in [66] and in [90].

B. CHAOTIC BAT ALGORITHM

In standard BA algorithm, the loudness and pulse emission rate are two parameters that can be adjusted to control exploration and exploitation of the BA algorithm and hence to improve its performance. Furthermore, the replacement of randomness by chaotic maps in adjusting the aforementioned parameters has proved to be more effective [98].

In this paper, two variants of CBA are going to be considered:

1) CBA-III

In this CBA variant, the sinusoidal map has been used to adjust the loudness as follows [90], [98]:

$$A^{i}(t+1) = a \left\{ A^{i}(t) \right\}^{2} \sin\left(\pi A^{i}(t)\right)$$
(31)
where $a = 2.3$,
 $A^{i}(0) \in [0, 1]$, and is randomly generated.

2) CBA-IV

In this CBA variant, the sinusoidal map has been used to adjust the pulse rate emission as follows [98]:

$$R^{i}(t+1) = a \left\{ R^{i}(t) \right\}^{2} \sin\left(\pi R^{i}(t)\right)$$
(32)
where $a = 2.3$,
 $R^{i}(0) = 52.43$, $R^{i}(0) =$

 $R^{l}(0) \in [0, 1]$, and is randomly generated.

IV. IMPLEMENTATION OF CBA TO ORPD PROBLEM

This section discusses the procedure of applying CBA to solve the ORPD problem.

A. CBA-III

Step 1: The initial position (bat solution) X^i and its velocity V^i of each bat are randomly generated within the specified limits. The bat solution X^i is a vector of the control variables of the ORPD problem such as generator voltages, tap setting transformers and shunt compensators.

For each *i* th bat, define pulse frequency f^i at X^i .

Initialize the pulse rate emission R^i and loudness A^i and specify their lower and upper limit values.

Step 2: Evaluate the fitness values of all the bats using the objective function of the problem defined in (6) or (7) or (8) in accordance with the results of New-Raphson power flow analysis [114].

Step 3: For each bat *i*, adjust the frequency f^i using (24).

Step 4: Update the velocity and position of each bat in accordance with (25) and (26) respectively.

Step 5: Generate a new solution by randomly walk based on (27).

Step 6: Select the best among the old and new solutions, with a probability $A^{i}(t)$, by using (28).

Step 7: Update the values of R^i and A^i in accordance with (30) and (31), respectively.

Step 8: Check for the equality constraints of the problem (12)-(13) and the inequality constraints (14)-(19) of the problem.

Step 9: Go to step 2 until stopping criteria is met.

B. CBA-IV

Step 1: The initial position (bat solution) X^i and its velocity V^i of each bat are randomly generated within the specified limits. The bat solution X^i is a vector of the control variables of the ORPD problem such as generator voltages, tap setting transformers and shunt compensators.

For each *i* th bat, define pulse frequency f^i at X^i .

Initialize the pulse rate emission R^i and loudness A^i and specify their lower and upper limit values.

Step 2: Evaluate the fitness values of all the bats using the objective function of the problem defined in (6) or (7) or (8) in accordance with the results of New-Raphson power flow analysis [114].

Step 3: For each bat *i*, adjust the frequency f^i using (24).

Step 4: Update the velocity and position of each bat in accordance with (25) and (26) respectively.

Step 5: Generate a new solution by randomly walk based on (27)

Step 6: Select the best among the old and new solutions, with a probability $A^{i}(t)$, by using (28).

Step 7: Update the values of A^i and R^i in accordance with (29) and (32), respectively.

Step 8: Check for the equality constraints of the problem (12)-(13) and the inequality constraints (14)-(19) of the problem.

Step 9: Go to step 2 until stopping criteria is met.

V. RESULTS AND DISCUSSION

In order to verify the performance and effectiveness of the CBA algorithm for ORPD problem, the standard BA and two different variants of CBA namely CBA-III and CBA-IV have been tested on standard IEEE 14-bus, IEEE 39 New England bus, IEEE 57-bus, IEEE 118-bus and IEEE 300-bus test systems for the minimization of the active power loss, total voltage deviations and voltage stability index objective functions. All these algorithms were implemented in MAT-LAB 2014a incorporated with MATPOWER 6.0 [115] and run on PC i.e Lenovo Ideapad 100-15IKB, 2.1 GHz Intel Pentium with 4GB RAM.

The population size is set to 120 ($N_b = 120$). The iteration numbers are set to 100 for IEEE 14-bus test system; 200 for IEEE 39 New England bus and IEEE 57-bus test systems; and 300 for IEEE 118-bus and IEEE 300-bus test systems. In this simulation, 30 independent test trial runs were carried out for all the test cases.

The performance capability of the CBA-IV for each objective function is indicated by bold faced results.

A. IEEE 14-BUS TEST SYSTEM

The IEEE 14-bus test system constists of five generators at bus 1, 2, 3, 6, and 8; 20 branches 17 branches are transmission lines and 3 branches are tap changing transformers; and 2 shunt VAR compensators installed at buses 9 and 14. In total, this system contains 10 control variables including five generators, three tap changing transformers and two shunt capacitors. Further details are found in [52], [58], [116].

The total demand of the system are

 $P_{load} = 259MW$ (active power demand),

 $Q_{load} = 73.5 MVAR$ (reactive power demand).

The initial total generations and power losses are given by

 $\sum PG = 272.39MW$ (active power of generators),

 $\sum QG = 82.44 MVAR$ (reactive power of generators),

 $P_{LOSS} = 13.49MW$ (active power losses),

 $Q_{LOSS} = -54.54 MVAR$ (reactive power losses).

The control variable limits are listed in Table 1 in p.u. [58], [116].

TABLE 1. Control variable limits for IEEE 14-bus test systems [58], [116].

V_G^{\max}	V_G^{\min}	V_{PQ}^{\max}	V_{PQ}^{\min}	T_k^{\max}	T_k^{\min}	Q_C^{\max}	Q_{C}^{\min}
1.1	0.95	1.05	0.95	1.1	0.9	0.3	0

1) CASE 1: MINIMIZATION OF ACTIVE POWER LOSS

The standard BA, CBA-III and CBA-IV algorithms are implemented on the IEEE 14-bus test system for the minimization of the active power loss defined in (6) taking into account the penalty terms defined in (20). Table 2 illustrates the best results yielded by the standard BA, CBA-III, CBA-IV and those yielded by other algorithms reported in the literature including DE [52], MTLA-DDE [116], MGBTLBO [58], SARGA [49], etc.

According to simulation results from Table 2, the best (minimum) power loss P_{Loss} offered by the CBA-IV algorithm is **12.2923 MW** which is also better than that obtained by the other algorithms compared with it.

Table 3 provides the comparison of statistical results such as the best active power loss (Best), the worst active power loss (Worst), the mean power loss (Mean), the standard deviation (Std) and the percent of power loss reduction (% P save); yielded by different algorithms available in the literature for IEEE 14- bus system. From Table 3, it is seen that the CBA-IV algorithm outperforms other algorithms as it can achieve a power loss reduction of 8.88% (from the initial power loss) compared with 8.75% by CBA-III, 8.74% by MGBTLBO [58], 8.7% by BA, 8.68% by JAYA [117], 8.3% by DE-ABC [118], 8.1% by IGSA-CSS [24], 8% by CSSP4 [119], 7.72% by DE [120], 7.717% by DEEP [121],7.05% by GSAPSO [54], 4.39% by MTLA-DDE [116], 3.86% by PSO [122], 2.03% by SARGA [49], 1.87% by PSO-AM [123], 1.86 by DE [52] and 1.68% by PSO-CM [123].

The outperformance of the CBA-IV over CBA-III and the standard BA is due to the sinusoidal map that has been employed to adjust the pulse rate emission as discussed in (32).

The convergence characteristics of active power loss (over 100 iterations) yielded by (the best solutions of) the standard BA, CBA-III and CBA-IV algorithms for IEEE 14-bus are shown in Fig.1. From this figure, it observed that the CBA-IV

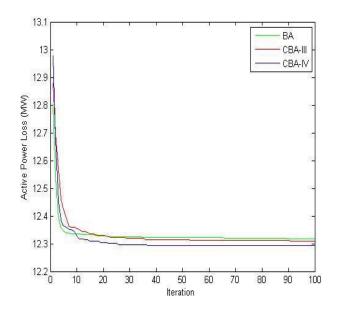


FIGURE 1. The convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 14-bus test power system with the minimization of P_{Loss} objective function.

CDAIN

TABLE 2. Optimal settings of control variables for IEEE 14-bus test systems with the minimization of P_{Loss} . Objective function.

Variable	Base Case (Initial)	MTLA- DDE [116]		MGB- TLBO [58]	DE [52]
Generator voltage					
V _{G1}	1.06	1	.07531	1.1	1.06
V _{G2}	1.045	1.05734		1.0791	1.0449
V _{G3}	1.01	1	.02847	1.0485	1.0416
V _{G6}	1.07	0	0.05057	1.0552	1.1
V_{G8}	1.09	1	.03535	1.0326	1.1
Transformer tap 1				4.64	1.0.0
<i>T</i> ₄₋₇	0.9467	1	.08	1.01	1.06
T ₄₋₉	0.9524	0	.91	1.01	1.04
T ₅₋₆	0.9091	1	.01	1.03	1.1
Capacitor banks (
Q_{C9}	0.18		0.3	0.3	0.18
Q_{C14}	0.18		0.08	0.07	0.06
P _{Loss} (MW)		12.8978		12.3106	13.239
TVD (p.u)	-	-		-	-
L - index	-	-		-	-
(p.u) Variable	PSO [122]		GSAPSO [54]	IGSA- CSS	JAYA [117]
Generator voltage	e (p.u)			[24]	
17	1.0917		1.1	1.1	1.10
V_{G1}	1.0862		1.076853	1.07657	
V _{G2}					
V _{G3}	1.0876		1.046118	1.04678	
V_{G6}	1.0577	1.025647		1.06230	
<i>V</i> _{<i>G</i>8}	1.0278	1.096356		1.09786	1 1.092
Transformer tap 1	atio (p.u)		0.96	1.02	1.045
<i>T</i> ₄₋₇	1.0039			0.94	0.90
<i>T</i> ₄₋₉			1.1		1.003
T ₅₋₆			1.04	1.00	1.003
Capacitor banks ((p.u) 0.0513		0.045	0.050	0.18
Q_{C9}	0.0313		0.012	0.050	0.18
Q_{C14}	12.9691	1	- 12.44901	12.3970	
P _{Loss} (MW)		L			
TVD (p.u)	-		-	-	-
L - index					

Variable	Base	BA	CBA-III	CBA-IV
	Case			
Computer	(Initial)			
Generator	voltage (p.	·		
V_{G1}	1.06	1.0990	1.0981	1.0921
V_{G2}	1.045	1.0325	1.0832	1.0884
V _{G3}	1.01	1.0123	1.0587	1.0558
V _{G6}	1.07	1.0631	1.0527	1.0325
V_{G8}	1.09	1.0368	1.0991	1.0951
Transform	er tap ratio	(p.u)		
<i>T</i> ₄₋₇	0.9467	0.9562	1.0055	0.9746
T ₄₋₉ T ₅₋₆	0.9524	1.0251	1.0059	1.0676
<i>T</i> ₅₋₆	0.9091	0.9730	1.0679	1.0599
Capacitor	banks (p.u)	•		
Q_{C9}	0.18	0.0863	0.1597	0.2208
Q_{C14}	0.18	0.6490	0.0827	0.0786
P _{Loss}	13.49	12.3171	12.3092	12.2923
(MW)		0.06108	0.05490	0.05209
TVD	-	0.06108	0.05489	0.05308
(p.u) <i>L</i> –	-	0.03769	0.03678	0.03642
L – index				
(p.u)				
L (P·u)	I	L		

TABLE 2. (Continued.) Optimal settings of control variables for IEEE

DA

Variable Dage

14-bus test systems with the minimization of PLoss. Objective function.

CDA III

gives better convergence characteristics than CBA-III and BA algorithms.

2) CASE 2: MINIMIZATION OF TOTAL VOLTAGE DEVIATION

For the present case, the minimization of the Total Voltage Deviations (TVD) discussed in (7) is considered as the objective function together with the penalty factors defined in (20).

The best results yielded by the CBA algorithm are tabulated in Table 4 together with the results of the CBA-III, BA and IGSA-CSS [24]. According to Table 4, it can be observed that the CBA-IV yielded a TVD value of **0.0330** compared with the results 0.0332, 0.0336 and 0.0339 achieved by CBA-III, CBA and IGSA-CSS [24] respectively. It is also recognized from Table 5 that an improvement of **31.10%** in TVD has been achieved by using CBA-IV in comparison to 30.69% with CBA-III, 29.85% with BA and 29.22% with IGSA-CSS [24]. The Fig.2 illustrates the comparison of the convergence characteristics of TVD for IEEE 14-bus system. From this figure, it is clear that the convergence characteristics of TVD for the CBA-IV outperforms those from the compared algorithms.

TABLE 3.	Statistical results for IEEE 14-bus test system with the
minimiza	tion of P _{Loss} . Objective function.

Algorithm	Best	Worst	Mean	Std	%
_	(MW)	(MW)	(MW)		Psave
DE [52]	13.239	13.275	13.25	0.000161	1.86
MTLA-	12.8978	12.8986	12.8982	0.000006486	4.39
DDE					
[116]					
MGB-	12.3106	12.3111	12.3106	0.0000086	8.74
TLBO					
[58]					
GSAPSO	12.44901	-	-	-	7.05
[54]					
IGSA-	12.39706	12.90281	12.46443	0.094	8.1
CSS [24]					
SARGA	13.21643	13.23891	13.22317	0.000024	2.03
[49]					
DE-ABC	12.3712	12.3712	12.3712	7.29E-08	8.3
[118]					
JAYA	12.3192	-	-	-	8.68
[117]					
PSO-CM	13.2634	13.3142	13.2671	0.00009	1.68
[123]					
PSO-AM	13.2371	13.2550	13.2395	0.00006	1.87
[123]					
DEEP	12.4489	12.4507	12.4494	0.0005	7.717
[121]					
CSSP4	12.4087	12.4974	12.4393	0.228	8
[119]					
DE [120]	12.4486	12.4496	12.4486	0.0018	7.72
PSO [122]	12.9691	-	-	-	3.86
CBA	12.3171	12.3456	12.3230	0.0099	8.7
CBA-III	12.3092	12.3212	12.3161	0.0034	8.75
CBA-IV	12.2923	12.3098	12.3042	0.0046	8.88

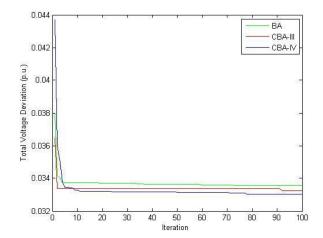


FIGURE 2. The convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 14-bus test power system with the minimization of TVD bjective function.

3) CASE STUDY 3: IMPROVEMENT OF VOLTAGE STABILITY INDEX

In this case, the CBA algorithm is employed to minimize the Voltage Stability Index (VSI) discussed in (8) as the objective function together with the penalty terms defined in (20).

TABLE 4. Optimal settings of control variables for IEEE 14-bus test
systems with the minimization of TVD objective function.

Variable	IGSA-CSS [24]	BA	CBA-III	CBA-IV
Generator volta	ge (p.u)			
V_{G1}	1.060879	0.9993	0.9998	0.9958
V _{G2}	1.040856	1.0004	0.9969	1.0189
V _{G3}	1.011222	1.0006	0.9972	1.0008
V _{G6}	1.016776	0.9999	0.9945	1.0102
V_{G8}	1.035129	0.9993	1.0015	1.0501
Transformer tap	ratio (p.u)			
<i>T</i> ₄₋₇	1.04	1.0307	1.0581	1.0121
T ₄₋₉	0.9	1.0455	0.9085	1.0975
T ₅₋₆	0.92	0.9966	0.9457	1.0370
Capacitor banks	s (p.u)	•	•	•
Q_{C9}	0.050	0.2308	0.1542	0.0903
$Q_{c_{14}}$	-	0.2168	0.1458	0.0637
P _{Loss} (MW)	-	16.2499	16.0172	15.9506
TVD (p.u)	0.03390	0.0336	0.0332	0.0330
L - index (p.u)	-	0.0435	0.0425	0.0414

 TABLE 5.
 Statistical results for IEEE 14-bus test system with the minimization of TVD objective function.

Algorithm	Best	Worst	Mean	Std	% TVD
	(p.u.)	(p.u.)	(p.u.)		Improve
IGSA- CSS [24]	0.03390	0.09056	0.04583	0.017	29.22
BA	0.0336	0.0515	0.0410	0.0058	29.85
CBA-III	0.0332	0.0489	0.0399	0.0054	30.69
CBA-IV	0.0330	0.0425	0.0368	0.0029	31.10

The table 6 shows the best results obtained using the CBA-IV alongside with the CBA-III and the BA. From this table, it can be seen that an L- index value of **0.0170** is obtained by the CBA-IV compared to 0.0174 with the CBA-III and 0.0179 with the BA. The statistical results for the present case are summarized in Table 7. According to Table 7, it is found that the CBA-IV also provides the smallest Best, Mean and Std values than the CBA-III and the BA. The comparison of convergence characteristics of VSI for IEEE 14-bus system is depicted in Fig.3. According to this figure, it can be recognized that the convergence

TABLE 6. Optimal settings of control variables for IEEE 14-bus test systems with the minimization of VSI objective function.

Variable	BA	CBA-III	CBA-IV
Generator voltage (p.	u)		
V_{G1}	1.0400	0.9895	1.100
V _{G2}	1.0292	0.9600	1.0964
V _{G3}	0.9779	1.0092	0.9270
V _{G6}	1.0344	1.0212	1.0187
V ₆₈	1.0337	1.0249	1.0003
Transformer tap ratio	(p.u)		
T ₄₋₇	0.9649	0.9639	0.9975
T ₄₋₉	1.0151	1.0576	1.0413
T ₅₋₆	0.9836	0.9399	0.9495
Capacitor banks (p.u)	, ,		
Q_{C9}	0.1416	0.0156	0.0572
$Q_{c_{14}}$	0.2630	0.2665	0.0903
P _{Loss} (MW)	18.3607	17.9569	17.6910
TVD (p.u)	0.07563	0.07348	0.07159
L - index (p.u)	0.0179	0.0174	0.0170

 TABLE 7. Statistical results for IEEE 14-bus test system with the minimization of VSI objective function.

Algorithm	Best (p.u.)	Worst (p.u.)	Mean (p.u.)	Std
BA	0.0179	0.0204	0.0189	7.8736× 10 ⁻⁴
CBA-III	0.0174	0.0187	0.0179	3.8850× 10 ⁻⁴
CBA-IV	0.0170	0.0182	0.0175	3.6334 × 10 ⁻⁴

characteristics from the CBA-IV outperforms the convergence characteristics from the CBA-III and the BA.

B. IEEE 39 NEW ENGLAND BUS TEST SYSTEM

The IEEE 39 New England bus test system comprises 10 generators at the buses 30, 31, 32, 33, 34, 35, 36, 37, 38 and 39; 46 lines; 5 tap changing transformers at lines 12-11, 10-32, 22-35, 2-30 and 19-20; and 6 shunt VAR compensators installed at buses 1, 5, 10, 14, 22 and 27. Totally, the IEEE 39-bus test system contains 21 control variables including ten generators, five tap changing transformers and six shunt capacitors. See [124] and [125] for further details.

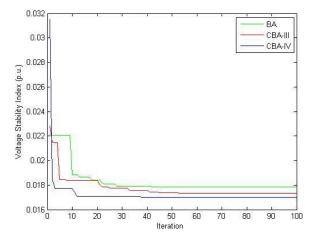


FIGURE 3. The convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 14-bus test power system with the minimization of VSI objective function.

TABLE 8. Control variable limits for IEEE 39-bus test systems [124].

ſ	V_G^{\max}	V_G^{\min}	V_{PQ}^{\max}	V_{PQ}^{\min}	T_k^{\max}	T_k^{\min}	Q_C^{\max}	Q_{C}^{\min}
ſ	1.1	0.9	1.1	0.9	1.07	0.93	0.25	0

The control variable limits are listed in Table 8 in p.u. [124].

1) CASE 1: MINIMIZATION OF ACTIVE POWER LOSS

Table 9 presents the best results yielded by different algorithms for the minimization of the active power loss of the IEEE 39-bus test system. It can be observed that the CBA-IV gives the best active power loss which is **35.9971 MW**. The comparative statistical results are illustrated in Table 10. It can be seen from Table 6 that a **17.44** % decrease (from the initial power loss of 43.6 MW) in active power loss is achieved with the CBA-IV algorithm, which performs better than other algorithms compared with it. The comparative convergence characteristics of active power loss over 200 iterations yielded by the standard BA, CBA-III and CBA-IV are presented in Fig.4. It can be observed from Fig.4 that the CBA-IV gives better convergence characteristics than the CBA-III and BA algorithms.

2) CASE 2: MINIMIZATION OF TOTAL VOLTAGE DEVIATION

Table 11 presents the solution of the ORPD problem obtained by using the CBA-IV algorithm in order to minimize the TVD of the IEEE 39-bus system. It is seen that the CBA-IV out-performs other algorithms compared with because the CBA-IV provides a TVD value of **0.0739** against 0.0796 with the CBA-III and 0.0825 with the BA. The statistical analysis of the present case is illustrated in Table 12. It is observed that the CBA-IV offers the minimum Best, Mean and Std values in comparison to the CBA-III and the BA. In addition, Fig.5 depicts the comparison of the convergence characteristics for TVD obtained by CBA-IV with the other approaches.

TABLE 9. Optimal settings of control variables for IEEE 39-bus test	
systems with the minimization of <i>P_{Loss}</i> . Objective function.	

Variable	Base Case	PSO [122]	BA	CBA-III	CBA-IV
Generator vol	(Initial) tage (p.u)				
V _{G30}	1.027	1.0496	1.0906	1.0668	1.0810
V _{G31}	1.023	1.0888	1.0999	1.0968	1.0999
V _{G32}	1.033	1.0931	1.0998	1.0989	1.1000
V _{G33}	1.046	1.0431	1.0943	1.0857	1.0952
V _{G34}	1.026	1.0944	1.0956	1.0941	1.0999
V _{G35}	1.068	1.0268	1.1000	1.0976	1.1000
V _{G36}	1.090	1.0856	1.0989	1.1000	1.0992
V _{G37}	1.031	1.0925	1.1000	1.1000	1.1000
V _{G38}	1.065	0.9767	1.0992	1.0988	1.0996
V _{G39}	1.005	1.0435	1.1000	1.1000	1.1000
Transformer t	ap ratio (p.	u)			I
<i>T</i> ₁₂₋₁₁	0	0.9917	1.0250	1.0461	1.0337
<i>T</i> ₁₀₋₃₂	0	1.0394	1.0700	1.0656	1.0703
<i>T</i> ₂₂₋₃₅	0	1.0304	1.0034	0.9792	1.0072
T ₂₋₃₀	0	1.0039	1.0478	1.0591	1.0485
<i>T</i> ₁₉₋₂₀	0	0.9857	1.0700	1.0645	1.0701
Capacitor ban		-	-		
Q_{C1}	1.000	0.1710	0.1985	0.2392	0.2092
Q_{C5}	1.000	0.1428	0.1266	0.0729	0.1188
Q_{C11}	1.000	0.1248	0.1985	0.2392	0.2092
Q_{C14}	1.000	0.1094	0.1266	0.0729	0.1188
Q_{C22}	1.000	0.0841	0.1985	0.2392	0.2092
Q_{C27}	1.000	0.1449	0.1985	0.2392	0.2092/
P _{Loss}	43.6	41.8892	36.7317	36.3125	35.9971
(MW) TVD	-	-	0.275	0.271	0.269
(p.u) <i>L</i> —	-	-	0.0829	0.0824	0.0821
index (p.u)					

3) CASE STUDY 3: IMPROVEMENT OF VOLTAGE STABILITY INDEX

The solution of the ORPD problem for minimizing the VSI of the IEEE 39-bus system by using the CBA-IV algorithm

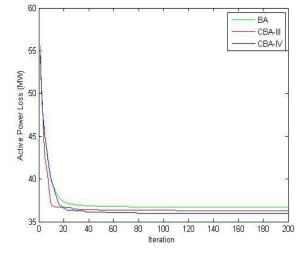


FIGURE 4. The convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 39-bus test power system with the minimization of P_{Loss} objective function.

TABLE 10.	Statistical	results for	IEEE 39-bus	test system with the
minimizati	on of PLoss	. Objective	function.	

Algorithm	Best (MW)	Worst (MW)	Mean (MW)	Std	% Psave
PSO [122]	41.8892	-	-	-	3.92
СВА	36.7317	38.2621	36.8847	0.3309	15.75
CBA-III	36.3125	37.7435	36.4527	0.3272	16.71
CBA-IV	35.9971	36.6986	36.1028	0.2122	17.44

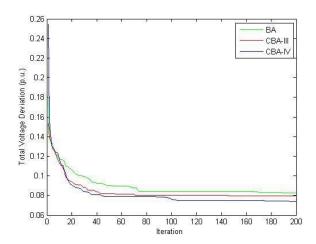


FIGURE 5. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 39-bus test power system with the minimization of TVD objective function.

is shown in Table 13. It is observed that the CBA-IV out-performs other algorithms compared with it because the CBA-IV offers an L- index value of **0.0113** against 0.0130 with the CBA-III and 0.0136 with the BA. Table 14 shows the statistical results for this case. Furthermore,

TABLE 11. Optimal settings of control variables for IEEE 39-bus test systems with the minimization of TVD objective function.

Variable	Base Case (Initial)	BA	CBA-III	CBA-IV
Generator voltage ((p.u)			
V _{G30}	1.027	0.9395	1.0286	0.9423
V _{G31}	1.023	1.0999	1.0723	0.9124
V _{G32}	1.033	0.9995	0.9782	1.0483
V _{G33}	1.046	1.0690	1.0084	0.9723
V _{G34}	1.026	0.9474	0.9797	0.9983
V _{G35}	1.068	0.9151	1.0668	0.9011
V _{G36}	1.090	0.9740	0.9279	0.9866
V _{G37}	1.031	0.9001	0.9512	0.9588
V _{G38}	1.065	0.9468	0.9362	1.0161
V _{G39}	1.005	1.0995	1.0077	0.9772
Transformer tap rat	tio (p.u)			
<i>T</i> ₁₂₋₁₁	0	0.9833	0.9731	0.9318
<i>T</i> ₁₀₋₃₂	0	1.0559	0.9972	0.9017
<i>T</i> ₂₂₋₃₅	0	0.9989	0.9457	0.9061
T ₂₋₃₀	0	0.9563	1.0325	1.0105
<i>T</i> ₁₉₋₂₀	0	0.9124	0.9636	0.9193
Capacitor banks (p				I
Q_{C1}	1.000	0.0533	0.2198	0.1234
Q_{C5}	1.000	0.1485	0.0554	0.2607
Q_{C11}	1.000	0.0533	0.2198	0.1234
Q_{C14}	1.000	0.1485	0.0554	0.2607
Q_{C22}	1.000	0.0533	0.2198	0.1234
Q_{C27}	1.000	0.0533	0.2198	0.1234
P _{Loss (MW)}	43.6	53.74922	53.4682	51.6495
TVD (p.u)	-	0.0825	0.0796	0.0739
L - index (p.u)	-	0.0880	0.0874	0.0868

Fig.6 illustrates the comparison of the convergence characteristics for VSI obtained by CBA-IV with the other methods.

Algorithm	Best (p.u.)	Worst (p.u.)	Mean (p.u.)	Std
СВА	0.0825	0.0967	0.0864	0.0038
CBA-III	0.0796	0.0921	0.0846	0.0034
CBA-IV	0.0739	0.0864	0.0763	0.0027

TABLE 12. Statistical results for IEEE 39-bus test system with the

minimization of TVD objective function.

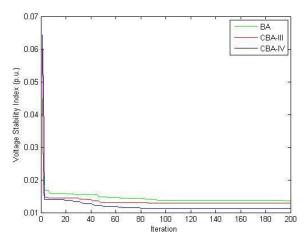


FIGURE 6. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 39-bus test power system with the minimization of VSI objective function.

C. IEEE 57 BUS TEST SYSTEM

The IEEE 57 bus test system comprises 7 generators at the buses 1,2,3,6,8,9 and 12; 80 transmission lines; 15 tap changing transformers and 3 shunt VAR compensators installed at buses 18, 25 and 53. Totally, the IEEE 57-bus test system contains 25 control variables including 7 generators, 15 tap changing transformers and 3 shunt capacitors.

Refer to [53], [63] and [118] for further details.

The total demand of the system are

 $P_{load} = 1250.8MW$ (active power demand),

 $Q_{load} = 336.4 MVAR$ (reactive power demand).

The initial total generations and power losses are given by

 $\sum PG = 1279.26MW$ (active power of generators),

 $\sum QG = 345.45MVAR$ (reactive power of generators),

 $P_{LOSS} = 28.462 MW$ (active power losses),

 $Q_{LOSS} = -124.27 MVAR$ (reactive power losses).

The control variable limits are listed in Table 15 in p.u. [118].

1) CASE 1: MINIMIZATION OF ACTIVE POWER LOSS

Table 16 illustrates the best results obtained by different algorithms to minimize the active power loss of the IEEE 57-bus test system. From Table 16, the CBA-IV algorithm leads to **21.9627 MW** active power loss which is better than the active power losses obtained by the other compared algorithms. The comparative statistical results are illustrated

TABLE 13. Optimal settings of control variables for IEEE 39-bus test systems with the minimization of VSI objective function.

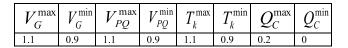
Variable	Base Case (Initial)	BA	CBA-III	CBA-IV
Generator voltage (J		•		
V _{G30}	1.027	0.9445	1.0346	0.9418
<i>V</i> _{G31}	1.023	1.0350	0.9322	1.0642
<i>V</i> _{G32}	1.033	1.0062	1.0853	1.0396
<i>V</i> _{G33}	1.046	0.9684	0.9507	1.0288
<i>V</i> _{G34}	1.026	0.9898	1.0492	1.0595
V _{G35}	1.068	1.0183	1.0566	0.9294
V _{G36}	1.090	0.9685	1.0177	1.0481
V _{G37}	1.031	1.0649	1.0512	0.9604
V _{G38}	1.065	1.0839	1.0534	0.9323
V _{G39}	1.005	1.0428	1.0981	1.0565
Transformer tap rati	o (p.u)	1		
<i>T</i> ₁₂₋₁₁	0	0.9863	1.0483	0.9484
<i>T</i> ₁₀₋₃₂	0	1.0249	0.9787	1.0282
<i>T</i> ₂₂₋₃₅	0	1.0569	1.0299	0.9397
T ₂₋₃₀	0	0.9817	1.0213	0.9326
T_{19-20}	0	0.9496	0.9689	0.9570
Capacitor banks (p.	1)		1	1
Q_{C1}	1.000	0.1522	0.2214	0.0093
Q_{C5}	1.000	0.0773	0.0577	0.2351
Q_{C11}	1.000	0.1522	0.2214	0.0093
$Q_{c_{14}}$	1.000	0.0773	0.0577	0.2351
Q_{C22}	1.000	0.1522	0.2214	0.0093
Q_{C27}	1.000	0.1522	0.2214	0.0093
P _{Loss (MW)}	43.6	56.2912	55.5923	55.1812
TVD (p.u)	-	0.2842	0.2809	0.2776
L - index (p.u)	-	0.0136	0.0130	0.0113

in Table 17. It can be seen from Table 17 that a **22.84** % reduction (from the initial loss of 28.462 MW) in active power loss is achieved with the CBA-IV algorithm, which performs

TABLE 14. Statistical results for IEEE 39-bus test system with the minimization of VSI objective function.

Algorithm	Best (p.u.)	Worst (p.u.)	Mean (p.u.)	Std
CBA	0.0136	0.0161	0.0147	7.3272× 10 ⁻⁴
CBA-III	0.0130	0.0156	0.0140	7.1840× 10 ⁻⁴
CBA-IV	0.0113	0.0142	0.0124	7.8366× 10 ⁻⁴

TABLE 15. Control variable limits for IEEE 57-bus test systems [118].



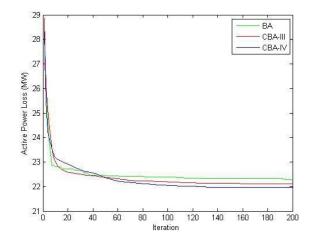


FIGURE 7. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 57-bus test power system with the minimization of P_{Loss} objective function.

better than other algorithms. The comparative con-vergence characteristics of active power loss over 200 iterations yielded by the standard BA, CBA-III and CBA-IV are presented in Fig.7. It can be observed from Fig. 7 that the CBA-IV gives better convergence characteristics than the CBA-III and BA algorithms.

2) CASE 2: MINIMIZATION OF TOTAL VOLTAGE DEVIATION

The best results of TVD minimization for IEEE 57-bus system achieved by the CBA-IV, CBA-III and CBA algorithms are tabulated in Table 18. According to this table, it can be seen that the CBA-IV outperforms other algorithms compared with it as the CBA-IV gives a TVD value of **0.6399** against 0.6413 with CBA-III, 0.6434 with BA, 0.6501 with NGBWCA [127], 0.6634 with ALC-PSO [130] and 0.6982 with OGSA [128]. The statistical results obtained by different methods are presented in Table 19. According to this table, it can be seen that the CBA-IV is able to enhance the TVD value by **48.13%** with respect to initial TVD value, in comparison to 48.01% with CBA-III, 47.84% with BA,

Variable	CBA-IV	CBA-III	BA	ABC [126]
Generator	voltage (p.u)			[120]
V_{G1}	1.0964	1.0919	1.0989	1.0808
V_{G2}	1.0999	1.0905	1.0960	1.0637
V _{G3}	1.0906	1.0903	1.0927	1.0467
V_{G6}	1.0838	1.0841	1.0863	1.0337
V_{G8}	1.1000	1.0902	1.0933	1.0464
V_{G9}	1.0869	1.0886	1.0873	1.0253
<i>V</i> _{<i>G</i>12}	1.0822	1.0825	1.0839	1.0526
	er tap ratio (p.u	 ນ		
T ₄₋₁₈	0.9002	1.0249	0.9071	0.97
T ₄₋₁₈	0.9005	0.9006	0.9067	0.94
T ₂₁₋₂₀	0.9958	1.0481	0.9776	0.97
T ₂₄₋₂₅	1.0086	0.9849	1.0257	0.97
T ₇₋₂₉	0.9061	0.9997	0.9703	0.94
T ₃₄₋₃₂	0.9990	1.0160	1.0192	0.93
T_{11-41}	0.9087	0.9001	0.9620	0.90
T_{15-45}	0.9003	0.9017	0.9101	0.99
<i>T</i> ₁₄₋₄₆	0.9002	0.9030	0.9135	0.96
<i>T</i> ₁₀₋₅₁	0.9123	0.9175	0.9454	0.98
<i>T</i> ₁₃₋₄₉	0.9002	0.9995	0.9007	0.93
<i>T</i> ₁₁₋₄₃	0.9000	0.9738	0.9026	0.92
T ₄₀₋₅₆	1.0267	1.0549	1.0240	0.96
T ₃₉₋₅₇	0.9729	1.0119	0.9742	0.94
T ₉₋₅₅	0.9220	0.9035	0.9853	0.94
	banks (p.u)	1		
Q_{C18}	0.1827	0.0826	0.1053	0.15
Q_{C25}	0.1335	0.0529	0.0573	0.17
Q_{C53}	0.0858	0.1806	0.0853	0.13
P _{Loss}	21.9627	22.0162	22.2716	23.9666
(MW) TVD	1.1661	1.1668	1.1674	-
(p.u)				
L- index (p.u)	0.3508	0.3639	0.3817	-
_	0.3508	0.3639	0.3817	-

TABLE 16. Optimal settings of control variables for IEEE 57-bus testsystems with the minimization of P_{Loss} . Objective function.

Variable	NGBWCA [127]	MICA- IWO [63]	CKHA [59]	PSO-ICA [6]
Generator voltage				
V_{G1}	1.0600	1.06	1.0600	1.0395
V _{G2}	1.0591	1.05841	1.0590	1.0259
V _{G3}	1.0492	1.04568	1.0487	1.077
V_{G6}	1.0399	1.03969	1.0431	0.9982
V_{G8}	1.0586	1.06	1.0600	1.0158
V _{G9}	1.0461	1.02737	1.0447	0.985
<i>V</i> _{<i>G</i>12}	1.0413	1.03499	1.0410	0.9966
Transformer tap ra				
T ₄₋₁₈	0.9712	1.01	0.9179	0.9265
T ₄₋₁₈	0.9243	0.95	1.0256	0.9532
T ₂₁₋₂₀	0.9123	1.02	0.9000	1.0165
T ₂₄₋₂₆	0.9001	1.01	0.9020	1.0071
T ₇₋₂₉	0.9112	0.96	0.9104	0.9414
<i>T</i> ₃₄₋₃₂	0.9004	0.98	0.9005	0.9555
<i>T</i> ₁₁₋₄₁	0.9128	0.9	0.9000	0.9032
T_{15-45}	0.9000	0.95	0.9000	0.9356
<i>T</i> ₁₄₋₄₆	1.0218	0.94	1.0797	0.9172
<i>T</i> ₁₀₋₅₁	0.9902	0.95	0.9887	0.9337
T_{13-49}	0.9568	0.91	0.9914	0.9
<i>T</i> ₁₁₋₄₃	0.9000	0.95	0.9000	0.9206
T ₄₀₋₅₆	0.9000	0.1	0.9002	1.0042
<i>T</i> ₃₉₋₅₇	1.0118	0.97	1.0173	1.0297
<i>T</i> ₉₋₅₅	1.0000	0.96	1.0023	0.9294
Capacitor banks (p		r	T	
$Q_{_{C18}}$	0.0914	0.1	0.0994	0.099846
$Q_{_{C25}}$	0.0587	0.059	0.0590	0.10
Q_{C53}	0.0634	0.063	0.0630	0.10
P _{Loss} (MW)	23.27	24.25684	23.38	25.5856
TVD (p.u)	-	-	-	-
L - index (p.u)	-	-	-	-

TABLE 16. (Continued.) Optimal settings of control variables for IEEE 57-bus test systems with the minimization of P_{Loss}. Objective function.

Variable	SOA [53]	BBO [57]	OGSA [128]	GSA [129]
Generator voltage (p				
V_{G1}	1.06	1'06	1.06	1.060000
V _{G2}	1.0580	1.0504	1.0594	1.060000
V _{G3}	1.0437	1.0440	1.0492	1.060000
V _{G6}	1.0352	1.0376	1.0433	1.008102
<i>V</i> _{<i>G</i>8}	1.0548	1.0550	1.0600	1.054955
V_{G9}	1.0369	1.0229	1.0450	1.009801
<i>V</i> _{<i>G</i>12}	1.0336	1.0323	1.0407	1.018591
Transformer tap ratio				
T ₄₋₁₈	1.00	0.96693	0.9000	1.100000
T ₄₋₁₈	0.96	0.99022	0.9947	1.082634
T ₂₁₋₂₀	1.01	1.0120	0.9000	0.921987
T ₂₄₋₂₆	1.01	1.0087	0.9001	1.016731
T ₇₋₂₉	0.97	0.97074	0.9111	0.996262
T ₃₄₋₃₂	0.97	0.96869	09000	1.100000
<i>T</i> ₁₁₋₄₁	0.90	0.90082	09000	1.074625
<i>T</i> ₁₅₋₄₅	0.97	0.96602	09000	0.954340
<i>T</i> ₁₄₋₄₆	0.95	0.95079	1.0464	0.937722
<i>T</i> ₁₀₋₅₁	0.96	0.96414	0.9875	1.016790
<i>T</i> ₁₃₋₄₉	0.92	0.92462	0.9638	1.052572
<i>T</i> ₁₁₋₄₃	0.96	0.95022	09000	1.100000
T_{40-56}	1.00	0.99666	09000	0.979992
T ₃₉₋₅₇	0.96	0.96289	1.0148	1.024653
T ₉₋₅₅	0.97	0.96001	0.9830	1.037316
Capacitor banks (p.u		·	·	ı
Q_{C18}	0.09984	0.09782	0.0682	0.078254
Q_{C25}	0.05808	0.05904	0.0590	0.005869
Q_{C53}	0.06288	0.06288	0.0630	0.046872
P _{Loss} (MW)	24.26548	24.544	23.43	23.461194
TVD (p.u)			1.1907	-
L — index (p.u)	-	-	0.4120	-

Variable	MFO [4]	ALC-	MGBICA	CPVEIHBMO
Generator	voltage (p.u)	PSO [130]	[131]	[132]
Generator				
V_{G1}	1.06000	1.0600	1.06	1.076
V_{G2}	1.05870	1.0593	1.0492	1.054
V_{G3}	1.04690	1.0491	1.0388	1.035
V_{G6}	1.04210	1.0432	1.0353	1.013
V_{G8}	1.06000	1.0600	1.0558	1.044
V_{G9}	1.04230	1.0451	1.0212	1.093
V_{G12}	1.03730	1.0411	1.0295	0.989
	er tap ratio (p.u)	0.0(11	0.05	1.020
T ₄₋₁₈	0.95011	0.9611	0.95	1.029
T_{4-18}	1.00760	0.9109	1	1.034
<i>T</i> ₂₁₋₂₀	1.00630	0.9000	1.01	0.989
<i>T</i> ₂₄₋₂₆	1.00760	0.9004	1.02	1.016
<i>T</i> ₇₋₂₉	0.97523	0.9106	0.99	0.994
<i>T</i> ₃₄₋₃₂	0.97218	0.9000	0.93	1.100
<i>T</i> ₁₁₋₄₁	0.90000 0.97186	0.9000	0.91	1.072 1.000
T_{15-45}				
<i>T</i> ₁₄₋₄₆	0.95355	1.0275	0.96	0.987
<i>T</i> ₁₀₋₅₁	0.96736	0.9876	0.96	0.933
<i>T</i> ₁₃₋₄₉	0.92788	0.9756	0.92	1.029
<i>T</i> ₁₁₋₄₃	0.96406	0.9000	095	1.0923
T_{40-56}	0.99980	0.9000	1.03	0.996
<i>T</i> ₃₉₋₅₇	0.96060	1.0121	0.98	1.0645
T_{9-55}	0.97899	0.9944	0.99	0.9847
Capacitor l		0.0001	0.04	
Q_{C18}	0.099968	0.0994	0.04	0.0653
$Q_{_{C25}}$	0.059000	0.0590	0.06	0.0084
$Q_{_{C53}}$	0.063000	0.0630	0.03	0.0763
P_{Loss}	24.25293	23.39	24.8863	22.78
(MW) TVD	-	1.2697	1.0283	-
(p.u) L —	-	-	-	-
index (p.u)				

Algorithm	Best	Worst	Mean	Std	%
	(MW)	(MW)	(MW)		Psave
SOA [53]	24.26548	24.28046	24.27078	4.2081E-	14.7443
				005	
MICA-	24.25684	24.28364	24.27561	2.3361E-	14.7746
IWO [63]				004	
ALC-PSO	23.39	23.44	23.41	82×10^{-5}	17.82
[130]					
SR-DE	23.3550	24.1656	23.4392	0.1458	17.94
[133]					
BA	22.030	23.2109	22.6017	0.3482	22.6
CBA-III	22.0162	22.9062	22.4721	0.2735	22.65
CBA-IV	21.9627	22.7961	22.0853	0.2241	22.84

TABLE 17. Statistical results for IEEE 57-bus test system with the minimization of P_{Loss} . Objective function.

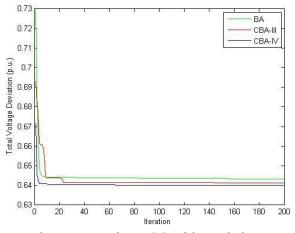


FIGURE 8. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 57-bus test power system with the minimization of TVD objective function.

47.29% with NGBWCA [127], 46.22% with ALC-PSO [130] and 43.40% with OGSA [128].

Moreover, the comparison of the convergence characteristics for TVD obtained by CBA-IV with the other methods is shown in Fig.8.

3) CASE STUDY 3: IMPROVEMENT OF VOLTAGE STABILITY INDEX

Table 20 presents the best results of minimization of the VSI for IEEE 57-bus system by using the CBA-IV algorithm. It is seen that the CBA-IV outperforms other algorithms compared with it, because the CBA-IV provides an L- index value of **0.1608** against 0.1789 with CBA-III, 0.19001 with OGSA [128], 0.190709 with SOA [53] and 0.1917 with BA. Table 21 sums up the statistical results performed by different algorithms for the present case.

Besides, the comparison of the convergence characteristics for VSI yielded by CBA-IV with the other approaches is illustrated in Fig.9. According to this figure, it can be recognized that the CBA-IV convergence characteristic outperforms the CBA-III as well as the BA convergence characteristics.

Variable	CBA-IV	CBA-III	BA	OGSA [128]	NGBWCA [127]
Generator	voltage (p.u)		1	[]	[]
V_{G1}	1.0014	0.9966	1.0013	1.0138	1.0151
V _{G2}	1.0010	0.9983	0.9995	0.9608	0.9810
V _{G3}	1.0002	0.9948	1.0008	1.0173	1.0002
V_{G6}	0.9995	0.9998	0.9996	0.9898	1.0039
V_{G8}	0.9998	1.0026	1.0006	1.0362	1.0198
V_{G9}	1.0003	0.9991	0.9997	1.0241	1.0254
<i>V</i> _{<i>G</i>12}	0.9994	1.0001	0.9991	1.0136	1.0081
	er tap ratio (j	p.u)		L	
T ₄₋₁₈	0.9638	0.9443	0.9524	0.9833	1.0185
T ₄₋₁₈	0.9172	1.0247	1.0919	0.9503	0.9601
<i>T</i> ₂₁₋₂₀	0.9997	0.9874	0.9145	0.9523	0.9458
<i>T</i> ₂₄₋₂₅	0.9418	1.0211	0.9848	1.0036	0.9919
T ₇₋₂₉	1.0699	0.9584	0.9439	0.9778	0.9951
<i>T</i> ₃₄₋₃₂	0.9352	1.0910	1.0324	0.9146	0.9000
<i>T</i> ₁₁₋₄₁	1.0632	1.0773	0.9868	0.9454	0.9622
<i>T</i> ₁₅₋₄₅	1.0644	1.0919	1.0874	0.9265	0.9058
<i>T</i> ₁₄₋₄₆	0.9109	1.0915	1.0031	0.9960	0.9764
<i>T</i> ₁₀₋₅₁	0.9224	0.9388	0.9938	1.0386	1.0600
<i>T</i> ₁₃₋₄₉	1.0108	0.9777	0.9899	0.9060	0.9100
<i>T</i> _{11–43}	0.9982	1.0160	1.0118	0.9234	0.9302
T ₄₀₋₅₆	1.0809	1.0518	1.0523	0.9871	0.9770
<i>T</i> _{39–57}	1.0111	0.9294	1.0352	1.0132	1.0271
<i>T</i> ₉₋₅₅	0.9112	1.0521	0.9454	0.9372	0.9000
	banks (p.u)				
Q_{C18}	0.0992	0.219	0.1491	0.0463	0.0550
Q_{C25}	0.1393	0.998	0.0203	0.0590	0.0590
Q_{C53}	0.0191	0.0865	0.1910	0.0628	0.0381
P_{Loss}	27.0982	28.1872	28.8669	32.34	29.20
(MW)	0.000	0.6412	0.6424	0.000	0.6501
<i>TVD</i> (p.u)	0.6399	0.6413	0.6434	0.6982	0.6501
L-	0.4246	0.4369	0.4472	0.5123	-
index (p.u)					
/	•	•		•	•

TABLE 18. Optimal settings of control variables for IEEE 57-bus test systems with the minimization of TVD Objective function.

Variable	ALC-PSO [130]	MGBICA [131]	СКНА [59]
Generator volt	age (p.u)		
V_{G1}	1.0172	1.0555	10236
V _{G2}	0.9819	1.0339	1.0121
V _{G3}	1.0044	1.0086	1.0030
V _{G6}	1.0050	1.0067	1.0052
V_{G8}	1.0208	1.0462	1.0180
V_{G9}	1.0258	1.0067	1.0427
<i>V</i> _{<i>G</i>12}	1.0080	1.0059	1.0030
Transformer ta			
T ₄₋₁₈	1.0197	0.93	0.9650
T ₄₋₁₈	0.9609	1.01	0.9877
T ₂₁₋₂₀	0.9465	0.98	0.9584
T ₂₄₋₂₆ T ₇₋₂₉	0.9923	1.07	1.0085
<i>T</i> ₇₋₂₉	0.9960	0.96	1.0112
T_{34-32}	0.9000	0.91	0.9000
<i>T</i> ₁₁₋₄₁	0.9622	0.9	0.9784
<i>T</i> ₁₅₋₄₅	0.9059	0.95	0.9000
T_{15-45} T_{14-46}	0.9764	0.95	0.9809
<i>T</i> ₁₀₋₅₁	1.0622	0.98	1.0388
<i>T</i> ₁₃₋₄₉	0.9106	0.94	0.9041
<i>T</i> ₁₁₋₄₃	0.9289	0.97	0.9119
T ₄₀₋₅₆	0.9771	1.04	0.9899
<i>T</i> ₃₉₋₅₇	1.0281	0.93	1.0213
T ₉₋₅₅	0.9001	0.98	0.9078
Capacitor banl	ks (p.u)		
$Q_{_{C18}}$	0.0585	0.03	0.0601
Q_{C25}	0.0587	0.06	0.0590
Q_{C53}	0.0381	0.03	0.0630
P _{Loss} (MW)	29.31	26.4618	28.46
TVD	0.6634	0.77461	0.6484
$(\mathbf{p}.\mathbf{u})$ L - index	-	-	0.1899
(p.u)	1		

TABLE 18. (Continued.) Optimal settings of control variables for IEEE 57-bus test systems with the minimization of TVD Objective function.

Algorithm	Best	Worst	Mean	Std	% TVD
	(p.u.)	(p.u.)	(p.u.)		Improve
OGSA	0.6982	-	-	-	43.40
[128]					
NGBWCA	0.6501	-	-	-	47.29
[127]					
ALC-PSO	0.6634	0.6689	0.6636	89 × 10 ⁻⁵	46.22
[130]					
BA	0.6434	0.6609	0.6499	0.0045	47.84
CBA-III	0.6413	0.6542	0.6440	0.0031	48.01
CBA-IV	0.6399	0.6516	0.6424	0.0028	48.13

TABLE 19. Statistical results for IEEE 57-bus test system with the

minimization of TVD objective function.

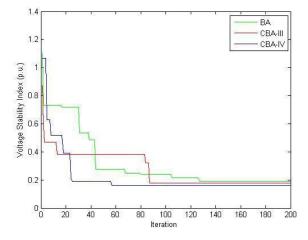


FIGURE 9. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 57-bus test power system with the minimization of VSI objective function.

D. IEEE 118 BUS TEST SYSTEM

The IEEE 118 bus test system comprises 54 generators, 186 transmission lines, 9 tap changing transformers and 14 shunt VAR compensators. Totally, the IEEE 118-bus test system contains 77 control variables. Refer to [53], [63], [115] and [116] for further details.

The total demand of the system are

 $P_{load} = 4242MW$ (active power demand),

 $Q_{load} = 1438MVAR$ (reactive power demand).

The initial total generations and power losses are given by

 $\sum PG = 4375.36MW$ (active power of generators), $\sum QG = 881.92MVAR$ (reactive power of generators),

 $P_{LOSS} = 133.357 MW$ (active power losses),

 $Q_{LOSS} = -785.11 MVAR$ (reactive power losses).

The control variable limits are listed in Table 22 in p.u. [53], [116].

1) CASE 1: MINIMIZATION OF ACTIVE POWER LOSS

The best results achieved by different algorithms for the minimization of the active power loss are tabulated in the Appendix section. The comparative statistical results are

TABLE 20. Optimal settings of control variables for IEEE 57-bus test systems with the minimization of VSI objective function.

Variable	CBA-IV	CBA-III	BA
Generator volta	ge (p.u)		
V_{G1}	1.0981	1.0887	1.0431
V _{G2}	1.0441	1.0492	0.9996
V _{G3}	1.0859	1.0953	1.0452
V _{G6}	1.0235	1.0219	1.0354
V_{G8}	1.0607	1.0609	0.9327
V_{G9}	0.9399	0.9466	0.9936
V _{G12}	0.9541	0.9444	1.0140
Transformer tap			
T ₄₋₁₈	1.0466	1.0442	1.0253
T ₄₋₁₈	1.0263	1.0149	0.9288
T ₂₁₋₂₀	0.9607	0.9540	0.9683
T ₂₄₋₂₅	0.9087	0.9103	1.0959
T ₇₋₂₉	0.9013	0.9047	1.0446
<i>T</i> _{34–32}	1.0804	1.0826	0.9646
<i>T</i> _{11–41}	1.0106	1.0156	1.0371
<i>T</i> ₁₅₋₄₅	0.9982	0.9815	1.0539
<i>T</i> ₁₄₋₄₆	1.0253	1.0222	1.0908
<i>T</i> ₁₀₋₅₁	0.9611	0.9615	0.9209
<i>T</i> ₁₃₋₄₉	1.0233	1.0193	0.9541
<i>T</i> ₁₁₋₄₃	0.9150	0.9235	1.0356
T ₄₀₋₅₆	0.9122	0.9041	0.9183
T ₃₉₋₅₇	0.9790	0.9715	1.0502
T ₉₋₅₅	0.9426	0.9357	0.9616
Capacitor banks			
Q_{C18}	0.1784	0.1818	0.1025
Q_{C25}	0.0391	0.0441	0.1912
Q_{C53}	0.1714	0.1704	0.1685
P _{Loss} (MW)	40.3425	40.982	42.4878
TVD (p.u)	1.2347	1.2641	1.4414
L - index (p.u)	0.1608	0.1789	0.1917

TABLE 21. Statistical results for IEEE 57-bus test system with the minimization of VSI objective function.

Algorithm	Best (p.u.)	Worst (p.u.)	Mean (p.u.)	Std
SOA [53]	0.190709	0.176374	0.187451	2.6388 × 10 ⁻³
OGSA [128]	0.190010	0.175935	0.184112	2.8188×10^{-3}
BA	0.1917	0.2154	0.1991	0.0075
CBA-III	0.1789	0.1898	0.1851	0.0035
CBA- IV	0.1608	0.1712	0.1674	0.0027

TABLE 22. Control variable limits for IEEE 118-bus test systems [53], [116].

V_G^{\max}	V_G^{\min}	V_{PQ}^{\max}	V_{PQ}^{\min}	T_k^{\max}	T_k^{\min}	Q_C^{\max}	$Q_{\scriptscriptstyle C}^{\scriptscriptstyle { m min}}$
1.06	0.94	1.1	0.95	1.1	0.9	0.2	0

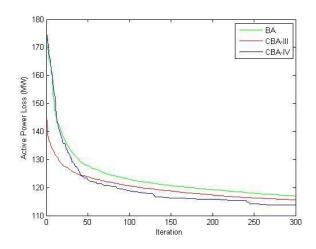


FIGURE 10. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 118-bus test power system with the minimization of P_{Loss} objective function.

presented in Table 23. According to Table 23, the CBA-IV algorithm leads to **113.7040 MW** active power loss which is better than the active power losses obtained by the other algorithms compared with it. In addition, it can also be seen from Table 23 that a **14.74%** reduction (from the initial loss of 133.357 MW) in active power loss is achieved with the CBA-IV algorithm, which performs better than other algorithms compared with it. The comparative convergence characteristics of active power loss over 300 iterations obtained by the standard BA, CBA-III and CBA-IV are shown in Fig.10. It can be observed from Fig.10 that the CBA-IV yields better convergence characteristics than the CBA-III and BA algorithms.

2) CASE 2: MINIMIZATION OF TOTAL VOLTAGE DEVIATION The best results of TVD minimization for IEEE 118-bus system yielded by the CBA-IV, CBA-III and CBA algorithms

TABLE 23. Statistical results for IEEE 118-bus test system with the minimization of P_{Loss} . Objective function.

Algori-	Best	Worst	Mean	Std	%
thm MICA-	(MW) 114.04568	(MW) 114.97562	(MW) 114.44837	2.4288× 10 ⁻⁴	Psave 14.48
IWO	117.07500	117.77502	117.7037		17.70
[63]					
MTLA-	113.9814	114.4975	114.0852	2.8755× 10 ⁻⁴	14.53
[116]	119.7792				9.847
ALO [62]		-	-	-	9.847
OGSA [128]	126.99	131.99	127.14	88×10^{-6}	4.4203
GSA [129]	127.7603	-	-	-	4.2
QOTL- BO [134]	112.2789	115.4516	113.7693	0.0244	15.8
FAHCL- PSO [112]	116.2479	-	-	-	12.83
GWO [61]	120.65	-	-	-	9.53
CLPSO [135]	130.96	132.74	131.15	85 × 10 ⁻⁶	1.8
EMA [3]	126.2243	130.98	127.0111	87.2×10^{-6}	5.35
ALC- PSO [130]	121.53	132.99	123.14	91 × 10 ⁻⁶	8.245
HFA [22]	134.24	134.8499	134.96	0.008814	-
BA	116.9329	119.2345	117.1409	0.2287	12.32
CBA- III	115.5989	116.3561	115.6542	0.1569	13.32
CBA- IV	113.7040	114.1689	114.0108	0.1218	14.74

TABLE 24. Statistical results for IEEE 118-bus test system with the minimization of TVD objective function.

Algorithm	Best	Worst	Mean	Std
	(p.u.)	(p.u.)	(p.u.)	
ALC-PSO	0.3262	0.3743	0.3281	95 × 10 ^{−6}
[130]				
BA	0.3172	0.3206	0.3187	9.7474× 10 ⁻⁴
CBA-III	0.3059	0.3072	0.3062	3.2319×10^{-4}
CBA-IV	0.3032	0.3041	0.3036	3.0558×10^{-4}

are shown in the Appendix section. The statistical results of this case are summarized in Table 24. According this table, it is seen that the CBA-IV outperforms other algorithms compared with it as the CBA-IV gives a TVD value of **0.3032** against 0.3059 with CBA-III, 0.3172 with BA and

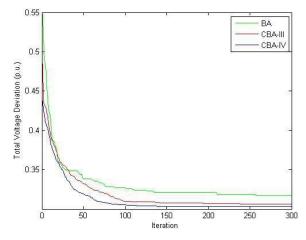


FIGURE 11. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 118-bus test power system with the minimization of TVD objective function.

 TABLE 25. Statistical results for IEEE 118-bus test system with the minimization of VSI objective function.

Algorithm	Best	Worst	Mean	Std
	(p.u.)	(p.u.)	(p.u.)	
QOTLBO [134]	0.0608	0.06311	0.0616	0.0476
BA	0.06126	0.0654	0.0632	0.0010
CBA-III	0.0587	0.0626	0.0607	0.0013
CBA-IV	0.0570	0.0609	0.0587	0.0012

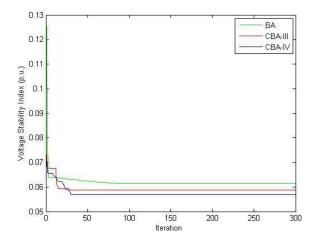


FIGURE 12. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 118-bus test power system with the minimization of VSI objective function.

0.3262 with ALC-PSO [130]. Futhermore, the CBA provides the smallest Best, Mean and Std values in comparison to CBA-III and BA.

The comparison of the convergence characteristics for TVD obtained by CBA-IV with the other methods is depicted in Fig.11 and shows the robust performance of the CBA-IV for larger dimension systems.

TABLE 26. Control variable limits for IEEE 300-bus test systems [62], [136].

V_G^{\max}	V_G^{\min}	V_{PQ}^{\max}	V_{PQ}^{\min}	T_k^{\max}	T_k^{\min}	Q_C^{\max}	Q_C^{\min}
1.1	0.9	1.1	0.9	1.1	0.9	325	0

TABLE 27. Statistical results for IEEE 300-bus test system with the minimization of P_{Loss} objective function.

Algorithm	Best (MW)	Worst (MW)	Mean (MW)	Std
MVMO [137]	385.6284	-	-	-
DEEPSO [137]	394.4343	-	-	-
ALO [62]	384.9224	-	-	-
BA	384.6609	419.145	387.7787	5.0014
CBA-III	379.9265	392.973	385.5485	2.1774
CBA-IV	373.6675	380.065	375.5762	1.6047

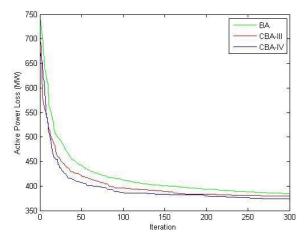


FIGURE 13. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 300-bus test power system with the minimization of P_{Loss} objective function.

 TABLE 28.
 Statistical results for IEEE 300-bus test system with the minimization of TVD objective function.

Algorithm	Best (p.u.)	Worst (p.u.)	Mean (p.u.)	Std
BA	1.1585	1.1712	1.1608	0.0030
CBA-III	1.1462	1.1521	1.1488	9.5050× 10 ⁻⁴
CBA-IV	1.1182	1.1216	1.1233	7.5829× 10 ⁻⁴

3) CASE STUDY 3: IMPROVEMENT OF VOLTAGE STABILITY INDEX

The best results of minimization of the VSI for IEEE i118-bus system achieved by the CBA-IV algorithm are illustrated in

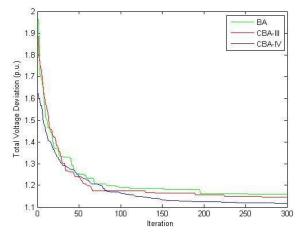


FIGURE 14. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 300-bus test power system with the minimization of TVD objective function.

 TABLE 29. Statistical results for IEEE 300-bus test system with the minimization of VSI objective function.

Algorithm	Best (p.u.)	Worst (p.u.)	Mean (p.u.)	Std
ALO [62]	0.3663	-	-	-
BA	0.3570	0.3611	0.3582	0.0011
CBA-III	0.3556	0.3592	0.3570	9.2311×10 ⁻⁴
CBA-IV	0.3536	0.3568	0.3551	9.6273× 10 ⁻⁴

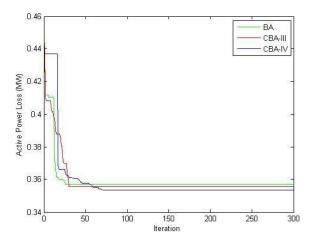


FIGURE 15. The Convergence characteristics of the standard BA, CBA-III and CBA-IV for IEEE 300-bus test power system with the minimization of VSI objective function.

the Appendix section. Table 25 provides the comparison in terms of statistical results for this case. It is observed that the CBA-IV outperforms other algorithms compared with it because the CBA-IV provides an L- index value of **0.0570** against 0.0587 with CBA-III, 0.0608 with QOTLBO [134] and 0.06126 with BA.

TABLE 30. Optimal settings of control variables for IEEE 118-bus test systems with the minimization of P_{Loss} . Objective function.

Variable	MICA- IWO [63]	MTLA- DDE [116]	ALO [62]	OGSA [128]
Generator voltage	(p.u)			
V_{G1}	1.04189	1.04307	1.0164	1.0350
V _{G4}	1.06	1.06	1.0299	1.0554
V _{G6}	1.05423	1.05397	1.0355	1.0301
<i>V</i> _{<i>G</i>8}	1.05817	1.05813	1.0247	1.0175
<i>V</i> _{<i>G</i>10}	1.06	1.06	1.0469	1.0250
<i>V</i> _{<i>G</i>12}	1.0511	1.05108	1.0259	1.0410
<i>V</i> _{<i>G</i>15}	1.04978	1.04993	1.0526	0.9973
V _{G18}	1.05246	1.05213	1.0580	1.0047
<i>V</i> _{<i>G</i>19}	1.04898	1.04921	1.0565	0.9899
<i>V</i> _{<i>G</i>24}	1.05276	1.0523	1.0549	1.0287
V _{G25}	1.06	1.06	1.0600	1.0600
V _{G26}	1.06	1.06	1.0457	1.0855
V _{G27}	1.04656	1.0464	1.0583	1.0081
<i>V</i> _{G31}	1.04299	1.04256	1.0573	0.9948
V _{G32}	1.04526	1.04514	1.0455	0.9993
V _{G34}	1.06	1.06	1.0322	0.9958
V _{G36}	1.05838	1.05826	1.0264	0.9835
V _{G40}	1.03842	1.0387	1.0124	0.9981
<i>V</i> _{<i>G</i>42}	1.03979	1.03968	1.0321	1.0068
V _{G46}	1.04695	1.04665	1.0446	1.0355
V _{G49}	1.06	1.06	1.0572	1.0333
V _{G54}	1.03876	1.03948	1.0313	0.9911
V _{G55}	1.038	1.03848	1.0305	0.9914
V _{G56}	1.03816	1.0378	1.0292	0.9920
V _{G59}	1.06	1.06	1.0269	0.9909
V _{G61}	1.06	1.06	1.0373	1.0747
V _{G62}	1.05593	1.05613	1.0217	1.0753

V_{G65}	1.06	1.06	1.0582	0.9814
V_{G66}	1.06	1.06	1.0591	1.0487
V ₆₆₉	1.06	1.06	1.0600	1.0490
<i>V</i> _{<i>G</i>70}	1.03774	1.03763	1.0577	1.0395
<i>V</i> _{<i>G</i>72}	1.04296	1.0419	1.0592	0.9900
V _{G73}	1.03744	1.03733	1.0348	1.0547
V _{G74}	1.02747	1.02741	1.0533	1.0167
V _{G76}	1.02525	1.02506	1.0382	0.9972
<i>V_{G77}</i>	1.04728	1.0474	1.0395	1.0071
V _{G80}	1.06	1.06	1.0508	1.0066
V _{G85}	1.06	1.06	1.0529	0.9893
V _{G87}	1.06	1.06	1.0510	0.9693
V _{G89}	1.06	1.06	1.0600	1.0527
V _{G90}	1.04449	1.04399	1.0382	1.0290
<i>V</i> _{<i>G</i>91}	1.0496	1.049	1.0223	1.0297
V _{G92}	1.06	1.06	1.0532	1.0353
V _{G99}	1.056	1.05529	1.0447	1.0395
V _{G100}	1.06	1.06	1.0445	1.0275
V _{G103}	1.06	1.05243	1.0385	1.0158
<i>V</i> _{G104}	1.05184	1.04431	1.0218	1.0165
V _{G105}	1.05166	1.04094	1.0376	1.0197
V_{G107}	1.03894	1.02781	1.0285	1.0408
V_{G110}	1.05083	1.03739	1.0458	1.0288
V_{G110}	1.06	1.0452	1.0254	1.0194
V_{G111}	1.03578	1.02184	1.0275	1.0132
	1.05998	1.05891	1.0567	1.0386
<i>V</i> _{<i>G</i>113}	1.06	1.06	1.0577	0.9724
V _{G116}	(m.)			
Transformer tap ration T	1.0	0.99	1.00	0.9568
T ₅₋₈				
<i>T</i> ₂₅₋₂₆	1.1	1.06	0.99	1.0409

<i>T</i> ₁₇₋₃₀		0.99		0.99		1.00	0.9963
T ₃₇₋₃₈		0.98		0.98		1.01	0.9775
T ₅₉₋₆₃		0.98		0.98		1.03	0.9560
T ₆₁₋₆₄		1.0		1.0		1.02	0.9956
T ₆₅₋₆₆		0.9		0.9		0.97	0.9882
T ₆₈₋₆₉		0.95		0.95		0.94	0.9251
T ₈₀₋₈₁		0.99		0.99		1.00	1.0661
Capacitor bank	s (p.u	L)					
Q_{C5}	<u> </u>	-0.4		-0.025	9	-0.19	-0.3319
Q_{C34}		0.14		0.0		0.06	0.0480
Q_{C37}		0.0		-0.000	4	-0.19	-0.2490
Q_{C44}		0.0429)	0.0387	7	0.03	0.0328
Q_{C45}		0.0073	3	0.0236	5	0.06	0.0383
Q_{C46}		0.0697	7	0.0575	5	0.05	0.0545
Q_{C48}		0.0368	3	0.0348	3	0.09	0.0181
Q_{C74}		0.0		0.113		0.07	0.0509
Q_{C79}		0.0		0.0		0.06	0.1104
Q_{C82}		0.0		0.0		0.12	0.0965
Q_{C83}		0.0		0.0001		0.06	0.0263
Q_{C105}		0.0001	l	0.1043	3	0.04	0.0442
Q_{C107}		0.0016	5	0.0004	ł	0.03	0.0085
Q_{C110}		0.0002	2	0.0		0.03	0.0144
P _{Loss} (MW)		114.04	568	113.98	314	119.7792	126.99
TVD (p.u)		-		-		-	1.1829
		_		-		0.0642	0.1400
L - index (p)		A	007	FLDO	EA	HCLPSO	GWO
Variable	GS [12	9]	QO [134	ΓLBO ·]	FA [11		GWO [61]
Generator volta	ge (p	o.u)					
V_{G1}	0.9	600	1.02	54	1.0	0120	1.0204
V_{G4}	0.9	620	1.04	51	1.0	0523	1.0257
V _{G6}	0.9	729	1.03	42	1.0	0666	1.0208

TABLE 30. (Continued.) Optimal settings of control variables for IEEE
118-bus test systems with the minimization of PLoss. Objective function.

V_{G8}	1.0570	1.0817	1.0597	1.0419
<i>V</i> _{<i>G</i>10}	1.0885	1.0886	1.0725	1.0413
<i>V</i> _{<i>G</i>12}	0.9630	1.0277	1.0333	1.0232
<i>V</i> _{<i>G</i>15}	1.0127	1.0245	1.0012	1.0207
<i>V</i> _{<i>G</i>18}	1.0069	1.0284	1.0058	1.0270
V_{G19}	1.0003	1.0224	1.1000	1.0204
V _{G24}	1.0105	1.0487	1.0971	1.0137
V _{G25}	1.0102	1.0821	1.0899	1.0270
V _{G26}	1.0401	1.0995	1.1000	1.0386
V _{G27}	0.9809	1.0368	1.0654	1.0188
V _{G31}	0.9500	1.0313	1.0318	1.0138
V _{G32}	0.9552	1.0380	1.0322	1.0135
V _{G34}	0.9910	1.0333	0.9999	1.0261
V _{G36}	1.0091	1.0321	0.9998	1.0261
V _{G40}	0.9505	1.0145	1.0501	1.0125
V _{G42}	0.9500	1.0181	1.0231	1.0233
V _{G46}	0.9814	1.0391	1.0005	1.0272
V _{G49}	1.0444	1.0448	0.9897	1.0401
V _{G54}	1.0379	1.0183	0.9998	1.0230
V _{G55}	0.9907	1.0194	1.0222	1.0221
V _{G56}	1.0333	1.0188	1.0008	1.0226
V _{G59}	1.0099	1.0341	1.0731	1.0379
<i>V</i> _{G61}	1.0925	1.0276	1.0258	1.0241
V _{G62}	1.0393	1.0289	1.0059	1.0199
V _{G65}	0.9998	1.0973	1.0630	1.0465
V _{G66}	1.0355	1.0581	1.0312	1.0378
V _{G69}	1.1000	1.0746	1.0636	1.0501
V _{G70}	1.0992	1.0289	1.1000	1.0243
0.0			1	1

<i>V</i> _{<i>G</i>72}	1.0014	1.0345	1.0500	1.0187
V _{G73}	1.0111	1.0297	1.0981	1.0397
<i>V</i> _{<i>G</i>74}	1.0476	1.0225	1.0444	1.0170
V _{G76}	1.0211	1.0290	1.0037	1.0080
V _{G77}	1.0187	1.0520	1.0559	1.0192
V _{G80}	1.0462	1.0630	0.9999	1.0329
V _{G85}	1.0491	1.0903	1.0882	1.0224
V _{G87}	1.0426	1.0949	1.0303	1.0361
V _{G89}	1.0955	1.0995	1.0001	1.0558
V _{G90}	1.0417	1.0695	1.0018	1.029
V _{G91}	1.0032	1.0697	1.0298	1.0127
V _{G92}	1.0927	1.0833	1.1005	1.036
V _{G99}	1.0433	1.0530	1.0498	1.0297
<i>V</i> _{G100}	1.0786	1.0600	1.0565	1.036
V _{G103}	1.0266	1.0422	1.0413	1.0232
<i>V</i> _{G104}	0.9808	1.0234	1.0189	1.018
V _{G105}	1.0163	1.0211	1.1000	1.0176
<i>V</i> _{G107}	0.9987	1.0021	1.0222	1.0201
<i>V</i> _{<i>G</i>110}	1.0218	1.0215	1.0115	1.0207
<i>V</i> _{<i>G</i>111}	0.9852	1.0267	1.1000	1.0261
V _{G112}	0.9500	1.0095	1.0500	1.0066
V _{G113}	0.9764	1.0450	1.0099	1.0251
<i>V</i> _{<i>G</i>116}	1.0372	1.0983	1.0500	1.0342
Transformer tap		I	_I	I
T ₅₋₈	1.0659	1.0381	1.0214	1.0208
<i>T</i> ₂₅₋₂₆	0.9534	1.0885	1.0533	1.0279
<i>T</i> ₁₇₋₃₀	0.9328	1.0202	1.0555	1.0323
<i>T</i> ₃₇₋₃₈	1.0884	1.0319	0.9995	1.0209
T ₅₉₋₆₃	1.0579	1.0294	1.0619	1.0091

TABLE 30. (Continued.) Optimal settings of control variables for IEEE
118-bus test systems with the minimization of P_{Loss} . Objective function.

<i>T</i> ₆₁₋₆₄	0.9493	1.0711	1.0318	1.0366
T ₆₅₋₆₆	0.9975	0.9139	1.0490	1.0301
T ₆₈₋₆₉	0.9887	0.9622	0.9660	1.0234
T ₈₀₋₈₁	0.9801	1.0182	0.9732	1.0211
Capacitor bank	s (n 11)			
Q_{C5}	0.00	0.0012	0.003500	-0.3976
Q_{C34}	0.0746	0.0690	0.101922	0.1379
Q_{C37}	0.00	0.0130	0.017500	-0.2473
Q_{C44}	0.0607	0.0987	0.044000	0.099571
Q_{C45}	0.0333	0.0976	0.069894	0.098678
Q_{C46}	0.0651	0.0134	0.071289	0.099186
Q_{C48}	0.0447	0.0740	0.066668	0.1489
Q_{C74}	0.0972	0.1154	0.110952	0.11972
Q_{C79}	0.1425	0.1915	0.150000	0.19649
Q_{C82}	0.1749	0.1733	0.105509	0.1989
Q_{C83}	0.0428	0.0907	0.055540	0.099515
Q_{C105}	0.1204	0.1803	0.151895	0.19968
Q_{C107}	0.0226	0.0007	0.044140	0.059136
Q_{C110}	0.0294	0.0366	0.022310	0.058834
P_{Loss}	127.7603	112.2789	116.2479	120.65
$\frac{(MW)}{TVD}$	-	2.6091	-	-
(p.u) L - index	-	0.0645	-	-
(p.u) Variable	CLPSO [135]	EMA [3]	NGBWCA [127]	ALC- PSO [130]
Generator volta	ge (p.u)			
V_{G1}	1.0332	0.97937	1.0215	1.0218
V_{G4}	1.0550	1.020732	1.0431	1.0432
V_{G6}	0.9754	0.981264	1.0312	1.0224
V_{G8}	0.9669	0.959845	1.0539	1.0543

<i>V</i> _{<i>G</i>10}	0.9811	0.983124	1.0271	1.0901
<i>V</i> _{<i>G</i>12}	1.0092	0.991302	1.0316	1.0325
<i>V</i> _{G15}	0.9787	0.987832	1.0129	1.0140
V _{G18}	1.0799	0.967977	1.0075	1.0080
<i>V</i> _{<i>G</i>19}	1.0805	1.034628	1.0102	1.0104
<i>V</i> _{<i>G</i>24}	1.0286	1.043718	1.0208	1.0200
V _{G25}	1.0307	1.01929	1.0531	1.0551
V _{G26}	0.9877	1.099976	0.9941	0.9932
V _{G27}	1.0157	1.013128	1.0291	1.0288
V _{G31}	0.9615	0.986824	1.0275	1.0288
V _{G32}	0.9851	1.031214	1.0201	1.0248
V _{G34}	1.0157	1.016642	1.0014	1.0362
V _{G36}	1.0849	1.016492	1.0412	1.0407
V _{G40}	0.9830	1.03021	1.0400	1.0391
V _{G42}	1.0516	1.032577	1.0512	1.0507
V _{G46}	0.9754	0.984638	1.0170	1.0171
V _{G49}	0.9838	0.998811	1.0510	1.0492
V ₆₅₄	0.9637	1.021781	1.0392	1.0424
V ₆₅₅	0.9716	0.960354	1.0331	1.0339
V ₆₅₆	1.0250	1.027817	1.0372	1.0393
V _{G59}	1.0003	1.029155	1.0564	1.0585
V _{G61}	1.0771	1.021323	1.0565	1.0569
V ₆₆₂	1.0480	1.051103	1.0489	1.0491
V _{G65}	0.9684	1.03287	1.0435	1.0437
V ₆₆₆	0.9648	1.073126	1.0435	1.0716
V _{G69}	0.9574	1.024818	1.0489	1.0535
V _{G70}	0.9765	1.049563	1.0113	1.0111
<i>V</i> _{<i>G</i>72}	1.0243	1.052822	1.0382	1.0389
<i>V</i> _{<i>G</i>73}	0.9651	1.041061	0.9926	0.9932
	-1			

TABLE 30. (Continued.) Optimal settings of control variables for IEEE 118-bus test systems with the minimization of P_{Loss} . Objective function.

<i>V</i> _{<i>G</i>74}	1.0733	1.016658	0.9934	0.9912
V _{G76}	1.0302	0.99704	1.0324	1.0335
V _{G77}	1.0275	1.048287	1.0185	1.0191
V _{G80}	0.9857	0.988072	1.0021	1.0247
V _{G85}	0.9836	0.999756	1.0312	1.0324
V _{G87}	1.0882	0.974953	1.0212	1.0243
V ₆₈₉	0.9895	1.044042	1.0387	1.0393
V _{G90}	0.9905	1.00067	1.0071	1.0072
V_{G91}	1.0288	1.026417	0.9989	0.9997
V_{G91}	0.9760	1.01533	1.0001	1.0012
V _{G92}	1.0880	1.049947	1.0467	1.0481
	0.9617	1.031294	1.0213	1.0332
V_{G100}	0.9611	1.053139	1.0416	1.0422
V_{G103}	1.0125	1.04564	1.0174	1.0183
<i>V</i> _{G104}	1.0684	1.033351	1.0223	1.0226
<i>V</i> _{G105}	0.9769	1.03335	1.0340	1.0344
<i>V</i> _{G107}	1.0414	1.046059	1.0103	1.0349
<i>V</i> _{<i>G</i>110}	0.9790	1.063805	1.0345	1.0425
V_{G111}	0.9764	1.067452	1.0160	1.0162
V_{G112}				
<i>V</i> _{<i>G</i>113}	0.9721	1.006101	1.0181	1.0188
<i>V</i> _{<i>G</i>116}	1.0330	1.024147	1.0330	1.0331
Transformer tap	ratio (p.u) 1.0045	0.941949	1.0051	1.0065
T ₅₋₈				
T ₂₅₋₂₆	1.0609	1.003201	0.9614	0.9617
T_{17-30}	1.0008	0.983478	0.9961	0.9745
<i>T</i> ₃₇₋₃₈	1.0093	0.923284	0.9523	0.9404
T_{59-63} T_{61-64}	0.9922	1.020838	1.0521	1.0531
T ₆₁₋₆₄	1.0074	0.985547	0.9520	0.9539
T ₆₅₋₆₆	1.0611	0.980426	0.9812	0.9448
			-	

TABLE 30. (Continued.) Optimal settings of control variables for IEEE 118-bus test systems with the minimization of P_{Loss} . Objective function.

V	0.987	5	1.05	59387	1.0	600
V _{G85} V _{G87}	1.021	3	1.04	12988	1.0	599
V _{G89}	1.0069	9	1.087995		1.0	600
V _{G90}	1.029	8	1.06497		1.0	9431
V _{G91}	0.9839		1.06	60584	1.0	496
V _{G92}	1.0021		1.07	7217	1.0	600
V _{G99}	0.9853	3	1.05	5034	1.0	551
V _{G100}	1.028	1	1.04	17629	1.0	584
V _{G103}	0.9802	2	1.03	39985	1.0	442
V _{G104}	1.018	7	1.04	2173	1.0	333
V _{G105}	1.0209	9	1.02	28912	1.0	281
V _{G107}	1.0234	4	1.00	9039	1.0	161
V _{G110}	0.9842	2	1.030944		1.0215	
V _{G111}			1.027332		1.0280	
V _{G112}	0.993		0.995456		1.0042	
<i>V</i> _{<i>G</i>113}	1.02		1.029401		1.0	350
V _{G116}	1.001	5	1.058149		1.0	484
Transformer tap	ratio (p.u)					
T ₅₋₈	1.025:	5	1.01	8418	1.0	1360
T ₂₅₋₂₆	1.0609	1.0032	01	0.9614		0.9617
<i>T</i> ₁₇₋₃₀	1.0008	0.9834	78	0.9961		0.9745
<i>T</i> ₃₇₋₃₈	1.0093	0.9232	84	0.9523		0.9404
T ₅₉₋₆₃	0.9922	1.0208	38	1.0521		1.0531
T ₆₁₋₆₄	1.0074	0.9855	47	0.9520		0.9539
T ₆₅₋₆₆	1.0611	0.9804	0.980426		0.9812	
T ₆₈₋₆₉	0.9307	0.9216	72	0.9510		0.9502
T ₈₀₋₈₁	0.9578	0.9963	89	0.9754		0.9747
Capacitor banks	1	0.0-5	100	0.0		0.0750
Q_{C5}	0.0000	-0.3584		-0.0723		-0.0750
Q_{C34}	0.117135	0.0437	0472	0.0483		0.0677

0.0000	-0.2156	53	-0.2390		-0.2399
0.098932	0.0097		0.0032		0.0038
0.094169	0.07321	17	0.0372		0.0179
	0.06216	.6 0.0624			0.0780
	0.04477	74	0.0172		0.0789
	0.09084	18	0.0013		0.0000
0.148532	0.12324	ł	0.0621		0.0717
0.194270	0.16927	7	0.0463		0.0589
0.069824	0.04266	55	0.0560		0.0561
0.090291	0.20		0.0653		0.0641
0.049926	0.04780)9	0.0072		0.0000
0.022086	0.00202	24	0.0108		0.0110
130.96	126.22		121.47		121.53
1.8525	1.798		1.452		1.4651
0.1461	0.162		-		-
	НВМО	HFA	[22]	Mł	FO[4]
Generator voltage (p.u)					
V _{G1} 0.9926		1.02	6576	1.0	173
1.0108		1.051842		1.0	402
1.0037		1.03	3094	1.0	292
0.9976		1.03	7189	1.0	600
1.0215		1.050354		1.0	374
1.0093		1.020722		1.0	250
1.0075		1.059803		1.0	268
1.0259		1.041598		1.0	298
G18 G19 0.9943		1.025998		1.0	275
1.0179		1.06	7503	1.0	483
1.0177		1.07	9399	1.0	600
0.999		1.09	00031	1.0	600
5 0.999 5 1.0084 7		1.06328 1.		1.0	267
	0.098932 0.094169 0.026719 0.028546 0.005471 0.148532 0.194270 0.0059824 0.002091 0.022086 130.96 1.8525 0.1461 [132] e (p.u) 0.9926 0.9926 1.0108 1.0215 1.0108 1.0037 0.9926 0.9926 0.9926 0.1461 [132] e (p.u) 1.0037 1.00108 1.00175 1.0075 1.0075 1.0259 0.9943 1.0177 0.9993	0.098932 0.0097 0.098932 0.007321 0.094169 0.06216 0.026719 0.06216 0.026719 0.004477 0.005471 0.00982 0.005471 0.00982 0.148532 0.12322 0.194270 0.16927 0.069824 0.04266 0.00202086 0.00202 0.022086 0.00202 0.022086 0.00202 0.022086 126.222 1.8525 1.798 0.1461 0.162 $(PVEIHBMO)$ [132] e (p.u) 1.0108 1.0108 1.0162 1.0108 1.0215 1.0093 1.0215 1.0093 1.0259 1.0259 0.99943 1.0177 0.9999	0.098932 0.0097 0.098932 0.073217 0.026719 0.06216 0.028546 0.0044774 0.005471 0.0090848 0.005471 0.0090845 0.005471 0.0090845 0.005471 0.0090845 0.005471 0.12324 0.194270 0.12324 0.0090291 0.042665 0.0090291 0.0010024 0.022086 0.002024 0.022086 0.002024 130.96 126.22 130.96 126.22 130.96 126.22 130.96 126.22 130.96 126.22 130.96 126.22 130.96 126.22 130.96 1.02 1.0108 1.02 1.0108 1.02 1.0108 1.03 1.0037 1.03 1.0018 1.02 1.0017 1.04 1.00259 1.04 1.0179 1.06 1.0179 1.06 1.0177 <td< td=""><td></td><td>Image: line Image: line <thimage: line<="" th=""> <thimage: line<="" th=""></thimage:></thimage:></td></td<>		Image: line Image: line <thimage: line<="" th=""> <thimage: line<="" th=""></thimage:></thimage:>

$\begin{array}{c c c c c c c c c c c c c c c c c c c $	
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	
V_{G46} 1.0084 0.984532 1.0450	
' G49	
V _{G54} 0.9806 1.022769 1.0284	
V _{G55} 0.9969 1.021025 1.0289	
$V_{G56} \qquad \qquad 0.9881 \qquad 1.059781 \qquad 1.0283$	
V_{G59} 1.0197 1.061729 1.0512	
V _{G61} 0.9956 1.032754 1.0534	
V_{G62} 1.0064 1.023555 1.0506	
V _{G65} 0.9883 1.052105 1.0596	
V_{G66} 1.0101 1.06896 1.0600	
V _{G69} 0.9931 1.06172 1.0600	
V_{G70} 1.0127 1.034243 1.0600	
V_{G72} 1.0145 1.037312 1.0526	
V_{G73} 1.0174 1.03344 1.0600	
V_{G74} 1.0025 1.036665 1.0600	
V _{G76} 0.9842 1.026668 1.0390	
V _{G77} 0.9914 1.060286 1.0502	
V_{G80} 1.0257 1.035531 1.0600	
V_{G85} 0.9876 1.059387 1.0600	
V _{G87} 1.0213 1.042988 1.0599	
V_{G89} 1.0069 1.087995 1.0600	
V _{G90} 1.0298 1.06497 1.0431	
V _{G91} 0.9839 1.060584 1.0496	
V_{G92} 1.0021 1.077217 1.0600	

TABLE 30. (Continued.) Optimal settings of control variables for IEEE
118-bus test systems with the minimization of P_{Loss} . Objective function.

V _{G99}	0.9853	1.05034	1.0551
V _{G100}	1.0281	1.047629	1.0584
V _{G103}	0.9802	1.039985	1.0442
V _{G104}	1.0187	1.042173	1.0333
V _{G105}	1.0209	1.028912	1.0281
	1.0234	1.009039	1.0161
V _{G107}	0.9842	1.030944	1.0215
V_{G110}			
<i>V</i> _{<i>G</i>111}	1	1.027332	1.0280
<i>V</i> _{<i>G</i>112}	0.993	0.995456	1.0042
<i>V</i> _{<i>G</i>113}	1.02	1.029401	1.0350
V _{G116}	1.0016	1.058149	1.0484
Transformer tap rati	L o (p.u)	1	
T_{5-8}	1.0255	1.018418	1.01360
T ₂₅₋₂₆	0.9891	0.910392	1.10000
T ₁₇₋₃₀	0.9932	0.991845	1.00380
T_{37-38}	0.9873	0.989105	0.98263
	0.9868	0.97365	0.98430
T ₅₉₋₆₃	1.0235	0.984136	1.01390
T ₆₁₋₆₄	1.0235	0.984130	1.01390
T ₆₅₋₆₆	1.009	0.975929	1.10000
T ₆₈₋₆₉	1.0075	0.918179	1.10000
T ₈₀₋₈₁	0.9872	0.947921	0.96831
Capacitor banks (p.	1)	1	<u> </u>
Q_{C5}	0	0	0
Q_{C34}	0.060111	0.06780651	0
Q_{C37}	0	0	-0.0003126
Q_{C44}	0.060057	0.0495166	0.10
Q_{C45}	0.030001	0.0478067	0
Q_{C46}	5.9838	0.05927768	0
Q_{C48}	0.03992	0.06336509	0.00000842
Q_{C74}	0.079862	0.03791729	0.00220540
	1	1	1

Q_{C79}	0.139892	0.07138172	0.020
Q_{C82}	0.17992	0.1097975	0
Q_{C83}	0.040009	0.04017512	0.10
Q_{c105}	0.109825	0.07750165	0
$Q_{c_{107}}$	0.020251	0.02947845	0.06
Q_{C110}	0.020272	0.01913037	0.06
P _{Loss} (MW)	124.0983	134.24	116.4254
TVD (p.u)	-	-	-
L - index (p.u)	-	-	-
Variable	BA	CBA-III	CBA-IV
Generator voltage (p	o.u)		
V_{G1}	0.9954	1.0080	0.9810
V_{G4}	1.0238	1.0334	1.0286
V_{G6}	1.0097	1.0234	1.0006
V_{G8}	1.0522	1.0679	1.0995
V _{G10}	1.0679	1.0838	1.0826
V _{G12}	1.0086	1.0195	1.0155
<i>V</i> _{G15}	1.0103	1.0155	1.0245
V _{G18}	1.0133	1.0171	1.0217
V _{G19}	1.0110	1.0128	1.0180
<i>V</i> _{<i>G</i>24}	1.0269	1.0276	1.0281
V _{G25}	1.0632	1.0614	1.0823
V _{G26}	1.0561	1.0886	1.0991
V _{G27}	1.0156	1.0195	1.0404
<i>V</i> _{<i>G</i>31}	1.0038	1.0118	1.0182
V _{G32}	1.0124	1.0159	1.0303
V _{G34}	1.0260	1.0241	1.0446
V _{G36}	1.0245	1.0209	1.0411
V _{G40}	1.0030	0.9990	1.0053
V _{G42}	1.0058	1.0033	1.0090
V _{G46}	1.0277	1.0293	1.0414
1	1		

TABLE 30. (Continued.) Optimal settings of control variables for IEEE118-bus test systems with the minimization of P_{Loss} . Objective function.

r			
V_{G49}	1.0443	1.0459	1.0701
V ₆₅₄	1.0118	1.0174	1.0462
V _{G55}	1.0108	1.0145	1.0460
V _{G56}	1.0114	1.0151	1.0445
V _{G59}	1.0369	1.0399	1.0740
V _{G61}	1.0372	1.0464	1.0620
V _{G61}	1.0341	1.0439	1.0655
	1.0406	1.0768	1.0744
V _{G65}	1.0593	1.0638	1.0925
V _{G66}			
V_{G69}	1.0624	1.0613	1.0701
V _{G70}	1.0256	1.0199	1.0246
<i>V</i> _{<i>G</i>72}	1.0195	1.0145	1.0181
V _{G73}	1.0191	1.0103	1.0147
V _{G74}	1.0158	1.0122	1.0184
V _{G76}	1.0138	1.0134	1.0158
V _{G77}	1.0409	1.0393	1.0579
V _{G80}	1.0545	1.0524	1.0753
V _{G85}	1.0628	1.0554	1.0885
V _{G87}	1.0322	1.0051	1.0730
V _{G89}	1.0843	1.0795	1.1051
V _{G90}	1.0523	1.0454	1.0807
V _{G91}	1.0506	1.0436	1.0896
V _{G91}	1.0654	1.0604	1.1031
V _{G99}	1.0394	1.0346	1.0654
V _{G100}	1.0461	1.0443	1.0804
V _{G103}	1.0311	1.0275	1.0632
	1.0196	1.0142	1.0503
<i>V</i> _{G104}	1.0117	1.0090	1.0546
V _{G105}			
V _{G107}	0.9967	0.9930	1.0229

V_{G110}	0.9996	1.0050	1.0455
<i>V</i> _{<i>G</i>111}	1.0061	1.0095	1.0415
<i>V</i> _{<i>G</i>112}	0.9822	0.9888	1.0052
V _{G113}	1.0219	1.0263	1.0277
V_{G116}	1.0310	1.0679	1.0764
Transformer tap ra	1.0107	1.0145	1.0557
T_{5-8}	1.0107	1.0115	1.0007
T ₂₅₋₂₆	0.9661	0.9709	0.9069
<i>T</i> ₁₇₋₃₀	1.0020	1.0106	0.9964
T ₃₇₋₃₈	0.9959	1.0416	1.0062
T_{59-63}	0.9465	1.0254	0.9642
- 59-63	1.0214	1.0141	1.0350
T ₆₁₋₆₄			
T_{65-66}	0.9808	1.0361	1.0328
T ₆₈₋₆₉	0.9541	0.9533	0.9568
T ₈₀₋₈₁	0.9613	0.9932	0.9515
Capacitor banks (p	.u)	1	1
Q_{C5}	0.1394	0.1259	0.1641
Q_{C34}	0.0991	0.0789	0.1155
Q_{C37}	0.1095	0.0955	0.1031
Q_{C44}	0.0288	0.0457	0.0396
Q_{C45}	0.1329	0.1199	0.1811
Q_{C46}	0.1219	0.1305	0.0700
Q_{C48}	0.1866	0.1494	0.1687
Q_{C74}	0.1405	0.0795	0.899
Q_{C79}	0.1733	0.1167	0.0975
Q_{C82}	0.1227	0.0805	0.0569
Q_{C83}	0.0268	0.0418	0.0748
Q_{C105}	0.0959	0.1031	0.0808
Q_{C107}	0.1013	0.0496	0.1885
Q_{C110}	0.0379	0.1106	0.0745
P _{Loss} (MW)	116.9329	115.5989	113.7040
<i>TVD</i> (p.u)	1.1783	1.17279	1.15726
_	0.0982	0.0975	0.0969
L — index (p.u)			

TABLE 30. (Continued.) Optimal settings of control variables for IEEE 118-bus test systems with the minimization of PLoss. Objective function.

The comparison of the convergence characteristics for VSI yielded by CBA-IV with the other approaches is illustrated in Fig.12 and reveals that the CBA-IV is able to achieve a good performance for larger dimension systems.

E. IEEE 300 BUS TEST SYSTEM

The IEEE 300-bus test system comprises 69 generators, 411 transmission lines, 107 tap changing transformers and 14 shunt VAR compensators. Totally, the IEEE 300-bus test system contains 190 control variables. See [62] and [136] for further details.

The control variable limits are listed in Table 26 in p.u. [62] and [136].

1) CASE 1: MINIMIZATION OF ACTIVE POWER LOSS

Table 27 illustrates the comparative statistical results obtained by different algorithms in minimizing active power loss for IEEE 300 bus system. From Table 27, the CBA-IV algorithm leads to **373.6675 MW** active power loss which is better than the active power losses obtained by the other algorithms compared with it. The comparative convergence characteristics of active power loss over 300 iterations yielded by the standard BA, CBA-III and CBA-IV are presented in Fig.13. It can be observed from Fig. 13 that the CBA-IV gives better convergence characteristics than the CBA-III and BA algorithms.

2) CASE 2: MINIMIZATION OF TOTAL VOLTAGE DEVIATION

The comparative statistical results of TVD minimization for IEEE 300-bus system achieved by the CBA-IV, CBA-III and CBA algorithms are shown in Table 28. It is observed that the CBA-IV outperforms other algorithms compared with because the CBA-IV provides a TVD value of **1.1182** against 1.1462 with CBA-III, 1.1585 with BA. The comparison of the convergence characteristics for TVD yielded by CBA-IV with the other approaches is given in Fig.14 and reveals that the CBA-IV is able to achieve a good performance for larger dimension systems.

3) CASE STUDY 3: IMPROVEMENT OF VOLTAGE STABILITY INDEX

The comparative statistical results of VSI minimization for IEEE 300-bus system yielded by the CBA-IV, CBA-III and CBA algorithms are tabulated in Table 29. According to this table, it is seen that the CBA-IV outperforms other algorithms compared with it as the CBA-IV gives an L- index value of **0.3536** against 0.3556 with CBA-III and 0.3570 with BA.

The comparison of the convergence characteristics for VSI obtained by CBA-IV with the other methods is presented in Fig.15 and shows the robust performance of the CBA-IV for larger dimension systems.

VI. CONCLUSION

In this paper, two variants of chaotic bat algorithm (CBA), namely CBA III and CBA IV have been successfully applied

TABLE 31. Optimal settings of control variables for IEEE 118-bus test systems with the minimization of TVD objective function.

Variable	OGSA [128]	QOTLBO [134]	FAHCLPSO [112]	CLPSO [135]
Generator vo	ltage (p.u)			
V _{G1}	1.0388	1.0022	1.0310	1.0351
V_{G4}	0.9872	0.9883	1.0429	1.0473
V _{G6}	0.9925	0.9938	1.0181	1.0199
V_{G8}	0.9905	0.9723	1.0731	1.0274
V _{G10}	0.9919	1.0104	1.0912	1.0267
<i>V</i> _{<i>G</i>12}	1.0077	1.0128	1.0443	1.0274
V _{G15}	1.0034	1.0151	1.0198	1.0217
V _{G18}	0.9773	0.9570	1.0550	0.9901
<i>V</i> _{<i>G</i>19}	1.0324	1.0350	1.0905	1.0174
V _{G24}	1.0285	0.9980	1.0798	1.0868
V _{G25}	0.9705	1.0155	1.0636	0.9991
V _{G26}	1.0175	1.0164	1.1000	1.0217
V _{G27}	1.0117	1.0251	1.0218	1.0400
<i>V</i> _{<i>G</i>31}	1.0014	0.9993	1.0789	0.9932
<i>V</i> _{G32}	0.9988	0.9863	1.0505	1.0360
<i>V</i> _{G34}	1.0158	0.9959	1.0420	1.0012
<i>V</i> _{G36}	0.9916	1.0039	1.0099	1.0370
V _{G40}	1.0132	1.0090	1.0834	1.0162
<i>V</i> _{G42}	0.9892	1.0025	1.0917	1.0252
<i>V</i> _{G46}	1.0607	1.0580	0.9999	1.0271
V _{G49}	1.0031	0.9926	1.0088	1.0021
<i>V</i> _{<i>G</i>54}	1.0236	1.0254	1.0009	1.0491
<i>V</i> _{<i>G</i>55}	1.0176	1.0212	1.0414	1.0093
V _{G56}	1.0149	1.0194	1.0055	1.0108
V _{G59}	1.0584	1.0342	1.0639	1.0313
<i>V</i> _{<i>G</i>61}	0.9829	1.0070	1.0193	1.0514
<i>V</i> _{<i>G</i>62}	1.0562	0.9665	1.0123	1.0786

 TABLE 31. (Continued.) Optimal settings of control variables for IEEE

 118-bus test systems with the minimization of TVD objective function.

			-	
V_{G65}	0.9724	0.9842	1.0719	0.9792
V_{G66}	1.0020	1.0400	1.0818	1.0790
V_{G69}	0.9827	0.9519	1.0450	1.0492
V_{G70}	0.9997	0.9608	1.0890	0.9908
<i>V</i> _{<i>G</i>72}	1.0123	0.9947	1.0500	0.9838
<i>V</i> _{<i>G</i>73}	0.9960	1.0505	1.0360	0.9979
<i>V</i> _{<i>G</i>74}	1.0232	1.0376	1.0128	1.0449
V _{G76}	1.0015	1.0131	1.0085	1.0403
<i>V</i> _{<i>G</i>77}	1.0124	1.0158	1.0888	1.0570
V _{G80}	1.0226	1.0103	1.0111	1.0627
V _{G85}	1.0117	1.0090	0.9789	0.9997
V ₆₈₇	1.0058	1.0098	0.9955	1.0293
V ₆₈₉	1.0076	1.0077	1.0452	1.0275
V _{G90}	0.9753	0.9547	1.0215	1.0320
<i>V</i> _{<i>G</i>91}	0.9836	1.0980	1.0719	1.0275
V _{G92}	1.0272	0.9973	1.1000	1.0101
V_{G99}	0.9612	0.9511	1.0465	1.0666
<i>V</i> _{<i>G</i>100}	1.0032	1.0425	1.0333	1.0213
<i>V</i> _{<i>G</i>103}	0.9843	0.9520	1.0458	1.0317
<i>V</i> _{G104}	0.9880	1.0996	1.0127	1.0390
<i>V</i> _{<i>G</i>105}	1.0003	0.9811	-	1.0259
V _{G107}	1.0033	1.0777	-	1.0410
<i>V</i> _{<i>G</i>110}	1.0040	1.0195	1.0121	0.9993
<i>V</i> _{<i>G</i>111}	1.0331	0.9518	1.1000	1.0332
<i>V</i> _{<i>G</i>112}	0.9877	1.0994	1.0497	1.0302
<i>V</i> _{<i>G</i>113}	0.9705	0.9999	1.0111	1.0463
<i>V</i> _{G116}	1.0270	0.9896	1.0418	0.9905
Transformer 1	L tap ratio (p.u)			

TABLE 31. (Continued.) Optimal settings of control variables for IEEE 118-bus test systems with the minimization of TVD objective function.

T ₅₋₈	0.9841	0.9311	1.0763	1.0027		
T ₂₅₋₂₆	1.0377	0.9144	1.0339	1.0558		
<i>T</i> ₁₇₋₃₀	0.9573	0.9994	1.0812	1.0114		
<i>T</i> ₃₇₋₃₈	0.9952	0.9975	0.9985	1.0010		
T ₅₉₋₆₃	0.9622	0.9690	1.0435	0.9894		
<i>T</i> ₆₁₋₆₄	1.0320	1.0056	1.0512	1.0236		
T ₆₅₋₆₆	1.0137	1.0834	1.0340	0.9496		
T ₆₈₋₆₉	0.9795	1.0964	0.9797	0.9284		
T ₈₀₋₈₁	0.9985	0.9826	0.9800	0.9661		
Capacitor bar		•		1		
Q_{C5}	-0.2403	-0.3404	-	0		
Q_{C34}	0.0371	0.0731	0.128756	0.076593		
Q_{C37}	-0.0437	-0.2046	0.043150	0		
Q_{C44}	0.0375	0.0983	0.049981	0.027759		
Q_{C45}	0.0400	0.0854	0.081112	0.052321		
Q_{C46}	0.0749	0.0427	0.100292	0.069942		
Q_{C48}	0.0796	0.0084	0.055000	0.063363		
Q_{C74}	0.0883	0.0453	0.108156	0.073078		
Q_{C79}	0.1218	0.0059	0.150000	0.148257		
Q_{C82}	0.0380	0.1982	0.098256	0.133066		
Q_{C83}	0.0627	0.0997	0.050000	0.085155		
Q_{C105}	0.0830	0.0771	0.181298	0.112338		
Q_{C107}	0.0459	0.0210	0.050055	0.022546		
Q_{C110}	0.0221	0.0474	0.044497	0.026128		
P _{Loss}	157.72	205.5978	-	132.06		
(MW) TVD	0.3666	0.1910	0.2218	1.6177		
(p.u)						
L-	0.1562	0.0672	-	0.1210		
index (p.u)						
Variable	NG [12]			A [22]		
Generator vol		'J [130	~J			

TABLE 31. (Continued.) Optimal settings of control variables for IEEE
118-bus test systems with the minimization of TVD objective function.

V_{G1}	1.(0002		1.0011	1.00518
V _{G4}	1.0	1.0202		1.0191	0.9991
V _{G6}	0.9	0.9936).9934	1.01878
V ₆₈	0.9	0.9771).9762	0.95837
V _{G10}	1.(0051		1.0064	0.98375
<i>V</i> _{G12}	1.0	0120		1.0126	0.99357
V _{G15}	0.9	9853	().9865	1.02834
V _{G18}	1.0	0557		1.0560	1.02191
V _{G19}	1.0	0190		1.0188	1.00587
V _{G24}	1.0	0197		1.0202	1.04220
V _{G25}	1.0	0108	-	1.0111	1.02798
V _{G25}	0.9	9954	().9801	1.01574
V _{G27}	1.0	0204		1.0231	1.00351
V _{G31}	0.9	9990	().9994	1.00564
V _{G31}	0.9	9877	().9878	1.02537
V _{G32}	1.0	0211		1.0214	0.99338
V _{G36}	0.9	9656	().9655	0.98535
V G36	1.0	0031		1.0043	1.00575
V _{G40}	1.0	0012		1.0138	1.00147
V _{G42}	1.0	0512		1.0526	1.00842
V_{G49}	1.0	0001		1.0024	0.95483
V_{G54}	1.0	0227		1.0234	1.03193
V _{G55}	1.0	0323		1.0330	1.00171
V _{G55}	1.0	1.0139		1.0146	0.99912
V _{G59}	1.0	1.0084		1.0090	1.00540
V _{G61}	1.0	0001		1.0003	0.98207
Variable		BA		CBA-III	CBA-IV
Generator voltage	(p.u)				
I/		1.0039		0.9941	1.0001
V_{G1}					

TABLE 31. (Continued.) Optimal settings of control variables for IEEE 118-bus test systems with the minimization of TVD objective function.

V	0.9996	1.0022	1.0000
V _{G4}	0.9983	1.0075	0.9999
V_{G6}	0.9983	1.0075	0.9999
V_{G8}	1.0053	0.9943	1.0000
V _{G10}	0.9950	1.0015	1.0002
<i>V</i> _{G12}	0.9989	1.0035	1.0000
<i>V</i> _{G15}	1.0032	1.0011	1.0000
V _{G18}	0.9989	0.9999	1.0000
<i>V</i> _{G19}	1.0028	1.0057	0.9999
<i>V</i> _{<i>G</i>24}	1.0048	1.0011	1.0001
<i>V</i> _{G25}	0.9933	0.9946	0.9999
<i>V</i> _{G26}	0.9962	0.9981	1.0001
<i>V</i> _{<i>G</i>27}	1.0028	0.9994	1.0001
<i>V</i> _{<i>G</i>31}	0.9982	0.9943	1.0000
V _{G32}	0.9978	1.0005	0.9998
V _{G34}	1.0048	1.0002	1.0003
V _{G36}	0.9927	0.9974	1.0000
V_{G40}	1.0020	1.0039	1.0000
<i>V</i> _{G42}	1.0040	1.0029	1.0002
V_{G46}	1.0011	0.9985	1.0000
V _{G49}	1.0061	1.0012	0.9999
V_{G54}	1.0016	0.9988	1.0002
V_{G55}	0.9877	0.9969	0.9997
V_{G56}	0.9920	1.0016	1.0001
<i>V</i> _{<i>G</i>59}	0.9986	0.9932	0.9999
V ₆₆₁	1.0023	1.0038	1.0000
V _{G62}	1.0014	1.0048	0.9999
V _{G65}	1.0059	1.0310	0.9838
V ₆₆₆	0.9688	0.9923	1.0368
V ₆₆₉	1.0047	0.9988	0.9659

 TABLE 31. (Continued.) Optimal settings of control variables for IEEE

 118-bus test systems with the minimization of TVD objective function.

	1 0000	1 00 40	1 0055
V_{G70}	1.0099	1.0848	1.0055
<i>V</i> _{<i>G</i>72}	0.9709	1.0253	1.0143
<i>V</i> _{<i>G</i>73}	1.0139	1.0894	0.9803
<i>V</i> _{<i>G</i>74}	1.0458	1.0299	1.0076
<i>V</i> _{<i>G</i>76}	1.0072	0.9647	1.0152
<i>V</i> _{<i>G</i>77}	0.9602	0.9707	0.9929
V ₆₈₀	0.9904	0.9878	1.0108
V _{G85}	1.0010	0.9717	0.9918
V ₆₈₇	0.9604	1.0024	1.0292
V _{G89}	1.0467	0.9703	1.0435
V _{G90}	0.9876	1.0288	1.0148
<i>V</i> _{<i>G</i>91}	1.0248	0.9998	1.0357
V _{G92}	0.9779	1.0111	0.9963
V ₆₉₉	0.9802	0.9549	1.0097
<i>V</i> _{<i>G</i>100}	1.0475	0.9754	0.9835
V _{G103}	1.0316	1.0358	0.9771
<i>V</i> _{<i>G</i>104}	0.9577	0.9827	1.0016
V _{G105}	0.9573	1.0707	1.0226
V _{G107}	1.0038	1.0561	1.0270
<i>V</i> _{<i>G</i>110}	1.0566	1.0306	0.9864
<i>V</i> _{<i>G</i>111}	0.9962	1.0504	0.9744
<i>V</i> _{<i>G</i>112}	1.0129	0.9506	0.9861
V _{G113}	1.0042	1.0631	1.0189
V _{G116}	1.0220	0.9654	1.0289
Transformer tap ratio (p	 .u)	1	1
T ₅₋₈	1.0512	0.9007	1.0441
T ₂₅₋₂₆	1.0302	1.0953	0.9630
T ₂₅₋₂₆ T ₁₇₋₃₀	1.0855	1.0024	1.0094
T ₃₇₋₃₈	0.9501	0.9140	1.0776

TABLE 31. (Continued.) Optimal settings of control variables for IEEE 118-bus test systems with the minimization of TVD objective function.

		1	
T ₅₉₋₆₃	1.0085	1.0978	0.9270
T ₆₁₋₆₄	1.0220	0.9969	1.0492
T ₆₅₋₆₆	1.0703	0.9990	1.0400
T ₆₈₋₆₉	0.9932	1.0299	0.9521
T ₈₀₋₈₁	1.0385	1.0034	0.9729
Capacitor banks (p.u)			
Q_{C5}	0.1296	0.0320	0.0730
Q_{C34}	0.1271	0.0910	0.1303
Q_{C37}	0.0704	0.1889	0.1174
Q_{C44}	0.1642	0.1814	0.0902
Q_{C45}	0.1692	0.0583	0.1228
Q_{C46}	0.0652	0.0808	0.1283
Q_{C48}	0.1521	0.1373	0.1448
Q_{C74}	0.0397	0.1258	0.0810
Q_{C79}	0.0144	0.1636	0.1032
Q_{C82}	0.0989	0.1230	0.0709
Q_{C83}	0.0163	0.0645	0.0520
Q_{c105}	0.0717	0.1712	0.1013
Q_{C107}	0.0681	0.0382	0.1642
Q_{C110}	0.1969	0.1554	0.0719
P _{Loss (MW)}	156.2585	154.3767	142.1661
TVD (p.u)	0.3172	0.3059	0.3032
L - index (p.u)	0.1419	0.1209	0.1189

to solve the ORPD problem. Their performance as well as their effectiveness were evaluated on standard IEEE 14-bus, IEEE 39 New England bus, standard IEEE 57-bus, standard IEEE 118-bus and standard IEEE 300-bus test systems. The results yielded by CBA variants were compared to those available in the literature along with those of the standard version of BA. The simulation results approve that the CBA-IV outperforms other compared algorithms in terms of effectiveness and robustness for solving the ORPD problem.
 TABLE 32. Optimal Settings of control variables for IEEE 118-bus test systems with the minimization of VSI objective function.

Variable	OGSA [128]	QOTLBO [134]	CLPSO [135]	EMA [3]
Generator vo	ltage (nu)			
Generator vo	nage (p.u)			
V_{G1}	0.9881	0.9889	1.0414	0.95
V_{G4}	0.9924	0.9926	1.0596	1.1
V_{G6}	0.9873	0.9876	1.0940	0.95
V_{G8}	0.9984	0.9989	1.0833	1.1
<i>V</i> _{<i>G</i>10}	1.0138	1.0130	1.0912	0.95
<i>V</i> _{<i>G</i>12}	0.9855	0.9850	1.0691	0.95
<i>V</i> _{<i>G</i>15}	1.0314	1.0311	0.9959	0.95
<i>V</i> _{<i>G</i>18}	1.0373	1.0376	0.9513	1.1
<i>V</i> _{<i>G</i>19}	1.0211	1.0219	0.9607	1.1
<i>V</i> _{<i>G</i>24}	1.0294	1.0296	0.9651	1.1
V _{G25}	1.0515	1.0519	0.9750	1.1
<i>V</i> _{G26}	1.0260	1.0261	1.1000	0.95
<i>V</i> _{<i>G</i>27}	1.0306	1.0377	0.9578	1.1
<i>V</i> _{<i>G</i>31}	1.0307	1.0334	0.9585	0.96542
<i>V</i> _{G32}	1.0601	1.0613	1.0501	1.1
<i>V</i> _{G34}	1.0587	1.0543	1.0858	1.1
<i>V</i> _{G36}	1.0211	1.0209	1.0938	1.1
V _{G40}	1.0356	1.0377	1.1000	0.9601
<i>V</i> _{<i>G</i>42}	1.0734	1.0763	0.9500	1.1
V _{G46}	1.0989	1.0988	0.9906	1.1
V _{G49}	1.0442	1.0402	1.0827	1.09734
<i>V</i> _{G54}	1.0827	1.0877	0.9649	0.95
V _{G55}	1.0943	1.0953	0.9500	1.1
V _{G56}	0.9575	0.9584	1.0221	1.1
V _{G59}	0.9519	0.9510	1.0510	1.1

TABLE 32. (Continued.) Optimal Settings of control variables for IEEE 118-bus test systems with the minimization of VSI objective function.

	1	1		-
V_{G61}	0.9569	0.9587	0.9882	0.95
V _{G62}	1.0869	1.0851	1.0212	0.95
V _{G65}	1.0194	1.0196	1.0987	1.1
V _{G66}	0.9525	0.9505	1.0732	1.1
V _{G69}	0.9553	0.9554	0.9500	1.1
V _{G70}	0.9818	0.9808	0.9672	0.95009
V _{G72}	1.0314	1.0309	0.9966	0.95
	1.0161	1.0142	1.0117	0.95
V_{G73}	1.0002	1.0023	1.0946	1.1
V _{G74}	1.0100	1.0135	0.9652	1.1
<i>V</i> _{<i>G</i>76}	1.0057	1.0047	0.9539	0.95
V _{G77}				
V_{G80}	0.9844	0.9846	1.0628	1.1
V _{G85}	1.0006	1.0016	1.0020	1.1
V _{G87}	0.9520	0.9524	0.9669	0.95
V ₆₈₉	1.0661	1.0665	0.9570	1.1
V _{G90}	1.0152	1.0158	1.0767	0.95
V _{G91}	0.9565	0.9569	1.0681	0.95
V _{G92}	1.0172	1.0173	0.9731	1.1
V _{G99}	0.9802	0.9804	1.0391	1.1
V _{G100}	1.0461	1.0460	1.0824	1.1
<i>V</i> _{G103}	1.0864	1.0869	1.0860	0.95
V _{G104}	0.9825	0.9815	1.0799	1.1
V_{G105}	1.0158	1.0138	0.9607	1.1
V _{G107}	1.0087	1.0067	0.9575	1.1
V _{G110}	1.0538	1.0518	1.0852	1.1
V_{G110}	1.0403	1.0413	0.9746	1.1
	0.9941	0.9929	1.0770	1.1
<i>V</i> _{<i>G</i>112}	1.0455	1.0489	1.0852	1.1
<i>V</i> _{<i>G</i>113}	1.0751	1.0775	0.9510	1.1
<i>V</i> _{<i>G</i>116}		1.0775	0.9510	1.1
Transformer ta	ap ratio (p.u)			

 TABLE 32. (Continued.) Optimal Settings of control variables for IEEE

 118-bus test systems with the minimization of VSI objective function.

T ₅₋₈	0.9621	0.9633	1.0014	0.9
T ₂₅₋₂₆	1.0749	1.0765	1.0770	1.1
<i>T</i> ₁₇₋₃₀	1.0149	1.0125	1.0022	1.1
<i>T</i> ₃₇₋₃₈	0.9910	0.9921	0.9995	0.9
T ₅₉₋₆₃	1.0871	1.0847	0.9863	0.9
T ₆₁₋₆₄	0.9121	0.9172	1.0349	1.1
T ₆₅₋₆₆	1.0637	1.0638	1.1000	0.9
T ₆₈₋₆₉	0.9049	0.9069	0.9819	0.9
T ₈₀₋₈₁	0.9848	0.9898	0.9438	0.9
Capacitor banks	s (p.u)			
Q_{C5}	-0.3071	-0.3080	0	0
Q_{C34}	0.0165	0.0164	0.131777	0
Q_{C37}	-0.1891	-0.1899	0	0.001262
Q_{C44}	0.0814	0.0808	0.098018	0.10
Q_{C45}	0.0777	0.0778	0.046021	0.10
Q_{C46}	0.0454	0.0460	0.049648	0.047775
Q_{C48}	0.0042	0.0038	0.044910	0
Q_{C74}	0.1078	0.1027	0.064931	0
Q_{C79}	0.1361	0.1336	0.167886	0.20
Q_{C82}	0.1391	0.1390	0.159579	0
Q_{C83}	0.0434	0.0438	0.100000	0
Q_{C105}	0.0991	0.0996	0.182425	0
Q_{C107}	0.0520	0.0521	0.043269	0
Q_{C110}	0.0164	0.0165	0.054901	0.046343
P _{Loss} (MW)	295.1122	295.8979	132.08	-
TVD	1.4804	1.4820	2.8863	-
$\frac{(p.u)}{L - index}$	0.0600	0.0608	0.0965	0.04841
(p.u)				

TABLE 32. (Continued.) Optimal Settings of control variables for IEEE 118-bus test systems with the minimization of VSI objective function.

Variable	BA	CBA-III	CBA-IV		
Generator voltage (p.u)					
V_{G1}	1.0346	1.0636	1.0307		
V_{G4}	1.0061	0.9695	1.0197		
V _{G6}	0.9874	1.0193	1.0018		
V_{G8}	1.0351	1.0277	1.0379		
<i>V</i> _{<i>G</i>10}	0.9926	1.0477	0.9986		
<i>V</i> _{<i>G</i>12}	1.0009	0.9525	1.0097		
<i>V</i> _{<i>G</i>15}	1.0012	1.0217	1.0074		
<i>V</i> _{<i>G</i>18}	0.9582	1.0549	0.9739		
<i>V</i> _{<i>G</i>19}	0.9776	1.0540	0.9679		
<i>V</i> _{<i>G</i>24}	1.0271	0.9639	1.0131		
<i>V</i> _{<i>G</i>25}	0.9967	0.9696	1.0157		
<i>V</i> _{<i>G</i>26}	1.0333	1.0025	1.0337		
<i>V</i> _{<i>G</i>27}	1.0079	0.9790	1.0021		
<i>V</i> _{<i>G</i>31}	1.0544	0.9735	1.0539		
<i>V</i> _{G32}	0.9690	0.9611	0.9627		
<i>V</i> _{<i>G</i>34}	0.9595	0.9691	0.9711		
<i>V</i> _{<i>G</i>36}	1.0234	0.9847	1.0185		
V _{G40}	1.0420	09473	1.0371		
<i>V</i> _{<i>G</i>42}	1.0052	0.9996	1.0017		
V _{G46}	0.9516	0.9690	0.9504		
V _{G49}	1.0169	1.0425	0.9996		
V ₆₅₄	1.0360	1.0658	1.0300		
V _{G55}	0.9821	0.9798	0.9785		
V _{G56}	1.0432	0.9736	1.0294		
V _{G59}	1.0030	1.0282	0.9918		
<i>V</i> _{<i>G</i>61}	1.0176	1.0360	0.9987		
V _{G62}	0.9864	1.0094	0.9787		
L	I	1	1		

TABLE 32. (Continued.) Optimal Settings of control variables for IEEE
118-bus test systems with the minimization of VSI objective function.

V_{G65}	0.9995	1.0267	1.0009
		1.0207	1.0009
V_{G66}	0.9912	1.0364	1.0045
V ₆₆₉	1.0059	0.9751	0.9753
V _{G70}	1.0381	0.9957	1.0415
V _{G72}	0.9606	0.9541	0.9646
V _{G73}	0.9990	1.0645	1.0118
<i>V</i> _{<i>G</i>74}	1.0321	0.9758	1.0159
<i>V</i> _{<i>G</i>76}	0.9744	0.9957	0.9907
<i>V</i> _{<i>G</i>77}	0.9580	1.0089	0.9541
V _{G80}	0.9577	1.0054	0.9640
V _{G85}	0.9649	1.0273	0.9645
V _{G87}	1.0111	1.0259	0.9907
V _{G89}	1.0026	0.9537	1.0053
V _{G90}	1.0028	0.9496	0.9952
V _{G91}	0.9624	1.0101	0.9560
V _{G92}	1.0030	0.9599	1.0031
V _{G99}	1.0008	0.9899	1.0058
<i>V</i> _{G100}	1.0346	0.9544	1.0212
V _{G103}	1.0087	1.0473	1.0221
V_{G104}	1.0076	0.9538	1.0107
V _{G105}	1.0140	0.9989	1.0156
V_{G105}	1.0481	0.9722	1.0488
V_{G110}	1.0409	0.9721	1.0390
V_{G110}	0.9650	1.0010	0.9646
V_{G111}	1.0046	1.0003	0.9859
	1.0157	0.9872	1.0088
<i>V</i> _{G113}	1.0201	1.0024	1.0102
<i>V</i> _{G116}			
Transformer tap	ratio (p.u) 0.9613	0.9712	0.9633
T		···· / ···	
T_{5-8}			

TABLE 32. (Continued.) Optimal Settings of control variables for IEEE 118-bus test systems with the minimization of VSI objective function.

<i>T</i> ₁₇₋₃₀	0.9602	0.9571	0.9839
T ₃₇₋₃₈	1.0178	0.9769	1.0315
T ₅₉₋₆₃	0.9689	1.0313	0.9762
T ₆₁₋₆₄	0.9668	1.0599	0.9681
T ₆₅₋₆₆	0.9961	0.9244	1.0103
T_{68-69}	0.9618	1.0550	0.9323
T_{80-81}	1.0725	1.0062	1.0543
² 80–81 Capacitor banks (p.	 4)		
Q_{C5}	0.1299	0.0840	0.1155
Q_{C34}	0.1616	0.1243	0.1551
Q_{C37}	0.1689	0.1458	0.1724
Q_{C44}	0.0713	0.1367	0.0786
Q_{C45}	0.1236	0.0534	0.1032
Q_{C46}	0.1124	0.1325	0.1321
Q_{C48}	0.0963	0.1764	0.1125
Q_{C74}	0.1763	0.0316	0.1706
Q_{C79}	0.0547	0.0774	0.0539
Q_{C82}	0.0493	0.0998	0.0631
Q_{C83}	0.0990	0.1244	0.0723
Q_{C105}	0.1575	0.1243	0.1268
Q_{C107}	0.1553	0.0526	0.1564
Q_{C110}	0.0591	0.1615	0.0573
P _{Loss} (MW)	184.0111	178.7326	169.5624
<i>TVD</i> (p.u)	1.4904	1.4642	1.4538
L - index (p.u)	0.06126	0.0587	0.0570

For future researches, CBA will be utilized to address some other power system optimization problems such as the ORPD problem considering FACTS devices, optimum capacitor placement, optimum VAR sizing and allocation, design of power system stabilizers, short-term electricity load forecasting, short-term wind power forecasting, etc.

APPENDIX

See Tables 30–32.

REFERENCES

- J. Carpentier, "Contribution à l'étude du dispatching économique," Bull. de la Société Française des Electriciens, vol. 3, pp. 431–447, Aug. 1962.
- H. W. Dommel and W. F. Tinney, "Optimal power flow solutions," *IEEE Trans. Power Appar. Syst.*, vol. PAS-87, no. 10, pp. 1866–1876, Oct. 1968.
 A. Paine and T. Malalen, "Englang medical elemidar based entities."
- [3] A. Rajan and T. Malakar, "Exchange market algorithm based optimum reactive power dispatch," *Appl. Soft Comput.*, vol. 43, pp. 320–336, Jun. 2016.
- [4] R. Ng Shin Mei, M. H. Sulaiman, Z. Mustaffa, and H. Daniyal, "Optimal reactive power dispatch solution by loss minimization using mothflame optimization technique," *Appl. Soft Comput.*, vol. 59, pp. 210–222, Oct. 2017.
- [5] H. Sharifzadeh and M. Jazaeri, "Optimal reactive power dispatch based on particle swarm optimization considering FACTS devices," in *Proc. Int. Conf. Sustain. Power Gener. Supply*, Apr. 2009, pp. 1–7.
- [6] M. Mehdinejad, B. Mohammadi-Ivatloo, R. Dadashzadeh-Bonab, and K. Zare, "Solution of optimal reactive power dispatch of power systems using hybrid particle swarm optimization and imperialist competitive algorithms," *Int. J. Electr. Power Energy Syst.*, vol. 83, pp. 104–116, Dec. 2016.
- [7] K. L. Lo and S. P. Zhu, "A decoupled quadratic programming approach for optimal power dispatch," *Electr. Power Syst. Res.*, vol. 22, no. 1, pp. 47–60, Sep. 1991.
- [8] N. Grudinin, "Reactive power optimization using successive quadratic programming method," *IEEE Trans. Power Syst.*, vol. 13, no. 4, pp. 1219–1225, Nov. 1998.
- [9] V. H. Quintana and M. Santos-Nieto, "Reactive-power dispatch by successive quadratic programming," *IEEE Power Eng. Rev.*, vol. 4, no. 3, pp. 425–435, Sep. 1989.
- [10] J. A. Momoh, S. X. Guo, E. C. Ogbuobiri, and R. Adapa, "The quadratic interior point method solving power system optimization problems," *IEEE Trans. Power Syst.*, vol. 9, no. 3, pp. 1327–1336, Aug. 1994.
- [11] K. Aoki, M. Fan, and A. Nishikori, "Optimal VAr planning by approximation method for recursive mixed-integer linear programming," *IEEE Trans. Power Syst.*, vol. 3, no. 4, pp. 1741–1747, Nov. 1988.
- [12] S. Granville, "Optimal reactive dispatch through interior point methods," *IEEE Trans. Power Syst.*, vol. 9, no. 1, pp. 136–146, Feb. 1994.
- [13] W. Yan, J. Yu, D. C. Yu, and K. Bhattarai, "A new optimal reactive power flow model in rectangular form and its solution by predictor corrector primal dual interior point method," *IEEE Trans. Power Syst.*, vol. 21, no. 1, pp. 61–67, Feb. 2006.
- [14] J. Z. Zhu and X. F. Xiong, "Optimal reactive power control using modified interior point method," *Electr. Power Syst. Res.*, vol. 66, no. 2, pp. 187–192, Aug. 2003.
- [15] M. Bjelogrlic, M. S. Calovic, P. Ristanovic, and B. S. Babic, "Application of Newton's optimal power flow in voltage/reactive power control," *IEEE Trans. Power Syst.*, vol. 5, no. 4, pp. 1447–1454, Nov. 1990.
- [16] N. I. Deeb and S. M. Shahidehpour, "An efficient technique for reactive power dispatch using a revised linear programming approach," *Electr. Power Syst. Res.*, vol. 15, no. 2, pp. 121–134, Oct. 1988.
- [17] D. S. Kirschen and H. P. Van Meeteren, "MW/voltage control in a linear programming based optimal power flow," *IEEE Trans. Power Syst.*, vol. 3, no. 2, pp. 481–489, May 1988.
- [18] S. Sachdeva and R. Billinton, "Optimum network var planning by nonlinear programming," *IEEE Trans. Power App. Syst.*, vol. PAS-92, no. 4, pp. 1217–1225, Jul. 1973.
- [19] F.-C. Lu, "Reactive power/voltage control in a distribution substation using dynamic programming," *IEE Proc. Gener., Transmiss. Distrib.*, vol. 142, no. 6, pp. 639–645, 1995.
- [20] D. C. Yu, J. E. Fagan, B. Foote, and A. A. Aly, "An optimal load flow study by the generalized reduced gradient approach," *Electr. Power Syst. Res.*, vol. 10, no. 1, pp. 47–53, Jan. 1986.
- [21] N. Deeb and S. M. Shahidehpour, "Linear reactive power optimization in a large power network using the decomposition approach," *IEEE Trans. Power Syst.*, vol. 5, no. 2, pp. 428–438, May 1990.
- [22] A. Rajan and T. Malakar, "Optimal reactive power dispatch using hybrid Nelder-Mead simplex based firefly algorithm," Int. J. Electr. Power Energy Syst., vol. 66, pp. 9–24, Mar. 2015.
- [23] S. Mirjalili, S. M. Mirjalili, and A. Lewis, "Grey wolf optimizer," Adv. Eng. Softw., vol. 69, pp. 46–61, Mar. 2014.

- [24] G. Chen, L. Liu, Z. Zhang, and S. Huang, "Optimal reactive power dispatch by improved GSA-based algorithm with the novel strategies to handle constraints," *Appl. Soft Comput.*, vol. 50, pp. 58–70, Jan. 2017.
- [25] P. Jangir, S. A. Parmar, I. N. Trivedi, and R. H. Bhesdadiya, "A novel hybrid Particle Swarm Optimizer with multi verse optimizer for global numerical optimization and Optimal Reactive Power Dispatch problem," *Eng. Sci. Technol., Int. J.*, vol. 20, no. 2, pp. 570–586, Apr. 2017.
- [26] J. H. Holland, "Genetic algorithms and the optimal allocation of trials," SIAM J. Comput., vol. 2, no. 2, pp. 88–105, Jun. 1973.
- [27] R. Storn and K. Price, "Differential evolution: A simple and efficient heuristic for global optimization over continuous spaces," J. Global Optim., vol. 11, no. 4, pp. 341–359, 1997.
- [28] J. Kennedy and R. C. Eberhart, "Particle swarm optimization," in *Proc. IEEE Int. Conf. Neural Netw*, vol. 4, Nov./Dec. 1995, pp. 1942–1948.
- [29] X. S. Yang and X. He, "Bat algorithm: Literature review and applications," Int. J. Bio-Inspired Comput., vol. 5, no. 3, pp. 141–149, 2013.
- [30] X.-S. Yang, "Firefly algorithm," in *Engineering Optimization*. Hoboken, NJ, USA: Wiley, 2010, pp. 221–230.
- [31] X.-S. Yang, "Flower pollination algorithm for global optimization," in Unconventional Computation and Natural Computation. Berlin, Germany: Springer, 2012, pp. 240–249.
- [32] Z. Woo Geem, J. Hoon Kim, and G. V. Loganathan, "A new heuristic optimization algorithm: Harmony search," *Simulation*, vol. 76, no. 2, pp. 60–68, Feb. 2001.
- [33] S. Kirkpatrick, C. D. Gelatt, and M. P. Vecchi, "Optimization by simulated annealing," *Science*, vol. 220, no. 4598, pp. 671–680, 1983.
- [34] E. Rashedi, H. Nezamabadi-pour, and S. Saryazdi, "GSA: A gravitational search algorithm," *Inf. Sci.*, vol. 179, no. 13, pp. 2232–2248, Jun. 2009.
- [35] M. Dorigo, M. Birattari, and T. Stutzle, "Ant colony optimization," *IEEE Comput. Intell. Mag.*, vol. 1, no. 4, pp. 28–39, Nov. 2006.
- [36] R. Tang, S. Fong, X.-S. Yang, and S. Deb, "Wolf search algorithm with ephemeral memory," in *Proc. 7th Int. Conf. Digit. Inf. Manage. (ICDIM)*, Aug. 2012, pp. 165–172.
- [37] S. Mirjalili and A. Lewis, "The whale optimization algorithm," Adv. Eng. Softw., vol. 95, pp. 51–67, May 2016.
- [38] S. Deb, S. Fong, and Z. Tian, "Elephant search algorithm for optimization problems," in *Proc. 10th Int. Conf. Digit. Inf. Manage. (ICDIM)*, Oct. 2015, pp. 249–255.
- [39] S. Mirjalili, "The ant lion optimizer," Adv. Eng. Softw., vol. 83, pp. 80–98, May 2015.
- [40] S. Mirjalili, A. H. Gandomi, S. Z. Mirjalili, S. Saremi, H. Faris, and S. M. Mirjalili, "Salp swarm algorithm: A bio-inspired optimizer for engineering design problems," *Adv. Eng. Softw.*, vol. 114, pp. 163–191, Dec. 2017.
- [41] S. Mirjalili, "Dragonfly algorithm: A new meta-heuristic optimization technique for solving single-objective, discrete, and multi-objective problems," *Neural Comput. Appl.*, vol. 27, no. 4, pp. 1053–1073, May 2016.
- [42] S. Mirjalili, "Moth-flame optimization algorithm: A novel natureinspired heuristic paradigm," *Knowl.-Based Syst.*, vol. 89, pp. 228–249, Nov. 2015.
- [43] A. A. Heidari, S. Mirjalili, H. Faris, I. Aljarah, M. Mafarja, and H. Chen, "Harris hawks optimization: Algorithm and applications," *Future Gener. Comput. Syst.*, vol. 97, pp. 849–872, Aug. 2019.
- [44] S. Saremi, S. Mirjalili, and A. Lewis, "Grasshopper optimisation algorithm: Theory and application," *Adv. Eng. Softw.*, vol. 105, pp. 30–47, Mar. 2017.
- [45] S. Mirjalili, S. M. Mirjalili, and A. Hatamlou, "Multi-verse optimizer: A nature-inspired algorithm for global optimization," *Neural Comput. Appl.*, vol. 27, no. 2, pp. 495–513, Feb. 2016.
- [46] S. Mirjalili, "SCA: A sine cosine algorithm for solving optimization problems," *Knowl.-Based Syst.*, vol. 96, pp. 120–133, Mar. 2016.
- [47] Q. H. Wu, Y. J. Cao, and J. Y. Wen, "Optimal reactive power dispatch using an adaptive genetic algorithm," *Int. J. Electr. Power Energy Syst.*, vol. 20, no. 8, pp. 563–569, Nov. 1998.
- [48] D. Devaraj, "Improved genetic algorithm for multi-objective reactive power dispatch problem," *Eur. Trans. Electr. Power*, vol. 17, no. 6, pp. 569–581, Nov. 2007.
- [49] P. Subbaraj and P. N. Rajnarayanan, "Optimal reactive power dispatch using self-adaptive real coded genetic algorithm," *Electr. Power Syst. Res.*, vol. 79, no. 2, pp. 374–381, Feb. 2009.
- [50] H. Yoshida, K. Kawata, Y. Fukuyama, S. Takayama, and Y. Nakanishi, "A particle swarm optimization for reactive power and voltage control considering voltage security assessment," *IEEE Trans. Power Syst.*, vol. 15, no. 4, pp. 1232–1239, Nov. 2000.

- [51] A. A. A. Esmin, G. Lambert-Torres, and A. C. Z. de Souza, "A hybrid particle swarm optimization applied to loss power minimization," *IEEE Trans. Power Syst.*, vol. 20, no. 2, pp. 859–866, May 2005.
- [52] M. Varadarajan and K. S. Swarup, "Differential evolution approach for optimal reactive power dispatch," *Appl. Soft Comput.*, vol. 8, no. 4, pp. 1549–1561, Sep. 2008.
- [53] C. Dai, W. Chen, Y. Zhu, and X. Zhang, "Seeker optimization algorithm for optimal reactive power dispatch," *IEEE Trans. Power Syst.*, vol. 24, no. 3, pp. 1218–1231, Aug. 2009.
- [54] G. Chen, L. Liu, and S. Huang, "Enhanced GSA-based optimization for minimization of power losses in power system," *Math. Problems Eng.*, vol. 2015, pp. 1–13, Dec. 2015.
- [55] M. R. Babu and M. Lakshmi, "Gravitational search algorithm based approach for optimal reactive power dispatch," in *Proc. 2nd Int. Conf. Sci. Technol. Eng. Manage. (ICONSTEM)*, Mar. 2016, pp. 360–366.
- [56] N. Sinsuphan, U. Leeton, and T. Kulworawanichpong, "Optimal power flow solution using improved harmony search method," *Appl. Soft Comput.*, vol. 13, no. 5, pp. 2364–2374, May 2013.
- [57] A. Bhattacharya and P. K. Chattopadhyay, "Solution of optimal reactive power flow using biogeography-based optimization," *Int. J. Energy Power Eng.*, vol. 3, pp. 269–277, Nov. 2010.
- [58] M. Ghasemi, M. Taghizadeh, S. Ghavidel, J. Aghaei, and A. Abbasian, "Solving optimal reactive power dispatch problem using a novel teaching-learning-based optimization algorithm," *Eng. Appl. Artif. Intell.*, vol. 39, no. 3, pp. 100–108, Mar. 2015.
- [59] A. Mukherjee and V. Mukherjee, "Solution of optimal reactive power dispatch by chaotic krill herd algorithm," *IET Gener, Transmiss. Distrib.*, vol. 9, no. 15, pp. 2351–2362, Nov. 2015.
- [60] M. Tripathy and S. Mishra, "Bacteria foraging-based solution to optimize both real power loss and voltage stability limit," *IEEE Trans. Power Syst.*, vol. 22, no. 1, pp. 240–248, Feb. 2007.
- [61] M. H. Sulaiman, Z. Mustaffa, M. R. Mohamed, and O. Aliman, "Using the gray wolf optimizer for solving optimal reactive power dispatch problem," *Appl. Soft Comput.*, vol. 32, pp. 286–292, Jul. 2015.
- [62] S. Mouassa, T. Bouktir, and A. Salhi, "Ant lion optimizer for solving optimal reactive power dispatch problem in power systems," *Eng. Sci. Technol., Int. J.*, vol. 20, no. 3, pp. 885–895, Jun. 2017.
- [63] M. R. Nayak, K. R. Krishnanand, and P. K. Rout, "Optimal reactive power dispatch based on adaptive invasive weed optimization algorithm," in *Proc. IEEE Int. Conf. Energy Autom. Signal*, Dec. 2011, pp. 1–7.
- [64] E. Agbugba, "Hybridization of particle swarm optimization with bat algorithm for optimal reactive power dispatch," M.S. thesis, Dept. Electr. Eng., Univ. South Afr., Pretoria, South Africa, Jun. 2017.
- [65] M. H. Sulaiman and Z. Mustaffa, "Cuckoo search algorithm as an optimizer for optimal reactive power dispatch problems," in *Proc. 3rd Int. Conf. Control, Autom. Robot. (ICCAR)*, Apr. 2017, pp. 735–739.
- [66] X.-S. Yang, "A new metaheuristic bat-inspired algorithm," in *Nature Inspired Cooperative Strategies for Optimization (NICSO)*, vol. 284. Berlin, Germany: Springer, 2010, pp. 65–74.
- [67] X. Yang and A. Hossein Gandomi, "Bat algorithm: A novel approach for global engineering optimization," *Eng. Computations*, vol. 29, no. 5, pp. 464–483, Jul. 2012.
- [68] A. Kaveh and P. Zakian, "Enhanced bat algorithm for optimal design of skeletal structures," Asian J. Civil Eng., vol. 15, no. 2, pp. 179–212, 2014.
- [69] X. S. Yang, "Bat algorithm for multi-objective optimisation," Int. J. Bio-Inspired Comput., vol. 3, no. 5, pp. 267–274, 2011.
- [70] J. T. Jose, "Economic load dispatch including wind power using Bat algorithm," in *Proc. Int. Conf. Adv. Electr. Eng. (ICAEE)*, Vellore, India, Jan. 2014, pp. 1–4.
- [71] X.-S. Yang, M. Karamanoglu, and S. Fong, "Bat algorithm for topology optimization in microelectronic applications," in *Proc. 1st Int. Conf. Future Gener. Commun. Technol.*, London, U.K., Dec. 2012, pp. 150–155.
- [72] P. Musikapun and P. Pongcharoen, "Solving multi-stage multi-machine multi-product scheduling problem using bat algorithm," in *Proc. 2nd Int. Conf. Manage. Artif. Intell. (IPEDR)*, vol. 35. Singapore: IACSIT Press, 2012, pp. 98–102.
- [73] J. H. Lin, C. W. Chou, C. H. Yang, and H. L. Tsai, "chaotic Lévy flight bat algorithm for parameter estimation in nonlinear dynamic biological systems," J. Comput. Inf. Technol., vol. 2, no. 2, pp. 56–63, 2012.
- [74] J. W. Zhang and G. G. Wang, "Image matching using a bat algorithm with mutation," *Appl. Mech. Mater.*, vol. 203, no. 1, pp. 88–93, Oct. 2012.

- [75] G. Komarasamy and A. Wahi, "An optimized K-means clustering technique using bat algorithm," *Eur. J. Sci. Res.*, vol. 84, no. 2, pp. 263–273, 2012.
- [76] K. Khan and A. Sahai, "A fuzzy C-means bi-sonar-based Metaheuristic optimization algorithm," *Int. J. Interact. Multimedia Artif. Intell.*, vol. 1, no. 7, pp. 26–32, 2012.
- [77] K. Khan, A. Nikov, and A. Sahai, "A fuzzy bat clustering method for ergonomic screening of office workplaces," in *Proc. 3rd Int. Conf. Softw. Services Semantic Technol. (ST)*, 2011, vol. 101, no. 1, pp. 59–66.
- [78] K. Khan and A. Sahai, "A comparison of BA, GA, PSO, BP and LM for training feed forward neural networks in e-Learning context," *Int. J. Intell. Syst. Appl.*, vol. 4, no. 7, pp. 23–29, Jun. 2012.
- [79] S. Mishra, K. Shaw, and D. Mishra, "A new meta-heuristic bat inspired classification approach for microarray data," *Procedia Technol.*, vol. 4, no. 1, pp. 802–806, 2012.
- [80] A. Natarajan, S. Subramanian, and K. Premalatha, "A comparative study of cuckoo search and bat algorithm for Bloom filter optimisation in spam filtering," *Int. J. Bio-Inspired Comput.*, vol. 4, no. 2, pp. 89–99, 2012.
- [81] R. Damodaram and M. L. Valarmathi, "Phishing Website detection and optimization using modified bat algorithm," *Int. J. Eng. Res. Appl.*, vol. 2, no. 1, pp. 870–876, 2012.
- [82] R. Y. M. Nakamura, L. A. M. Pereira, K. A. Costa, D. Rodrigues, J. P. Papa, and X.-S. Yang, "BBA: A binary bat algorithm for feature selection," in *Proc. 25th SIBGRAPI Conf. Graph., Patterns Images*, Aug. 2012, pp. 291–297.
- [83] J. Xie, Y. Zhou, and H. Chen, "A novel bat algorithm based on differential operator and Lévy flights trajectory," *Comput. Intell. Neurosci.*, vol. 2013, pp. 1–13, Mar. 2013.
- [84] E. Osaba, X.-S. Yang, F. Diaz, P. Lopez-Garcia, and R. Carballedo, "An improved discrete bat algorithm for symmetric and asymmetric traveling salesman problems," *Eng. Appl. Artif. Intell.*, vol. 48, pp. 59–71, Feb. 2016.
- [85] B. Zhu, W. Zhu, Z. Liu, Q. Duan, and L. Cao, "A novel quantumbehaved bat algorithm with mean best position directed for numerical optimization," *Comput. Intell. Neurosci.*, vol. 2016, pp. 1–17, 2016.
- [86] F. P. Mahdi, P. Vasant, M. Abdullah-Al-Wadud, V. Kallimani, and J. Watada, "Quantum-behaved bat algorithm for many-objective combined economic emission dispatch problem using cubic criterion function," *Neural Comput. Appl.*, vol. 31, no. 10, pp. 5857–5869, Oct. 2019.
- [87] F. P. Mahdi, P. Vasant, M. Abdullah-Al-Wadud, J. Watada, and V. Kallimani, "Quantum-behaved bat algorithm for combined economic emission dispatch problem with valve-point effect," in *Recent Advances in Electrical Engineering and Related Sciences: Theory and Application—AETA* (Lecture Notes in Electrical Engineering), vol. 465. Cham, Switzerland: Springer, 2018, pp. 923–933.
- [88] G. Kaur and S. Arora, "Chaotic whale optimization algorithm," J. Comput. Design Eng., vol. 5, no. 3, pp. 275–284, Jul. 2018.
- [89] R. Caponetto, L. Fortuna, S. Fazzino, and M. G. Xibilia, "Chaotic sequences to improve the performance of evolutionary algorithms," *IEEE Trans. Evol. Comput.*, vol. 7, no. 3, pp. 289–304, Jun. 2003.
- [90] B. R. Adarsh, T. Raghunathan, T. Jayabarathi, and X.-S. Yang, "Economic dispatch using chaotic bat algorithm," *Energy*, vol. 96, pp. 666–675, Feb. 2016.
- [91] Y. Li-Jiang and C. Tian-Lun, "Application of chaos in genetic algorithms," *Commun. Theor. Phys.*, vol. 38, no. 2, pp. 168–178, 2002.
- [92] B. Alatas, "Chaotic harmony search algorithms," Appl. Math. Comput., vol. 216, no. 9, pp. 2687–2699, Jul. 2010.
- [93] D. Jia, G. Zheng, and M. Khurram Khan, "An effective memetic differential evolution algorithm based on chaotic local search," *Inf. Sci.*, vol. 181, no. 15, pp. 3175–3187, Aug. 2011.
- [94] B. Liu, L. Wang, Y.-H. Jin, F. Tang, and D.-X. Huang, "Improved particle swarm optimization combined with chaos," *Chaos, Solitons Fractals*, vol. 25, no. 5, pp. 1261–1271, Sep. 2005.
- [95] S. Talatahari, B. F. Azar, R. Sheikholeslami, and A. H. Gandomi, "Imperialist competitive algorithm combined with chaos for global optimization," *Commun. Nonlinear Sci. Numer. Simul.*, vol. 17, no. 3, pp. 1312–1319, Mar. 2012.
- [96] G.-G. Wang, L. Guo, A. H. Gandomi, G.-S. Hao, and H. Wang, "Chaotic krill herd algorithm," *Inf. Sci.*, vol. 274, pp. 17–34, Aug. 2014.
- [97] S. Mirjalili and A. H. Gandomi, "Chaotic gravitational constants for the gravitational search algorithm," *Appl. Soft Comput.*, vol. 53, pp. 407–419, Apr. 2017.
- [98] A. H. Gandomi and X.-S. Yang, "Chaotic bat algorithm," J. Comput. Sci., vol. 5, no. 2, pp. 224–232, Mar. 2014.

- [99] A. H. Gandomi, X.-S. Yang, S. Talatahari, and A. H. Alavi, "Firefly algorithm with chaos," *Commun. Nonlinear Sci. Numer. Simul.*, vol. 18, no. 1, pp. 89–98, Jan. 2013.
- [100] G.-G. Wang, S. Deb, A. H. Gandomi, Z. Zhang, and A. H. Alavi, "Chaotic cuckoo search," *Soft Comput.*, vol. 20, no. 9, pp. 3349–3362, Sep. 2016.
- [101] S. Arora and S. Singh, "An improved butterfly optimization algorithm with chaos," J. Intell. Fuzzy Syst., vol. 32, no. 1, pp. 1079–1088, Jan. 2017.
- [102] M. Kohli and S. Arora, "Chaotic grey wolf optimization algorithm for constrained optimization problems," *J. Comput. Design Eng.*, vol. 5, no. 4, pp. 458–472, Oct. 2018.
- [103] Y. Xu, H. Chen, A. A. Heidari, J. Luo, Q. Zhang, X. Zhao, and C. Li, "An efficient chaotic mutative moth-flame-inspired optimizer for global optimization tasks," *Expert Syst. Appl.*, vol. 129, no. 1, pp. 135–155, Sep. 2019.
- [104] Q. Zhang, H. Chen, A. A. Heidari, X. Zhao, Y. Xu, P. Wang, Y. Li, and C. Li, "Chaos-induced and mutation-driven schemes boosting salp chains-inspired optimizers," *IEEE Access*, vol. 7, pp. 31243–31261, 2019.
- [105] G. I. Sayed, G. Khoriba, and M. H. Haggag, "A novel chaotic salp swarm algorithm for global optimization and feature selection," *Appl. Intell.*, vol. 48, no. 10, pp. 3462–3481, Oct. 2018.
- [106] A. A. Ateya, A. Muthanna, A. Vybornova, A. D. Algarni, A. Abuarqoub, Y. Koucheryavy, and A. Koucheryavy, "Chaotic salp swarm algorithm for SDN multi-controller networks," *Eng. Sci. Technol., Int. J.*, vol. 22, no. 4, pp. 1001–1012, 2019.
- [107] A. Mukherjee and V. Mukherjee, "Chaotic krill herd algorithm for optimal reactive power dispatch considering FACTS devices," *Appl. Soft Comput.*, vol. 44, pp. 163–190, Jul. 2016.
- [108] A. Mukherjee and V. Mukherjee, "Chaos embedded krill herd algorithm for optimal VAR dispatch problem of power system," *Int. J. Electr. Power Energy Syst.*, vol. 82, pp. 37–48, Nov. 2016.
- [109] B. Bentouati, S. Chettih, and R. A.-A. El-Sehiemy, "A chaotic krill herd algorithm for optimal solution of the economic dispatch problem," *Int. J. Eng. Res. Afr.*, vol. 31, pp. 155–168, Jul. 2017.
- [110] P. Lu, J. Zhou, H. Zhang, R. Zhang, and C. Wang, "Chaotic differential bee colony optimization algorithm for dynamic economic dispatch problem with valve-point effects," *Int. J. Electr. Power Energy Syst.*, vol. 62, pp. 130–143, Nov. 2014.
- [111] J. Cai, X. Ma, L. Li, Y. Yang, H. Peng, and X. Wang, "Chaotic ant swarm optimization to economic dispatch," *Electr. Power Syst. Res.*, vol. 77, no. 10, pp. 1373–1380, Aug. 2007.
- [112] E. Naderi, H. Narimani, M. Fathi, and M. R. Narimani, "A novel fuzzy adaptive configuration of particle swarm optimization to solve largescale optimal reactive power dispatch," *Appl. Soft Comput.*, vol. 53, pp. 441–456, Apr. 2017.
- [113] M. Ghasemi, S. Ghavidel, M. M. Ghanbarian, and A. Habibi, "A new hybrid algorithm for optimal reactive power dispatch problem with discrete and continuous control variables," *Appl. Soft Comput.*, vol. 22, pp. 126–140, Sep. 2014.
- [114] X. F. Wang, Y. Song, and M. Irving, *Modern Power System Analysis*. New York, NY, USA: Springer, 2008.
- [115] R. D. Zimmerman, C. E. Murillo-Sanchez, and D. Gan. MATPO-WER: A Matable Power System Simulation Package. Accessed: May 4, 2019. [Online]. Available: http://www.pserc.cornell.edu/matpower/
- [116] M. Ghasemi, M. M. Ghanbarian, S. Ghavidel, S. Rahmani, and E. M. Moghaddam, "Modified teaching learning algorithm and double differential evolution algorithm for optimal reactive power dispatch problem: A comparative study," *Inf. Sci.*, vol. 278, no. 10, pp. 231–249, Sep. 2014.
- [117] T. Das and R. Roy, "Optimal reactive power dispatch using JAYA algorithm," in Proc. Emerg. Trends Electron. Devices Comput. Techn. (EDCT), Mar. 2018, pp. 1–6.
- [118] Y. Li, Y. Wang, and B. Li, "A hybrid artificial bee colony assisted differential evolution algorithm for optimal reactive power flow," *Int. J. Electr. Power Energy Syst.*, vol. 52, pp. 25–33, Nov. 2013.
- [119] C. H. Liang, C. Y. Chung, K. P. Wong, and X. Z. Duan, "Comparison and improvement of evolutionary programming techniques for power system optimal reactive power flow," *IEE Proc. Gener., Transmiss. Distrib.*, vol. 153, no. 2, pp. 228–236, 2006.
- [120] C. H. Liang, C. Y. Chung, K. P. Wong, X. Z. Duan, and C. T. Tse, "Study of differential evolution for optimal reactive power flow," *IET Gener.*, *Transmiss. Distrib.*, vol. 1, no. 2, pp. 253–260, 2007.

- [121] C. Y. Chung, C. H. Liang, K. P. Wong, and X. Z. Duan, "Hybrid algorithm of differential evolution and evolutionary programming for optimal reactive power flow," *IET Gener., Transmiss. Distrib.*, vol. 4, no. 1, pp. 84–93, 2010.
- [122] S. Pandya and R. Roy, "Particle swarm optimization based optimal reactive power dispatch," in *Proc. IEEE Int. Conf. Electr., Comput. Commun. Technol. (ICECCT)*, Mar. 2015, pp. 1–5.
- [123] P. Subbaraj and P. N. Rajnarayanan, "Optimal reactive power dispatch with Cauchy and adaptive mutations," in *Proc. IEEE Comput. Soc. Int. Conf. recent trends Inf. Telecommun. Comput.*, Mar. 2010, pp. 110–115.
- [124] A. Rabiee, M. Vanouni, and M. Parniani, "Optimal reactive power dispatch for improving voltage stability margin using a local voltage stability index," *Energy Convers. Manage.*, vol. 59, pp. 66–73, Jul. 2012.
- [125] N. K. Patel and B. N. Suthar, "Optimal reactive power dispatch using particle swarm optimization in deregulated environment," in *Proc. Int. Conf. Electr., Electron., Signals, Commun. Optim. (EESCO)*, Jan. 2015, pp. 298–309.
- [126] S. Mouassa and T. Bouktir, "Artificial bee colony algorithm for discrete optimal reactive power dispatch," in *Proc. Int. Conf. Ind. Eng. Syst. Manage. (IESM)*, Oct. 2015, pp. 654–662.
- [127] A. A. Heidari, R. A. Abbaspour, and A. R. Jordehi, "Gaussian bare-bones water cycle algorithm for optimal reactive power dispatch in electrical power systems," *Appl. Soft Comput.*, vol. 57, pp. 657–671, Aug. 2017.
- [128] B. Shaw, V. Mukherjee, and S. P. Ghoshal, "Solution of reactive power dispatch of power systems by an opposition-based gravitational search algorithm," *Int. J. Electr. Power Energy Syst.*, vol. 55, pp. 29–40, Feb. 2014.
- [129] S. Duman, Y. Sönmez, U. Güvenç, and N. Yörükeren, "Optimal reactive power dispatch using a gravitational search algorithm," *IET Gener, Transmiss. Distrib.*, vol. 6, no. 6, pp. 563–576, 2012.
- [130] R. P. Singh, V. Mukherjee, and S. P. Ghoshal, "Optimal reactive power dispatch by particle swarm optimization with an aging leader and challengers," *Appl. Soft Comput.*, vol. 29, pp. 298–309, Apr. 2015.
- [131] M. Ghasemi, S. Ghavidel, M. M. Ghanbarian, and M. Gitizadeh, "Multi-objective optimal electric power planning in the power system using Gaussian bare-bones imperialist competitive algorithm," *Inf. Sci.*, vol. 294, pp. 286–304, Feb. 2015.
- [132] A. Ghasemi, K. Valipour, and A. Tohidi, "Multi objective optimal reactive power dispatch using a new multi objective strategy," *Int. J. Electr. Power Energy Syst.*, vol. 57, pp. 318–334, May 2014.
- [133] R. Mallipeddi, S. Jeyadevi, P. N. Suganthan, and S. Baskar, "Efficient constraint handling for optimal reactive power dispatch problems," *Swarm Evol. Comput.*, vol. 5, pp. 28–36, Aug. 2012.
- [134] B. Mandal and P. K. Roy, "Optimal reactive power dispatch using quasioppositional teaching learning based optimization," *Int. J. Electr. Power Energy Syst.*, vol. 53, pp. 123–134, Dec. 2013.
- [135] K. Mahadevan and P. S. Kannan, "Comprehensive learning particle swarm optimization for reactive power dispatch," *Appl. Soft Comput.*, vol. 10, no. 2, pp. 641–652, Mar. 2010.
- [136] W. Villa-Acevedo, J. López-Lezama, and J. Valencia-Velásquez, "A novel constraint handling approach for the optimal reactive power dispatch problem," *Energies*, vol. 11, no. 9, p. 2352, 2018.
- [137] A. P. Mazzini, G. G. Lage, and E. N. Asada, "Solving control-constrained reactive power dispatch with discrete variables," in *Proc. 18th Int. Conf. Intell. Syst. Appl. Power Syst. (ISAP)*, Sep. 2015, pp. 1–6.



ZHAOYANG QU received the M.S. degree in electrical engineering from the Dalian University of Technology, Dalian, China, in 1988, and the Ph.D. degree in electrical engineering from North China Electric Power University, Beijing, China, in 2010.

He is currently a Professor and a Doctoral Advisor with the School of Information Engineering, Northeast Electric Power University, Jilin City, China. His research interests include smart grid

and power information processing, virtual reality, and network technology. Prof. Qu is also a member of the China Electric Engineering Society Power

Information Committee, the Vice-President of the Jilin Province Image and Graphics Society, the Head of the Power Big Data Intelligent Processing Engineering Technology Research Center, and the Jilin Changbai Mountain Scholar. He is also a top-notch innovative talent young in Jilin Province and middle-aged professional and technical talent with outstanding contributions. He has presided over the completion of two national natural science funds, received the second prize of Jilin Province Science and Technology Progress Award, and wrote 46 articles on electric power information SCI/EI search.



FRANÇOIS XAVIER RUGEMA received the B.S. degree in electronics and telecommunication engineering from the Kigali Institute of Science and Technology, Kigali, Rwanda, in 2005, and the M.S. degree in telecommunication engineering from the National University of Rwanda, Huye, Rwanda, in 2010. He is currently pursuing the Ph.D. degree in electrical engineering with the Northeast Electric Power University, Jilin City, China.

From 2009 to 2015, he was an Assistant Lecturer with the National University of Rwanda. His research interests include intelligent information processing, power cyber physical systems, smart grid communications, and smart grid cyber security.



SYLVERE MUGEMANYI received the B.S. degree in electronics and telecommunication engineering from the Kigali Institute of Science and Technology, Kigali, Rwanda, in 2005, and the M.S. degree in telecommunication engineering from the National University of Rwanda, Huye, Rwanda, in 2010. He is currently pursuing the Ph.D. degree in electrical engineering with the Northeast Electric Power University, Jilin City, China.

From 2009 to 2015, he was an Assistant Lecturer with the IPRC-North/College of Tumba, Rulindo, Rwanda. His research interests include intelligent information processing, power cyber physical systems, smart grid communications, and smart grid cyber security.



YUNCHANG DONG received the B.S. degree from Northeast Electric Power University, Jilin City, China, in 2016, where he is currently pursuing the M.S. degree.

His research interests include power cyber physical systems and information processing of smart grid.

IEEE Access



power systems.

CHRISTOPHE BANANEZA received the B.S. degree in electrical power engineering from the National University of Rwanda, Huye, Rwanda, in 2011. He is currently pursuing the M.S. degree in electrical engineering with the College of Energy and Electrical Engineering, Hohai University, Nanjing, China.

From 2012 to 2016, he was an Assistant Lecturer with the IPRC-North/College of Tumba, Rulindo, Rwanda. His research interest includes



LEI WANG received the M.S. degree from Northeast Electrical Power University, Jilin City, China, where he is currently pursuing the Ph.D. degree in electrical engineering. He is currently an Associate Professor with the School of Information Engineering, Northeast Electric Power University. His research interest includes information processing in smart grid.

...