

An innovative Hybridized optimization perspective for optimal allocation of UPFC, FACTS and SVC

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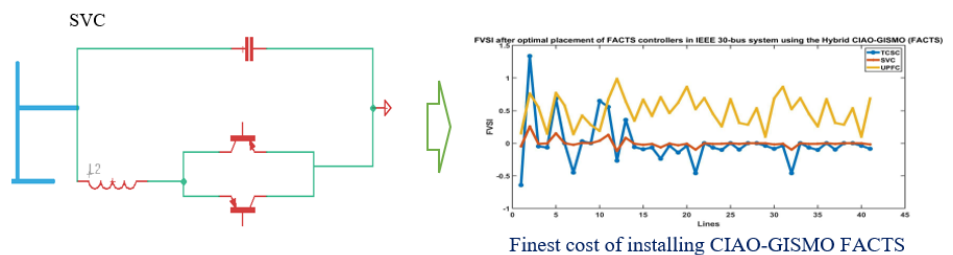
Article

ABSTRACT

Long-term changes in the arousal of electric power systems now routinely dodge the situation. The regulators for the Flexible AC Transmission System (FACTS) can assist improve power structure stability. Because of the high level of capital speculation, it would be ideal to

integrate these regulators in the power structure. The work has prompted the development of a revolutionary multi-objective hybrid advancement model for power security's optimal configuration, which includes a static voltage compensator, thyristor-controlled sequence capacitor, and a united power flow controller. The suggested approach integrates its decision regarding the power stream in each line to position the FACTS device in a line or at the remote end of the line. Two scenarios for the best position and types of FACTS devices are favored using a hybrid optimization-based multi-objective capacity. The several objectives include carelessness, actual power misfortune, minimal responsibility, little absolute voltage fluctuation, line voltage strength, rapid voltage constancy, and running expense. Additionally, the traditional Confidence Interval-based Aquila Optimizer (CI-AO) and Glorot initialization-based Spider Monkey Optimization are used to produce the Hybrid optimization model (GI-SMO). An exploratory result demonstrates that, in contrast to current methods, the suggested methodologies achieve minor power and receptive losses during transmission and continue to be very successful.

Keywords: FACTS devices, Reactive power; SVC; TCSC, UPFC, Confidence interval-based Aquila Optimizer (CI-AO), Glorot initialization-based Spider Monkey Optimization (GI-SMO)



INTRODUCTION

The weight of the power framework is growing as interest in electrical power increases. Power framework activity gets more troublesome, the power stream becomes less reliable, and disasters rise.^{1,2} The intricacy of power stability has long confounded power framework engineers. Power framework security concerns are a compelling and exciting topic of discussion since force frameworks are always being established to meet the needs of an increasing

population.^{3,4} When working on a massive power transmission network close to the border of voltage dependability, controlling the receptive power interest for that system becomes difficult. Voltage security is a major issue for the reliability of the power system.^{5,6} The hang in responsive power at different areas in an interrelated power framework network is the essential driver of voltage unsteadiness. In power frameworks that are vigorously stacked, have a responsive power lack, or are blamed, voltage dependability is an issue.^{7,8}

Many FACTS devices, including the shunt-series UPFC, SVC, and STATCOM, as well as the series-related TCSC and SSSC, are of interest to power specialists. As previously indicated^{9,10} their area, kind, and size all play key roles in their activity and how it impacts the network. This has led to numerous attempts over a very long period of time to simplify the integration of such FACTS devices into different IEEE transport frameworks using various conventional advancement procedures like straight programming

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procedures, non-direct programming strategies, and mixed number non-direct programming techniques.

It is challenging to determine the appropriate locations and FACTS device designs in power systems, and substantial data collection is frequently required.¹³ The intricacy of the objective time or space may lead a smearing computation that provides the exact perfect answer to the problem to fail in a reenactment scenario.^{14,15} In the existing research, the methodologies and strategies utilized for formative the most advantageous positions and scenery of FACTS strategy were divided into four categories: scientific methodologies, number juggling programming strategies or traditional enhancement strategies, meta-heuristic streamlining procedures, and half-and-half techniques.^{16,17} They are acknowledged by the FACTS regulators' acceptance of control calculations set up to accomplish a range of goals. A novel multi-objective hybrid optimization improvement model combining CI-AO and GI-SMO techniques has been created by the effort to aid in power security.

Challenges

- High loss of active power system, operational costs of the system, and voltage deviation
- Increased network complexity and Early convergence
- The FACT device's optimal placement is complex
- Poor rate performance and more expensive to implement as well as time-consuming

To proficiently enhance the power security, there is a great essential to develop an innovative multi-objective hybrid advancement model. Thus it consists of the perfect combination of a static voltage compensator, a unified power flow controller, and a series capacitor controlled by a thyristor. The Hybrid optimization model is created using the conventional Confidence Interval-based Aquila Optimizer (CI-AO) and Glorot initialization-based Spider Monkey Optimization (GI-SMO). Due to the use of hybrid optimization techniques, it is prompted to implement a balanced exploration and exploitation rate in order to regulate network complexity and early convergence, improve rate performance, and save costs and time.

This study has following objectives:

- The precise location of various FACTS regulator types has been used to MSL for power consistency and load interest in the power infrastructure. The FACTS device lowers voltage volatility, active power loss, and operational expenses for power systems.
- For the optimization of single- and multi-objective functions, this non-optimized site yields unacceptable target values. Since FACTS regulators are expensive, raising SL is probably going to cut down on their speculative expenses. Working with both series and shunt FACTS regulators, the UPFC aims to reduce transmission loss in the influence framework.
- The optimum places to add FACTS devices within the linked power framework network in order to raise the framework's general exchange limit and obtain the lowest working expense attainable under various stacking scenarios.

Study contributes:

- The work has developed a innovative hybrid heuristic algorithm indulged on Confidence Interval based Aquila Optimizer

(CI-AO) and Glorot initialization-based Spider Monkey Optimization (GI-SMO) through multi-objective function for the FACTS controllers distribution in power grids for minimizing the losses of operational power system, operational costs of the system, and voltage deviation.

- Balanced exploration and exploitation rate for controlling the network complexity and early convergence

- Better rate performance and expensive low cost as well as time due to the usage of hybrid optimization techniques

LITERATURE SURVEY

To confine movements in single and multimachine power structures, Paital et al.,¹⁸ presented a HHO changed twofold territory type-2 padded lead-slack (Double IT2FLL) based bound together power stream controller (UPFC). The damping regulator was organized considering speed deviation, a tricky data signal for strength improvement, and headings between the equilibrium list (MI) and stage point of shunt and series converters of UPFC simultaneously. Different execution records (PIs), like overshoots, SD, settling time, and mean, were sent to show that the HHO-changed twofold IT2FLL-based UPFC outsmarted others in assorted working circumstances.

Murugan et al.,¹⁹ formed a strategy for computing age expenses and catastrophes. The Crossover Power Stream controller was the primary Realities gadget to be thought about. A GA controller was then used to decide the cost limit and power misfortunes in different vehicles comparable to the Half and half Power Stream Regulator. The Hereditary Calculation was habitually utilized to foster prevalent arrangements or results for development and search issues. At last, the IEEE-14 vehicle movement test system gave its endorsement to the strategy. At last, the created voltage profile and age cost work utilizing the Hereditary Calculation with and without a controller came to the front.

J. Mahadevan et al.,²⁰ used Hybrid ABCDE to work on various objectives, including line stacking records, cost, and voltage deviation decrease. The IEEE 30 vehicle engineering and three micro processor controlled Realities controllers, including the TCSC, SVC, and bound-together power stream controller, are where we are best arranged. The consequences of atomic multitude smoothing out were contrasted with the effects of the mutt ABC-DE (PSO). It was shown that ABC/DE to some degree outsmarted PSO when it came to apportioning Realities controllers by all the while progressing numerous objectives.

Shehata et al.²¹ inspected a Realities part in a power framework is worked utilizing heuristic calculations. The paper advanced an AGPSO method for the ideal case and assessed the SVC to lessen outright unique power misfortunes in transmission lines. Moreover, correlations with existing heuristic improvement computations were finished to affirm the recommended estimation's appropriateness. As a result, as indicated by the generation discoveries, the estimation had the best presentation, minimal measure of dynamic power misfortunes, and the quickest mix rate.

Biswas et al.,²² set off different gadgets on Realities, integrating a static VAR compensator, a series compensator driven by a thyristor, and a phase shifter constrained by a thyristor, were utilized to address the OPF that joined stochastic breeze power. The

improvement point incorporated the warm age cost, the quick expense of arranged breeze power, the result cost for mistake, and the saving cost for misjudging wind power. For limiting the expense or cost of power age, the Realities gadget's areas and appraisals were refreshed. Various calculations were investigated, including the crossover approach of the SF technique with a couple of notable metaheuristic computations.

Padmavathi et al.²³ used the two primary goals of node voltage deviations and line apparent power flow factors, which we recently established, the integrity of energy systems is tackled. The study investigates the employment of several power devices, including thyristor-controlled series compensation (TCSCs), distribution Static compensators (SVCs), and integrated energy stream processors, in the pursuit of these goals (UPFCs). The main objective of the study is to evaluate the performance of various placement techniques for these devices on the IEEE 30 bus system across varying flow scenarios. The hybridized multiple objective particles swarm optimization (DEPSO) and fuzzy adaptive force of gravity search technique are indeed the two techniques that were examined in this study (FAGSA). This study intends to add to the body of knowledge by elucidating the efficacy of various optimization techniques for addressing energy system safety issues.

Aryaet al.²⁴ have suggested using the force of gravity search algorithm (GSA), a metaheuristic optimization method, to analyze distribution networks (DS) with precise placement and measurement of decentralized synchronous condensers attenuator (D-STATCOM) to accomplish line losses lowering, lower limit overall voltage ratio, advancement in voltage level, and yearly basis energy savings for the operators of distribution networks. For IEEE 33 and IEEE 69 bus systems, the suggested algorithm's efficacy is evaluated, and its performance is contrasted with that of other methods for D-STATCOM allocation, including the immune algorithm, bat algorithm, and sensitivity approach method. The study sheds light on the use of metaheuristic optimization techniques for the best placement of D-STATCOM in DS and emphasizes the value of GSA in enhancing system performance. The study's findings show how the suggested strategy could raise the effectiveness and productivity of electricity distribution networks.

Darebin et al.²⁵ have sent SCESS to control the GSC active power and conveyed SSSC to diminish low-recurrence motions. For SCESS and SSSC, RSC alongside the damping regulator has been controlled at the same time utilizing the executed methodology in view of prescient control. Besides, the information ways of Laguerre capabilities were picked by the capability used in the prescient control technique for decreasing computational entanglement. Moreover, outstanding information weighting has been conveyed to limit the examining time. The trial results have in this manner assessed and affirmed the viability of the executed procedure by upgrading dependability.

METHODS

Increased transactions in newly developed electric power systems usually lead to situations where the system does not operate widely in a certain operational region. The FACTS controllers can aid in enhancing the safety of the control structure. These regulator

Author [citation]	Technique used	Features	Challenges
Patel et al. [18]	HHO	Better oscillation-damping performance	Need to carry out this work in complex power system
Murugan et al. [19]	Genetic algorithm	Minimize power loss in the system	Expensive to implement and is time-consuming
J. Mahadevan et al. [20]	ABC-DE	Maximizes the system safety Efficiency in optimizing multi-modal, multi-objective and discrete system	More development is needed on microprocessor-based control procedures
Shehata et al.[21]	AGPSO	Higher convergence rate Last active power loss	Further plans to apply for this work in real network
Biswaset al. [22]	SHADE	The entire constrained optimization tasks are solved	Increased cost Increased network complexity
Padmavathiet al. [23]	FAGSA	Improved power system security Reduced power losses	Complexity of the system Computational complexity
Aryaet al. [24]	GSA	Improve power quality issues	simulation results are based on a single distribution system Increased network complexity
Darabian et al. [25]	Firefly Algorithm	Progress the electrical energy contour of the control structure. provide better control and regulation of the STATCOM device	Require significant investments in hardware and software Poor rate performance

should, however, be positioned as effectively as possible in the control structure due to the large resources disbursement. One of the main issues is also the transmission loss brought on by UPFC, TCSC, and SVC. For identifying TCSC, SVC, and UPFC, the research offers a novel multi-objective hybrid optimization approach.

FACTS CONTROLLER STATIC MODELLING

The facts controller model design comprises three primary power streams, i.e. TCSC, SVC, and UPFC²⁶

TCSC MODEL

The basic concept behind power stream regulation for TCSC is to add a capacitive or inductive response in series to reduce or increase the overall lines' persuasive transmission impedance. The line partners bus-i and bus-j have an equal reactance, which is

represented by the TCSC as a variable impedance system as follows:

$$A_{ij} = A_{line} + A_{TCSC} \quad (1)$$

Where A_{line} it indicates the line reactance original transmission and A_{TCSC} the reactance of TCSC. Instalment of TCSC, the new line reactance is aluded by:

$$A_{ij} = (1 - \mathfrak{R}_c) A_{line} \quad (2)$$

While \mathfrak{R}_c illustrates the reactance compensation percentage. The Compensation levels of the TCSC generally vary among 80% capacitive and 20% inductive.

MODEL OF SVC

A shunt compensator by definition, the SVC is used to replace inductive or capacitive reactance in order to monitor or regulate the explicit limits of an electrical power structure, most frequently a vehicle voltage.

The SVC equation that expresses the real-time power reactive injected at the bus-i is denoted as:

$$P_i = B_{SVC} \mathcal{Q}_i^2 \quad (3)$$

$$B_{SVC}^{min} \leq B_{SVC} \leq B_{SVC}^{max} \quad (4)$$

UPFC MODEL

VSI share a typical dc interface capacitor in the UPFC. The UPFC technique is tended to by Series voltage V_s and shunt power source I_{SC} s. Here, UPFC is believed to be set at bus-i and in the line associated among bus-i and bus-j. The current of the shunt and the voltage source in the series wellspring of UPFC are portrayed as follows:

$$V'_s = (V_r + V_s) e^{j\Phi} = t V'_i e^{j\chi} \quad (5)$$

$$i'_{SC} = (I_r + I_s) e^{j\Theta} \quad (6)$$

The UPFC observes the apparent power via bus-i and bus-j, i.e. given by $H_{iju} + jG_{iju}$ and $H_{jiu} + jG_{jiu}$, respectively, and given as follows.

$$H_{iju} = \beta k V_i V_j \sin(\chi + \Theta_i - \Theta_j) \quad (7)$$

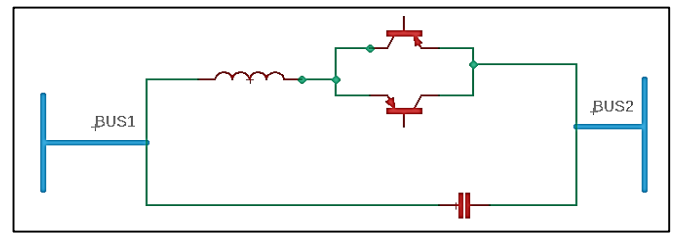
$$G_{iju} = \beta k V_i^2 \cos \chi - I_q V_i \quad (8)$$

$$H_{jiu} = -\beta k V_i V_j \sin(\chi + \Theta_i - \Theta_j) \quad (9)$$

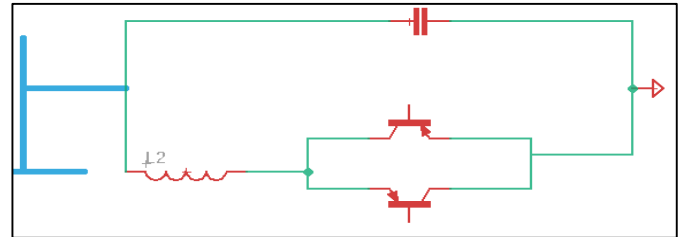
$$G_{jiu} = -\beta k V_i V_j \cos(\chi + \Theta_i - \Theta_j) \quad (10)$$

MULTI-OBJECTIVE FUNCTION

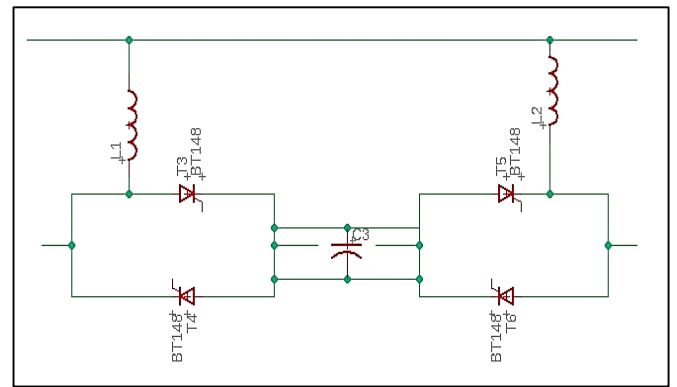
By reducing the cost of two key arrangements—the cost of energy loss due to dynamic influence—it is possible to increase the base working expenditure under various stacking situations. Here is a description of the objective work and its restrictions:



(a)



(b)



(c)

Figure 1: State Modelling of (a) TCSC (b) SVC (c) UPFC

MINIMIZATION OF ACTUAL POWER LOSS

$$\min \nabla_1(\alpha_1, \alpha_2) = \Omega_{loss} = \sum_{k=1}^n [\zeta_k (v_s^2 + v_r^2 - 2v_s v_r \cos \Gamma_{sr})] \quad (11)$$

Here α_1 and α_2 is denoted by the following equations:

$$\alpha_1 = [\mathfrak{S}_{q1}, \dots, \mathfrak{S}_{qn_{pv}}, v_{l1}, \dots, v_{ln_{pq}}, s_{l1}, \dots, s_{ln_l}] \quad (12)$$

$$\alpha_2 = [T_1, \dots, T_n, v_{g1}, \dots, v_{gn_p}, q_{c1}, \dots, q_{cn_j}, B_{SVC}, \dots, B_{NSVC}, A_{TCSC}, \dots, A_{NTCSC}, V'_{UPFC}, \dots, V'_{NUPEC}] \quad (13)$$

Where $\nabla_1(\alpha_1, \alpha_2)$ illustrates the minimization active power loss function, ζ_k illustrates branch conductance of branch k , v_s and v_r illustrates sending bus voltage magnitude and getting bus respectively, Γ_{sr} illustrates the difference among phase angle s^{th} r^{th} and bus, α_1 illustrates the vector of ward factors comprising of receptive power age of generator ($\mathfrak{S}_{q1}, \dots, \mathfrak{S}_{qn_{pv}}$), load voltages ($v_{l1}, \dots, v_{ln_{pq}}$), and line loading transmission (s_{l1}, \dots, s_{ln_l}), α_2 denotes the

transformer tap settings ($T_1, \dots, T_{n_\varphi}$) control variables, generator voltages magnitude ($V_{g1}, \dots, V_{gn_{pv}}$), reactive power injections ($q_{c1}, \dots, q_{cn_{cl}}$), static Var compensator (B_{SVC}, \dots, B_{NSVC}), ($A_{TCSC}, \dots, A_{NTCSC}$) and ($V'_{UPFC}, \dots, V'_{NUPFC}$).

MINIMIZATION OF VOLTAGE DEVIATION

Maintaining a consistent voltage profile is considered one of the most troublesome assignments in power framework security. The deviation of voltage minimization can be expressed as:

$$V_D = \sum_{B=1}^N |V_B - 1.0| \tag{14}$$

Where N is the total number of buses, and V_B is the bus voltage

MINIMIZATION OF OPERATING COST

It is separated into two sections: the first is the expense of energy misfortune, and the second is the expense of FACTS device venture. Thus, the goal work should not just diminish the expense of energy misfortune by limiting dynamic influence misfortune with TCSC and SVC but also lessen the TCSC and SVC venture costs. Subsequently, the goal work is to limit total working expenses, which can be composed as:

$$C_{tot} = C_{energy} + C_{facts} \tag{15}$$

Where, $C_{energy} = p_{loss} \times 0.06 \times 100000 \times 365 \times 24$

Energy loss cost=0.06 \$/KWhr, installation shunt capacitor cost= 1000 \$, year days=365, hours in a day= 24.

The cost of FACTS devices (C_{facts}) may be formulated as follows:

$$C_{facts} = \hbar s^2 + \tilde{\lambda} s + \xi \tag{16}$$

Where, S indicates MVAR operating cost range and $\hbar, \tilde{\lambda}, \xi$ indicates the coefficient cost of the FACTS devices that depends on different FACTS types.

POWER SYSTEM STABILITY CONSTRAINTS

POWER STABILITY SYSTEM

Different nonlinear differential conditions for each generator regulate the coordinated exciter, control mechanisms, and machine. Each generator also includes a number of mathematical requirements that link its consistent state working point power injection into the system with the appropriate placement of FACTS regulators for enhancing the system's liabilities 17 factors. The Kirchhoff's regulatory circuit criteria that must be met at the consistent state operating point are the power framework network conditions, which are as follows:

$$E = Y_x - Y_y Z_y^{-1} Y_y \tag{17}$$

Where Y_x, Y_y, Z_y, Z_x illustrates the power flow Jacobian matrices. Assuming the linearised framework's intricate Eigenvalues have genuine negative parts, the power framework can endure minor annoyances and is, in this manner, considered stable in the little sign sense. The condition integrates the eigenvalue solidness examination into the imperative.

$$E(Y_x, Y_y, Z_y, Z_x) = 0 \tag{18}$$

The stability assures grid stability on the basis of eigenvalue under diverse levels of SL.

FAST VOLTAGE STABILITY INDEX

The FVSI helps to maintain safe bus loading.

$$\nabla_{FVSI} = \frac{4C^2 \Psi_j}{v_i^2 x} \tag{19}$$

The line that displays ∇_{FVSI} near 1.00 states that the shakiness point was accomplished ∇_{FVSI} ; if it goes After 1.00, causing the framework to fail. FVSI's list joining the regulator guarantees that no transport will fall due to over-burdening.

LINE STABILITY FACTOR

The System constancy index guaranteed the line stability factor. Maintaining an LQP of fewer than 1.00 helps to preserve a constant method.

$$\nabla_{LQP} = 4 \left(\frac{x}{v_i^2} \right) \left(\frac{x}{v_i^2} p_i^2 + \Psi_j \right) \tag{20}$$

∇_{LQP} guarantee the regulator that no line is overburdened under any network circumstance.

WEAK BUSES DETECTION FOR SERIES AND SHUNT FACTS DEVICE PLACEMENT

FACTS's primary purpose is to improve the constant state of communicable power, regulate the voltage profile along the lines, and impact the normal electrical properties of transmission lines. The goal of identifying weak transportation is to locate the ideal locations for FACTS equipment. The voltage profile is further developed, power misfortune is reduced, and enough receptive power assistance is provided at the appropriate locations. But it also addresses the problem of voltage shakiness. These processes require a single line of electricity to flow. (a) The branches with the most unique receptive power are identified by the power stream analysis, which also identifies reactive power streaming in all branches. The branch's terminus or the transport where the branch joins are considered to be fragile transports, and TCSCs are put on these powerless transports. The accompanying advances are utilized to decide the area of TCSCs:

Stage 1: Read the test framework's line and bus information.

Step 2: Create a Y-bus network.

Stage 3: Using the Newton-Raphson strategy, compute the voltage and point.

Stage 4: Using the heap stream strategy, compute dynamic and responsive power in each branch.

Stage 5: Choose the branch that has the most reactive power.

Stage 6: Verify that the picked branch is associated with the generator or the slack bus. If indeed, continue to stage 5; in any case, proceed to the following stage.

Stage 7: The TCSC is not set in stone by the branch or transport endpoint.

VOLTAGE COLLAPSE PROXIMITY INDICATION

A line's most extreme power move hypothesis is utilized to foster the voltage breakdown closeness sign technique. Let internal impedance $\eta_s \angle \theta$ take care of the load impedance $\eta_l \angle \Phi$. When the proportion η_l / η_s is equivalent to 1.0, the most incredible power will be moved to the load. After summing up the organization into a single line, this proportion is utilized as a voltage breakdown indicator for that bus. Think about shifting burden impedance while keeping the load steady. This suspicion will not just keep up with exactness. However, it will likewise work on the issue. Load request increments, bringing about diminishes of η_l current increments. Thus, the voltage at the less-than-desirable end drops.

$$v_r = \eta_l I \tag{21}$$

Where,

$$I = \frac{v_s}{\sqrt{[(\eta_s \cos \varphi + \eta_l \cos \beta)^2 + (\eta_s \sin \varphi + \eta_l \sin \beta)^2]}}$$

$$v_r = \eta_r / \eta_s \frac{v_s}{\sqrt{[1 + (\eta_r / \eta_s)^2 + 2(\eta_r / \eta_s) + \cos(\beta - \varphi)]}} \tag{22}$$

Receiving end's active power,

$$P_r = v_r I \cos \beta \tag{23}$$

$$P_r = \left[\frac{v_s^2 \eta_s}{1 + (\eta_r / \eta_s)^2 + 2(\eta_r / \eta_s) + \cos(\varphi - \theta)} \right] \eta_r / \eta_s \cos \Phi \tag{24}$$

Likewise, control thrashing in the lineup is

$$P_l = \left[\frac{v_s^2 \eta_s}{1 + (\eta_l / \eta_s)^2 + 2(\eta_l / \eta_s) + \cos(\varphi - \theta)} \right] \cos \beta \tag{25}$$

By applying the boundary condition, i.e. $\frac{\partial P_r}{\partial \eta_l} = 0$ the Maximum real power can be obtained, which lead $\eta_l / \eta_s = 1_s$ to. Substitute it in Equation (24); utmost moveable control is:

$$P_{r(max)} = v_s^2 \eta_s \frac{\cos \Phi}{4 \cos^2 \left(\frac{\varphi - \beta}{2} \right)} \tag{26}$$

Given that VCPI depends upon the maximum power transferred via a line. Hence VCPI is given as,

$$\forall_{vcpi} = \frac{P_r}{P_{r(max)}} \tag{27}$$

VCPI should be less than unity in a voltage stability system. If the assessment is close to 1.0, it indicates that it is forthcoming to the point of unsteadiness. Weak buses are those that are approaching the point of instability. These buses were chosen for SVC's candidate locations.

FITNESS FUNCTION EVALUATION

By taking the expenses of device setup into consideration, the best region and FACTS rating are still at the top of the list. The study seeks to reduce transmission adversity while still meeting the standards. The minimization is accomplished using meta-heuristic methods, where the wellness work is used to choose or ignore arbitrarily selected components. Using following Eq, the wellness capacity may be expressed numerically.

$$SD = c \times s \times 1000 \tag{28}$$

Where, SD indicates the setting upcharge of procedure, C illustrates the charge of FACTS gadget in and S is formulated as.

$$s = |q_2 - q_1| \tag{29}$$

Where q_1 and q_2 are the receptive power streaming in the lines when the establishment of FACTS gadgets happens individually? The expense C relies on the sort of FACTS devices introduced. The expense for SVC, TCSC, and UPFC can be composed as Eq.

$$e_{TCSC} = 0.0016s^2 - 0.7140s + 154.80 \tag{30}$$

$$e_{svc} = 0.00040s^2 - 0.3061s + 128.14 \tag{31}$$

$$e_{UPFC} = 0.00035s^2 - 0.2684s + 177.11 \tag{32}$$

HYBRID OPTIMIZATION

The best FACTS device allocation issue is solved using hybrid optimization techniques. The suggested hybrid approach combines Spider Monkey Optimization based on Glorot initialization with the traditional Aquila Optimizer. The goal of hybrid optimization is to determine which local and global rates are the most stable. The optimum and most limited location for the FACTS controller is made possible by balancing the exploration and exploitation rates. To find the best and most ideal answer, the algorithm discards the poorest results.

The Aquila Optimizer was originally a population-based optimization technique encouraged via the natural behaviors of the Aquila when chasing prey. Confidence interval-based population initialization is utilized to get over this issue since random initialization of the upper and lower limits results in significant diversity in local search solutions, causing the convergence rate in local solutions to be trapped. The proposed Hybrid enhancement calculation's advancement technique is thus divided into four categories: selecting the key space by high start - up with longitudinal, and the new ideal place.

The number of residents in the newcomer arrangements is initially established using the confidence interval so as to is determined by:

$$\alpha = \begin{bmatrix} \alpha_{1,1} & \dots & \alpha_{1,Dim} \\ \vdots & \dots & \vdots \\ \alpha_{N,1} & \alpha_{N,J} & \alpha_{N,Dim} \end{bmatrix} \tag{33}$$

$$\alpha_{ij} = rand \times (vc_j - mc_j) + mc_j, i = 1, 2, \dots, Dim$$

$$vc_j = \alpha_{ij} + \mathfrak{I} \frac{\sigma}{\sqrt{o}}$$

$$mc_j = \alpha_{ij} - \mathfrak{I} \frac{\sigma}{\sqrt{o}}$$

Without missing a beat, the CI-AO uses a high takeoff and vertical stoop to determine the optimal hunting area. The inspection then starts by taking the Aquila high takeoff with superb stoop's acting strategy into consideration. Scientifically, this direct is represented as in Eq.

$$\alpha_i(u+1) = \alpha_{best}(u) \times \left(1 - \frac{u}{U}\right) + (\alpha_M(u) - \alpha_{best}(u) * rand) \quad (34)$$

While search technique $(\alpha_i, \alpha_{best}(u))$ is the best-acquired arrangement until t th emphasis mirrors the inexact spot of the prey

Focusing on the prey inside the Aquila rings completes a form trip with a brief float assault. Here, in preparation for the attack, AO just scans the selected area of the target victim. This behaviour is described mathematically in Eq.

$$\alpha_2(u+1) = \alpha_{best}(u) \times Levy(E) + \alpha_R(u) + (Z - Y) * rand \quad (35)$$

Which is strong-minded applying Eq. (35). $\alpha_R(u)$ It is an asymmetrical procedure taken in the scope of $[1, N]$ i that iteration

$$levy(E) = a \times \frac{v \times m}{|w|^{\frac{1}{2}}} \quad (36)$$

$$m = \left(\frac{\kappa(1 + \varepsilon) \times \sin\left(\frac{\pi\chi}{2}\right)}{\kappa\left(\frac{1 + \varepsilon}{2}\right) \times \varepsilon \times 2^{\left(\frac{\beta-1}{2}\right)}} \right) \quad (37)$$

$$Z = r \times \cos(\varphi) \quad (37)$$

$$Y = r \times \sin(\varphi)$$

where,

$$s = s_1 + v \times e_1 \quad (38)$$

$$\varphi = -\alpha \times e_1 + \varphi_1$$

$$\varphi_1 = \frac{3 \times \pi}{2}$$

From the above equation r_1 takes a value someplace in the scope of 1 and 20 for the set number of search cycles and is roughly worth 0.00565. Represents total numbers ranging from 1 to the length of the pursuit space (Dim), and is set to 0.005.

The Aquila is ready for assault after the exact indication of the target zone; it lowers upward with a primer assault to discover the prey response. Finally, a numerical introduction of the Aquila's behaviour in the low trip with the slow drop assault is made.

$$\alpha_3(u+1) = (\alpha_{best}(u) - \alpha_M(u)) \times \mu - rand + ((vc - mc) \times rand + mc) \times \tau \quad (39)$$

While that is operated by the third pursuit strategy $(\alpha_3(u+1))$, $\alpha_{best}(u)$ alludes to the rough area of the prey until i^{th} emphasis arrangement at emphasis, which is calculated using Eq (39). is an arbitrary value alteration borders that are established using GI-SMO, indicating the lower limit and indicating the upper bound of the given problem.

The GI-SMO process was made more interesting by the spider monkeys' clever foraging behavior. The foraging behaviour of spider monkeys is influenced by a complex web of social organization. The social relationship of a group where a female leader decides whether to split or consolidate determines the features. However, FACTS are presented incorrectly due to the lengthy process and high SMO assembly rate. The effort has made a Glorot statement more familiar with avoiding the previously mentioned problems. Utilizing the GI-SMO approach conduct, the exploitation change in CI-AO is completed.

Following the stage of inquiry, the calculation moves on to the stage of double-dealing, when it updates potential chance that a wellness component may be ascertained from the objective task. f_t

$$g_u = \begin{cases} 1 & \text{if } g_j \geq 0 \\ 1 + g_j & \\ 1 + abs(g_j), & \text{if } g_{j_i} \leq 0 \end{cases} \quad (40)$$

The choice likelihood Pb_i is resolved on the roulette wheel determination. If it g_u is the wellness of i^{th} SM, Its chances of being picked in the global pioneer stage are calculated using both of the following two formulae:

$$Pb_i = \frac{g_j}{\sum_{i=1}^N fitness_i} \quad (41)$$

In order to update the location, SM makes use of data on the global pioneer, adjacent SM experience, and its resolve. The position update requirement at this step is as follows:

$$\alpha_{nemi_j} = \alpha_{ij} + (vc_k(0, te)) \times (\alpha_{GLkj} - \alpha_{ij}) + vc_j(-1, 1) \times (\alpha_{ij} - \alpha_{ij})$$

$$sd = \sqrt{\frac{2}{\alpha_{ij} + \alpha_{GLkj}}} \quad (42)$$

This status change constraint is broken down into three parts: the first shows how attentive the parent (current) SM is, the second shows how curious the parent SM is about the global pioneer, and the final is utilised to retain the stochastic structure of the computation.

At the global pioneer learning stage, the computation notices the best solution for the entire huge number. The well-known SM was regarded as the global leader of the enormous number. The worldwide trailblazer's location is additionally evaluated, and in the event that it has not been revived, the counter associated using the worldwide trailblazer, called As far as feasible Global Limit Count (GLC), is extended by one; nonetheless, it is set to 0. The Worldwide limit computation is validated and distinguished from the Global Leader.

Neighborhood Pioneer Learning Stage: Utilizing a greedy assurance among the get-together people, the spot of the close by pioneer is revived in this piece of the estimation. The close by pioneer doesn't revive what is happening, a counter called Nearby Breaking point Count (LLC) connected with the local boss is expanded by one; regardless, the counter is reset to 0. This method is used to recognize the close by pioneer for each social affair. Close

by Breaking point Count is an expanded counter until it shows up at which is a right edge Local Leader Limit (LLL).

Neighborhood Pioneer Choice Stage: Nearby and overall trailblazers have been perceived going before this stage. Expecting any local boss fails to upgrade to a specific edge, known as the Nearby Pioneer Cutoff. All people from that social occasion ought to invigorate their situations through inconsistent presentation or using the overall trailblazer's knowledge. The aggravation rate is resolved using Equation (44).

$$\alpha_{nemij} = \alpha_{ij} + (vc_j(0,1)) \times (\alpha_{Lj} - \alpha_{ij}) + vc_j(0,1) \times (\alpha_{rj} - \alpha_{ij}) \quad (43)$$

Because the present nearby pioneer is exhausted (not renewed up to The number of focus), placements, as demonstrated by this scenario. Finally, the inquiry bearings and positions permit proceeding towards alternative FACT regulator configurations till an ideal arrangement is discovered, and the power misery and expenditures are calculated and examined in light of the situation.

Hence, the overall outline of the hybrid optimization technique is illustrated in the pseudocode form in figure 2.

```

Input: Multi-objective features based on Fitness evaluated  $c_{TCSC}, c_{svc}, c_{UPFC}$ 
Output: Optimal positioning of FACT controller

Begin
  Initialize the population using CI
   $\zeta_{ij} = rand \times (ub_j - lb_j) + lb_j, i=1,2,\dots,Dim$ 
  For  $i=1$  to  $Dim$ 
    Evaluate the exploration phase using,
     $\zeta_1(t+1) = \zeta_{best}(t) \times \left(1 - \frac{t}{T}\right) + (\zeta_M(t) - \zeta_{best}(t)) * rand$ 
    Update the Contour flight with short glide attack behavior using,
     $\zeta_2(t+1) = \zeta_{best}(t) \times Levy(D) + \zeta_R(t) + (Y - X) * rand$ 
    Update the low flight with a slow descent attack behavior using,
     $\zeta_3(t+1) = (\zeta_{best}(t) - \zeta_M(t)) \times \lambda - rand + ((ub - lb) \times rand + lb) \times \tau$ 
    Evaluate the exploitation rate
  For  $i=1$  to  $Dim$ 
    For  $j=1$  to  $Dim$ 
      Generate an best Exploitation using GI-SMO
       $\zeta_{nemij} = \zeta_{ij} + (ub_j(0,1)) \times (\zeta_{GLBj} - \zeta_{ij}) + ub_j(-1,1) \times (\zeta_{ij} - \zeta_{ij})$ 
      Finally update the solution using
       $\zeta_{nemij} = \zeta_{ij} + (ub_j(0,1)) \times (\zeta_M - \zeta_{ij}) + ub_j(0,1) \times (\zeta_{ij} - \zeta_{ij})$ 
    End for
  End for
  If  $\zeta_{nemij} \leq \zeta_{GBEST}$ 
    Return the best solution
  Else
    Initialize the AO population and repeat the steps.
  End if
End for
End begin
  
```

Figure 2: Pseudocode for Hybrid optimization

RESULTS:

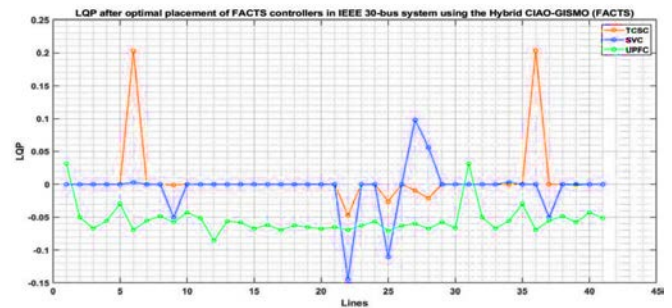
The standard IEEE 30 bus test framework has been applied to verify the relevance and legitimacy of the generated Hybrid advancement computation for responsive power arranging with TCSC and SVC devices positioned on weak buses. The suggested hybrid optimization computation was run for 500 emphases in each of the test frameworks to demonstrate its capacity for improvement, with the results of interest being listed in the tables 1. Table 1 shows, using the PSO technique, the MSL, optimal establishment cost, and least number of regulators anticipated for a 30-transport architecture.

Table 1: Evaluation of finest position, constraint setting, MSL, and finest cost of installing FACTS controllers in IEEE 30-bus system for the proposed method

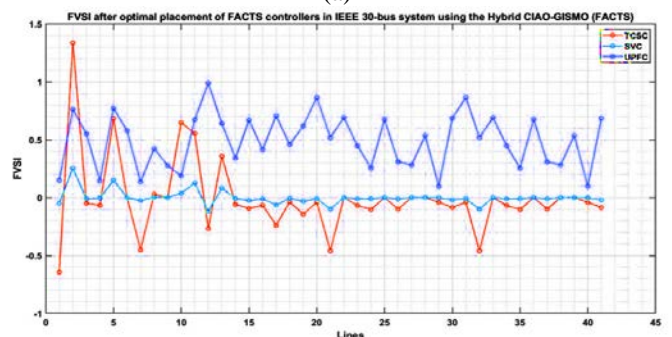
Considered stability	FACTS controller	Location	V _{facts}	MSL(PU)	C(F2)(10 ^{^4})
	NO FACTS	-	-	2.3220	-
Yes	TCSC	21-23	-	2.7764	0.1166
	SVC	30	1.06	2.3967	0.0965
	UPFC	5-7	1.06	2.3609	0.0950
NO	NO FACTS	-	-	2.1716	-
	TCSC	26-25	-	2.5833	0.2393
	SVC	29	1.01	2.5004	0.0273
	UPFC	12-13	1.01	2.6037	0.1948

If there should arise an occurrence devoid of and through constancy limitations, the improvisation of UPFC for the SL is from 2.3609pu to 2.6037pu. Also, in the 2belongings, TCSC gives an MSL of 2.7764pu and 2.5833pu, and SVC works on the SL to 2.3967pu and 2.5004pu, separately. Therefore, while contrasting the expenses by counting constancy imperatives, TCSC is the ideal choice. The proposed framework is steady at the greatest SL, utilizing a wide range of FACTS regulators (TCSC, SVC, and UPFC), as displayed in Figure 3.

Without FACTS, the framework has a voltage fluctuations of 2.3220 p.u. and a planned and actual power loss of 2.3220 p.u. In order to reduce the required multi-function, the ideal positioning and size of the SVC, TCSC, and UPFC in the IEEE bus system have also been assessed. As a result, the suggested strategy has been used, using a multi-objective function (F) strategy to choose the ideal weighting component values.



(a)



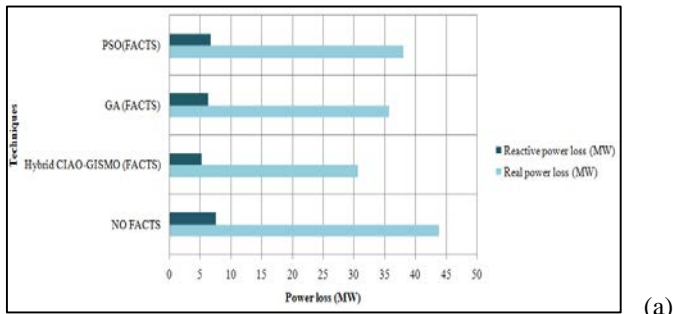
(b)

Figure 3: Optimal placement of FACTS controllers using proposed hybrid optimization based on (a) After FVSI (B) After LQP

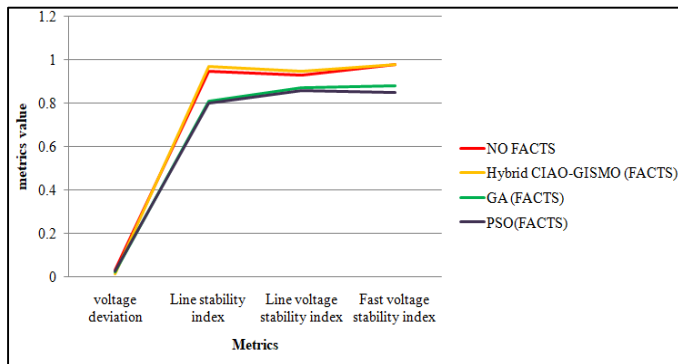
The proposed hybrid optimization for the FACT regulator location using the fast line stability factor and voltage stability index is displayed in Figure 3 as an exhibition research. Even if the voltage profile shifts as SL increases, the FACTS regulators can maintain a satisfactory cutoff point range for the power stream. All voltages are inside the cutoff points when the proposed hybrid optimization procedures are used, too. To ensure matrix strength at various SL levels while applying the suggested techniques, the energy and line constancy indices, FVSI and LPQ, are acceptable within the allowable reach. As a consequence, SL is increased and the cost of implementing FACTS regulators is reduced.

Table2: Evaluation of power loss and other factors for the proposed and existing techniques

Techniques	Actual power loss (MW)	Reactive power loss (MW)	voltage deviation	Line stability index	Line voltage stability index	Fast voltage stability index
NO FACTS	43.767	7.545	0.033	0.95	0.93	0.98
Hybrid CIAO-GISMO (FACTS)	30.637	5.282	0.014	0.97	0.95	0.98
GA (FACTS)	35.671	6.312	0.021	0.81	0.87	0.88
PSO(FACTS)	37.987	6.765	0.027	0.80	0.86	0.85



(a)



(b)

Figure 4: Active power loss variation with FACTS device in IEEE 30 bus system

Table 2 compares the proposed hybrid optimization strategy with other current methods like NO FACTS, GA, PSO, and others while evaluating the power loss and other factor changes. According to

the aforementioned statistics, the suggested hybrid approach results in an active power loss of 30.637 MW and a reactive power of 5.282 MW. In contrast, the current methods frequently result in considerable power loss, with real losses ranging from 43.767 MW to 35.671 MW. Additionally, compared to the existing approaches, the suggested method achieves a voltage variation of 0.014, LSI of 0.97, LVSI of 0.95, and FVSI of 0.98.

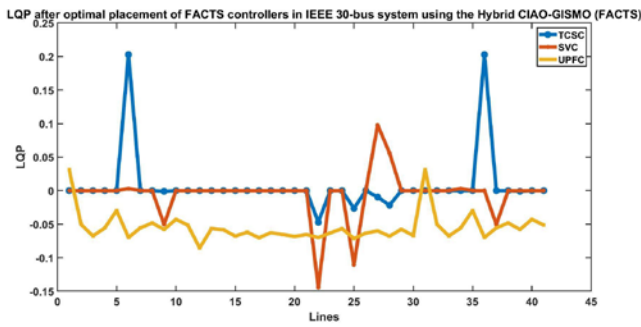
Using the FACTS method inside the IEEE 30 bus architecture, Figure 4 addresses the graphical comparison of the proposed strategy with current techniques given a range of power misfortune. The figures illustrate the best FACTS mixture for minimizing power loss when TCSC and SVC are combined. As a result, the suggested solutions reduce dynamic power loss during transmission.

Table 3: Evaluation of finest position, constraint setting, MSL, and finest cost of installing CIAO-GISMO FACTS

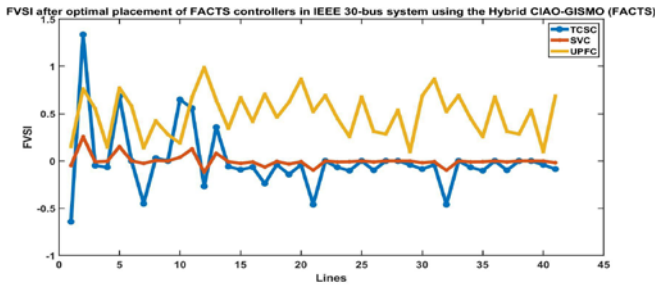
Considered stability	FACTS controller	Location	Vfacts	MSL(pu)	C(F2)(10^4)
	NO FACTS	-	-	2.3220	-
Yes	TCSC	21-23	-	2.7764	0.1166
	SVC	30	1.06	2.3967	0.0965
	UPFC	5-7	1.06	2.3609	0.0950
NO	NO FACTS	-	-	2.1716	-
	TCSC	26-25	-	2.5833	0.2393
	SVC	29	1.01	2.5004	0.0273
	UPFC	12-13	1.01	2.6037	0.1948

If an occurrence should occur that is both independent of and independent of consistency constraints, the UPFC improvisation for the SL is between 2.3609pu and 0.0950pu. Additionally, TCSC provides an MSL of 2.7764pu and 0.1166pu for the alternative possessions, while SVC works on the SL to 2.3967pu and 0.0965pu, independently. As a result, TCSC is the best option when comparing the costs using consistency imperatives. Figure 3 illustrates the suggested framework, which is stable at the highest SL and uses a variety of FACTS regulators (TCSC, SVC, and UPFC).

The system's voltage deviation and overall real power loss would be 2.3220 p.u. and 2.3220 p.u., respectively, without the installation of FACTS. In order to reduce the required multi-function, the ideal evaluated. As a result, the proposed method has been used to discover the most acceptable values for the weighting components utilising the multi-objective function (F) technique.



(a)



(b)

Figure 5: Graphical representation of finest position, constraint setting, MSL, and finest cost of installing CIAO-GISMO FACTS

Table 4: Evaluation of power loss and other factors for the proposed and existing techniques

Techniques	Real power loss (MW)	Reactive power loss (MW)	voltage deviation	Line stability index	Line voltage stability index	Fast voltage stability index
NO FACTS	43.767	7.545	0.033	0.95	0.93	0.98
Hybrid CIAO-GISMO (FACTS)	30.637	5.282	0.014	0.97	0.95	0.98
GA (FACTS)	35.671	6.312	0.021	0.81	0.87	0.88
PSO(FACTS)	37.987	6.765	0.027	0.80	0.86	0.85

Table 4 assesses the power loss and other factor changes while contrasting the suggested hybrid optimization strategy with other existing techniques like NO FACTS, GA, PSO, and others. The aforementioned statistics show that the proposed hybrid strategy causes a loss of active power of 30.637 MW and a loss of reactive power of 5.282 MW. The real losses from the present technologies, on the other hand, range from 43.767 MW to 7.545 MW, and usually cause significant power loss. The recommended method also achieves a voltage variation of 0.014, LSI of 0.97, LVSI of 0.95, and FVSI of 0.98 when compared to the existing approaches.

According to Table 1 and 3, the integration of the SVC into the power system, which is based on a variety of tested methodologies, is done in the scenario of reducing as a single goal function. As a result, the proposed approach offers the least size of SVC for obtaining the total active power loss's global minimum value scrutinized to the other approaches.

Additionally, the voltage deviation is reduced in the power system using the SVC and TCSC installation that is on the basis of proposed algorithms.

DISCUSSION

In power systems, determining the proper locations and FACTS device designs is difficult, and significant data collection is typically needed. In a reenactment scenario, the complexity of the objective time or space may cause a smearing computation that yields the problem's exact, ideal solution to fall short. Science based methodologies, number struggling to balance scripting approaches or classical make a positive difference, meta-heuristic process improvement methods, and half-and-half strategies were the four kinds into which the strategies and strategies can use for figuring out the best positions and settings of FACTS devices were divided in the existing research. These are recognized by the FACTS regulator' acceptance of control calculations designed to achieve a variety of objectives. The attempt to support power security has resulted in a novel multi-objective hybrid optimization improvement model combining CI-AO and GI-SMO methods. The standard Confidence Interval-based Aquila Optimizer (CI-AO) and Glorot initialization-based Spider Monkey Optimization are used to develop the hybrid optimization model (GI-SMO). The implementation of a balanced exploration and exploitation rate is necessitated by the employment of hybrid optimization approaches in order to control network complexity and early convergence, enhance rate performance, and save money and time. The primary goal of the work is the exact positioning of different FACTS regulator types has been utilized to MSL for load interest and power consistency. Voltage volatility, active power loss, and operating costs for power systems are reduced using the FACTS device. This non-optimized site produces undesirable goal values for the optimization of single- and multi-objective functions. Raising SL is likely to reduce FACTS regulators' speculative costs because they are expensive. The UPFC seeks to lower transmission loss in the influence framework by collaborating with both series and shunt FACTS regulators. The best locations within the linked power framework network to add FACTS devices in order to increase the framework's general exchange limit and achieve the lowest working expense possible under various stacking circumstances.

LIMITATION

- The Higher conductivity power system loss, system overhead cost, and power dissipation should be considered
- The Premature completion and increased network complexity has to be limited in terms of generation
- The ideal location of the FACT gadget is complicated and taken into consideration as well.
- Low rate performance, higher implementation costs, and time-consuming should be made on limits.

CONCLUSION AND FUTURE WORK

The paper offers the most robust evolutionary hybrid optimization method, CIAO-based GI-SMO. It has been used to MSL by the precise positioning of several types of FACTS regulators for power consistency and load interest in the power framework. Voltage volatility, active power loss, and power system operating costs are reduced by using the FACTS device. Single- and multi-objective optimization approaches are used to achieve these aims. The evaluation's findings show how effectively the special tactic (CIAO-GISMO technique) optimises both single- and non-linear and non-functions. Additionally, the FACTS device's locations and ratings have been established concurrently. The results demonstrate how reduced voltage variance, system running expenses, and power loss result from enhanced FACTS device allocation. Additionally, employing this non-optimized location results in unsatisfactory goal values for the optimization of single and multi-objective functions. Because FACTS regulators are expensive, increasing SL is likely to reduce their speculative costs. To decrease transmission loss in the influence framework, the UPFC is working with both series and shunt FACTS regulators. The work determines the best locations for FACTS devices inside the linked power framework network to be installed in order to increase the general exchange limit of the framework and achieve the lowest possible working expense under various stacking situations. In trials comparing the constructed method to other evolutionary optimization methods, it performed satisfactorily.

This approach has several benefits, including a high convergence rate, balanced exploration and exploitation characteristics, and superior computing efficiency. It can be optimized and improved considering the development of algorithms. The results also manipulated that the system's liability may be raised successfully while preserving respectable levels of stability and security. Every gained result supports and endorses the suggested technique.

CONFLICT OF INTEREST STATEMENT

Authors declare that there is no CoI (academic or financial).

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