



An-Najah National University
Faculty of Graduate Studies

**A HYBRID FIREFLY-GENETIC ALGORITHM
FOR THE OPTIMAL COORDINATION OF
DIRECTIONAL OVERCURRENT RELAYS**

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of Master of Electrical Power Engineering, Faculty of Graduate Studies, An-Najah
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Dedication

This research is dedicated to:

My great parents, who never stop giving of themselves in countless ways;

Our friends and colleagues who support us through all obstacles and issues;

An-Najah National University; my second magnificent home

The dear supervisor Dr. Maher Khammash on their generous efforts in the service of
this study.

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Declaration

I, the undersigned, declare that I submitted the thesis entitled:

A HYBRID FIREFLY-GENETIC ALGORITHM FOR THE OPTIMAL COORDINATION OF DIRECTIONAL OVERCURRENT RELAYS

I declare that the work provided in this thesis, unless otherwise referenced, is the researcher's own work, and has not been submitted elsewhere for any other degree or qualification.

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Date:

19/02/2023

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Abstract

Theoretical background: Directional overcurrent relays are applied for power system protection to ensure safe, reliable, and efficient operation. The coordination of directional overcurrent relays is non-linear and highly constrained optimization problem. The main goal of the optimization is to minimize the summation of operating times of primary relays, by setting optimal values for decision variables as time multiplier setting (TMS) and plug setting (PS).

Aims: The main objective of this research is to develop a hybrid optimization algorithm which consists of modified firefly algorithm and genetic algorithm to find better solutions.

Methodology: First, this study modified the original firefly to obtain a global solution by updating the firefly's brightness and to avoid the distance between individual fireflies from being too far. Additionally, the randomized movements were controlled to produce a high convergence rate. Second, the optimization problem is solved using standard genetic algorithm. Finally, the solution obtained from the modified firefly algorithm is used as the initial population for the standard genetic algorithm. The modified firefly algorithm, genetic algorithm and hybrid firefly-genetic algorithm have been tested on IEEE 3-bus, 8-bus, 9-bus and 15-bus networks.

Main Results: The results indicate the effectiveness and superiority of the proposed algorithms in minimizing the overall operating time of primary relays compared to other algorithms mentioned in the literature for directional overcurrent relays coordination.

Conclusion: Compared to modified firefly algorithm and standard genetic algorithm, the proposed hybrid algorithm has minimized coordination interval time between primary and backup relay pairs.

Keywords: Directional overcurrent relays optimization, Hybrid algorithms, Firefly algorithm, Genetic Algorithm.

Chapter One

Introduction and Theoretical Background

1.1 General Background

Because of the accelerated expansion of the distribution electric power networks resulting from increasing consumer consumption and the requirement for more reliable service that meets regulatory standards, the distribution electric power system requires robust and effective protection system [1]. The distribution system is mainly protected by overcurrent protection devices such as solid-state power switches, conventional reclosers, fuses, and relays. These devices have the problem of being unable to determine the direction of current, despite the fact that relays have been stated as being able to do so [2].

Directional overcurrent relays are utilized for power system protection to ensure safe, reliable, and efficient operation. The relay monitors the current flow in the circuit that is being protected. When the short circuit current flow in a specific direction exceeds a preset value, A trip signal is sent to the circuit breaker by the relay [3]. These relays have been used to the design of economical alternatives for primary and backup power system protection [4]. When a fault occurs, primary relays activate and are supported by backup relays, which activate when the primary relay fails [5]; this ensures that just the defective area of the network is isolated. This may be accomplished by appropriately determining the operating time of the relays [6].

Generally, these relays have two settings: time multiplier setting (TMS) or time dial (TD) and plug setting (PS) or pickup current (IP). The operating time of each relay is determined by these parameters [7,8]. The relay coordination studies attempt to determine each relay's TMS, PS, and type of tripping characteristics. The overall operating times of the primary relays should be reduced to ensure minimal network outage. To ensure that the selectivity study is valid, the specified coordination time interval (CTI) between the primary and backup protections must be maintained [9].

For the coordination of the directional overcurrent relays, the total operational time of primary relays should be minimized as much as possible by maintaining coordination among other relays. A robust time multiplier setting is also necessary to accommodate

all possible operational conditions [10]. As a result, the coordination problem can be formulated as an optimization problem with the purpose of decreasing overall total operational time of primary relays while taking into account various constraints and boundary limits [11].

1.2 Problem Statement

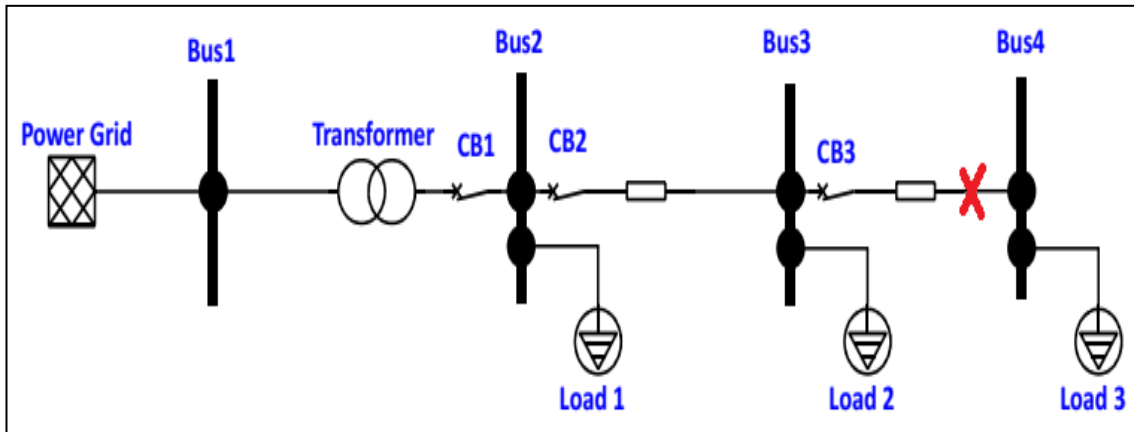
The process of modifying protection device settings such that they operate in a specific order to minimize power service disruption caused by a power system disturbance is known as coordination of protection devices [12]. When a fault occurs in the power system, the relay closest to the fault must trip first. If the closest protection relay fails to detect the fault due to a longer operating time, the backup protection relay will trip quickly to prevent the fault from propagating throughout the power system. Thus, coordination between the primary and backup relays is necessary for avoiding malfunctions.

To demonstrate the coordination problem, consider Figure 1.1, which illustrates a radial power system with a fault at point (X). CB3 will act as the primary protection, while CB2 will operate as backup protection. For this fault, CB3 must be opened, but CB2 and CB1 must remain closed. Under these conditions, only load 3 is interrupted. We may choose a longer time delay for the relay at CB2, allowing CB3 to operate first. We also choose the CB1 relay with a higher time delay than the CB2 relay, so that the CB2 relay opens first. There is an interval between the relays called coordination time interval (CTI). That is the amount of time it takes for circuit breakers to clear the fault under primary relaying after backup relaying has started operating [13]. This emphasizes protective relaying coordination, which involves selecting a proper settings for each relay so that their primary protection function is accomplished while maintaining the existing features of protective relaying, including reliability, speed, sensitivity, and selectivity [14].

To obtain the optimum settings of the directional overcurrent relays, it is possible to define the coordination problem as an optimization problem. Various techniques and methods have previously been developed to solve the problem.

Figure 1.1

Radial Network



1.3 Significance of Research

The significance of this research derives from the importance of integrating two optimization algorithms to solve the coordination problem. The main purpose of this study is to investigate the coordination problem of the directional overcurrent relays and to solve it by developing modified firefly algorithm, standard genetic algorithm and new hybrid optimization technique to obtain better solution.

1.4 Research Objectives

The objectives of this study can be summarized as follows:

1. Review the optimization algorithms utilized to solve the coordination problem.
2. Develop a modified version of the firefly algorithm to solve the coordination problem.
3. Use the standard Genetic algorithm to solve the coordination problem.
4. Solve the coordination problem by combining two metaheuristic algorithms, firefly algorithm and genetic algorithm to get better solution.
5. Accessing the performance of modified firefly algorithm, genetic algorithm and hybrid firefly genetic algorithm by implementing them to the standard IEEE 3-bus, 6-bus, 9-bus and 15-bus test networks.
6. Verify the proposed optimization techniques by comparing them to up-to-date optimization algorithms that have been utilized to solve the coordination problem.

1.5 Research Scope

The scope of this research will be focused on electrical protection systems and the application of optimization algorithms to solve electrical protection system problems such as the overcurrent coordination problem.

1.6 Overcurrent Protection in Power System

To better understand the overcurrent protection concepts, this section reviews the fundamentals of the overcurrent protection, beginning with the major principles of the protection system. After that, discusses the causes of overcurrent in power system and the need for overcurrent protection. Moreover, the various overcurrent relay types and their directionality characteristic. Finally, explains the protection coordination of the overcurrent relays.

1.6.1 Protection System Concepts

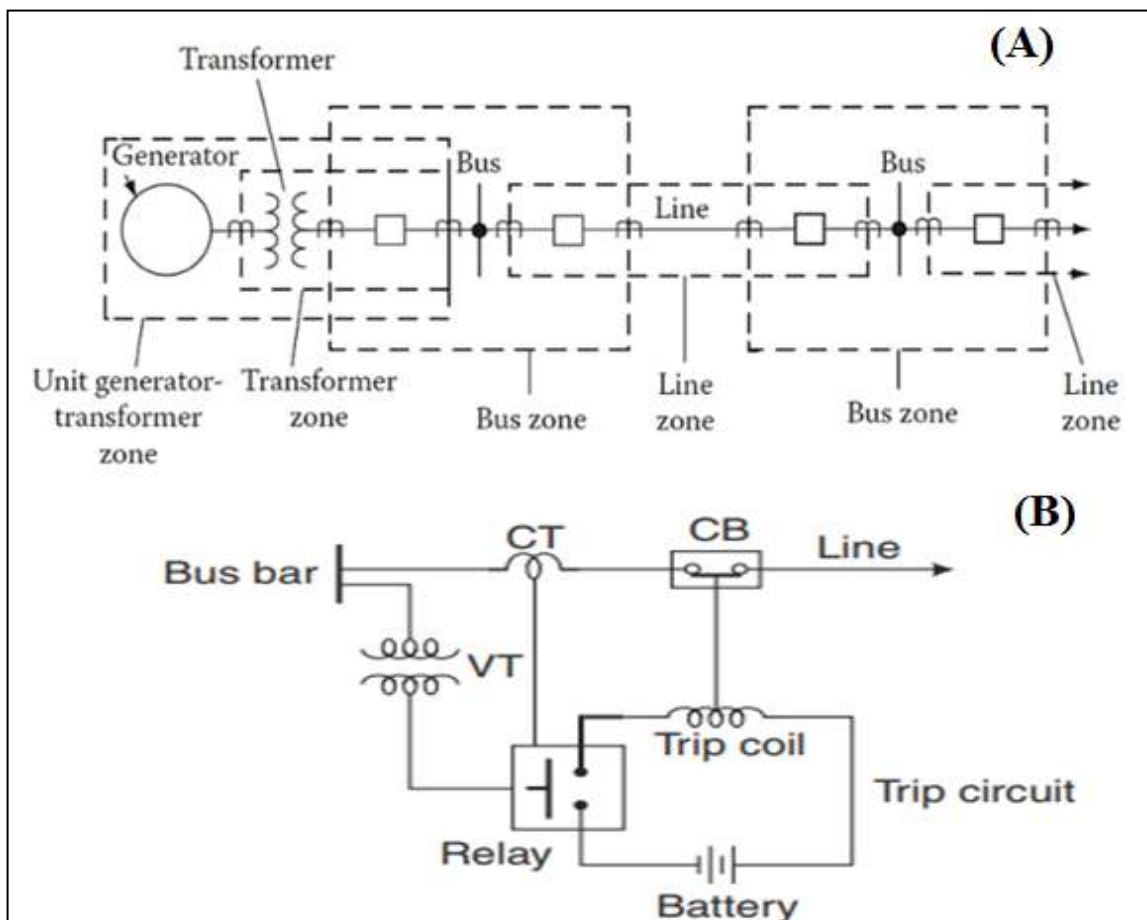
Protection system must operate properly (reliable), isolate the fault by disconnecting only the system component that is faulty (selectivity), clear the fault in the shortest amount of time, and ensure coordination at the lowest feasible cost. The general philosophy to implementing protection to power systems is to divide it into separate zones. Protection devices are chosen and assigned to protect a specific portion of the power system. The protective device is in responsibility of this section, which is referred to as the primary zone of protection [15]. These protective zones include the whole power system, protecting each part. Protection of sections that are on the boundaries of two protected zones as well as sections that may not be in either of these zones is made possible by the overlap of these zones as shown in Figure 1.2 (A) [16,17]. In the design of a protection system, the primary protection is backed up by a backup protection that must trip if the first one fails [18]. In order to allow the primary protection enough time to operate, The primary protection's operational time must be longer than the backup relay [15].

The main components of a protective system are presented in Figure 1.2 (B), which are as follows [19]:

1. Transducers: Current transformers (CT) and potential transformers (PT) provide the information necessary for fault detection at standard lower levels. Additionally, in order to protect protective relays from the high voltages of the power system, currents are typically reduced to either 5-A or 1-A and voltages are minimized to 110-V or 120-V [15,19].
2. Protection Relay: a relay whose purpose is to detect abnormal or dangerous power system circumstances or other faulty equipment conditions and to initiate the proper control circuit action [17].
3. Circuit breakers and trip circuit containing trip coil and battery: to isolate the faulty parts [15].

Figure 1.2

Protection System Concepts



Note: Subfigure (A) represents the protection zones in power system and it is adapted from [17], subfigure (B) represents the components of a protection system and it is adapted from [19].

The relay that is connected to the CT and VT activates and closes its contacts when there is a short circuit in the protected circuit, completing the trip circuit. In the trip circuit, current is being supplied by the battery. The circuit breaker's trip coil activates the operating mechanism, which powers the opening process to disconnect the defective element [19].

1.6.2 Overcurrent Protection

Overcurrent may result from either abnormal system circumstances, such as overload and short circuits, or from normal system conditions, including transformer inrush current and motor starting. Protective relays are included in power systems so that switching equipment can only activate in response to abnormal system circumstances [15].

1.6.2.1 Overcurrent Relays

The least expensive and easiest type of protection is provided by overcurrent relays [19]. Any power system element that has a fault will nearly always experience a short circuit current that is larger than the pre-fault load current. Using the current magnitude as a fault indicator is a very basic and efficient relaying principle [20]. Overcurrent relays respond to the amount of the input current and activate when the input current rises above the specified value. The relay closes its trip contacts and energizes the circuit breaker trip coils when this threshold is reached [16].

1. Overcurrent Relays Types

Based on their operating time characteristics, overcurrent relays can be divided into the following categories:

- a. Instantaneous overcurrent relay (definite current): When the current reaches a specific level, the relay functions instantaneously as shown in Figure 1.4 (A). The only operating condition is current magnitude, and it operates in 0.1 seconds or less [16].
- b. Definite time overcurrent relay: As illustrated in Figure 1.3 (B), after a specified current level is reached, these relays have an operating time that is unaffected by the current magnitude. This type of relay has two settings: a time-dial setting to select

the actual time the relay functions, and a current setting (also called as a pickup, plug, or tap setting) to choose the current value at which the relay will operate [15].

- c. Inverse time overcurrent relay: An inverse-time overcurrent relay activates when the current is greater than its pick-up value. The operating time is determined by the operating current's magnitude. As the current rises, the operating time reduces. The inverse time-current characteristic of these types of relays is shown in Figure 1.3 (C) [19].
- d. Inverse definite minimum time (IDMT) overcurrent relay: The IDMT overcurrent relay combines the characteristics of definite current and inverse time for discrimination. Figure 1.3 (D) shows the curve representing its operating characteristic [16]. In general, if the plug setting multiplier value is less than 10, an inverse-time characteristic is obtained. The characteristic tends to become a straight line, or toward the definite time characteristic, for plug setting multiplier values between 10 and 20. Relays for IDMT are frequently used to protect distribution lines [19].

The further classes listed below are based primarily on the inverse characteristics of overcurrent relays shown in Table 1.1 and are defined by IEC 60255-3 and IEEE C37.112-1996 standards [21].

Figure 1.4 shows the overcurrent relay's standard tripping characteristics curves, Figure 1.4 (A) depicts the IEC standard characteristics, whereas Figure 1.4 (B) represents the IEEE standard characteristics (B).

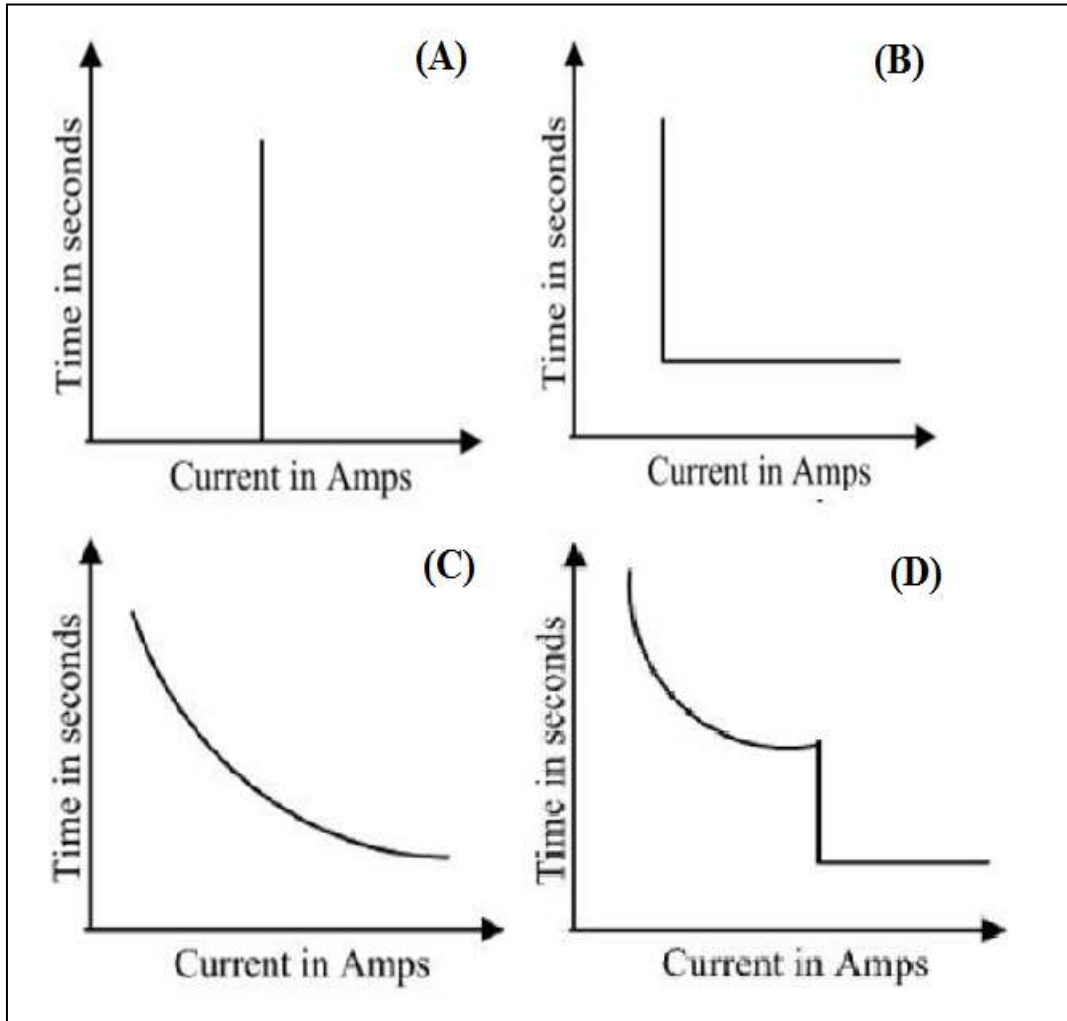
Table 1.1

Standard inverse characteristics of overcurrent relays defined by IEC and IEEE.

Standard characteristics			
IEC		IEEE	
IEC short time inverse	STI	Moderately inverse	MI
IEC normal/standard inverse	SI	Very inverse	VI
IEC very inverse	VI	Extremely inverse	EI
IEC extremely inverse	EI	Inverse	I
IEC long-time inverse	LTI	Short Inverse	SI
		Long Inverse	LI

Figure 1.3

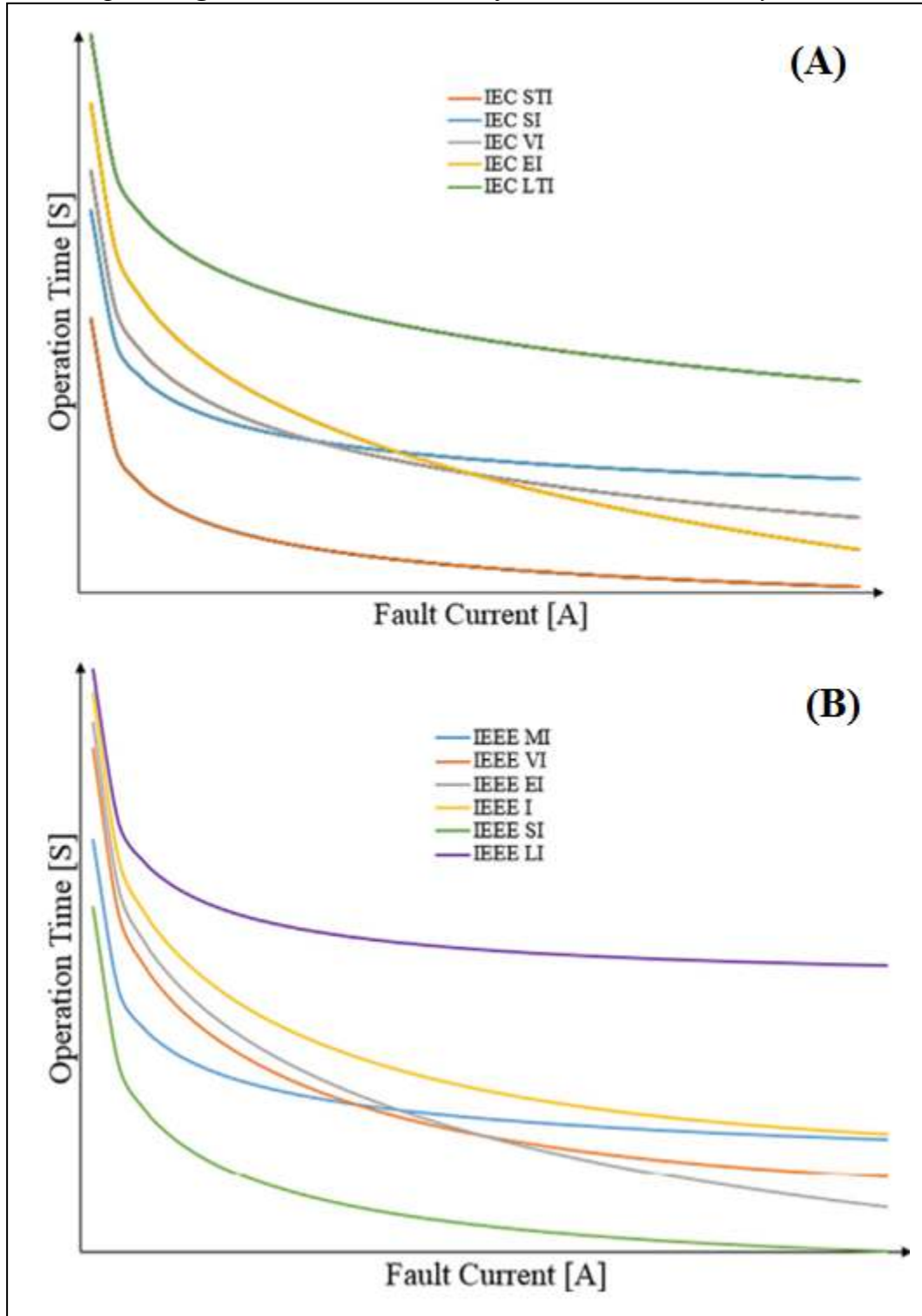
Overcurrent relay operating time characteristics.



Note: Adapted from [16]. Subfigure (A) represents definite current operating characteristic of the overcurrent relay, subfigure (B) represents definite time operating characteristic of the overcurrent relay, subfigure (C) represents inverse time operating characteristic of the overcurrent relay and subfigure (D) represents IDMT operating characteristic of the overcurrent relay.

Figure 1.4

Standard operating characteristic curves of the overcurrent relay



Note: Adapted from [21]. Subfigure (A) represents the IEC standard operating characteristic curves of the overcurrent relay and Subfigure (B) represents the IEEE standard operating characteristic.

2. Operating Time of Overcurrent Relays

IEC 60255 and IEEE C37.112 specify the overcurrent relay's operational time in Equation (2.1) as follows:

$$t = \frac{T_D \alpha}{\left(\frac{I_{sc}}{I_{PU}}\right)^\beta - 1} + C \quad (2.1)$$

Where:

t: the relay operating time in seconds

T_D : the time dial, or time multiplier setting

I_{sc} : the fault current level in secondary amperes

I_{PU} : the tap or pickup current setting

C: the constant

α : the slope constant

β : the slope constant

The slope of the relay characteristics is specified by α and β . In Table A.1, the values of α , β , and C for several standard overcurrent relay types manufactured in accordance to IEC and IEEE Standards are given [22].

3. Current (plug) Pickup Setting

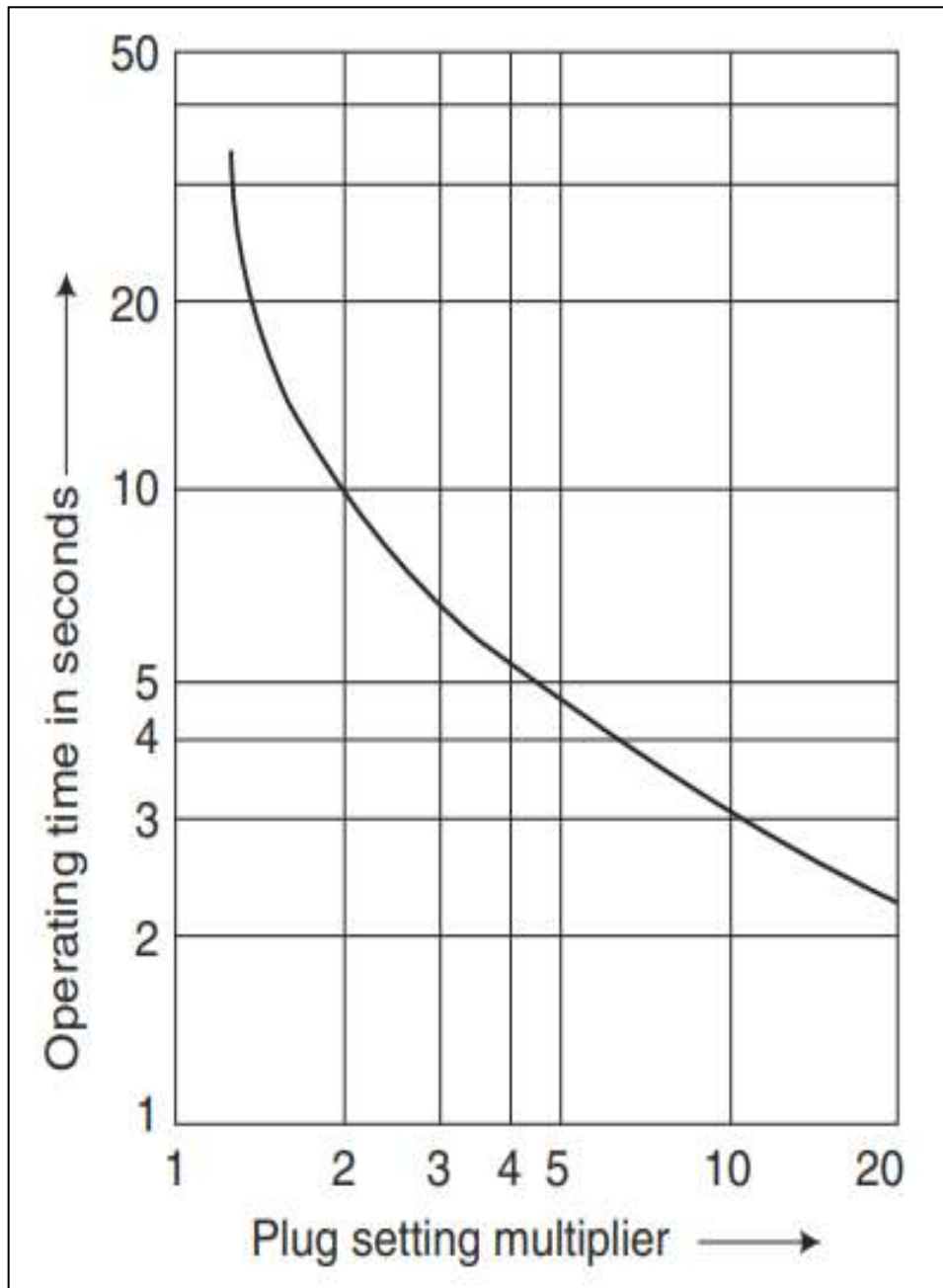
The level of current at which the relays will operate is determined by this setting [23]. The plug setting multiplier (PSM) measures the actual RMS current passing through the relay as a multiple of the setting current. However, PSM can be expressed as:

$$\begin{aligned} PSM &= \frac{\text{Secondary Current}}{\text{Relay current setting}} \\ &= \frac{\text{Primary Current during fault}}{\text{Relay current setting} * CT \text{ ratio}} \end{aligned} \quad (2.2)$$

A time-current characteristic with PSM on the X-axis is shown in Figure 1.5. Typically, a log/log graph is used to plot the curve. The operating time for various relay settings can only be specified by this curve [19].

Figure 1.5

Standard I.D.M.T. characteristics



Note: Adapted from [19].

The following expression illustrates how to determine the pickup setting for phase relays by allowing an overload margin exceeding the nominal current [15]:

$$\text{Pickup setting} = \frac{k_{ld} * I_{nom}}{CTR} \quad (2.3)$$

Where:

K_{ld} : overload factor that based on the element being protected.

I_{nom} : nominal rated current.

CTR: current transformer ratio.

When choosing the pickup setting for earth-fault relays, the maximum unbalance that could possibly exist in the system under normal operating conditions is taken into consideration. 20% is a typical imbalance allowance, thus the expression in (2.3) becomes [22]:

$$Pickup\ setting = \frac{0.2 * I_{nom}}{CTR} \quad (2.4)$$

4. Time Setting

Time dial setting modifies the time delay before the relay functions whenever the short circuit current equals or exceeds the relay current setting. In electromechanical relays, where the time delay is frequently obtained by adjusting the physical distance between the moving and stationary contacts, a lower time dial value results in shorter operating times. The time dial setting is also known as the time multiplier setting (TMS) [22]. There are ten steps in which time can be set. For these time-setting steps, the term TMS is used.

1.6.2.2 Directional Overcurrent Relay

Nowadays, with the expansion of the electric grid and the increasing complexity of the power system, the protective system has become an essential part of the power system. Directional type overcurrent relays are employed for bidirectional power flow feeders. [24]. Utilizing the directionality function, which operates in the forward or reverse value depending on the current and voltage values and the polarization angle, can prevent incorrect operation. The phase shift of the current angle relative to the voltage angle determines the directionality torque function, therefore the directional protection relay has to be fed from both the secondary sides of the line current and voltage transformers [25].

1.6.3 Protection Coordination of Overcurrent Relays

The primary relay must function initially to protect its own zone through an appropriate time interval, with all the fault-free parts remaining in operation. This is the basis for the coordination of overcurrent protection relays [15]. If the primary relay fails to isolate a fault within its own zone, the backup relay should eventually be able to do so by isolating the fault but with a larger area of the faulted zone, which includes part of the non-faulted zone [11]. The settings for the relays are modified such that the primary relay nearest to the defect responds quickly to operation, followed by all of the backup relays [26]. The different types of coordination are time-based coordination, current-based coordination and logic coordination.

1.6.3.1 Coordination Time Interval

Time-current coordination of the overcurrent relays is required to ensure selectivity and isolate only the faulty part. With a specific coordination time interval (CTI), a downstream device must operate more rapidly than the upstream device. In order for series-connected protection devices to operate sequentially, a minimum CTI must be maintained throughout the operational range. Typically, a CTI on the order of 0.20-0.50 seconds should be utilized between two sequential time/current characteristics. Due to one or more of the following, this value prevents selectivity from being lost [18]:

- Circuit breaker interrupting time.
- Relay overrun time after the fault has been cleared.
- A safety factor that takes CT saturation and setting errors into consideration [22].

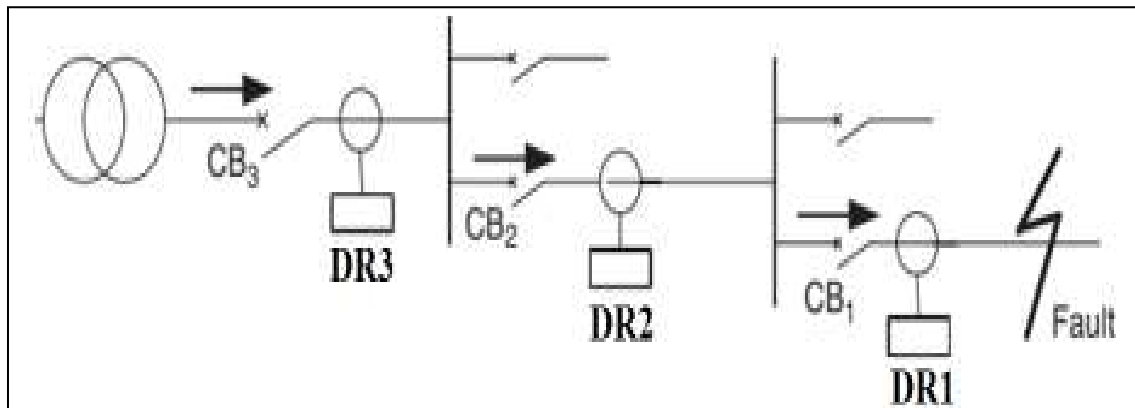
1.6.3.2 Time-Based Coordination

During fault conditions, accurate discrimination could be achieved by utilizing relays configured to operate at different time delays. It is obvious that the timing differences between the relays connected to adjacent sections could be sufficiently large to allow the relevant circuit-breaker to open and isolate the fault on its section before the relay connected to the adjacent section closer to the source could start opening its circuit-breaker [27]. As depicted in Figure 1.6, the fault is detected by relays DR1, DR2 and DR3. The DR1 functions faster than DR2, which in turn functions faster than DR3. Protection units DR2 and DR3 return to the standby state once the CB1 is tripped and the problem is resolved. The difference in operating time Δt between two subsequent

protection relays is CTI. This coordination system benefits from being simple and providing its own backup; for instance, if DR1 fails, DR2 is activated after Δt [15]. This type of coordination, meanwhile, would fall short in networks where the fault current varies based on where the fault lies [16].

Figure 1.6

Radial Power System Protection



Note: Adapted from [15].

According to the protection characteristics, three different time delay types are used: definite time, inverse-time, and combined inverse-time and instantaneous protection systems [15].

- Definite-time protection: as seen by the protection tripping curves shown in Figure B.1, the time delay is constant and independent of the current [15]. For this type of protection, it is recognized that relays utilized close to the sources in networks with different sections connected in series may have unacceptable high operation times for faults in the sections they protect since such faults could be at current levels which can only be permitted to exist for short durations [27].
- Inverse-time protection: As illustrated in Figure B.2, inverse curves operate faster at greater fault currents and slower at lower fault currents [23]. For a fault at the end of the section closer to the power source, a relay whose delay is inversely proportional to the current would trip faster in this case. In the event where the selection intervals are the same, inverse time relay can offer faster clearing times than definite time relay [28].

- Combined inverse-time and instantaneous protection: It operates on the same IDMT concept and trips instantly at the highest downstream short circuit current. This type of tripping curve is shown in Figure B.3 [15].

1.6.3.3 Current-Based Coordination

In circuits where there will be a significant variation in the ratio of fault current to rated current in different parts of a network, current-based coordination can be employed to achieve proper discrimination. As illustrated in Figure B.4 [27], the operating time is maintained constant for all relays employed to protect various feeder segments. This coordination is based on the concept that discrimination is made possible by decreasing the current setting as one moves from the load to the power source. Thus, the relay nearest to the fault will always trip first [16]. The benefit of this system over the time coordination technique is that it requires less operating time when close to the power source [19]. This approach has the drawback that the fault current does not necessarily change with the location. As a result, accurate relay discrimination is not achievable [16].

1.6.3.4 Time-Current Based Coordination

The most common method of coordination is the time-current coordination scheme [16]. They combine the characteristics of time-and current grading. IDMT relays are employed in this scheme. Time and current settings are available for IDMT relays. According to the short circuit current level of the specific zone that has to be protected, the relay's current setting is set. The relays are configured to pick up gradually at greater current levels, in the direction of the source. Furthermore, time setting is also described in progressively higher order away from the source [19].

1.7 Optimization Techniques Applied For Overcurrent Relays Coordination Problem

This section provides a survey of the conventional optimization techniques, heuristic techniques, and hybrid optimization techniques that have been used for solving the coordination problem.

1.7.1 Optimization Techniques

For solving the coordination problem, different techniques have been developed. These techniques can be divided into four categories: dual setting protection schemes, new constraints for optimal coordination, non-standard characteristics, and optimization techniques. The optimization algorithms applied for the coordination problem are presented in this section. Relay coordination is characterized as an optimization problem that can be resolved by employing optimization techniques that fall under the following categories represented in Figure C.1.

1.7.1.1 Mathematical Based Optimization Techniques

Several electrical engineering and relay coordination problems were solved using various optimization techniques. Conventional techniques have the idea that fault studies, system emergencies, and abnormal conditions should be predetermined [5,24,29]. The conventional techniques involve trial and error, and topological techniques [30]. Power system engineers have previously used a trial-and-error methodology to determine the best relay settings. However, the rate of convergence for this methodology is slow due to the several iterations required to find a suitable relay setting, and it may not always produce acceptable relay settings [8,12,30–37]. Breakpoint technique was applied to solve this problem [38]. The break points are determined using topological methods based on graph theory and functional dependency to decrease the number of iteration needed for relay coordination process [5,24,37,39,40]. The solutions obtained from these techniques are among the best of all feasible solutions, but they are not optimal. In other words, the relays' TMS are quite large [24,37,40–42].

The coordination problem in such complex networks was solved using a computer with a graphical user interface. In [43], the coordination problem of the inverse definite minimum time over current relays has been solved using Computer Aided Protection Engineering (CAPE). The graphical user interface of CAPE is useful since it enables protection engineers to make any required adjustments, revisions, or even upgrades to the current setting to ensure good coordination with the network system. Matlab Graphical User Interface (GUI) has been implemented to solve the inverse-time and instantaneous overcurrent relays coordination problems as presented in [44]. The GUI provides curves of various standardizations that can generate coordination charts by

varying the slopes of inverse-time curves and the adjustment parameters of instantaneous and timed overcurrent functions.

Deterministic methods, which are characterized as mathematically-based optimization techniques which are utilized extensively to solve overcurrent relays coordination problems, are classified as shown in Figure C.2 [11,29].

Depending on the nature of settings in the coordination problem, the protective relay coordination problem typically modelled as either an LP, NLP, MILP or MINLP problem [44]. LP has some benefits such as fast computational time and simplicity to resolve, but it needs expertise to set the initial guess for pickup current settings and may get stuck in local minima [34,44]. The Big-M method, in which the PS are considered to be known and fixed, has been suggested in [45] to obtain the optimum value of time multiplier settings of directional overcurrent relays. Due to the non-linear nature of the relay configuration, it is extremely inefficient and complex [29,46]. The directional overcurrent relays problem was stated as a NLP problem to overcome the challenges by designating TMS and PS as decision variables [2]. The optimum relay settings have been obtained applying the non-linear programming Rosenbrock-Hillclimb technique in [47]. Both the decision variables are calculated and optimized in MINLP [48]. As a result of the coordination problem's non-linearity and non-convexity, NLP and MINLP techniques are susceptible to become trapped in local minima. Therefore, a formulation based on MILP has been suggested for the coordination problem in [49] to overcome these problems by taking into account the PS as a discrete optimization variable and the TMS as a continuous optimization variable. The problem has been solved using the branch and bound approach. Therefore, at each branch, the proposed formulation converts the nonlinear and nonconvex coordination problem into a linear and convex problem. This matter assures convergence to global optimal settings and makes the problem easier to solve.

These methods' main drawback is that they are highly dimensional, and in order to solve the coordination problem, they require a lot of computational time and computer memory [29]. The problem of a minimum solution plagues conventional methods, especially in the case of complex systems, despite the fact that they provide a considerable contribution to the problem's solution. Convergence is also difficult to achieve [11]. Table 1.2 compares between these mathematical algorithms.

Authors in [39] suggested analytical method to determine the optimal directional overcurrent relays settings. The suggested numerical technique converges to the global optimal values, which are independent by the relay setting order and initial values. To accurately determine the critical fault point to coordinate overcurrent relays, authors in [53] presented an analytical method for calculating the network's impedance matrix in a fault condition. Relay coordination problems for mid line faults were successfully optimized using a gradient search-based method in [54]. These techniques often use a lot of iterations to calculate the relay settings and fall short of determining the optimal setting of the relay for an interconnected power network. Nevertheless, for radial systems, analytical methods are very effective [48,55,56].

Table 1.2

Comparison between mathematically-based optimization methods.

Reference	Mathematical algorithm	Remarks
[45,50,51]	LP	The only design variable considered is TMS, whereas PS values are fixed between allowable overload and minimum short circuit currents. As a result, each relay's operational time is computed as a linear function of time dial setting.
[2]	NLP	Both the TMS and the PS are regarded as decision factors and are regarded to be continuous variables.
[52]	MILP	The PS values are defined as discrete variables and the TMS as continuous variables.
[48]	MINLP	TMS values are considered as continuous variables and PS values are regarded to be integer variables.

1.7.1.2 Meta-Heuristic Optimization Techniques

Researchers and scholars have recently used various heuristic, metaheuristic, and evolutionary algorithms to obtain solutions to electrical engineering problems, such as the optimum settings of overcurrent relays coordination [33,57–59]. Artificial intelligence has recently attracted a lot of interest for its ability to optimize directional overcurrent relays settings [24,48,60,61]. They populate a number of solutions to start the optimization process rather of starting with just one, as in conventional ones. Additionally, for complex tasks, their gradient independence can increase their flexibility [55]. The protection system's speed can be improved by choosing the appropriate characteristics for

modern directional overcurrent relays, though. Artificial intelligence methods that take into account the operation characteristics of the relay as an optimization parameter together with time multiplier settings and plug settings can help achieve this [62]. The operator carefully examines the power system limits before selecting the relay settings in the conventional optimization of relay coordination. The maximum fault current, load current, fault clearing time, current transformers ratios, and other constraints must be checked in order to yield adaptive curves. These checks are not supported by many commercial computer software products. Relay coordination has been solved by artificial intelligence in terms of a constrained objective function [63].

Different metaheuristic algorithms has been applied to optimize the coordination problem, such as Genetic Algorithm (GA) [64], Particle Swarm Optimization (PSO) [65], Seeker Algorithm (SE) [48], Crow Search Algorithm (CrSA) [61], Firefly Algorithm (FA) [66], Honey Bee Algorithm (HBA) [67], Differential Evolution (DE) algorithm [32], Harmony Search Algorithm (HSA) [42], Artificial Bee Colony (ABC) [68], Cuckoo Search Algorithm (CSA) [69], Whale Optimization Algorithm (WOA) [70], Water Cycle algorithm (WCA) [71], Teaching Learning Based Optimization (TLBO) algorithm [72], Harris Hawk Optimization (HHO) [73], Grey Wolf Optimizer (GWO) [74], Improved Group Search Optimization (IGSO) [75], Sine Cosine Algorithm (SCA) [76], Shuffled Frog Leaping Algorithm (SFLA) [77], Symbiotic Organism Search technique (SSO) [78], JAYA algorithm [79], and Seagull Optimization Algorithm (SOA) [31].

1.7.1.3 Hybrid Optimization Techniques

There are significant drawbacks to artificial intelligence based optimization techniques, including premature convergence, higher computation time, initial solution sensitivity, and significant speed differences. Due to the fact that no single algorithm is capable of solving all possible varieties of optimization problems, there is always a chance of convergence to local minima [11,55,80,81]. Hybrid optimization techniques, which integrate different optimization algorithms, are employed to overcome these problems. They make it possible for each approach to work more effectively and accurately than it could on its own [11,55,80].

GA has the shortcoming of sometimes converging to values that are not optimal, whereas the drawback of NLP approaches is that they tend to converge to local optimal settings when the initial choice is closer to the local optimum. GA is a multipoint search technique that explores a large solution space, in contrast to conventional single point search techniques. In [82], the authors effectively combined the features of GA with NLP to find the optimal relays settings. The authors in [83] suggested a technique for adaptive coordination of overcurrent relays considering the topological changes of the radial and meshed networks using the existing relay setting groups. A hybrid GA and LP technique has been employed to solve the issue, where the GA categorizes the scenarios of network topology changes into a limited number of setting groups in a near-optimal manner, and the LP method optimally coordinates the overcurrent relays within the setting groups. In [84], the coordination problem has been solved using hybrid optimization method that combined the integer coded genetic algorithm ICGA and NLP techniques. The authors in [85] developed a method for optimizing overcurrent relay coordination using computational intelligence approaches applied to adaptive protection in the context of DG. The developed hybrid algorithm utilizes fuzzy logic to control the current settings and GA to determine the relay's TDS and curves.

The convergence rate of PSO is slow, and optimization problems with constraints cannot be solved efficiently. In [86], hybrid PSO has been utilized to efficiently coordinate DOCRs. The optimal IP setting is determined using PSO, and the TDS of each relay is determined using the LP approach. In [87], the same technique used in [86] was successfully applied to coordinate DOCRs in a microgrid system. The authors in [88] and [89] proposed a hybridization of Nelder-Mead simplex search approach with PSO to solve the coordination problem. Nelder-Mead simplex search with a faster reach to the optimum settings and PSO to achieve the globally optimum solution, as well as combining the two algorithms and the gradient-based repair approaches allows for the discovery of feasible optimal solutions that satisfy the constraint conditions. In addition to [88] and [89], the various network topologies have been successfully included in [90]. In [91], an efficient variant of the PSO algorithm known as Time Varying Acceleration Coefficient (PSO-TVAC) was developed to determine the optimal settings for directional overcurrent relays. TVAC's goal is to improve the global search and encourage particles to converge on the global optima at the end of the search. A hybridization between particle swarm optimization and differential evolution technique

(PSO-DE) has been performed in [92]. This technique yields the most globally optimal solution at a faster convergence rate. The coordination problem in [80] and [60] was successfully solved using a hybridized version of particle swarm optimization (HPSO). In order to avoid getting stuck in local optima and successfully search for a global optimal solution, the hybridization was accomplished by incorporating simulated annealing (SA) into the original PSO.

In [93], hybrid ABC with LP has been proposed to improve the conventional ABC algorithm performance. These decreased the search space, resulting in time consuming and computational efficiency in determining the optimal solutions. In [94], an effective combination of two optimization techniques, SFL and LP, has been proposed for the coordination problem. In [8], a new population-based evolutionary algorithm called biogeography-based optimization (BBO) has been proposed as an optimization method for optimal relay coordination problem. In addition, to improve convergence speed and accuracy, a hybrid BBO with LP has been developed, and the hybridization demonstrated that the needed number of populations and generations has been greatly reduced with better optimized fitness and required lower CPU time.

In [33], the authors performed a hybridized version of the WOA by establishing the SA into the WOA algorithm to enhance the optimal solution obtained after each iteration and improve exploitation by searching the most promising regions identified by the WOA algorithm, which results in a globally optimal solution. To gain the global search capability of HHO and the precise local search capability of sequential quadratic programming (SQP), the authors in [11] presented a hybrid HHO-SQP to give a globally optimum solution. Since the Moth-Flame Optimization (MFO) performs effectively in the exploitation phase. The water cycle algorithm's performance has been enhanced by the hybrid water cycle and moth flame (WCMF) algorithm, which has been developed in [95]. The Levy flight function was included to investigate randomization of the current hybrid model, which increased its performance. An efficient combination of gradient-based optimizer (GBO) and memory-based linear population size reduction technique of Success-History-based Adaptive Differential Evolution (LSHADE) algorithm has been presented in [55] to develop novel hybrid optimization model called GB-LSHADE. The GBO was used to explore potential

regions of search space, while the LSHADE algorithm was used as a local search scheme to improve the diversity of solutions and avoid premature convergence.

In [26], a new optimization strategy named FPSOGSA has been proposed by including the concept of fractional calculus inside the mathematical model of canonical PSO and integrating it with gravitational search algorithm (GSA) to improve the optimizer characteristics by enhancing the convergence rate and preventing premature convergence. Optimal coordination of directional overcurrent relays in modern distribution systems with a high concentration of DG successfully proposed in [96] using hybrid harmony search and simulated annealing (HS-SA) technique.

The optimal solution for the coordination problem in a microgrid under different modes has been solved using hybrid programming of interval linear programming (ILP) and DE in [46]. By presenting short circuit current magnitudes as intervals and formulating a linear objective function, the complexity of overcurrent relays' optimal coordination in a microgrid was successfully reduced. In [2], a hybrid FA and LP (FA-LP) has been combined to improve the solution quality and convergence rate of the FA, resulting in a technique that relaxes the search space by linearizing the directional overcurrent relays coordination problem and avoids it from being trapped at the local optima.

The recently developed FA, based on the flashing behavior of fireflies, is a nature-inspired optimization algorithm utilized for evaluating the complicated and highly nonlinear constraints [35]. In a highly efficient way, FA can simultaneously find the global and all local optima. It doesn't need a suitable initial guess to start its iteration process because of the fast convergence speed of firefly and the high probability of obtaining the global optimisation solution. Another benefit of FA is that each firefly will operate practically independently, making it especially well-suited for parallel implementation. In parallel implementations, there are very few interactions between the various subregions [97]. Although it has many drawbacks similar to other inspired algorithms. For example, in the case of high dimensions, the algorithm would easily reach the local minimum value, which lowers the performance of the solution by reducing its attractiveness [98].

In this research, a modified firefly algorithm is implemented to solve the coordination problem. Compared to standard firefly algorithm, the attractiveness and randomized

movements parameters are controlled to obtain a global solution and produce a good convergence rate. The study then proposed the combination of genetic algorithm with the modified firefly algorithm as a new generation which can obtain better solutions and make a balance between global and local search. Additionally, it can get rid of trapping into various local optimums. Besides, to the best of the author' knowledge, the hybrid firefly-genetic algorithm hasn't yet been optimized the directional overcurrent relay coordination problem.

1.8 Thesis Outline

This thesis is organized in four chapters and three appendices. The first chapter of this study begins with an introduction that includes an overview of the research, identification of the problem statement, description of the significance and scope of the research, review of overcurrent protection in the power system, and review of the optimization techniques applied to optimize the coordination problem. Chapter 2 describes the methodology used in this study; it consists of the formulation of the optimization problem, the description of techniques used to solve the coordination problem, and a description of the testing systems used in simulation. Chapter 3 presents the results achieved from the proposed algorithms for each test system with discussion. The results include the optimized relay settings, the minimum total operating time obtained by the proposed methods in comparison to the literature, the overall net gain and the percentage performance in the operating time obtained by the suggested methods and finally presents the objective function evaluation number results for the GA and the hybrid algorithm to illustrate the efficacy of the developed hybrid algorithm. Chapter 4 concludes the study with some future research directions. Appendix A includes the IEC and IEEE standards Constants for Overcurrent Relays. Appendix B presents the time-delay coordination characteristics for different types of overcurrent relays. Figures related to the optimization algorithms used in this study are shown in Appendix C. In Appendix D, figures related to the proposed method applied in this study are presented. Appendix E includes the relay pairs and related parameters for the IEEE 3-bus, 6-bus, 9-bus and 15-bus networks. The simulation results for the IEEE 6-bus, 9-bus, and 15-bus networks are provided in Appendix F. The Matlab codes of the proposed methods that used to solve the coordination problem of the IEEE 3-bus network are provided in Appendix G.

Chapter Two

Methodology

2.1 Introduction

Figure D.1 illustrates the general steps that will be taken for the optimum coordination of the directional overcurrent relays [99].

The first step is to collect all the necessary data regarding the given network. The necessary data are shown in Figure D.2. The second step involves utilizing ETAP software to conduct a load flow analysis to determine the maximum load current. Then all of the relay pairs are identified since coordinating relays imply that selectivity among relay groups is carried out appropriately and sequentially. If any relay malfunctions or exceeds the assigned chance, one or a set of backup relays should activate immediately. Each of these backup relays will also function as the primary relay for a different set of relays. Thus, identifying the P/B relay pairs of all relays is a very essential step for solving the coordination problem. An effective protection design should be able to detect both the largest overload current and the smallest severe faults, which are reflected in PS's permissible limits. Therefore, it is necessary to conduct a short-circuit analysis on the given network before starting the process of optimal relay coordination. The analysis of the near-end three-phase short-circuit is applied. The minimum and maximum short circuit currents were found using ETAP software. Since this study is a comparative study, all the results of the previous stages are obtained from previous works. The last stage, which applies the optimization method to determine the optimum relay settings, can then be initiated. The optimization algorithms are coded using MATLAB software.

2.2 Problem Formulation

The coordination of directional overcurrent relays is a complex, non-convex optimization problem that should be solved while taking a variety of different linear and nonlinear inequality constraints into consideration. Formulation of the directional overcurrent relays coordination problem can be divided as follows:

- Objective function,
- Relay settings constraints, and
- Coordination constraints.

2.2.1 Objective Function

In this study, minimizing the total operating times of all main relays in the system is the purpose of the problems for directional overcurrent relays. The objective function is presented by Equation (4.1);

$$OF = \min \sum_{i=1}^n t_i \quad (4.1)$$

Where n is the total number of primary relays (i) in the network and t is the primary relay's operating time.

To fairly compare the effectiveness of various algorithms presented in the literature, the IEC standard inverse (IECSI) characteristic is applied in this study. Refer to Table 2, the equation (2.1) will be reformulated as follows considering the European relays are used:

$$t = \frac{0.14 * TMS}{\left(\frac{I_{SC}}{PS}\right)^{0.02} - 1} \quad (4.2)$$

2.2.2 Relay Settings Constraints

The objective function minimization in equation (4.1) is bound by several sets of constraints. The first set of constraints relates to the setting of the relay. It includes the upper and lower limits of current settings (IP) and time multiplier settings (TMS), which are defined in equations (4.3) and (4.4).

$$IP_i^{min} \leq IP_i \leq IP_i^{max}; \forall i \in N \quad (4.3)$$

$$TMS_i^{min} \leq TMS_i \leq TMS_i^{max}; \forall i \in N \quad (4.4)$$

Where, TMS^{min} and TMS^{max} are the minimum and maximum limits of TMS which are generally set to 0.025 and 1.1-sec [61], respectively. IP^{min} and IP^{max} are the minimum and maximum limits of IP . The limits of IP are taken as follows:

$$k_{ld} * I_{nom} < IP < \frac{2}{3} I_{f,min} \quad (4.5)$$

Where, k_{ld} is the overload factor and set to 1.25.

Two factors determine plug setting (PS); the first is the short circuit current, and the second is the maximum load current. PS is mathematically bounded as follows:

$$PS_i^{min} \leq PS_i \leq PS_i^{max} \quad (4.6)$$

The range of PS is graphically described in Figure D.3. Where, PS_i^{max} and PS_i^{min} are the maximum and minimum limits of PS_i such that [26]:

$$PS_i^{max} = \frac{2 * I_{f,min}}{3 * CTR} \quad (4.7)$$

$$PS_i^{min} = \frac{k_{ld} * I_{nom}}{CTR} \quad (4.8)$$

As shown in equation (4.9), the t_{min} and t_{max} represent the time bound for the minimum and maximum operational time of the relay, The standard values for t_{min} and t_{max} are 0.1 and 1.1-sec, respectively [78]. Where the critical clearing time determines the maximum time and the minimum time is based on the manufacture of the relay [11].

$$t_{min} < t_i < t_{max} \quad (4.9)$$

2.2.3 Coordination Constraints

The coordination constraints serve to ensure that both the primary and backup relays function effectively and prevent any occurrences of unnecessary or uncoordinated relay trips. In order to ensure correct coordination, the back-up relay's operational time must exceed that of the primary relay by the pre-defined CTI shown in equation (4.10).

$$t_{backup} - t_{primary} \geq CTI_{min} \quad (4.10)$$

Where, t_{backup} and $t_{primary}$ represent the operational time of the backup and primary relays, respectively. CTI_{min} is the minimum coordination time interval, and is typically set at 0.20 seconds for reliable relay operation as mentioned previously. Here,

$$t_{i,k}^{back} = \frac{0.14 * TMS^b}{\left(\frac{I_{sc}^b}{PS^b}\right)^{0.02} - 1} \quad (4.11)$$

$$t_{i,k}^{pri} = \frac{0.14 * TMS^P}{\left(\frac{I_{sc}^P}{PS^P}\right)^{0.02} - 1} \quad (4.12)$$

When directional overcurrent relays are utilized, they won't need to coordinate with the relays behind them because they will work if the fault current flows in the direction that is intended for tripping [26].

2.3 Standard Firefly Algorithm

Firefly algorithm (FA) is a nature-inspired optimization algorithm applied for solving the complex and highly nonlinear constrained problems [35]. The FA was proposed by Xin She Yang in late 2007 and 2008 who was inspired by the movement of fireflies at Cambridge University [100]. This algorithm was developed using three idealized rules. First, all Fireflies are unisex making it possible for them to attract one another regardless of their gender. Second, the brightness and attractiveness are inversely correlated, and as the distance between them increases, both decrease. Finally, the brightness of each firefly is affected by the objective function's value [97]. Figure D.4 shows the arrangement of fireflies [35]. Based on these three principles, the fundamental procedures of the FA can be summarized into the pseudo code shown in Figure D.5.

In minimization problems, the firefly with larger light intensity has lesser objective function. Equations (4.13) and (4.14) provide mathematical descriptions of the second assumption [97,101]:

$$I(r_{ij,m}) = I_0 e^{-\gamma r_{ij,m}^2} \quad (4.13)$$

$$\beta(r_{ij,m}) = \beta_0 e^{-\gamma r_{ij,m}^2} \quad (4.14)$$

Where, I_0 is the actual intensity of light, β_0 is the attractiveness at $r = 0$, γ is the coefficient of light absorption that controls the variation in attractiveness and defines convergence. In most of cases, its value lie in range [0.01, 100], r is the

distance between two fireflies using Cartesian distance and m is the number of local optima of an optimization problem. The distance between the j^{th} and i^{th} fireflies is expressed as follows:

$$r_{ij} = \|X_{i,m} - X_{j,m}\| = \sqrt{\sum_{m=1}^k (x_{i,m} - x_{j,m})^2} \quad (4.15)$$

$$X_{i,m} = [x_{i,1}, x_{i,2}, x_{i,3}, \dots, x_{i,k}]$$

$$X_{j,m} = [x_{j,1}, x_{j,2}, x_{j,3}, \dots, x_{j,k}]$$

Where, $x_{j,m}$ is the m^{th} component of the spatial coordinate of the j^{th} firefly and k is the number of dimensions [102]. The movement of a firefly i is attracted to another more attractive (brighter) firefly j is computed by [100]:

$$x_{i,m+1} = x_{i,m} + \beta(r_{ij})(x_{j,m} - x_{i,m}) + \alpha_m (\text{rand} - 0.5) \quad (4.16)$$

Where, $x_{i,m+1}$ is the next generation of firefly, $x_{i,m}$ and $x_{j,m}$ are the current position of fireflies, α_m is the randomization parameter in interval $[0, 1]$ and rand is random number generator with numbers uniformly distributed in range $[0, 1]$ [2,102]. Three terms make up the movements of firefly, as demonstrated by equation (4.16). The first one describes the current location of the i^{th} firefly. The second describes the movement of a firefly, showing that the firefly with less brightness attracts to another brighter firefly. The third describes the firefly's random movement [36]. During the initialization phase of the algorithm, every agent in the population is assigned a solution to an optimization problem. Then, in the iterative stage, the light intensity of each agent i located in position x_i is compared to the light intensity of the other agents in the population. If the light intensity of agent i is less than agent j 's light intensity located in position x_j , then agent i moves towards agent j . The agents are ranked based on the fitness value of their solutions, and the global best is updated with the most recent one (if applicable). [100,101].

2.4 Modified Firefly Algorithm

The Firefly algorithm is a simple to use and effective technique. For multimodal problems, researches reveal that it is sluggish to converge and easily falls into a local optimum. Furthermore, updates are made solely based on the current performance without retaining any memory of previous best solutions or performances. That could result in the loss of better solutions. Moreover, because the parameters are fixed, the search behavior remains constant throughout all iterations for any condition [103]. The search of FA is evaluated by the attraction between any two different fireflies. Different settings of the randomization parameter α and attractiveness coefficient β_0 parameters may result in different performance. Authors in [104] proposed modifications on the attraction parameters of the standard FA to obtain global solution by updating brightness of the firefly and to avoid the distance between individual fireflies from being too far. However, the attractiveness between r^{th} and n^{th} fireflies is given as follows:

$$\beta(r_{ij}) = \beta_{\min r,j} + (\beta_{\max r,j} - \beta_{\min r,j}) e^{-\gamma r_{ij}^2} \quad (4.17)$$

Where β_{\min} and β_{\max} are user-supplied values and are taken as 0.2 and 1, respectively. Even if the distance is too far, $e^{-\gamma r_{ij}^2} \rightarrow 0$, the attraction between them can be the β_{\min} . The parameter α , which generally has values between 0 and 1, is essential in controlling the random movements to get a solution. A large value of α will have a low degree of accuracy while searching for an optimal value since the firefly's random movement spreads out and does not lead to the intended point. On the other hand, a small α will produce a good convergence rate for the displacement of the firefly moving in the desired direction [97]. Authors in [102] modified α using equation (4.19):

$$\text{delta} = 1 - \left(\frac{10^{-4}}{0.9} \right)^{\frac{1}{\text{maxgen}}} \quad (4.18)$$

$$\alpha = (1 - \text{delta}) * \alpha_0 \quad (4.19)$$

The flow chart of the modified FA is shown in Figure D.6.

2.5 Standard Genetic Algorithm

Genetic Algorithm (GA) was developed by Holland in the 1960s and further analyzed by Goldberg in 1989 [105]. GA is optimization algorithm inspired by the principles of natural evolution and natural selection and the idea of the “survival of the fittest” [40,82]. The process begins with a population of solutions randomly generated, where the ones with higher fitness are given more preference for being chosen as parents to create new solutions (offspring) for the subsequent generation [106]. Similar to any other optimization technique, GA starts by identifying the variables for optimization and the fitness function (objective function) [82]. The next step is to evaluate each chromosome's fitness value for the present generation. The GA selects a few chromosomes and utilizes them to produce the next generation in order to evolve the existing population and reach at an optimal solution, proportional to fitness value. The crossover and mutation operators are used to create new individuals in the decision space for the chosen pair of chromosomes [107]. After a predetermined number of generations, the process will be terminated. Depending on the complexity of the system and the size of the population, the required number of generations differs from system to system. The pseudocode demonstrated in Figure D.7 represents the basic steps of the GA. Figure D.8 depicts the GA's flowchart [108].

2.6 Constraint-Handling Technique

It is necessary to change the original constrained problem into an unconstrained one to solve a constrained optimization problem. The simplest and most widely used optimization techniques for handling constraints are transformation methods [82]. Different approaches, like Karush-Kuhn-Tucker (KKT) [109] and the death penalty function [24], have been suggested to address and satisfy the constraints of the coordination problem. Because the other approaches need derivations or are difficult to model, the penalty functions are frequently utilized [8].

The penalty method involves augmenting the objective function with a penalty term that penalizes infeasible solutions that violate the constraints [4]. Relay coordination constraints and relay characteristic constraints are incorporated

into the objective function of the coordination problem using the penalty method, as illustrated in equation (4.20). In case any constraint is violated, a penalty is added to the objective function value. The penalty factor is typically set to a large value as the objective function is in a minimization form.

$$\mathbf{OF} = \sum_{i=1}^n \mathbf{t}_i + \sum_{l=1}^k \mathbf{P}(l) \quad (4.20)$$

Where k is the number of relay pairs, the penalty term $\mathbf{P}(l)$ is given by the following equation:

$$P(l) = \mu \sum_{i=1}^m H_i[\varphi_i(l)] \varphi_i^2(l) + v \sum_{j=1}^r H_j[\psi_j(l)] \psi_j^2(l) \quad (4.21)$$

Depending on the required level of solution quality, the values of μ and v , which are penalty constants or penalty factors, should be sufficiently large with $\mu \gg 1$ and $v \geq 0$. φ_i and ψ_j represent the quality and inequality constraints, respectively. m is the number of quality constraints, while r is the number of inequality constraints. $H_i[\varphi_i(l)]$ and $H_j[\psi_j(l)]$ are index functions. More specifically,

$$H_i[\varphi_i(l)] = \begin{cases} \mathbf{1}, & \varphi_i(l) \neq \mathbf{0} \\ \mathbf{0}, & \varphi_i(l) = \mathbf{0} \end{cases} \quad (4.22)$$

Similarly,

$$H_j[\psi_j(l)] = \begin{cases} \mathbf{1}, & \psi_j(l) > \mathbf{0} \\ \mathbf{0}, & \psi_j(l) \leq \mathbf{0} \end{cases} \quad (4.23)$$

It is shown that the $P(l)$ has a value of zero when individuals are feasible, and a positive value when there is a violation of constraints. During the minimization process, μ and v increase the relevance of the objective function's value. To attain optimal solutions with zero penalties, μ and v are typically given relatively high values [4]. Their range usually set between 10^3 and 10^{15} [110].

2.7 Proposed FA-GA Approach for Coordination Problem

The optimization algorithm performance can be enhanced by transforming the current solution into one or more improved solutions. A combination between modified FA and GA techniques is used to perform this improvement. In this hybrid method, the master meta-heuristic is the modified FA and the GA is subordinate to it. It comprises of two stages. The modified FA operates in two stages. The first stage is aimed at exploring the search space to identify the most promising region. In the second stage, the GA is incorporated to investigate the search space further (beginning with the FA's solution) and generate improved solutions to improve global search while avoiding getting trapped in multiple local optima. The main idea behind using GA is due to its genetic operators, crossover and mutation in generating new solutions. The best solutions generated by GA are considered to be the best solutions at all. The pseudo-code in Figure D.9 illustrates the structure of the hybrid FA-GA. Figure D.10 illustrates the hybrid FA-GA flowchart.

2.8 Testing Systems

The proposed methods are applied to a four power systems IEEE 3-bus, 6-bus, 9-bus, and 15-bus networks. These networks are employed to verify the efficacy of the suggested optimization techniques.

2.8.1 System I: IEEE 3-bus Network

To assess the effectiveness of the proposed MFA, GA, and FAGA algorithms in minimizing the operating time of directional overcurrent relays, the standard IEEE 3-bus network is utilized as the first test case. Figure E.1 depicts its component parts, which are 3 buses, 3 power generators, 3 branches, and 6 relays. This case is presented as a formulation of linear and nonlinear programming. Tables E.1 and E.2, respectively, present the results of 3-short circuits and the CT ratios of the relays in the IEEE 3-bus system. The lower and upper values for TMS and PS are set at 0.1 and 1.1 and 1.5 and 5.0, respectively. The CTI value is set to 0.2 seconds, and the time bounds for the primary relays' minimum and maximum operating times are set to 0.1 and 0.5, respectively [26,33].

For this system, there will be 12 design variables and 36 constraints in the coordination problem, the standard optimization model of the optimization problem can be expressed as follows:

$$\min_{TMS,PS} Z(TMS_1, \dots, TMS_6, PS_1, \dots, PS_6) \quad \text{Objective Function/General expression}$$

Subjected to (refer to equations 4.4, 4.6 4.9 and 4.10) [26,33]:

$$0.1 - t_{i,k}^{pri} \leq 0$$

$$t_{i,k}^{pri} - 0.5 \leq 0$$

$$t_{i,k}^{pri} + 0.2 - t_{j,k}^{back} \leq 0$$

$$0.1 \leq TMS_i \leq 1.1$$

$$1.5 \leq PS_i \leq 5$$

The objective function can be expanded as shown in equation (4.1):

$$OF = \min (t_1^{pri} + t_2^{pri} + t_3^{pri} + t_4^{pri} + t_5^{pri} + t_6^{pri})$$

Where (refer to equation 4.12):

$$t_1^{pri} = \frac{0.14 * TMS_1}{\left(\frac{1978.9}{\left(\frac{300}{5} \right) PS_1} \right)^{0.02} - 1}$$

$$t_2^{pri} = \frac{0.14 * TMS_2}{\left(\frac{1525.7}{\left(\frac{200}{5} \right) PS_2} \right)^{0.02} - 1}$$

$$t_3^{pri} = \frac{0.14 * TMS_3}{\left(\frac{1683.9}{\left(\frac{200}{5} \right) PS_3} \right)^{0.02} - 1}$$

$$t_4^{pri} = \frac{0.14 * TMS_4}{\left(\frac{1815.4}{\left(\frac{300}{5} \right) PS_4} \right)^{0.02} - 1}$$

$$t_5^{pri} = \frac{0.14 * TMS_5}{\left(\frac{1499.66}{\left(\frac{200}{5}\right)PS_5}\right)^{0.02} - 1}$$

$$t_6^{pri} = \frac{0.14 * TMS_6}{\left(\frac{1766.3}{\left(\frac{400}{5}\right)PS_6}\right)^{0.02} - 1}$$

Subjected to (refer to equations 4.4, 4.6 4.9 and 4.10):

$$\frac{0.14 * TMS_1}{\left(\frac{1978.9}{\left(\frac{300}{5}\right)PS_1}\right)^{0.02} - 1} + 0.2 - \frac{0.14 * TMS_5}{\left(\frac{175}{\left(\frac{200}{5}\right)PS_5}\right)^{0.02} - 1} \leq 0$$

$$\frac{0.14 * TMS_2}{\left(\frac{1525.7}{\left(\frac{200}{5}\right)PS_2}\right)^{0.02} - 1} + 0.2 - \frac{0.14 * TMS_4}{\left(\frac{545}{\left(\frac{300}{5}\right)PS_4}\right)^{0.02} - 1} \leq 0$$

$$\frac{0.14 * TMS_3}{\left(\frac{1683.9}{\left(\frac{200}{5}\right)PS_3}\right)^{0.02} - 1} + 0.2 - \frac{0.14 * TMS_1}{\left(\frac{617.22}{\left(\frac{300}{5}\right)PS_1}\right)^{0.02} - 1} \leq 0$$

$$\frac{0.14 * TMS_4}{\left(\frac{1815.4}{\left(\frac{300}{5}\right)PS_4}\right)^{0.02} - 1} + 0.2 - \frac{0.14 * TMS_6}{\left(\frac{466.17}{\left(\frac{400}{5}\right)PS_6}\right)^{0.02} - 1} \leq 0$$

$$\frac{0.14 * TMS_5}{\left(\frac{1499.66}{\left(\frac{200}{5}\right)PS_5}\right)^{0.02} - 1} + 0.2 - \frac{0.14 * TMS_3}{\left(\frac{384}{\left(\frac{200}{5}\right)PS_3}\right)^{0.02} - 1} \leq 0$$

$$\frac{0.14 * TMS_6}{\left(\frac{1766.3}{\left(\frac{400}{5}\right)PS_6}\right)^{0.02} - 1} + 0.2 - \frac{0.14 * TMS_2}{\left(\frac{145.34}{\left(\frac{200}{5}\right)PS_2}\right)^{0.02} - 1} \leq 0$$

$$0.1 - \frac{0.14 * TMS_1}{\left(\frac{1978.9}{\left(\frac{300}{5}\right)PS_1}\right)^{0.02} - 1} \leq 0$$

$$0.1 - \frac{0.14 * TMS_2}{\left(\frac{1525.7}{\left(\frac{200}{5}\right)PS_2}\right)^{0.02} - 1} \leq 0$$

$$0.1 - \frac{0.14 * TMS_3}{\left(\frac{1683.9}{\left(\frac{200}{5}\right)PS_3}\right)^{0.02} - 1} \leq 0$$

$$0.1 - \frac{0.14 * TMS_4}{\left(\frac{1815.4}{\left(\frac{300}{5}\right)PS_4}\right)^{0.02} - 1} \leq 0$$

$$0.1 - \frac{0.14 * TMS_5}{\left(\frac{1499.66}{\left(\frac{200}{5}\right)PS_5}\right)^{0.02} - 1} \leq 0$$

$$0.1 - \frac{0.14 * TMS_6}{\left(\frac{1766.3}{\left(\frac{400}{5}\right)PS_6}\right)^{0.02} - 1} \leq 0$$

$$\frac{0.14 * TMS_1}{\left(\frac{1978.9}{\left(\frac{300}{5}\right)PS_1}\right)^{0.02} - 1} - 0.5 \leq 0$$

$$\frac{0.14 * TMS_2}{\left(\frac{1525.7}{\left(\frac{200}{5}\right)PS_2}\right)^{0.02} - 1} - 0.5 \leq 0$$

$$\frac{0.14 * TMS_3}{\left(\frac{1683.9}{\left(\frac{200}{5}\right)PS_3}\right)^{0.02} - 1} - 0.5 \leq 0$$

$$\frac{0.14 * TMS_4}{\left(\frac{1815.4}{\left(\frac{300}{5}\right)PS_4}\right)^{0.02} - 1} - 0.5 \leq 0$$

$$\frac{0.14 * TMS_5}{\left(\frac{1499.66}{\left(\frac{200}{5}\right)PS_5}\right)^{0.02} - 1} - 0.5 \leq 0$$

$$\frac{0.14 * TMS_6}{\left(\frac{1766.3}{\left(\frac{400}{5}\right)PS_6}\right)^{0.02}} - 0.5 \leq 0$$

$$0.1 \leq TMS_1 \leq 1.1$$

$$0.1 \leq TMS_2 \leq 1.1$$

$$0.1 \leq TMS_3 \leq 1.1$$

$$0.1 \leq TMS_4 \leq 1.1$$

$$0.1 \leq TMS_5 \leq 1.1$$

$$0.1 \leq TMS_6 \leq 1.1$$

$$1.5 \leq PS_1 \leq 5.0$$

$$1.5 \leq PS_2 \leq 5.0$$

$$1.5 \leq PS_3 \leq 5.0$$

$$1.5 \leq PS_4 \leq 5.0$$

$$1.5 \leq PS_5 \leq 5.0$$

$$1.5 \leq PS_6 \leq 5.0$$

2.8.2 System II: IEEE 6-bus Network

Figure E.2 depicts the single line diagram of IEEE 6-bus network. There are 7 branches, 4 power generators, and 14 relays in the single-line diagram. The point of the proposed calculation for this case is to arrange the settings of 14 relays and the main goal is to obtain the optimum TMS and PS settings. At the close end of each relay (close in faults) three phase short circuit are connected. Table E.3 presents the primary/backup relay (P/B) relay pairs and the close in short circuit currents. The relays' CT ratios are presented in Table E.4. The CTI value is selected as 0.2-sec. This test system has 76 constraints in total, including 14 inequality constraints for the minimum operating times, 14 inequality constraints for the maximum operating times, 20 inequality

constraints for selectivity criteria, 14 side constraints for the TMS, and 14 side constraints for the PS.

2.8.3 System III: IEEE 9-bus Network

The third system investigated in this study is the IEEE 9-bus network shown in Figure E.3. It consists of 9 buses, one power generator, 12 branches and 24 directional overcurrent relays. This system is considered as NLP formulation. Additionally, the lower and upper bounds of TMS have been set at 0.1 and 1.2, respectively. Meanwhile, the lower and higher bounds of PS have been defined at 0.5 and 2.5, respectively. Also, for a justified comparison, the value of the coordination time interval is set as 0.2-sec. The time bound for the minimum operational time of primary relays is set as 0.2. To complete the analysis, the three-phase short circuit currents for the primary and backup relays are provided in Table E.5 and All relays use a CT ratio of 500/1. There are a total of 128 constraints for this test system: 32 inequality constraints for selectivity criteria, 24 inequality constraints for the minimum and maximum operating times, 24 side constraints for the TMS, and 24 side constraints for the PS.

2.8.4 System IV: IEEE 15-bus Network

The system shown in Figure E.4 has a high level of distributed generation (DG) penetration, consisting of 15 buses, 21 branches, and 42 relays. The fault currents measured by primary and backup relays are given in Table E.6 [33]. For the directional overcurrent relays setting, the CT ratios are provided in Table E.7. Moreover, the upper and lower limits for TMS are set to 0.1 and 1.2, respectively, while the upper and lower limits for PS variable are set to 0.5 and 2.5, respectively [2]. A 0.2-sec coordination interval is considered. The test system comprises 250 constraints, which include 82 inequality constraints for selectivity criteria, 42 inequality constraints for the minimum allowable operating time, 42 inequality constraints for the maximum allowable operating time, 42 side constraints for TMS, and 42 side constraints for PS.

Chapter Three

Results and Discussions

3.1 Introduction

This chapter provides the results achieved for the directional overcurrent relays by the proposed algorithms. To demonstrate the efficacy of the suggested modified firefly algorithm (MFA), genetic algorithm (GA) and hybrid firefly-genetic algorithm (FAGA), a comparison with other optimization techniques is provided. Four standard networks, including the IEEE 3-bus, 6-bus, 9-bus, and 15-bus networks, are used to test the proposed methods. The overall net gain in time and the percentage improvement performance obtained by the suggested methods for each test system are presented. An accurate simulation program is developed using the MATLAB software to obtain the results.

3.2 Test System I: IEEE 3-bus Network

This network is experimented by LP, and NLP formulations. In LP formulation the only decision variable is TMS, which is continuous lying in [0.1, 1.1]. The PS values are fixed constants given in Table 3.1. PS and TMS are regarded as the design variables in the NLP formulation, which lies in [1.5, 5.0] and [0.1, 1.1] respectively, and both of them are continuous values.

Table 3.1

The IEEE 3-bus network's PS values for the relays

Relay Number	PS
1	5.0
2	1.5
3	5.0
4	4.0
5	2.0
6	2.5

3.2.1 Linear Programming Formulation

Table 3.2 provides the optimum settings of TMS achieved by MFA, GA and FAGA. Other methods that proposed for this system are presented to make comparison. It was observed that the proposed methods achieved superior results when compared to other algorithms. Figure 3.1 shows the graphical illustration of the optimized minimum total

operating time obtained by the proposed methods in comparison to the literature. The overall net gain in time achieved by the proposed methods is presented in Table 3.3, which reveals the superiority of the proposed methods over other algorithms mentioned in the literature. The percentage performance in the operating time at 3-bus test system yielded by the proposed methods is presented in Figure 3.2, which indicates an enhancement of relay operating time using the proposed methods.

Table 3.2

TMS values by LP formulation for the IEEE 3-bus network

Method	Time Multiplier Settings (TMS)						Fitness
	1	2	3	4	5	6	
Simplex [109]	0.100000	0.136400	0.1000	0.1000	0.129800	0.1000	1.92580
LP [37]	0.100000	0.136400	0.1000	0.1000	0.129800	0.1000	1.92580
PSO [37]	0.100000	0.136400	0.1000	0.1000	0.129800	0.1000	1.92580
Seeker [48]	0.100000	0.136400	0.1000	0.1000	0.129800	0.1000	1.92580
ABC [68]	0.100000	0.136400	0.1000	0.1000	0.129800	0.1000	1.92580
OJaya [4]	0.100000	0.100000	0.1000	0.1000	0.129800	0.1000	1.78040
HCSO [111]	0.100000	0.100000	0.1000	0.1000	0.100000	0.1000	1.78040
MFA	0.100000	0.100000	0.1000	0.1000	0.100000	0.1000	1.78039
GA	0.100016	0.100004	0.1000	0.1000	0.100002	0.1000	1.78047
FAGA	0.100000	0.100000	0.1000	0.1000	0.100000	0.1000	1.78039

Figure 3.1

Total operation time compared to the literature for the proposed methods



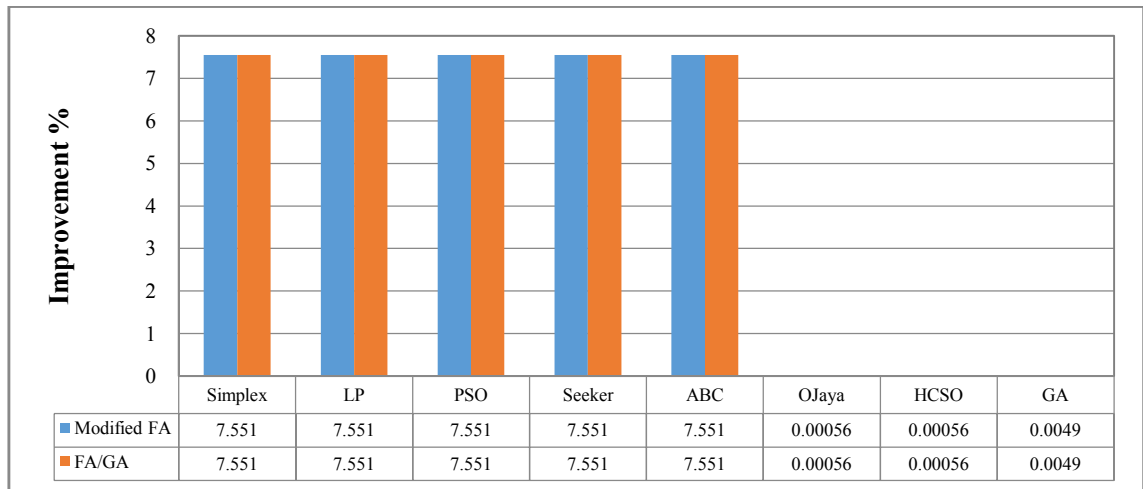
Table 3.3

Comparison of total net gain in time obtained by proposed methods with the algorithms used in the literature

Method	$\sum \Delta(t)$ (sec)	% Improvement	Method	$\sum \Delta(t)$ (sec)	% Improvement
MFA/ Simplex	0.14541	7.55100	FAGA/Simplex	0.14541	7.55100
MFA/ LP	0.14541	7.55100	FAGA /LP	0.14541	7.55100
MFA/ PSO	0.14541	7.55100	FAGA /PSO	0.14541	7.55100
MFA/ Seeker	0.14541	7.55100	FAGA /Seeker	0.14541	7.55100
MFA/ ABC	0.14541	7.55100	FAGA /ABC	0.14541	7.55100
MFA/ OJaya	0.00001	0.00056	FAGA /OJaya	0.00001	0.00056
MFA/ HCSO	0.00001	0.00056	FAGA /HCSO	0.00001	0.00056
MFA/ GA	0.00008	0.00490	FAGA /GA	0.00008	0.00490

Figure 3.2

Percentage improvement of the proposed methods compared to other algorithms in the literature for IEEE 3-bus network in terms of the net relay operational time



The CTI of each relay pair is presented in Table 3.4. It is demonstrated each backup and primary relay pair satisfy the CTI criteria and that all of them are greater than 0.2.

Table 3.4*CTI using the LP formulation for the IEEE 3-bus network*

P/R	MFA	GA	FAGA
	CTI	CTI	CTI
1	0.52319	0.52375	0.52308
2	0.63712	0.63806	0.63705
3	0.64169	0.64170	0.64286
4	0.48122	0.48166	0.48140
5	0.83420	0.83452	0.83414
6	0.46982	0.46985	0.47003

3.2.2 Non Linear Programming Formulation

The optimal settings of the decision variables TMS and PS are shown in Table 3.5. A comparison of results obtained by the proposed methods and other published algorithms are also provided in Table 3.6. Figure 3.3 presents a graphical representation of the total operating time achieved by the proposed methods compared to other published techniques. The figure displays the optimized values of total operating time. The overall net gain in time achieved by the proposed methods and other published algorithms reported in the literature is depicted in Table 3.7. Figure 3.4 indicates the improvement in the relay operating time using the proposed methods.

Table 3.5*Relay settings for IEEE 3-bus network by NLP formulation*

Variables	AFA [97]	Analytic [39]	Jaya [4]	IGWO [112]	MFA	GA	FAGA
TMS 1	0.1143	0.1000	0.1000	0.10000	0.10000	0.118970	0.100069
TMS 2	0.1000	0.1000	0.1000	0.10000	0.10000	0.100001	0.100000
TMS 3	0.1074	0.1000	0.1453	0.10010	0.10000	0.109758	0.100022
TMS 4	0.1000	0.1000	0.1000	0.10000	0.10000	0.107297	0.100033
TMS 5	0.1000	0.1000	0.1000	0.10000	0.10000	0.100004	0.100000
TMS 6	0.1125	0.1000	0.1000	0.10000	0.10000	0.100011	0.100000
PS 1	1.2500	2.7000	1.5000	1.50000	2.25484	1.500000	1.945890
PS 2	1.3400	2.1250	2.9780	2.61660	1.55414	1.500000	1.500000
PS 3	1.2500	2.8750	1.5000	2.97700	1.80029	1.528990	1.787530
PS 4	1.4300	2.3333	1.7841	1.58500	2.32436	1.500040	1.683100
PS 5	1.3900	2.2750	1.8601	2.81690	1.51354	1.500000	1.500000
PS 6	1.2500	1.2695	1.5000	1.50090	1.61407	1.500070	1.500050
Fitness	4.7636	1.5108	1.5109	1.47890	1.41858	1.401310	1.36501

Table 3.6

Comparison of the proposed methods' percentage improvement and the overall net gain in time over the literature's algorithms

Method	$\sum \Delta(t)$ (sec)	% Improvement	Method	$\sum \Delta(t)$ (sec)	% Improvement
MFA/ AFA	3.34502	70.22	FAGA/AFA	3.39859	71.34
MFA/ Analytic	0.09222	6.100	FAGA/Analytic	0.14579	9.649
MFA/ Jaya	0.09232	6.110	FAGA/Jaya	0.14589	9.656
MFA/ IGWO	0.06032	4.079	FAGA/IGWO	0.11389	7.701
MFA/ GA	-	-	FAGA /GA	0.03630	2.590

From the obtained results, it can be concluded that the suggested methods have once again outperformed their counterparts by computing a minimum value of fitness evaluation function in terms of the overall operational time of directional overcurrent relays in the IEEE 3-bus system, thereby confirming the superiority, efficiency, and robustness of the suggested algorithms.

Figure 3.3

Total operation time compared to the literature for the proposed methods

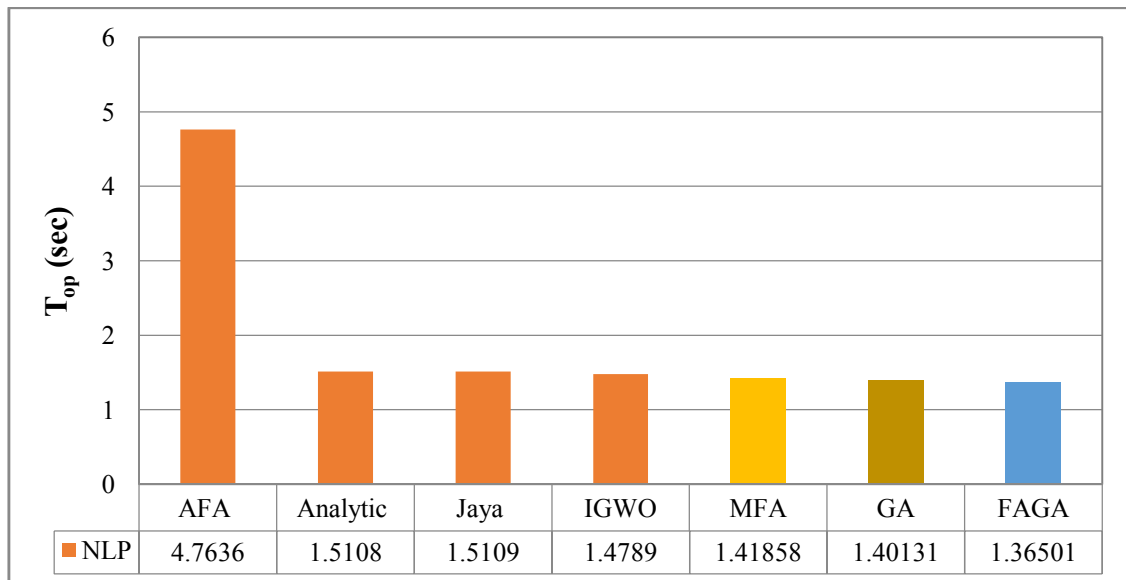
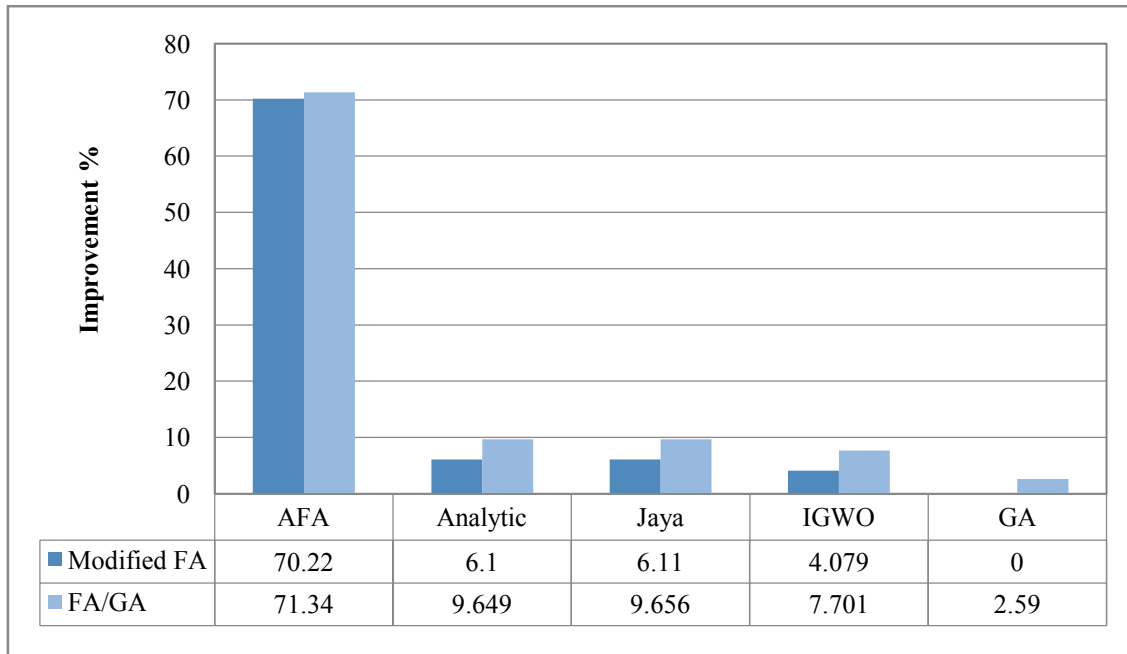


Figure 3.4

Percentage improvement of the proposed methods compared to other algorithms in the literature for IEEE 3-bus network in terms of the net relay operational time



The CTI for the six relay pairs for the IEEE 3-bus system is shown in Table 3.7. It is clear that the primary relays function first, followed by the backup relays in the case that the primary relays are not able to isolate the problem within the coordination time margin.

Table 3.7

CTI using the NLP formulation for the IEEE 3-bus network

P/R	MFA	GA	FAGA
	CTI	CTI	CTI
1	0.40020	0.38590	0.40602
2	0.29461	0.20003	0.20001
3	0.23977	0.20019	0.20000
4	0.27273	0.26659	0.27314
5	0.20000	0.20005	0.20000
6	0.55715	0.53164	0.53085

3.3 Test System II: IEEE 6-bus Network

The values of the continuous TMS in the LP formulation lie between [0.1, 1.1]. The PS values are fixed constants given in Table F.1. In the NLP formulation, the continuous variables PS and TMS both have ranges between [1.5, 5.0] and [0.1, 1.1], respectively.

3.3.1 Linear Programming Formulation

The optimal TMS settings obtained by the proposed methods are shown in Table F.2, along with a comparative analysis of those methods with other algorithms that have been published and are mentioned in the literature. Figure F.1 shows a graphical representation of the total operating time obtained by proposed methods with other published techniques. It is shown that the proposed methods have obtained improved results for the IEEE 6-bus test system. The assessment of total net gain in time achieved in this case by the proposed methods is shown in Table F.3 with respect to other techniques mentioned in the literature. It was found that the proposed algorithms outperform other methods in terms of net time gain while also producing results that are more satisfying and effective. The percentage improvement in the overall operating time, as compared to the literature, is shown graphically in Figure F.2, which shows that the effectiveness of the proposed techniques in improving relay operating time.

The CTI of each relay pair obtained by the proposed algorithms is shown in Table F.4. These results indicate that the CTI values are consistently higher than the lowest limit of CTI applied throughout this study.

3.3.2 Nonlinear Programming Formulation

This study formulates the case of the network as a nonlinear programming case. Table F.5 presents the TMS and PS's optimum settings. Figure F.3 shows the optimized values of the total operating time that result by the proposed methods.

The coordination margin achieved using the proposed techniques are shown in Table F.6. The backup relays will be activated if the primary relays fail to function, as can be observed from the table. It can be concluded that the proposed methods success to obtain the optimal relay setting that maintains the sequential operation between relay pairs.

3.4 Test System III: IEEE 9-bus Network

The suggested algorithms are applied to an IEEE 9-bus network. In this case, all relays are considered as numerical relays. Thus, this network is formulated as NLP. The continuous decision variables PS and TMS have ranges between [0.5, 2.5] and [0.1, 1.2], respectively. The optimized TMS and PS values obtained by the MFA, GA and

FAGA with other published techniques for this case are given in Table F.7. The overall operating time obtained using the proposed methods and other methods that have been published are depicted graphically in Figure F.4. Table F.8 presents an examination of the overall net gain in time that was achieved for this case using the proposed methods. It can be shown that the proposed methods outperform other algorithms in terms of net time gain. Figure F.5 shows the percentage improvement of the net relay operating time of the proposed methods compared to other methods in the literature.

The discrimination times for each relay pair are included in Table F.9. It has been demonstrated that the coordination time interval between the relays has been maintained, which yields good performance from the proposed techniques.

3.5 Test System IV: IEEE 15-bus Network

The last test system considered in this chapter is the IEEE 15-bus network. All relays are regarded as numerical relays in this case. As a result, this network is formulated as NLP. The continuous decision variables PS and TMS have ranges between [0.5, 2.5] and [0.1, 1.2], respectively. The optimized TMS and PS values obtained by the MFA, GA and FAGA with other published techniques for this case are given in Table F.10. The optimal solution found by the suggested approaches is superior to that obtained by using other techniques, as can be seen from Table F.11. Figure F.6 shows a graphical presentation of the total operating time that was achieved by utilizing the proposed techniques and other methods that have been published. Table B10 presents an examination of the overall net gain in time that was achieved for this case using the proposed methods. In comparison to other approaches mentioned in the literature, the proposed methods' percentage improvement in the net relay operating time is shown in Figure F.7. It is evident that the proposed methods provide the least objective function, which confirms to their effectiveness in resolving the coordination problem of directional overcurrent relays.

Table F.12 includes the discrimination times for each relay pair. It can be seen that the primary relays operate initially, and then, after a coordination time margin, the backup relays function in case the primary relays are failed in isolating the fault.

3.6 Number of Objective Function Evaluation

In order to illustrate the efficacy of the hybrid proposed algorithm, the objective function evaluation number is used to make comparison between the hybrid algorithm and the standard genetic algorithm. The number of the objective function evaluation and the percentage improvement for each simulated case are shown in Table 3.8. The results reveal the superiority of the proposed hybrid algorithm when compared to the standard GA.

Table 3.8

Number of fitness evaluation for the GA and hybrid algorithms

Case	Method	Number of objective function evaluation	%Improvement
IEEE 3-bus LP	GA	165432	48.345
	FAGA	85454	
IEEE 3-bus NLP	GA	173656	53.316
	FAGA	81070	
IEEE 6-bus LP	GA	266000	54.343
	FAGA	121448	
IEEE 6-bus NLP	GA	281200	43.282
	FAGA	159490	
IEEE 9-bus NLP	GA	604800	33.639
	FAGA	401350	
IEEE 15-bus NLP	GA	602432	47.971
	FAGA	313440	

Chapter Four

Conclusions and Future Works

4.1 Introduction

The main objective of this study was to find the optimal settings of the directional overcurrent relays using modified FA, GA and hybrid FA-GA algorithm. The algorithms were evaluated using four standard test systems. The proposed techniques were utilized to obtain the optimum relay settings and the overall operating time, which were then compared to other published techniques. To verify the superiority of the hybrid developed algorithm, the number of the objective function evaluation was used to make comparison with the standard genetic algorithm. The results indicated that the developed hybrid method outperformed the standard GA. In this chapter, the conclusions and the future works of this research are provided.

4.2 Conclusions

This research proposed a modified firefly algorithm and hybrid firefly-genetic algorithm to solve the directional overcurrent relays coordination problem. In modified firefly algorithm, the attractiveness coefficient and the randomization parameter are controlled to get good convergence rate. The hybrid firefly-genetic algorithm is proposed to get better solution, the two techniques are applied in a serial fashion, which divided into two stages, the modified firefly algorithm is applied in the first stage to obtain global solutions, and the results from the algorithm are used as the initial population for the genetic algorithm in the last stage to obtain better solutions. To compare the performance of the proposed methods in solving the problem, the methods are applied to the directional overcurrent relays coordination problem including IEEE 3-bus, 6-bus, 9-bus and 15-bus test systems. The results demonstrate that the proposed methods are superior to the previous approaches in obtaining the minimal total operating time of primary relays and maintaining the correct coordination between primary and backup relay pairs. The hybrid firefly-genetic algorithm achieves a better solution with fewer objective function evaluations than the standard genetic algorithm.

4.3 Future Works

The algorithms used in this work were tested using a single type of fault and the same standard inverse characteristic for all relays. It is advised that the proposed algorithms be tested for various standard characteristics and multiple types of faults. Additionally, the work can be expanded by using additional decision variables that are non-standard characteristics curve relays to provide the relay coordination issue more flexibility.

List of Abbreviations

Abbreviation	Meaning
TMS	Time Multiplier Setting
TD	Time Dial
PS	Plug Setting
IP	Pickup Current
CTI	Coordination Time Interval
CB	Circuit Breaker
CT	Current Transformers
PT	Potential Transformers
IDMT	Inverse Definite Minimum Time
t	Relay Operating Time In Seconds
TD	Time Dial, Or Time Multiplier Setting
Isc	Fault Current Level In Secondary Amperes
IPU	Tap Or Pickup Current Setting
C	Constant
α	Slope Constant
β	Slope Constant
PSM	Plug Setting Multiplier
RMS	Root Mean Square
Kld	Overload Factor That Based On The Element Being Protected.
Inom	Nominal Rated Current.
CTR	Current Transformer Ratio.
DR	Directional Overcurrent Relay
CAPE	Computer Aided Protection Engineering
GUI	Matlab Graphical User Interface
LP	Linear Programming
NLP	Non-Linear Programming
MILP	Mixed Integer Linear Programming
MINLP	Mixed Integer Non-Linear Programming
GA	Genetic Algorithm
PSO	Particle Swarm Optimization
SE	Seeker Algorithm

Abbreviation	Meaning
CrSA	Crow Search Algorithm
FA	Firefly Algorithm
HBA	Honey Bee Algorithm
DE	Differential Evolution Algorithm
HSA	Harmony Search Algorithm
ABC	Artificial Bee Colony
CSA	Cuckoo Search Algorithm
WOA	Whale Optimization Algorithm
WCA	Water Cycle Algorithm
TLBO	Teaching Learning Based Optimization Algorithm
GWO	Grey Wolf Optimizer
IGSO	Improved Group Search Optimization
HHO	Harris Hawk Optimization
Symbol	Meaning
SCA	Sine Cosine Algorithm
SFLA	Shuffled Frog Leaping Algorithm
SSO	Symbiotic Organism Search Technique
SOA	Seagull Optimization Algorithm
PSO-TVAC	Time Varying Acceleration Coefficient
HPSO	Hybridized Version Of Particle Swarm Optimization
SA	Simulated Annealing
BBO	Biogeography-Based Optimization
SQP	Sequential Quadratic Programming
MFO	Moth-Flame Optimization
WCMF	Water Cycle And Moth Flame Algorithm
GBO	Gradient-Based Optimizer
GSA	Gravitational Search Algorithm
HS-SA	Hybrid Harmony Search And Simulated Annealing
ILP	Interval Linear Programming
LSHADE	Memory-Based Linear Population Size Reduction Technique Of Success-History-Based Adaptive Differential Evolution Algorithm
IEC	International Electrotechnical Commission
IECSI	IEC Standard Inverse

Abbreviation	Meaning
OF	Objective Function
TMS^{\min}	Minimum Limit of TMS
TMS^{\max}	Maximum Limit of TMS
IP^{\min}	Minimum Limit of IP
IP^{\max}	Maximum Limit of IP
PS^{\max}	Minimum Limit of PS
PS^{\min}	Maximum Limit of PS
t_{\min}	Time Bound For The Minimum Operational Time of The Relay
t_{\max}	Time Bound For The Maximum Operational Time of The Relay
t_{backup}	Relay Operating Time of The Backup Relays
t_{primary}	Relay Operating Time of The Primary Relays
CTI_{\min}	The Minimum Coordination Time Interval
$I_{f,\min}$	Minimum Fault Current
KKT	Karush-Kuhn-Tucker
k	Number of Relay Pairs
μ and ν	Penalty Constants or Penalty Factors
m	Number of Quality Constraints
r	Number of Inequality Constraints
$H_i [\varphi_i (l)]$ and $H_j [\psi_j (l)]$	Index Functions
FAGA	Hybrid Firefly-Genetic Algorithm
AFA	Adaptive Firefly Algorithm
IGWO	Improved Grey Wolf Optimizer
MFA	Modified Firefly Algorithm
Symbol	Meaning
IFA	Improved Firefly Algorithm

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Appendices

Appendix A

IEC and IEEE standards Constants for Overcurrent Relays.

Table A.1

IEC and IEEE Constants for Standard Overcurrent Relays.

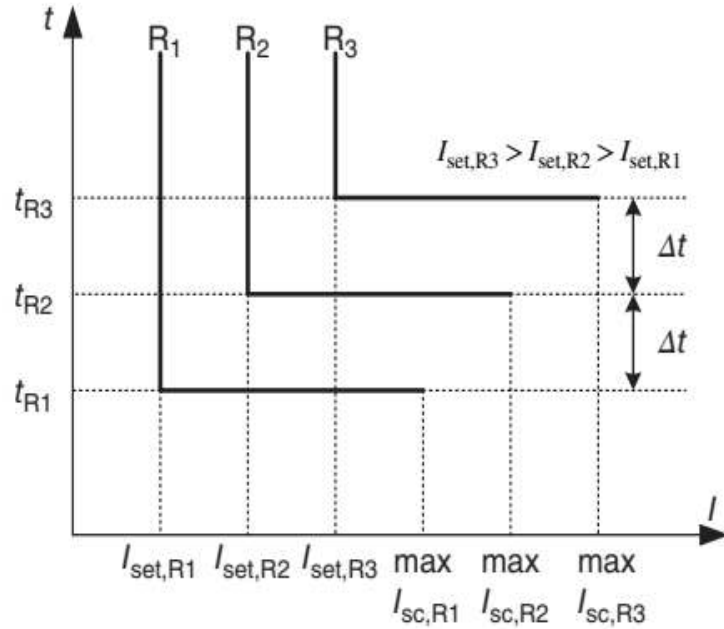
IDMT Curve Description	α	β	C
IEC			
STI	0.05	0.04	
SI	0.14	0.02	
VI	13.5	1	0
EI	80	2	
LTI	120	1	
IEEE			
MI	0.0515	0.02	0.114
VI	19.61	2	0.491
EI	28.2	2	0.1217
I	44.6705	2.0938	0.8983
SI	1.3315	1.2969	0.16965
LI	28.0715	1	10.9296

Appendix B

Time-delay Coordination for different types of overcurrent relay

Figure B.1

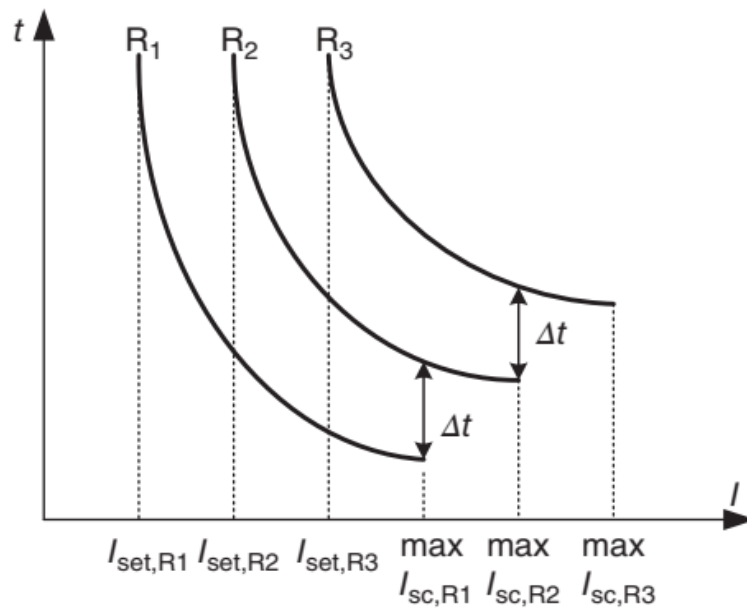
Time-delay Coordination for Definite-Time Protection



Note: Adapted from [15].

Figure B.2

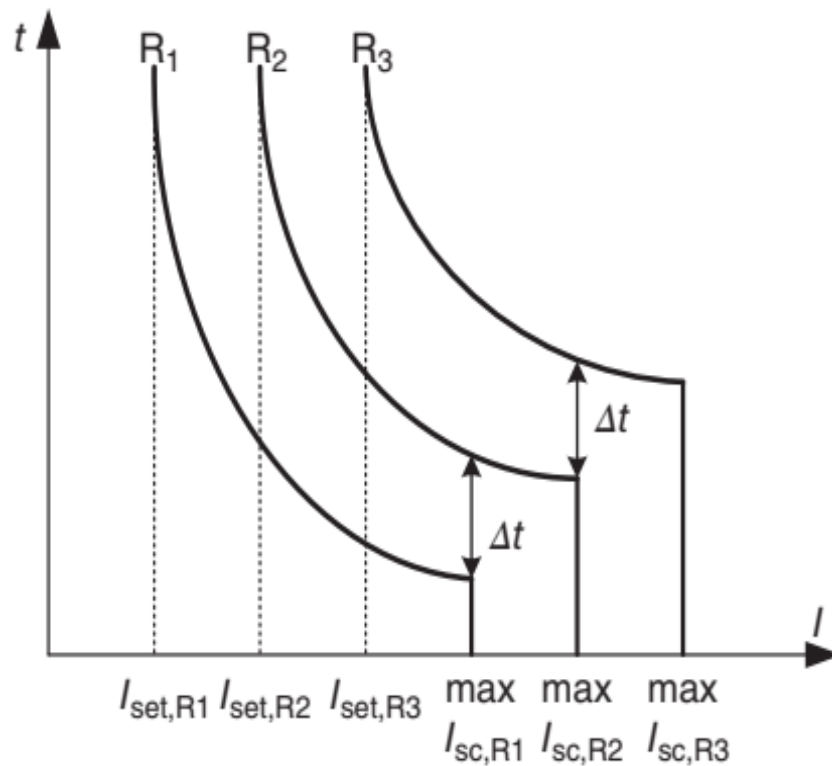
Time-delay Coordination for Inverse-Time Protection



Note: Adapted from [15].

Figure B.3

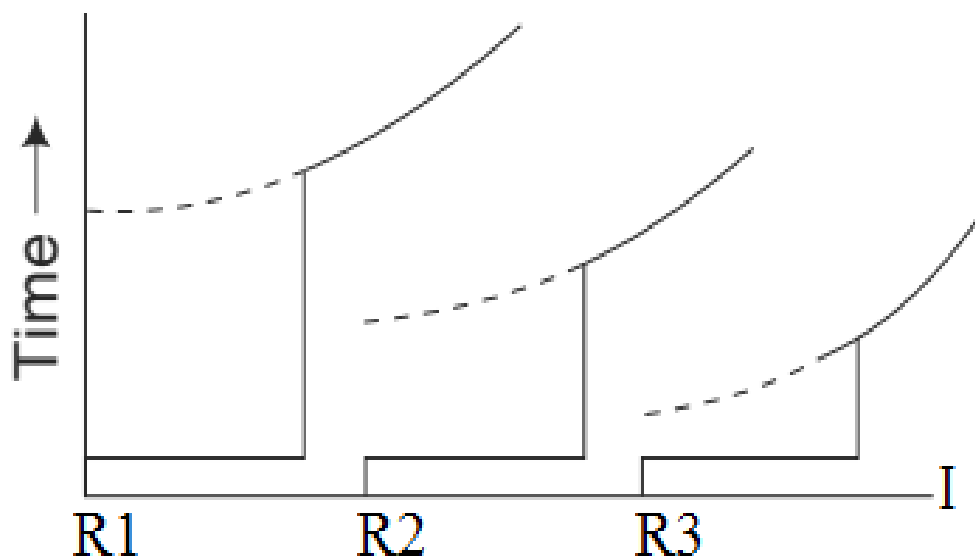
Time-delay Coordination for Combined Inverse-Time and Instantaneous Protection



Note: Adapted from [15].

Figure B.4

Current-based Coordination for Combined Inverse-Time and Instantaneous Protection



Note: Adapted from [19].

Appendix C

Figures related to optimization algorithms section

Figure C.1

Classifications of optimization techniques

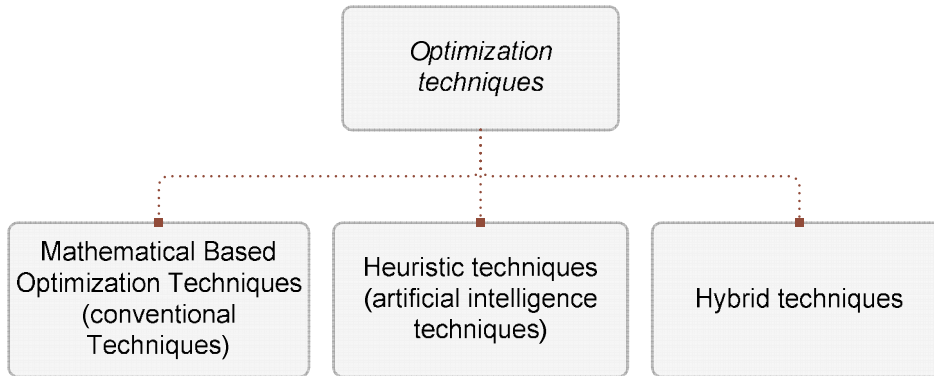
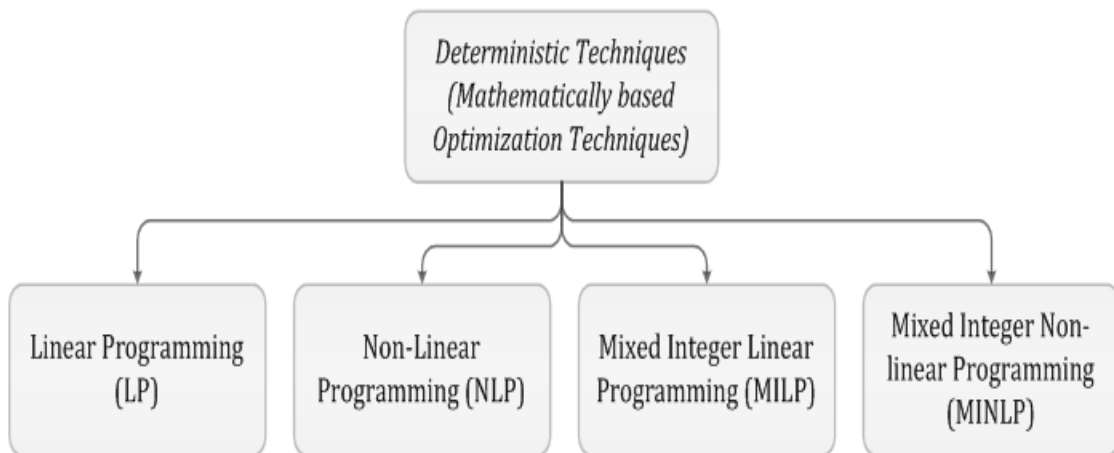


Figure C.2

Mathematically-based optimization methods

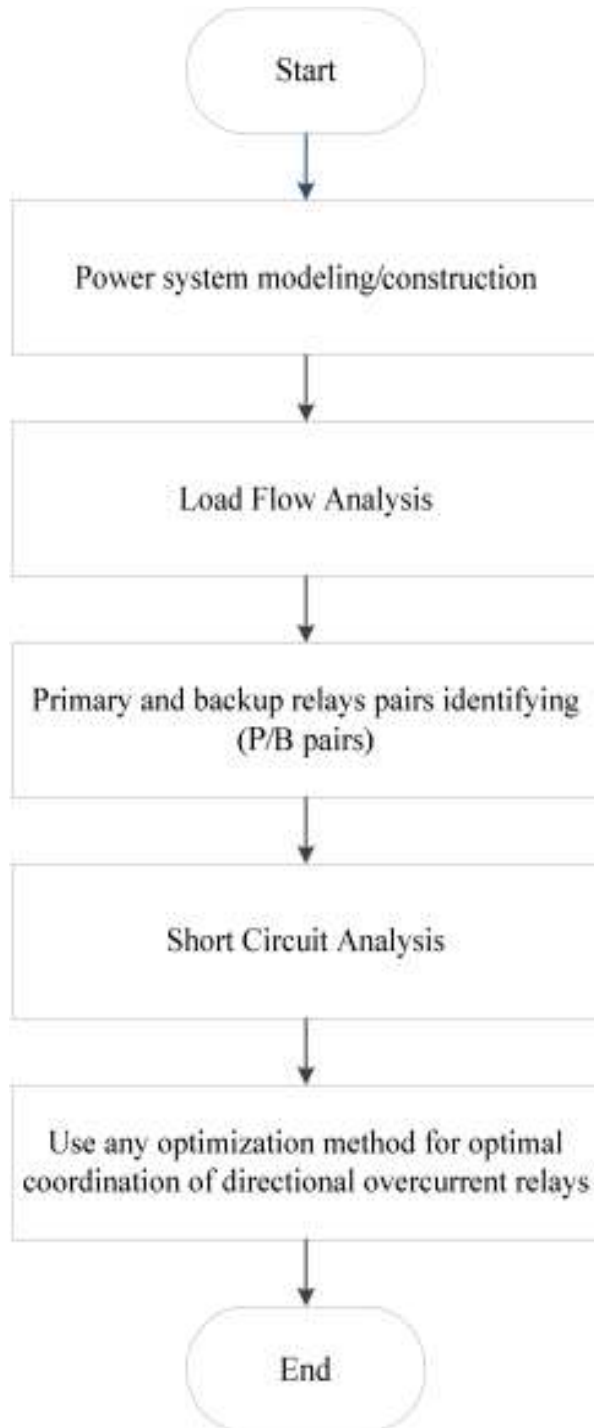


Appendix D

Figures related to Methodology Section

Figure D.1

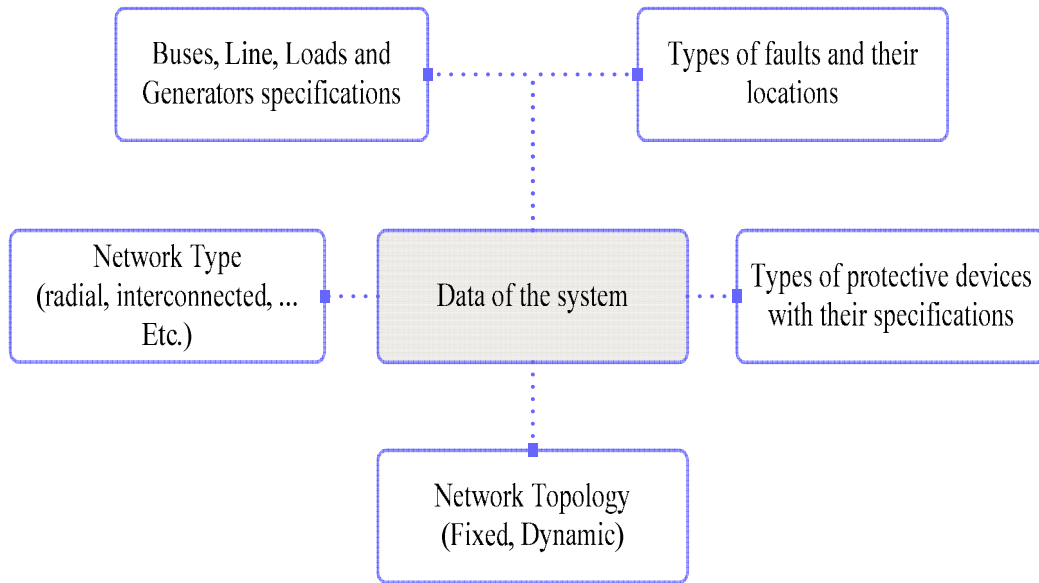
General procedures for optimally coordination directional overcurrent relays



Note: Adapted from [99].

Figure D.2

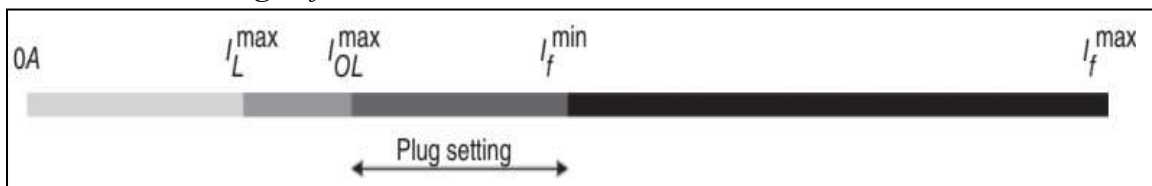
Information needed for modeling the power system



Note: Adapted from [99].

Figure D.3

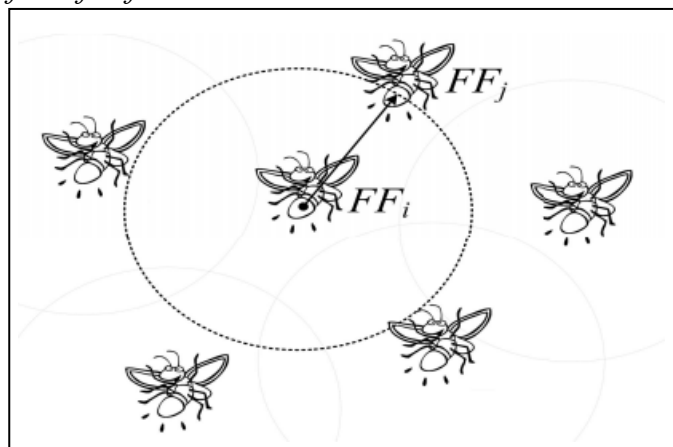
The available range of PS



Note: Adapted from [99].

Figure D.4

Representation of the fireflies



Note: Adapted from [35].

Figure D.5

FA Pseudo Code

-
1. Initialize algorithm parameters (number of fireflies (n), generation number ($nGer$), absorption coefficient (γ), attractiveness (β_{min} and β_{max}) and the randomness strength (α_0))
 2. Objective function $f(x)$, $x=(x_1, \dots, x_d)^T$
 3. Generate initial population of fireflies x_i ($i= 1, 2, \dots, n$)
 4. Light intensity I_i in x_i is determined by $f(x_i)$
 5. While ($t < nGer$)
 - for $i = 1:n$
 - for $j = 1:i$
 - if ($I_j > I_i$), then move i towards j in d -dimension; End if
 - Attractiveness varies with distance r via $\exp[-\gamma r]$
 - Evaluate solutions and update light intensity
 - End for j
 - End for i
 6. Rank the fireflies and find the best global solution
 7. End while
 8. Postprocess results and visualization
-

Figure D.6

Flowchart of the modified FA

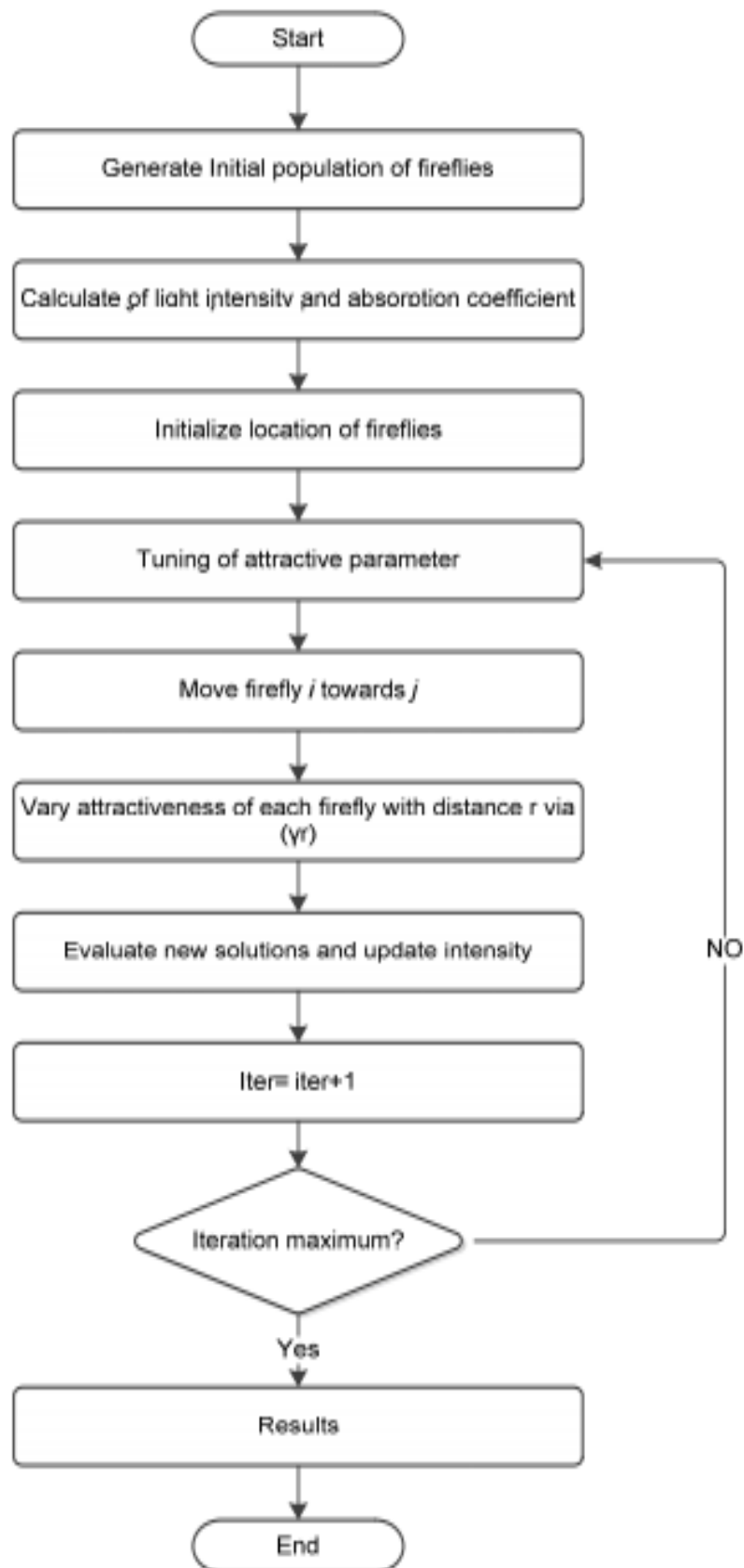


Figure D.7

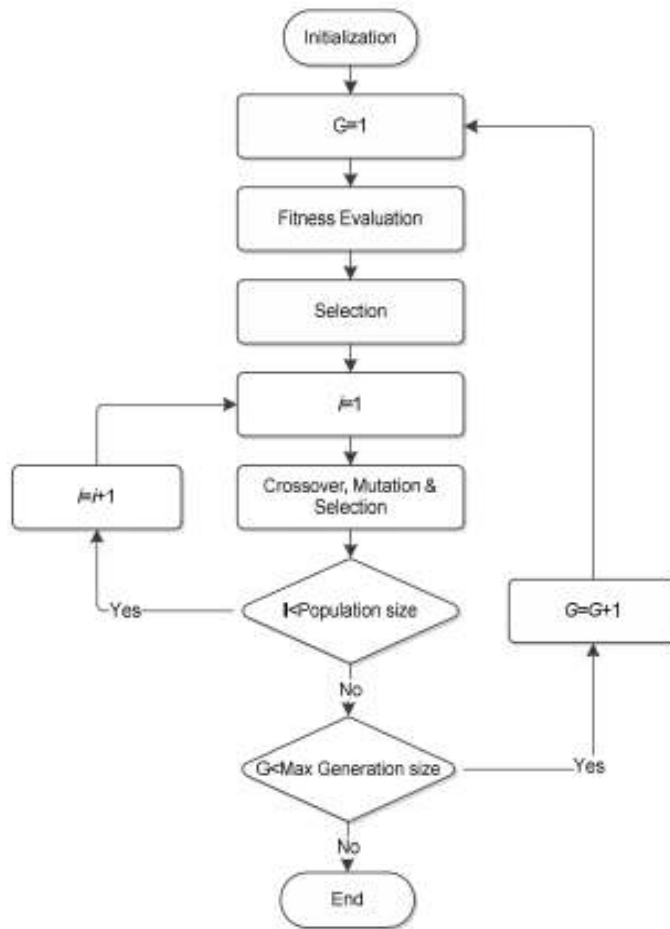
GA Pseudo Code

-
1. Initialize algorithm parameters (number of population (nPop), generation number (nGer), Crossover probability (Pc), Mutation probability (Pm))
 2. Objective function $f(x)$, $x=(x_1, \dots, x_d)^T$
 3. $i=0$
 4. Generate initial population of chromosomes x_i ($i= 1, 2, \dots, n$)
 5. Fitness function is determined by $f(x_i)$
 6. While ($t < nGer$)
 - $i=i+1$
 - Select parents from population
 - Apply crossover mechanism with probability Pc
 - Apply mutation mechanism with probability Pm
 7. Rank individuals and find the best global solution
 8. End while
 9. Postprocess results and visualization
-

Note: Adapted from [108].

Figure D.8

Flowchart of the GA



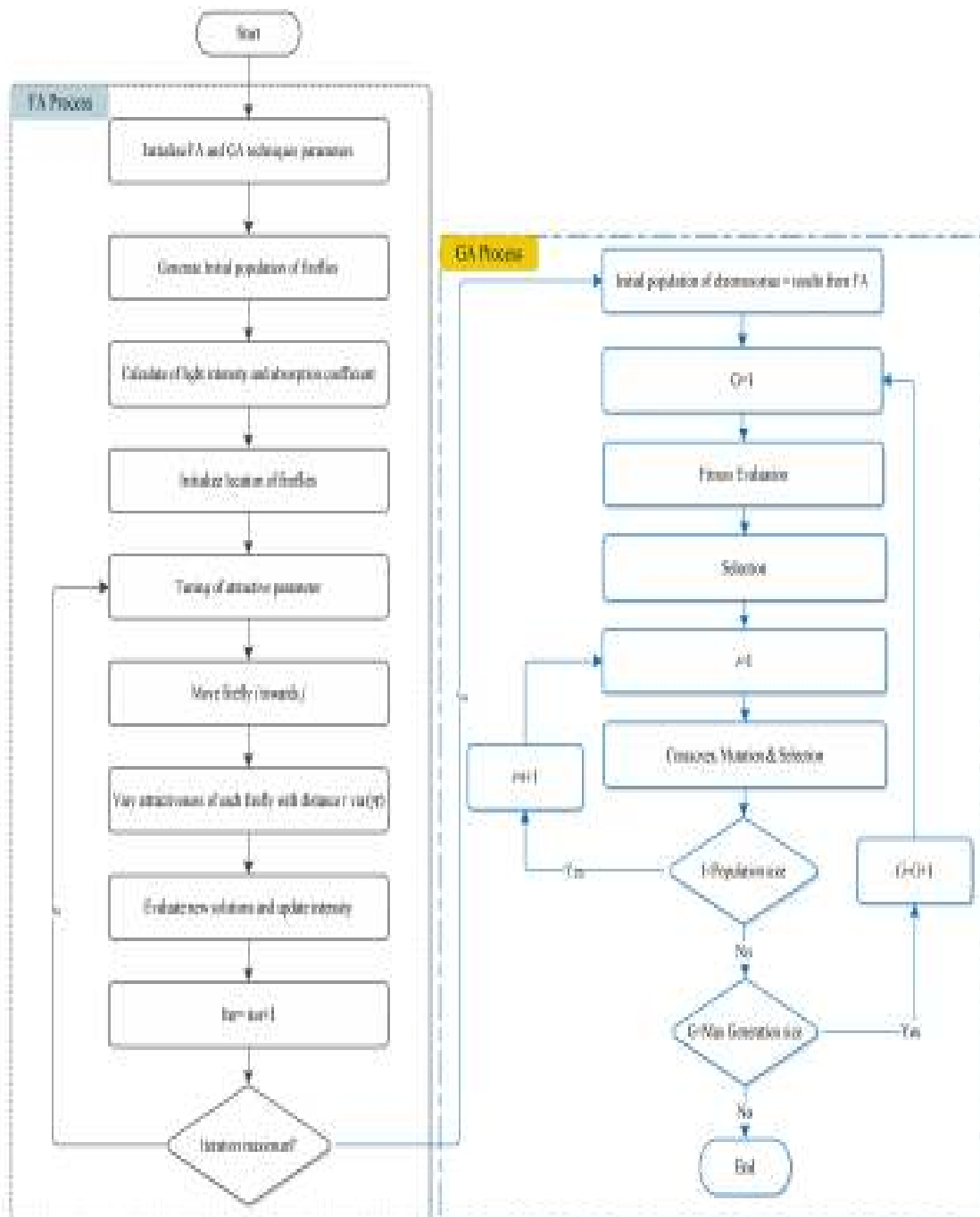
Note: Adapted from [108].

Figure D.9*Hybrid FA-GA Pseudo Code*

Stage 1: FA Starting	
1.	Initialize FA and GA parameters (number of fireflies (n), generation number of FA (nGerFA), absorption coefficient (γ), attractiveness (β_{\min} and β_{\max}) the randomness strength (α), number of population (nPop), generation number of GA (nGerGA), Crossover probability (Pc) and Mutation probability (Pm))
2.	Objective function $f(x)$, $x=(x_1, \dots, x_d)^T$
3.	Generate initial population of fireflies x_i ($i= 1, 2, \dots, n$)
4.	Determine light intensity I_i in x_i by $f(x_i)$
5.	While ($t < nGerFA$) for $i = 1:n$ for $j = 1:n$ if ($I_j > I_i$), then move i towards j in d-dimension; End if Attractiveness varies with distance r via $\exp[-\gamma r^2]$ Evaluate solutions and update light intensity End for j End for i
6.	Rank the fireflies and find the best global solution
7.	End while
8.	Results from FA
9.	End FA
Stage 2: GA Starting	
10.	$i=0$
11.	initial population of chromosomes $P(0) =$ results from FA
12.	Determine fitness function
13.	While ($t < nGerGA$) $i=i+1$ Select parents from population Apply crossover mechanism with probability Pc Apply mutation mechanism with probability Pm Fitness calculation
14.	Rank individuals and find the best global solution
15.	End while
16.	Postprocess results and visualization

Figure D.10

Flowchart of the hybrid FA-GA technique



Appendix E

Test Systems Specifications

Test System II: IEEE 3-bus

Table E.1

Three-phase short circuit current of IEEE 3-bus network

P/B	Primary Relay	Short circuit Current (A)	Backup Relay	Short circuit Current (A)
1	1	1978.90	5	175.00
2	2	1525.70	4	545.00
3	3	1683.90	1	617.22
4	4	1815.40	6	466.17
5	5	1499.66	3	384.00
6	6	1766.30	2	145.34

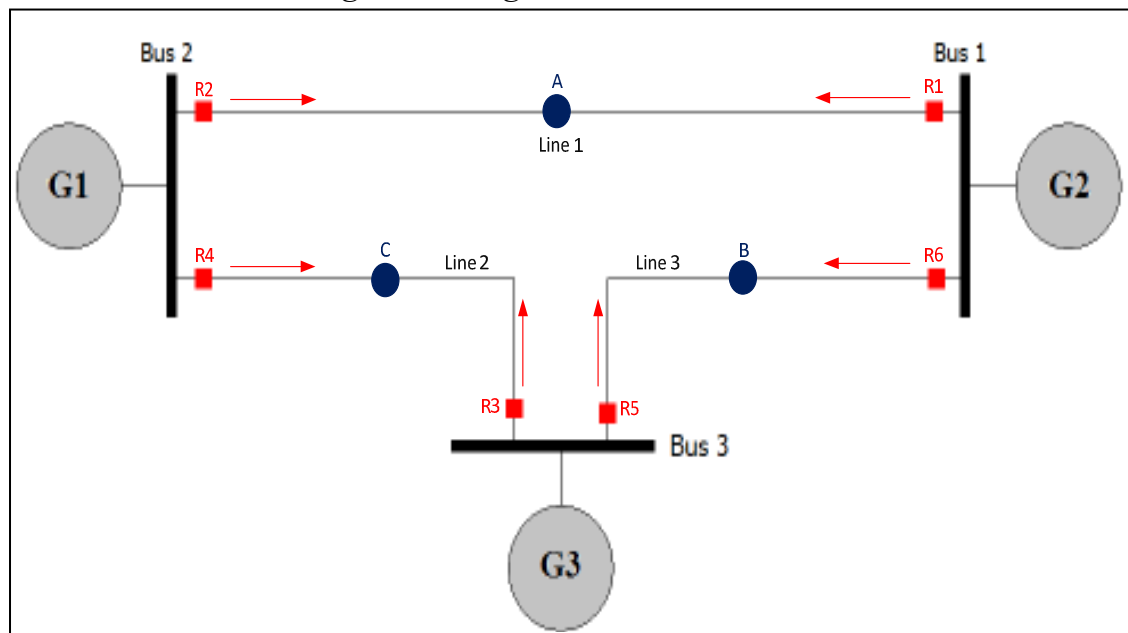
Table E.2

Current transformer ratio for primary relays in IEEE 3-bus network

Relay Number	CT ratio
1, 4	300/5
2, 3, 5	200/5
6	400/5

Figure E.1

IEEE 3-bus network single-line diagram



Test System II: IEEE 6-bus

Table E.3

Three-phase short circuit current of IEEE 6-bus network

Primary Relay	Short circuit Current (kA)	Backup Relay	Short circuit Current (kA)
1	18.172	13	0.6010
2	4.8030	3	1.3650
3	30.547	4	0.5528
4	5.1860	12	3.4220
4	5.1860	14	1.7640
5	2.8380	11	1.0740
5	2.8380	14	1.7640
6	18.338	8	0.7670
7	4.4960	11	1.0740
7	4.4960	12	3.4220
8	2.3510	2	0.8690
8	2.3510	7	1.4830
9	6.0720	1	4.5890
9	6.0720	7	1.4830
10	4.0770	9	0.6390
11	30.939	10	0.9455
12	17.705	6	0.8610
13	17.821	5	0.9770
14	5.4570	1	4.5890
14	5.4570	2	0.8680

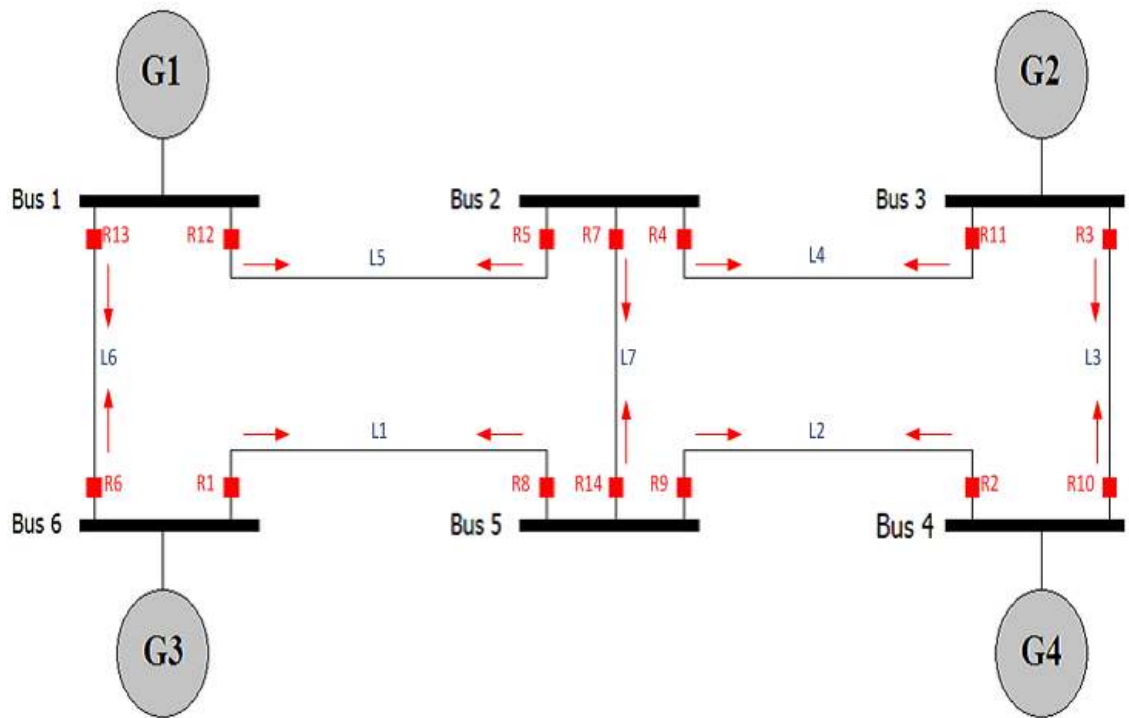
Table E.4

Current transformer ratio for primary relays in IEEE 6-bus network

Relay Number	CT ratio
1, 6, 13	1200/5
2, 3, 4, 5, 7, 8, 9, 11, 12, 14	800/5
10	600/5

Figure E.2

IEEE 6-bus network single-line diagram



Test System III: IEEE 9-bus

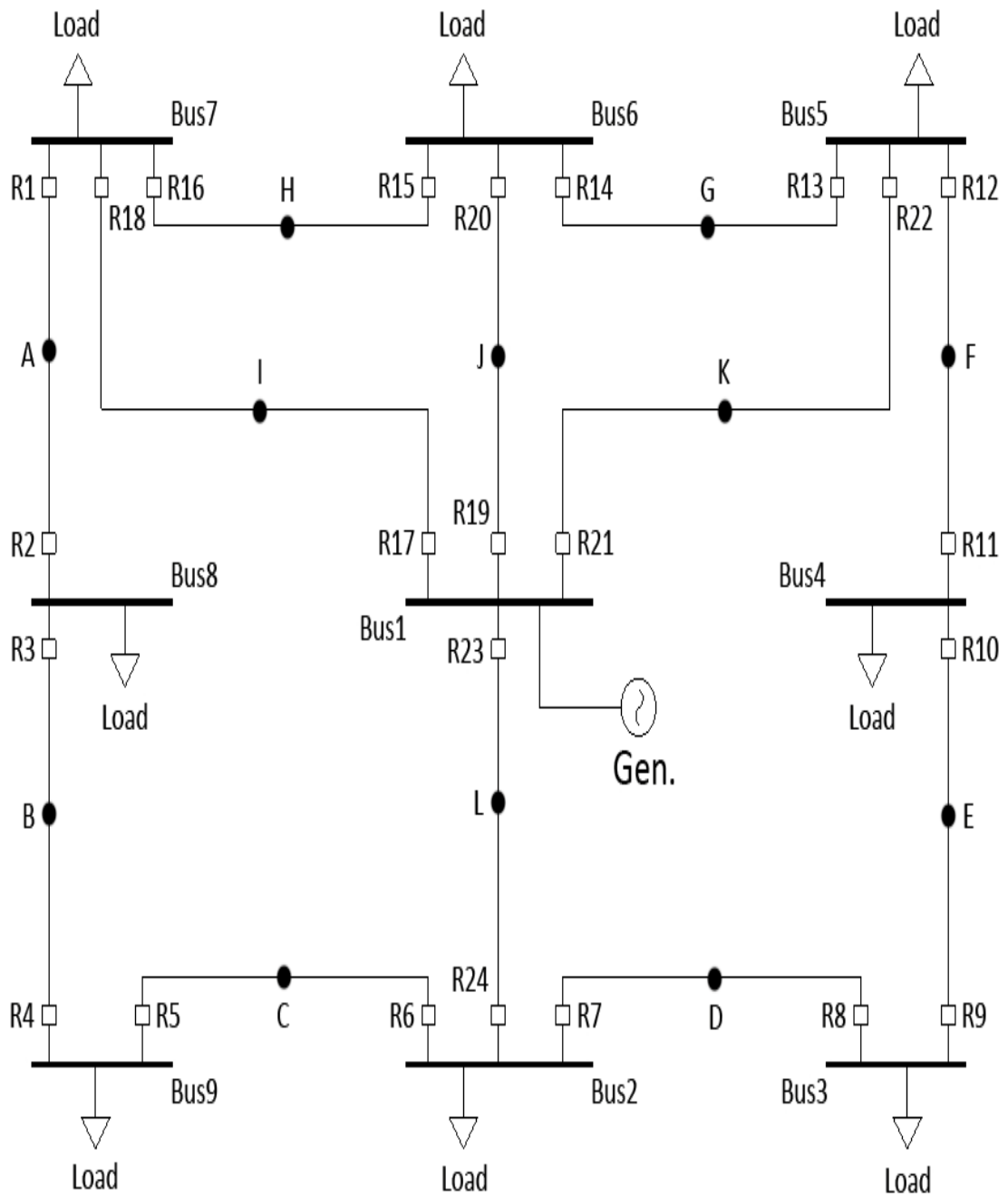
Table E.5

Three-phase short circuit current of IEEE 9-bus network

Primary Relay	Short circuit Current (A)	Backup Relay	Short circuit Current (A)
1	4863.6	15	1168.3
1	4863.6	17	1293.9
2	1634.4	4	1044.2
3	2811.4	1	1361.6
4	2610.5	6	1226.0
5	1778.0	3	1124.4
6	4378.5	8	711.20
6	4378.5	23	1345.5
7	4378.5	5	711.20
7	4378.5	23	1345.5
8	1778.0	10	1124.4
9	2610.5	7	1226.0
10	2811.4	12	787.20
11	1634.4	9	1044.2
12	2811.4	14	1168.2
12	2811.4	21	1293.9
13	3684.5	11	653.60
13	3684.5	21	1293.9
14	4172.5	16	1031.7
14	4172.5	19	1264.1
15	4172.5	13	1031.7
15	4172.5	19	1264.1
16	3684.5	2	653.60
16	3684.5	17	1293.9
17	7611.2	-	0
18	2271.7	2	653.60
18	2271.7	15	1168.3
19	7435.8	-	0
20	2624.2	13	1031.7
20	2624.2	16	1031.7
21	7611.2	-	0
22	2271.7	11	653.60
22	2271.7	14	1168.3
23	7914.7	-	0
24	1665.5	5	711.20
24	1665.5	8	711.20

Figure E.3

IEEE 9-bus network single-line diagram



Test System IV: IEEE 15-bus

Table E.6

Three-phase short circuit current of IEEE 15-bus network

Primary Relay	Short circuit Current (A)	Backup Relay	Short circuit Current (A)	Primary Relay	Short circuit Current (A)	Backup Relay	Short circuit Current (A)
1	3621	6	1233	20	7662	30	681
2	4597	4	1477	21	8384	17	599
2	4597	16	743	21	8384	19	1372
3	3984	1	853	21	8384	30	681
3	3984	16	743	22	1950	23	979
4	4382	7	1111	22	1950	34	970
4	4382	12	1463	23	4910	11	1475
4	4382	20	1808	23	4910	13	1503
5	3319	2	922	24	2296	21	1326
6	2647	8	1548	24	2296	34	970
6	2647	10	1100	25	2289	15	969
7	2497	5	1397	25	2289	18	1320
7	2497	10	1100	26	2300	28	1192
8	4695	3	1424	26	2300	36	1109
8	4695	12	1463	27	2011	25	903
8	4695	20	1808	27	2011	36	1109
9	2943	5	1397	28	2525	29	1828
9	2943	8	1548	28	2525	32	697
10	3568	14	1175	29	8346	17	599
11	4342	3	1424	29	8346	19	1372
11	4342	7	1111	29	8346	22	642
11	4342	20	1808	30	1736	27	1039
12	4195	13	1503	30	1736	32	697
12	4195	24	753	31	2867	27	1039
13	3402	9	1009	31	2867	29	1828
14	4606	11	1475	32	2069	33	1162
14	4606	24	753	32	2069	42	907
15	4712	1	853	33	2305	21	1326
15	4712	4	1477	33	2305	23	979
16	2225	18	1320	34	1715	31	809
16	2225	26	905	34	1715	42	907
17	1875	15	969	35	2095	25	903
17	1875	26	905	35	2095	28	1192
18	8426	19	1372	36	3283	38	882
18	8426	22	642	37	3301	35	910
18	8426	30	681	38	1403	40	1403
19	3998	3	1424	39	1434	37	1434
19	3998	7	1111	40	3140	41	1434
19	3998	12	1463	41	1971	31	809
20	7662	17	599	41	1971	33	1162
20	7662	22	642	42	3295	39	896

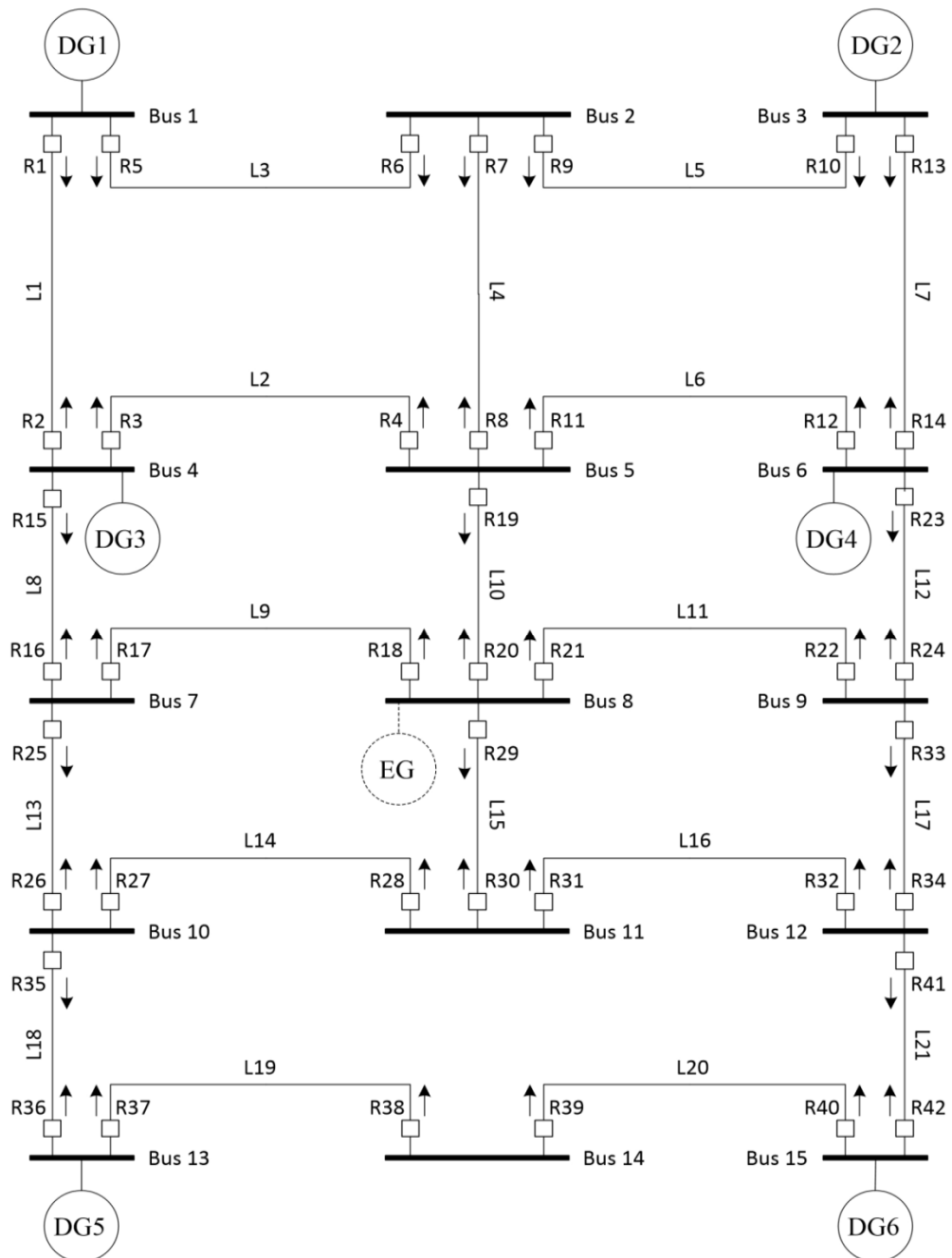
Table E.7

Current transformer ratio for primary relays in IEEE 15-bus network

Relay Number	CT ratio
17, 22, 30, 34, 38, 39, 41	400/5
6, 7, 9, 16, 24, 25, 26, 27, 28, 31, 32, 33, 35	600/5
1, 3, 5, 10, 13, 19, 36, 37, 40, 42	800/5
2, 4, 8, 11, 12, 14, 15, 23	1200/5
18, 20, 21, 29	1600/5

Figure E.4

IEEE 15-bus network single-line diagram



Appendix F

Results

IEEE 6-bus Linear Programming Formulation

Table F.1

PS values for relays in IEEE 6-bus network

Relay Number	PS	Relay Number	PS
1	0.8	8	0.8
2	0.8	9	0.5
3	1.0	10	1.0
4	0.5	11	1.0
5	0.5	12	1.5
6	0.5	13	0.5
7	1.0	14	1.0

Table F.2

LP-based relay settings for the IEEE 6-bus network

TMS	TLBO [24]	PSO- DE [92]	IA- PSO [113]	FA [35]	IFA [35]	MFA	GA	FAGA
R1	0.3780	0.4064	0.2602	0.1002	0.1000	0.281092	0.237645	0.237660
R2	0.3443	0.7506	0.4739	0.1002	0.1000	0.183928	0.141543	0.141515
R3	0.2553	0.3872	0.2406	0.1006	0.1296	0.342781	0.145152	0.145036
R4	0.3346	0.4031	0.2711	0.1102	0.1050	0.178586	0.107978	0.108029
R5	0.1005	0.2005	0.1268	0.1000	0.1007	0.144307	0.136076	0.136054
R6	0.2376	0.2011	0.1264	0.4070	0.3880	0.142615	0.142682	0.142615
R7	0.3000	0.2003	0.1265	0.1092	0.1016	0.141982	0.142060	0.141982
R8	0.4720	0.2133	0.1265	0.1000	0.1000	0.101237	0.101296	0.101237
R9	0.0414	0.2006	0.1268	0.1175	0.1128	0.133389	0.125504	0.125525
R10	0.3323	0.2265	0.1424	0.2860	0.1001	0.111931	0.111617	0.111619
R11	0.2518	0.2610	0.1647	0.1439	0.1143	0.135528	0.135385	0.135377
R12	0.2704	0.2039	0.1401	0.3404	0.2138	0.190353	0.190452	0.190353
R13	0.1735	0.2002	0.1265	0.1063	0.1899	0.144114	0.128572	0.128488
R14	0.2817	0.2837	0.1170	0.4070	0.2881	0.171252	0.160715	0.160723
Fitness	23.787	9.2671	8.1245	7.6866	6.4040	3.87061	3.29554	3.29480

Table F.3

Comparison of the proposed methods' percentage improvement and the overall net gain in time over the literature's algorithms

Method	$\Sigma \Delta(t)$ (sec)	% Improvement	Method	$\Sigma \Delta(t)$ (sec)	% Improvement
MFA/TLBO	19.9164	83.728	FAGA/TLBO	20.4922	85.149
MFA/PSO- DE	5.39649	58.233	FAGA/PSO- DE	5.97230	64.446
MFA/ IA- PSO	4.25389	52.359	FAGA/ IA- PSO	4.82970	59.446
MFA/ FA	3.81599	49.645	FAGA/ FA	4.39180	57.136
MFA/ IFA	2.53339	39.559	FAGA/ IFA	3.10920	48.551
MFA/GA	-	-	FAGA/GA	0.00074	0.0225

Table F.4

CTI using the LP formulation for the IEEE 6-bus network

P/R	MFA	GA	FAGA
	CTI	CTI	CTI
1	0.20481	0.20038	0.20045
2	0.75308	0.20013	0.20034
3	0.20096	0.20003	0.20028
4	0.20074	0.31476	0.31503
5	0.19955	0.28391	0.28374
6	0.21762	0.23064	0.23084
7	0.21437	0.20000	0.20004
8	0.20069	0.20050	0.20035
9	0.20195	0.20033	0.20026
10	0.20000	0.20047	0.20075
11	0.42290	0.27130	0.27113
12	0.20000	0.20000	0.20000
13	0.39363	0.31339	0.31337
14	0.22975	0.24197	0.24204
15	0.22601	0.20031	0.20044
16	0.20087	0.20027	0.20035
17	0.20050	0.20048	0.20035
18	0.20043	0.20039	0.20010
19	0.27309	0.20010	0.20020
20	0.33201	0.20049	0.20016

IEEE 6-bus Non Linear Programming Formulation

Table F.5

NLP-based relay settings for the IEEE 6-bus network

Variables	MFA	GA	FAGA
TMS1	0.436650	0.287456	0.249768
TMS2	0.100016	0.185867	0.100016
TMS3	0.121972	0.175585	0.101220
TMS4	0.100000	0.146315	0.100000
TMS5	0.100001	0.139980	0.100001
TMS6	0.100000	0.100125	0.100000
TMS7	0.137568	0.193363	0.136347
TMS8	0.152850	0.158478	0.151751
TMS9	0.113350	0.149016	0.100655
TMS10	0.178494	0.200589	0.176909
TMS11	0.100000	0.125244	0.100000
TMS12	0.272713	0.311442	0.271248
TMS13	0.100007	0.136331	0.100007
TMS14	0.122226	0.200200	0.117578
PS1	1.079420	0.501935	0.500038
PS2	1.755140	0.500968	1.400890
PS3	1.787890	0.793011	1.633960
PS4	1.458140	0.501100	0.696583
PS5	2.319870	0.500247	0.823288
PS6	2.077760	1.015010	0.948732
PS7	1.709690	0.727959	1.469700
PS8	0.820459	0.500476	0.527368
PS9	1.197200	0.500027	0.890075
PS10	0.574520	0.500830	0.555721
PS11	1.940750	1.428750	1.755200
PS12	0.678410	0.502007	0.606144
PS13	1.819770	0.500759	0.682573
PS14	2.298170	0.592759	1.621050
Fitness	4.04849	3.84454	3.22831

Table F.6*CTI using the NLP formulation for the IEEE 6-bus network*

P/R	MFA	GA	FAGA
	CTI	CTI	CTI
1	1.50040	0.20206	0.20051
2	0.29804	0.20011	0.20015
3	0.51849	0.20118	0.20004
4	0.31570	0.32398	0.33869
5	0.31853	0.23006	0.24667
6	0.22010	0.29435	0.29378
7	0.20006	0.20068	0.20095
8	0.20016	0.20140	0.20000
9	0.22295	0.20213	0.20087
10	0.20008	0.20240	0.20007
11	0.25300	0.21605	0.20138
12	0.20000	0.20132	0.20003
13	0.81153	0.30225	0.28175
14	0.33840	0.28794	0.32798
15	0.38422	0.20121	0.20002
16	0.20000	0.20000	0.20009
17	0.95586	0.20355	0.20941
18	0.53451	0.20041	0.20004
19	0.72445	0.20089	0.20000
20	0.30460	0.20131	0.24756

Figure F.1*Total operation time compared to the literature for the proposed methods*

Figure F.2

Percentage improvement of the proposed methods compared to other algorithms in the literature for IEEE 3-bus network in terms of the net relay operational time

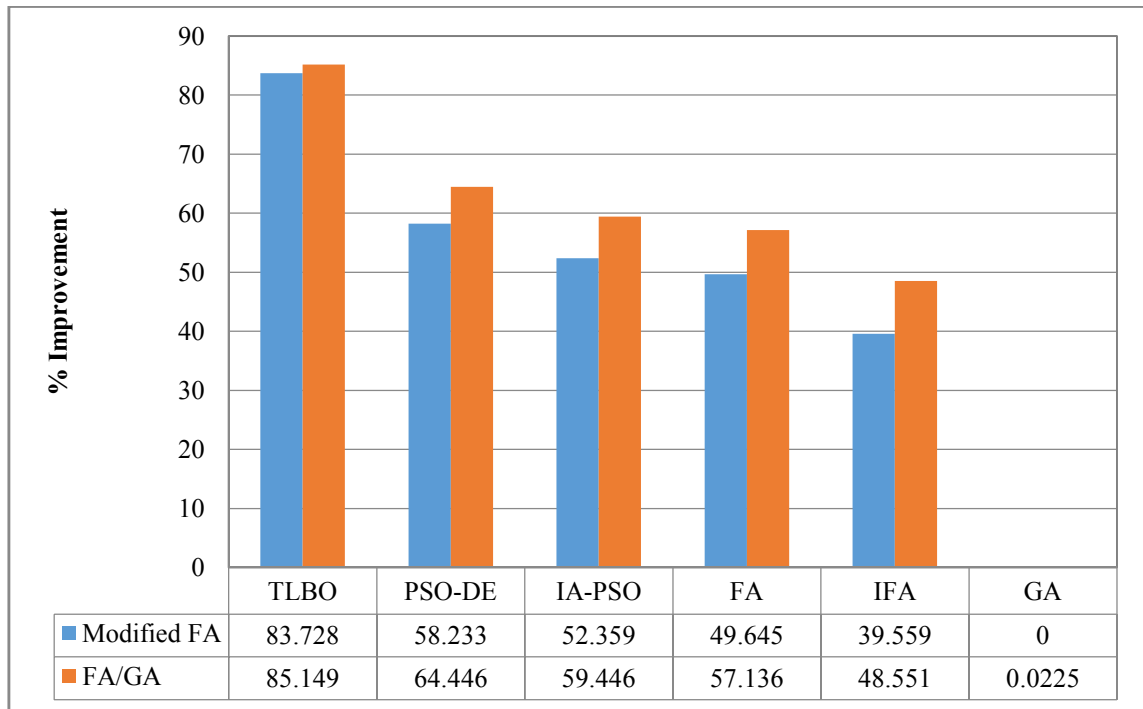
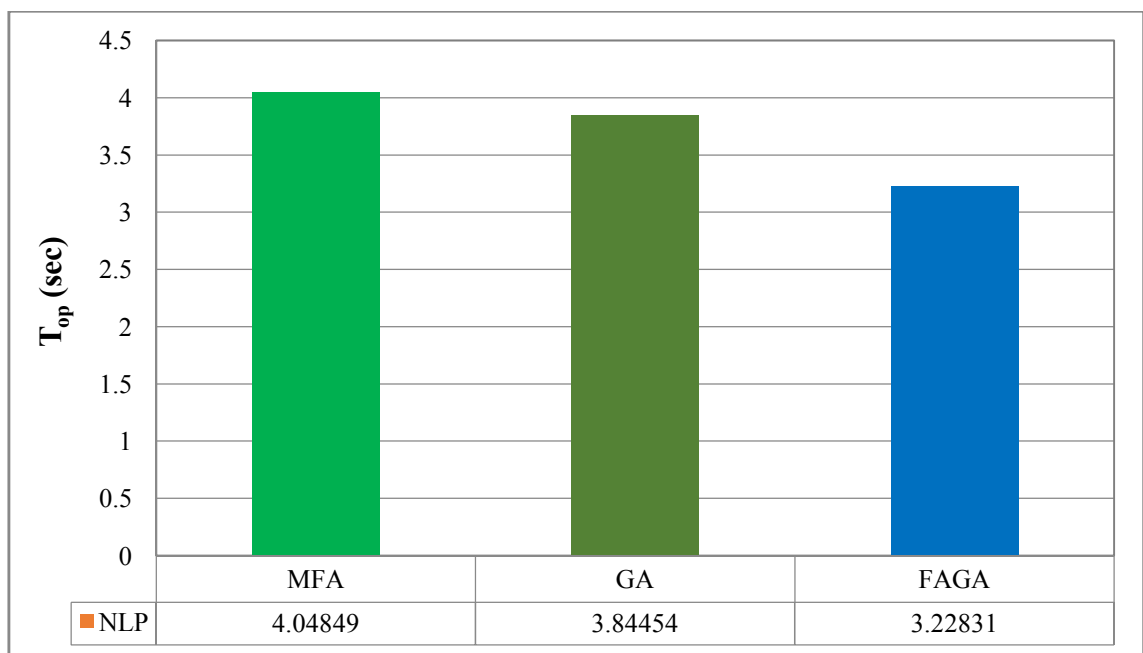


Figure F.3

Total operating time of the proposed methods.



IEEE 9-bus Non Linear Programming Formulation

Table F.7

NLP-based relay settings for the IEEE 9-bus network

Variables	Seeker [114]	FA [35]	IFA [35]	WOA [35]	MFA	GA	FAGA
TMS1	0.2662	1.0850	0.9788	0.1039	0.100000	0.100045	0.100000
TMS2	0.2076	1.0654	0.7945	0.5934	0.100000	0.100024	0.100000
TMS3	0.2928	0.5560	0.1309	0.6080	0.121327	0.100000	0.111683
TMS4	0.3192	0.4492	0.3740	0.1000	0.100006	0.100072	0.100000
TMS5	0.2879	0.7472	0.3880	0.1000	0.100000	0.100015	0.100000
TMS6	0.3677	1.0764	0.6056	1.1000	0.123330	0.100000	0.112346
TMS7	0.3006	0.9764	0.6240	0.3927	0.100022	0.100000	0.100022
TMS8	0.2905	1.0472	1.0285	1.1000	0.100000	0.100083	0.100000
TMS9	0.2476	1.0492	0.5377	0.1000	0.100000	0.117236	0.100000
TMS10	0.2480	0.8557	0.7623	0.1000	0.100000	0.119556	0.100000
TMS11	0.2578	0.7305	0.2139	1.1000	0.100035	0.100000	0.100035
TMS12	0.3665	0.9802	0.5964	0.2313	0.100100	0.100024	0.100100
TMS13	0.2581	1.0492	0.9327	0.5111	0.100000	0.100282	0.100000
TMS14	0.3117	1.0637	0.7761	0.1000	0.100026	0.100103	0.100026
TMS15	0.2921	1.0908	0.6309	0.1000	0.428972	0.101203	0.114164
TMS16	0.3633	1.0593	1.0295	0.8753	0.100138	0.100041	0.100016
TMS17	0.2560	0.9420	0.2740	0.1005	0.375384	0.100000	0.111377
TMS18	0.1038	1.0295	0.6019	0.1005	0.100000	0.100024	0.100000
TMS19	0.2589	1.0572	0.7807	0.1121	0.130127	0.115283	0.100098
TMS20	0.1002	0.3245	0.2164	0.2180	0.100114	0.100000	0.100114
TMS21	0.2758	0.5918	0.3072	0.1000	0.116052	0.100090	0.100061
TMS22	0.1010	0.6747	0.4921	0.1000	0.100000	0.100000	0.100000
TMS23	0.1757	0.6609	0.1011	0.1005	0.100036	0.113113	0.100036
TMS24	0.1014	0.9041	0.4305	0.1000	0.100000	0.100046	0.100000
PS1	1.2732	1.8150	1.6079	0.5196	1.480550	0.713501	0.749105
PS2	1.5200	1.2988	0.3692	1.3962	0.542203	0.500025	0.500089
PS3	1.1975	1.4980	0.9686	0.6714	0.5827590	0.639849	0.556856
PS4	0.6701	1.3920	0.3262	0.5000	0.8656400	0.667917	0.619558
PS5	1.0785	0.9480	0.4523	0.5000	0.7030500	0.500011	0.506492
PS6	0.6311	1.6430	1.5864	2.0000	0.5672920	0.696574	0.555085
PS7	0.9637	1.6430	1.3431	0.7141	0.8748260	0.668990	0.615549
PS8	1.1393	0.9480	0.2088	2.0000	0.5882030	0.500019	0.500068
PS9	1.1994	1.3920	0.4618	0.5000	0.8313440	0.500061	0.657638
PS10	1.7451	1.4980	0.8566	0.5000	1.0481100	0.502563	0.640949
PS11	0.8454	1.1369	1.1585	2.0000	0.9184340	0.500000	0.553641
PS12	0.6461	1.8150	1.7203	0.8336	0.6561500	0.500018	0.500022
PS13	0.9784	1.3740	0.9274	0.5283	1.5035700	0.503662	0.519555

Variables	Seeker [114]	FA [35]	IFA [35]	WOA [35]	MFA	GA	FAGA
PS14	0.8860	1.5560	0.6999	0.5000	1.9258400	0.785659	0.603911
PS15	0.8993	1.5560	0.5456	0.5000	0.5042720	0.594298	0.500122
PS16	0.5004	0.9639	0.9873	1.2390	1.3279600	0.507543	0.508474
PS17	0.9197	1.7200	1.3869	0.5129	0.5009380	0.982412	0.500328
PS18	0.5003	1.6347	1.9435	0.5028	0.6570340	0.500000	0.500284
PS19	0.7629	1.6800	1.2390	0.5000	1.2197700	0.500000	0.628484
PS20	0.5041	0.8006	0.7110	1.6530	0.6508340	0.500000	0.500077
PS21	0.8902	1.7200	1.6203	0.5000	1.2047200	0.622735	0.763314
PS22	0.5008	0.7000	0.5859	0.5000	0.9303220	0.500000	0.500055
PS23	1.5724	1.7900	1.4038	0.5025	1.1970700	0.512695	0.635455
PS24	0.5017	0.7441	0.7467	0.5000	0.9271820	0.500022	0.500281
Fitness	14.2338	14.2216	13.3409	13.7000	10.23700	7.08666	7.03106

Table F.8

Comparison of the proposed methods' percentage improvement and the overall net gain in time over the literature's algorithms

Method	$\sum \Delta(t)$ (sec)	% Improvement	Method	$\sum \Delta(t)$ (sec)	% Improvement
MFA/ TLBO [115]	72.6642	87.6516	FAGA/TLBO [115]	75.87014	91.5188
MFA/ IDE [115]	46.4101	81.9285	FAGA/ IDE [115]	49.61604	87.5879
MFA/BBO [8]	18.5978	64.4978	FAGA/BBO [8]	21.80374	75.6161
MFA/ FA	3.9968	28.0796	FAGA/ FA	7.20274	50.6031
MFA/ IFA	3.9846	28.0179	FAGA/ IFA	7.19054	50.5607
MFA/ WOA	3.1039	22.6562	FAGA/ WOS	6.30984	46.5722
MFA/GA	-	-	FAGA/GA	0.05560	0.78457

Table F.9*CTI using the NLP formulation for the IEEE 9-bus network*

P/R	MFA	GA	FAGA
	CTI	CTI	CTI
1	1.563200	0.24947	0.24435
2	1.207000	0.45453	0.20059
3	0.406350	0.24184	0.20319
4	0.777460	0.20103	0.20503
5	0.197750	0.21582	0.20009
6	0.194270	0.20000	0.20007
7	0.479920	0.39416	0.38535
8	0.550260	0.20010	0.20090
9	0.691380	0.39818	0.41417
10	0.559510	0.20473	0.22143
11	0.527580	0.20000	0.20082
12	0.361250	0.22789	0.20191
13	0.385170	0.26540	0.28854
14	0.207930	0.20006	0.21138
15	3.299800	0.35356	0.22792
16	0.735760	0.20282	0.28412
17	-	-	-
18	1.544000	0.46712	0.55129
19	-	-	-
20	0.620760	0.23049	0.30972
21	-	-	-
22	1.113300	0.20271	0.23325
23	-	-	-
24	0.770660	0.20048	0.23671
25	1.172100	0.22988	0.22458
26	0.20162	0.22884	0.22039
27	0.38757	0.46743	0.46660
28	1.16970	0.46100	0.21174
29	0.43498	0.41171	0.41121
30	1.57340	0.20021	0.20011
31	1.88400	0.20007	0.20945
32	1.25620	0.20126	0.20181
33	1.54420	0.41149	0.49803
34	3.18490	0.32560	0.20026
35	0.44749	0.30093	0.30860
36	0.24528	0.30154	0.30032

Figure F.4

Total operation time compared to the literature for the proposed methods

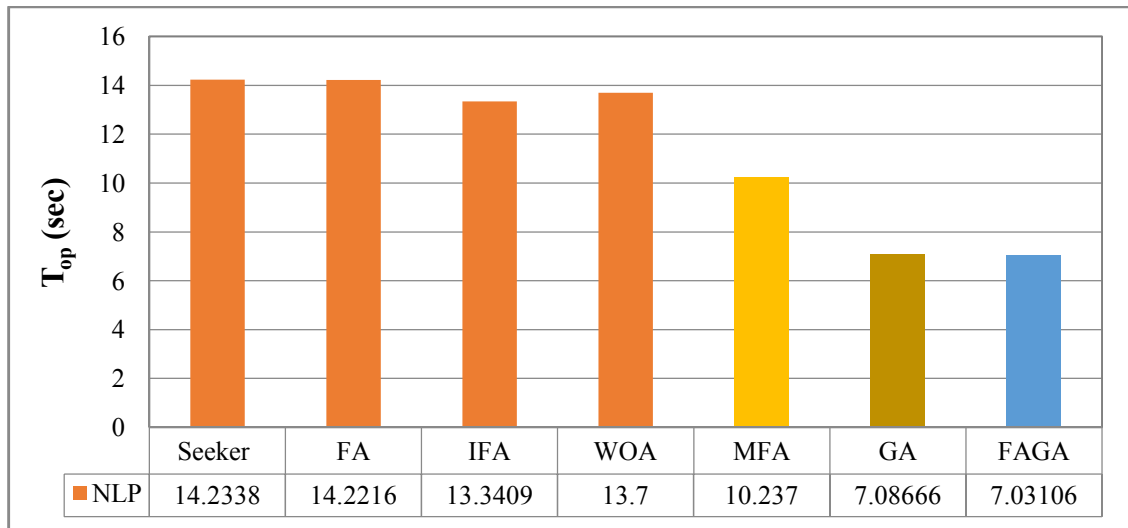
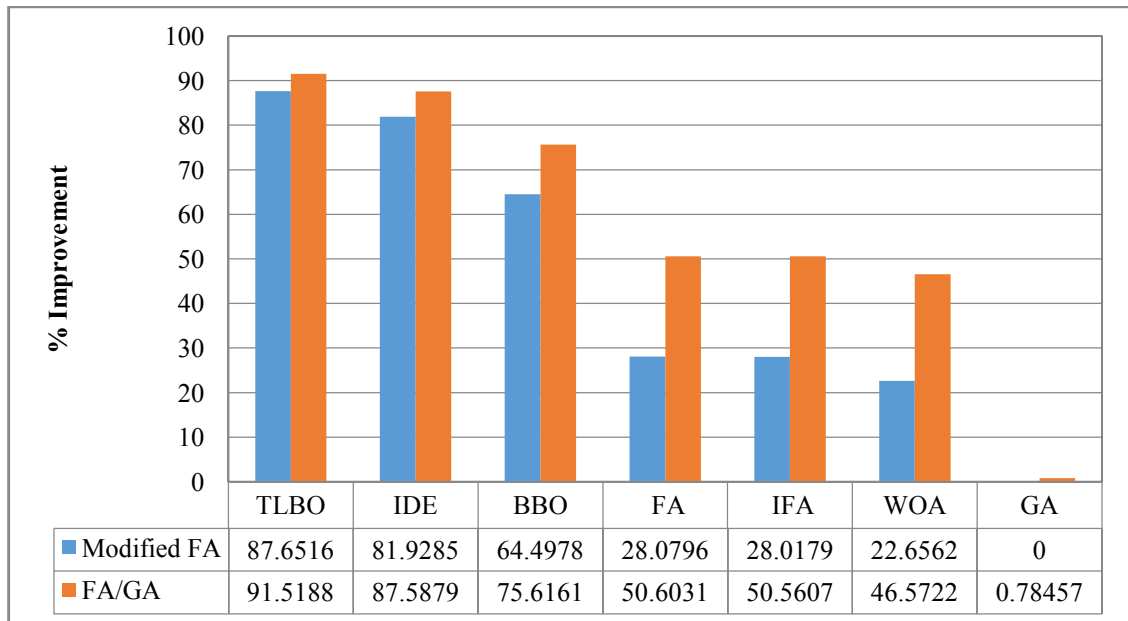


Figure F.5

Percentage improvement of the proposed methods compared to other algorithms in the literature for IEEE 3-bus network in terms of the net relay operational time



IEEE 15-bus Non Linear Programming Formulation

Table F.10

NLP-based relay settings for the IEEE 15-bus network

Relay	MFA		GA		FAGA	
	TMS	PS	TMS	PS	TMS	PS
1	0.100272	2.112490	0.206152	0.500055	0.100028	2.108580
2	0.199528	0.818450	0.208066	0.500000	0.168278	0.810638
3	0.148360	1.631750	0.133555	2.047870	0.148116	1.631510
4	0.102725	1.985600	0.217709	0.500111	0.102725	1.985600
5	0.132894	2.051690	0.252685	0.678760	0.128988	2.051690
6	0.143683	1.769940	0.123010	2.298000	0.143439	1.769940
7	0.126064	2.362570	0.135935	2.056640	0.125087	2.331320
8	0.147295	1.370410	0.145139	1.463750	0.138262	1.370410
9	0.157093	1.586560	0.256192	0.643388	0.157093	1.570940
10	0.216869	0.774944	0.236010	0.500244	0.216869	0.774944
11	0.160230	1.132450	0.154077	0.958190	0.160230	1.132450
12	0.100060	1.885810	0.206159	0.633033	0.100060	1.885810
13	0.170058	1.510500	0.263739	0.522666	0.169924	1.260500
14	0.145922	1.019050	0.115466	1.368390	0.145312	1.019050
15	0.118373	1.437680	0.213140	0.500702	0.118373	1.312190
16	0.133919	1.450150	0.124083	1.439820	0.133919	1.450150
17	0.190055	1.171030	0.101080	2.292620	0.174430	1.170910
18	0.179936	0.575983	0.214490	0.504973	0.179968	0.574035
19	0.154495	1.367610	0.244176	0.511433	0.154495	1.234020
20	0.165046	0.841027	0.222023	0.501041	0.164944	0.841021
21	0.107406	2.059560	0.205402	0.500000	0.107406	1.559560
22	0.205113	0.604595	0.129066	1.799800	0.205021	0.604586
23	0.145834	1.655810	0.202590	0.506614	0.144735	1.055440
24	0.114539	2.033670	0.205760	0.649283	0.112586	2.033670
25	0.166519	1.328160	0.248675	0.657970	0.166665	1.328170
26	0.120819	2.151390	0.213332	0.501492	0.119842	2.151390
27	0.151402	1.242180	0.228699	0.738998	0.151532	1.242250
28	0.151529	2.048490	0.137597	2.485720	0.151529	2.048490
29	0.136204	1.715440	0.228716	0.500117	0.134983	1.715440
30	0.126348	1.665180	0.113577	2.245970	0.126369	1.665180
31	0.100447	2.437110	0.219824	0.780606	0.100447	2.437110
32	0.100045	2.253360	0.100000	1.977600	0.100045	2.190860
33	0.197349	1.171590	0.153101	2.022740	0.197422	1.171600
34	0.211959	0.976432	0.231846	0.913722	0.211876	0.976428

Relay	MFA		GA		FAGA	
	TMS	PS	TMS	PS	TMS	PS
35	0.208108	0.811549	0.201486	1.036120	0.208439	0.811591
36	0.131359	1.518780	0.219030	0.805507	0.131493	1.518790
37	0.169650	1.471440	0.271151	0.510730	0.169830	1.471460
38	0.130993	2.100210	0.141771	2.497920	0.131124	2.100210
39	0.155660	1.647920	0.171159	1.290270	0.155806	1.647930
40	0.222847	0.765282	0.211496	1.103120	0.222801	0.765034
41	0.190438	1.305190	0.214445	1.139900	0.190462	1.305190
42	0.149375	1.205710	0.128640	1.586970	0.149451	1.205730
Fitness	16.0694		17.2657		15.7578	

Table F.11

Comparison of the results for IEEE 15-bus network with NLP formulation.

Algorithm	OF	Net gain		%Improvement
MATLBO [115]	52.5039	MFA	36.4345	69.394
		GA	35.2385	67.116
		FAGA	36.7461	69.987
PSO [114]	26.8093	MFA	10.7399	40.060
		GA	9.5436	35.598
		FAGA	11.0515	41.223
Jaya [4]	23.5579	MFA	7.4885	31.788
		GA	6.2922	26.709
		FAGA	7.8001	33.110
SA [114]	20.4068	MFA	4.3374	21.255
		GA	3.1411	15.392
		FAGA	4.649	22.782
CSA [116]	19.5521	MFA	3.4827	17.812
		GA	2.2864	11.694
		FAGA	3.7943	19.406
DJaya [4]	18.8404	MFA	2.7710	14.708
		GA	1.5747	8.3581
		FAGA	3.0826	16.362

Table F.12*CTI using the NLP formulation for the IEEE 15-bus network*

P/R	MFA	GA	FAGA	P/R	MFA	GA	FAGA
	CTI	CTI	CTI		CTI	CTI	CTI
1	0.27192	0.20250	0.27283	42	0.20000	0.20248	0.20000
2	0.20021	0.20691	0.26908	43	0.41564	0.24179	0.38690
3	0.20684	0.20169	0.27565	44	0.29074	0.24035	0.28809
4	0.38088	0.23080	0.37773	45	0.24635	0.24007	0.27433
5	0.26344	0.22217	0.26634	46	0.74960	0.32804	0.36524
6	0.32155	0.21447	0.30853	47	0.20017	0.27387	0.20000
7	0.27452	0.21430	0.27376	48	0.25592	0.20036	0.32076
8	0.27888	0.21752	0.27701	49	0.24526	0.25135	0.24884
9	0.49556	0.20007	0.36414	50	0.71526	0.25415	0.41813
10	0.26661	0.30249	0.22716	51	0.22361	0.20014	0.23031
11	0.29066	0.24161	0.29234	52	0.36697	0.20016	0.30300
12	0.23533	0.27319	0.22315	53	0.20098	0.20016	0.20000
13	0.28149	0.21233	0.28906	54	0.28378	0.29056	0.28606
14	0.22659	0.24462	0.24636	55	0.21609	0.30199	0.22190
15	0.21415	0.24106	0.23697	56	0.26451	0.20003	0.26351
16	0.21851	0.24428	0.24022	57	0.20000	0.20009	0.20000
17	0.24207	0.20005	0.22522	58	0.33770	0.20008	0.33260
18	0.26418	0.20007	0.22594	59	0.28796	0.20135	0.26646
19	0.20402	0.20727	0.20351	60	0.36205	0.20169	0.30750
20	0.21032	0.26991	0.20686	61	0.23714	0.20025	0.20869
21	0.2449	0.26651	0.23224	62	0.20000	0.20628	0.20243
22	0.20223	0.26957	0.20073	63	0.20000	0.29184	0.20053
23	0.33315	0.20015	0.27222	64	0.39788	0.29994	0.37551
24	0.39635	0.20001	0.38326	65	0.23487	0.20007	0.23525
25	0.20921	0.20000	0.23524	66	0.48308	0.20691	0.47639
26	0.31469	0.27174	0.31481	67	0.30347	0.35741	0.30810
27	0.36723	0.32258	0.35392	68	0.32832	0.38202	0.33295
28	0.44449	0.20353	0.44979	69	0.58471	0.20013	0.28149
29	0.32041	0.20012	0.33182	70	0.64250	0.20030	0.25892
30	0.26942	0.36868	0.26771	71	0.21768	0.20016	0.21860
31	0.30942	0.20487	0.30188	72	0.20075	0.20005	0.20000
32	0.36309	0.40260	0.33389	73	0.20056	0.21167	0.20000
33	0.23709	0.23879	0.26506	74	0.20215	0.20030	0.20026
34	0.26106	0.22444	0.23028	75	0.20123	0.20069	0.20000
35	0.22315	0.23047	0.22402	76	0.20053	0.20000	0.20000
36	0.21667	0.22416	0.21651	77	0.20096	0.20000	0.20060

P/R	MFA	GA	FAGA	P/R	MFA	GA	FAGA
	CTI	CTI	CTI		CTI	CTI	CTI
37	0.24265	0.20376	0.25201	78	0.20000	0.20053	0.20000
38	0.27723	0.20036	0.27739	79	0.20001	0.20014	0.20000
39	0.23020	0.20020	0.24263	80	0.24223	0.22486	0.24347
40	0.36906	0.20420	0.31257	81	0.20045	0.20014	0.20000
41	0.20624	0.20879	0.20751	82	0.20000	0.20014	0.20000

Figure F.6

Total operational time compared to the literature for the proposed algorithms.

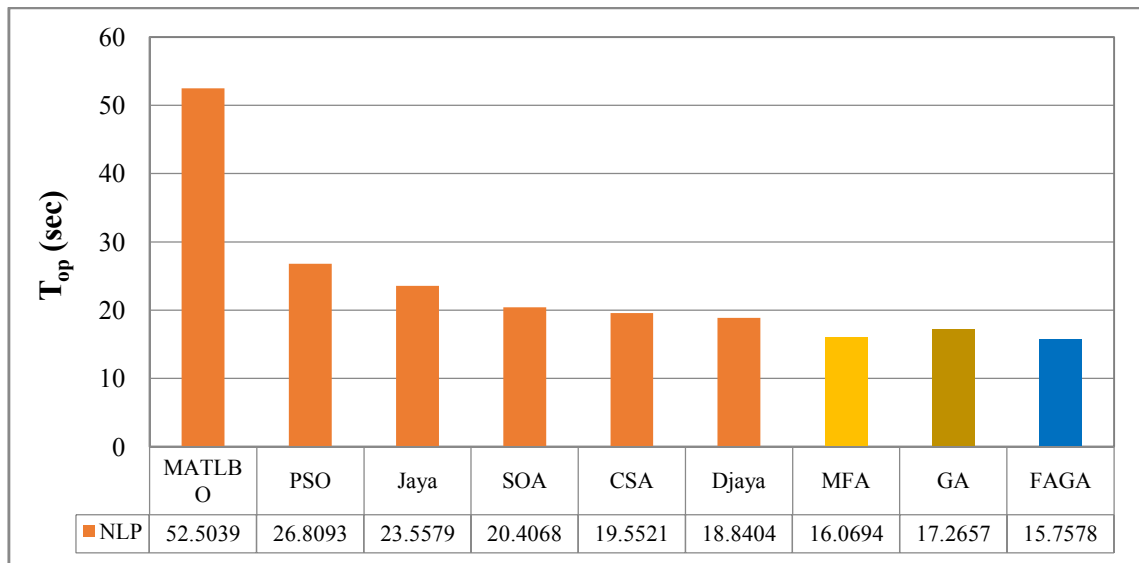
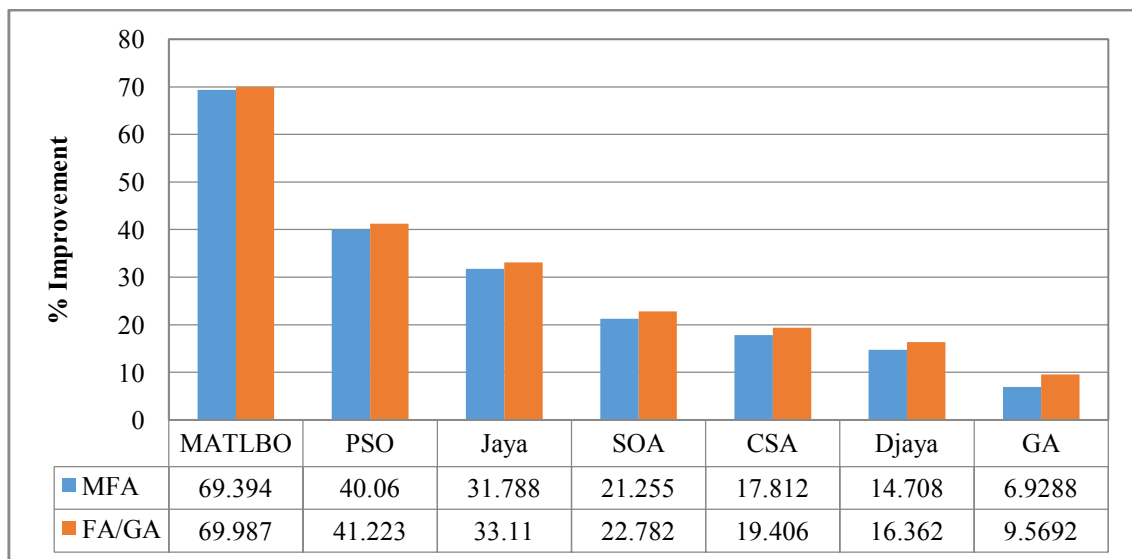


Figure F.7

Percentage improvement of the proposed methods compared to other algorithms in the literature for IEEE 15-bus network in terms of the net relay operational time



Appendix G

Matlab Codes: Test System I

Modified FA with LP formulation

```
function Tpr=cost(x)
CTR=[300/5, 200/5, 200/5, 300/5, 200/5, 400/5];
PS=[5 1.5 5 4 2 2.5];
FCpr=[1978.9, 1525.7, 1683.9, 1815.4, 1499.66, 1766.3];
FCbc=[175, 545, 617.22, 466.17, 384, 145.34];

Tpr=x(1)*0.14/(((FCpr(1)/(PS(1)*CTR(1)))^0.02)-1)+
x(2)*0.14/(((FCpr(2)/(PS(2)*CTR(2)))^0.02)-1)+ ...
x(3)*0.14/(((FCpr(3)/(PS(3)*CTR(3)))^0.02)-1)+
x(4)*0.14/(((FCpr(4)/(PS(4)*CTR(4)))^0.02)-1)+ ...
x(5)*0.14/(((FCpr(5)/(PS(5)*CTR(5)))^0.02)-1)+
x(6)*0.14/(((FCpr(6)/(PS(6)*CTR(6)))^0.02)-1);

T1pr=x(1)*0.14/(((FCpr(1)/(PS(1)*CTR(1)))^0.02)-1);
T2pr=x(2)*0.14/(((FCpr(2)/(PS(2)*CTR(2)))^0.02)-1);
T3pr=x(3)*0.14/(((FCpr(3)/(PS(3)*CTR(3)))^0.02)-1);
T4pr=x(4)*0.14/(((FCpr(4)/(PS(4)*CTR(4)))^0.02)-1);
T5pr=x(5)*0.14/(((FCpr(5)/(PS(5)*CTR(5)))^0.02)-1);
T6pr=x(6)*0.14/(((FCpr(6)/(PS(6)*CTR(6)))^0.02)-1);
TPR=[T1pr T2pr T3pr T4pr T5pr T6pr];

T5bc=x(5)*0.14/(((FCbc(1)/(PS(5)*CTR(5)))^0.02)-1);
T4bc=x(4)*0.14/(((FCbc(2)/(PS(4)*CTR(4)))^0.02)-1);
T1bc=x(1)*0.14/(((FCbc(3)/(PS(1)*CTR(1)))^0.02)-1);
T6bc=x(6)*0.14/(((FCbc(4)/(PS(6)*CTR(6)))^0.02)-1);
T3bc=x(3)*0.14/(((FCbc(5)/(PS(3)*CTR(3)))^0.02)-1);
T2bc=x(2)*0.14/(((FCbc(6)/(PS(2)*CTR(2)))^0.02)-1);
TBC= [T5bc T4bc T1bc T6bc T3bc T2bc];

for j=1:length(TBC)
Violation(j)=TBC(j)-TPR(j);
end

disp(['T1(pr) = ', num2str(TPR(1))])
disp(['T2(pr) = ', num2str(TPR(2))])
disp(['T3(pr) = ', num2str(TPR(3))])
disp(['T4(pr) = ', num2str(TPR(4))])
disp(['T5(pr) = ', num2str(TPR(5))])
disp(['T6(pr) = ', num2str(TPR(6))])

disp(['T1(BC) = ', num2str(TBC(1))])
disp(['T2(BC) = ', num2str(TBC(2))])
disp(['T3(BC) = ', num2str(TBC(3))])
disp(['T4(BC) = ', num2str(TBC(4))])
disp(['T5(BC) = ', num2str(TBC(5))])
disp(['T6(BC) = ', num2str(TBC(6))])
```

```

disp(['T5(bc) - T1(pr) = ', num2str(Violation(1))])
disp(['T4(bc) - T2(pr) = ', num2str(Violation(2))])
disp(['T1(bc) - T3(pr) = ', num2str(Violation(3))])
disp(['T6(bc) - T4(pr) = ', num2str(Violation(4))])
disp(['T3(bc) - T5(pr) = ', num2str(Violation(5))])
disp(['T2(bc) - T6(pr) = ', num2str(Violation(6))])
end

function [g,geq]=constraint(x)
CTR=[300/5, 200/5, 200/5, 300/5, 200/5, 400/5];
FCpr=[1978.9, 1525.7, 1683.9, 1815.4, 1499.66, 1766.3];
PS=[5 1.5 5 4 2 2.5];
FCbc=[175, 545, 617.22, 466.17, 384, 145.34];
CTI=0.2;
g=[
    x(1)^0.14/(((FCpr(1)/(PS(1)^CTR(1)))^0.02)-1)+CTI-
    x(5)^0.14/(((FCbc(1)/(PS(5)^CTR(5)))^0.02)-1);
    x(2)^0.14/(((FCpr(2)/(PS(2)^CTR(2)))^0.02)-1)+CTI-
    x(4)^0.14/(((FCbc(2)/(PS(4)^CTR(4)))^0.02)-1);
    x(3)^0.14/(((FCpr(3)/(PS(3)^CTR(3)))^0.02)-1)+CTI-
    x(1)^0.14/(((FCbc(3)/(PS(1)^CTR(1)))^0.02)-1);
    x(4)^0.14/(((FCpr(4)/(PS(4)^CTR(4)))^0.02)-1)+CTI-
    x(6)^0.14/(((FCbc(4)/(PS(6)^CTR(6)))^0.02)-1);
    x(5)^0.14/(((FCpr(5)/(PS(5)^CTR(5)))^0.02)-1)+CTI-
    x(3)^0.14/(((FCbc(5)/(PS(3)^CTR(3)))^0.02)-1);
    x(6)^0.14/(((FCpr(6)/(PS(6)^CTR(6)))^0.02)-1)+CTI-
    x(2)^0.14/(((FCbc(6)/(PS(2)^CTR(2)))^0.02)-1);

    0.1-x(1)^0.14/(((FCpr(1)/(PS(1)^CTR(1)))^0.02)-1);
    0.1-x(2)^0.14/(((FCpr(2)/(PS(2)^CTR(2)))^0.02)-1);
    0.1-x(3)^0.14/(((FCpr(3)/(PS(3)^CTR(3)))^0.02)-1);
    0.1-x(4)^0.14/(((FCpr(4)/(PS(4)^CTR(4)))^0.02)-1);
    0.1-x(5)^0.14/(((FCpr(5)/(PS(5)^CTR(5)))^0.02)-1);
    0.1-x(6)^0.14/(((FCpr(6)/(PS(6)^CTR(6)))^0.02)-1);

    x(1)^0.14/(((FCpr(1)/(PS(1)^CTR(1)))^0.02)-1)-0.5;
    x(2)^0.14/(((FCpr(2)/(PS(2)^CTR(2)))^0.02)-1)-0.5;
    x(3)^0.14/(((FCpr(3)/(PS(3)^CTR(3)))^0.02)-1)-0.5;
    x(4)^0.14/(((FCpr(4)/(PS(4)^CTR(4)))^0.02)-1)-0.5;
    x(5)^0.14/(((FCpr(5)/(PS(5)^CTR(5)))^0.02)-1)-0.5;
    x(6)^0.14/(((FCpr(6)/(PS(6)^CTR(6)))^0.02)-1)-0.5;
];
geq=[];

function Fireflyx
clc;
clear;
format long;
rng('default')
Lb=[0.1 0.1 0.1 0.1 0.1 0.1];
Ub=[1.1 1.1 1.1 1.1 1.1 1.1];
u0=(Lb+Ub)/2;

```

```

para=[25 2000 0.8 0.2 1];
[u,fval,NumEval]=ffa_mincon(@cost,@constraint,u0,Lb,Ub,para);
disp('Directional overcurrent relay coordination problem for six relay
system...');
fprintf('\n\nTMS 1 = %g\nTMS 2 = %g\nTMS 3 = %g\nTMS 4 = %g\nTMS 5 =
%g\nTMS 6 = %g\n'...
,u(1),u(2),u(3),u(4),u(5),u(6))
fprintf('\nBest overall min. operating time :%g\n\n',fval)
total_number_of_function_evaluations=NumEval

function [nbest,fbest,NumEval]...
=ffa_mincon(fhandle,nonhandle,u0, Lb, Ub, para)
if nargin<6,
para=[25 2000 0.8 0.2 1];
end

if nargin<5,
Ub=[];
end

if nargin<4,
Lb=[];
end

if nargin<3,
disp('Usage: FA_mincon(@cost, @constraint,u0,Lb,Ub,para)');
end

n=para(1);
MaxGeneration=para(2);
alpha=para(3);
betamin=para(4);
gamma=para(5);

NumEval=n*MaxGeneration;

if length(Lb) ~=length(Ub),
disp('Simple bounds/limits are improper!');
return
end

d=length(u0);
zn=ones(n,1)*10^100;
[ns,Lightn]=init_ffa(n,d,Lb,Ub,u0);
for k=1:MaxGeneration,
alpha=alpha_new(alpha,MaxGeneration);

for i=1:n,
zn(i)=Fun(fhandle,nonhandle,ns(i,:));
Lightn(i)=zn(i);
end

[Lightn,Index]=sort(zn);
ns_tmp=ns;
for i=1:n,
ns(i,:)=ns_tmp(Index(i),:);
end

nso=ns;
Lighto=Lightn;

```

```

nbest=ns(1,:);
Lightbest=Lightn(1);
fbest=Lightbest;
% Move all fireflies to the better locations
[ns]=ffa_move(n,d,ns,Lightn,nso,Lighto,nbest,...
    Lightbest,alpha,betamin,gamma,Lb,Ub);
    plot(k,fbest,'b-x')% this plots the graph
        xlabel('Iteration number');
        ylabel('fitness value');
        hold on
    grid on
end

    plot(k,fbest,'b-x')% this plots the graph
        xlabel('Iteration number');
        ylabel('fitness value');
        hold on
    grid on
function alpha=alpha_new(alpha,NGen)
delta=1-(10^(-4)/0.9)^(1/NGen);
alpha=(1-delta)*alpha;
end
function [ns]=ffa_move(n,d,ns,Lightn,nso,Lighto,...
    nbest,Lightbest,alpha,betamin,gamma,Lb,Ub)
scale=abs(Ub-Lb);

for i=1:n,
    for j=1:n,
        r=sqrt(sum((ns(i,:)-ns(j,:)).^2));
        % Update moves
    if Lightn(i)>Lighto(j),
        beta0=1;
        beta=(beta0-betamin)*exp(-gamma*r.^2)+betamin;
        tmpf=alpha.*(rand(1,d)-0.5).*scale;
        ns(i,:)=ns(i,).* (1-beta)+nso(j,).*beta+tmpf;
        end
    end
end
[ns]=findlimits(n,ns,Lb,Ub);
function [ns]=findlimits(n,ns,Lb,Ub)
for i=1:n,
    ns_tmp=ns(i,:);
    I=ns_tmp<Lb;
    ns_tmp(I)=Lb(I);
    J=ns_tmp>Ub;
    ns_tmp(J)=Ub(J);
    ns(i,:)=ns_tmp;
end
function z=Fun(fhandle,nonhandle,u)
z=fhandle(u);
z=z+getnonlinear(nonhandle,u);
end
function H=getH(g)
if g<=0,
    H=0;
else
    H=1;
end
function H=geteqH(g)
if g==0,
    H=0;

```

```

else
    H=1;
end
function Z=getnonlinear(nonhandle,u)
Z=0;
lam=40; lameq=40;
[g,geq]=nonhandle(u);
for k=1:length(g),
    Z=Z+ lam*g(k)^2*getH(g(k));
end
for k=1:length(geq),
    Z=Z+lameq*geq(k)^2*geteqH(geq(k));
end
function [ns,Lightn]=init_ffa(n,d,Lb,Ub,u0)
if length(Lb)>0,
    for i=1:n,
        ns(i,:)=Lb+(Ub-Lb).*rand(1,d);
    end
else
    for i=1:n,
        ns(i,:)=u0+randn(1,d);
    end
end
Lightn=ones(n,1)*10^100;

```

GA with LP formulation

```
rng default % For reproducibility

ObjFcn= @cost;
nvars=6;
lb=[0.1 0.1 0.1 0.1 0.1 0.1]; %Lower bound of the decision
variables
ub=[1.1 1.1 1.1 1.1 1.1 1.1]; %Upper bound of the
decision variables
ConsFcn= @constraint;
options =
optimoptions('ga','PlotFcn',{@gaplotbestf,@gaplotchange},'PopulationSi
ze',200,...
'Display','iter','PenaltyFactor',10000,'MaxStallGenerations',200,'MaxG
enerations',400,'ConstraintTolerance',1.0000e-06);
[x,fval,exitFlag,output,population,scores] =
ga(ObjFcn,nvars,[],[],[],[],lb,ub,ConsFcn,options);
fprintf('The number of generations is: %d\n', output.generations);
fprintf('The number of function evaluations is: %d\n',
output.funccount);
fprintf('The best function value found is: %g\n', fval);
fprintf('\n\nTMS 1 = %g\nTMS 2 = %g\nTMS 3 = %g\nTMS 4 = %g\nTMS 5 =
%g\nTMS 6 = %g\nPS 1 = %g\nPS 2 = %g\nPS 3 = %g\nPS 4 = %g\nPS 5 =
%g\nPS 6 = %g\n'...
,x(1),x(2),x(3),x(4),x(5),x(6))
function state = gaplotchange(options, state, flag)

if(strcmp(flag,'init')) % Set up the plot
    xlim([1,options.MaxGenerations]);
    axx = gca;
    axx.YScale = 'log';
    hold on;

        xlabel Generation
        title('Log Absolute Change in Best Fitness Value')
    end

    best = min(state.Score);
    if state.Generation == 0
        last_best = best;
    else
        change = last_best - best;
        last_best = best;
        if change > 0
            plot(state.Generation,change,'xr');
        end
    end
end
```

FAGA with LP formulation

Only the main m.file is modified

```
rng default % For reproducibility

ObjFcn= @cost;
nvars=6;
lb=[0.1 0.1 0.1 0.1 0.1 0.1];
ub=[1.1 1.1 1.1 1.1 1.1 1.1];
ConsFcn= @constraint;

options =
optimoptions('ga','PlotFcn',{@gaplotbestf,@gaplotchange},'PopulationSi
ze',200,...
'Display',
'iter','PenaltyFactor',100,'MaxStallGenerations',200,'MaxGenerations',
400,...
'InitialPopulation',[0.1 0.1 0.1 0.1 0.1
0.1],'ConstraintTolerance',1.0000e-06);
[x,fval,exitFlag,output,population,scores] =
ga(ObjFcn,nvars,[],[],[],[],lb,ub,ConsFcn,options);
fprintf('The number of generations is: %d\n', output.generations);
fprintf('The number of function evaluations is: %d\n',
output.funccount);
fprintf('The best function value found is: %g\n', fval);
fprintf('\n\nTMS 1 = %g\nTMS 2 = %g\nTMS 3 = %g\nTMS 4 = %g\nTMS 5 =
%g\nTMS 6 = %g\nPS 1 = %g\nPS 2 = %g\nPS 3 = %g\nPS 4 = %g\nPS 5 =
%g\nPS 6 = %g\n'...
,x(1),x(2),x(3),x(4),x(5),x(6))
```



جامعة النجاح الوطنية
كلية الدراسات العليا

خوارزمية اليراع-الجينية المهجنة من أجل التنسيق الأمثل لمرحلات زيادة التيار

إعداد

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إشراف

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قدمت هذه الرسالة استكمالاً لمتطلبات الحصول على درجة الماجستير في هندسة القوى الكهربائية، من كلية الدراسات العليا، في جامعة النجاح الوطنية، نابلس-فلسطين.

2023

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الملخص

الخلفية النظرية: تُستخدم مرحلات زيادة التيار الزائدة الاتجاهية لحماية نظام الطاقة لضمان التشغيل الآمن والموثوق والفعال. إن تنسيق المرحلات الزائدة التيار الاتجاهية هو مشكلة أمثل غير خطية ومقيدة للغاية. الهدف الأساسي من التحسين هو تقليل تجميع أوقات التشغيل للمرحلات الأولية، من خلال تحديد القيم المثلى لمتغيرات القرار كإعداد (TMS) Time multiplier setting وإعداد plug (PS) setting.

هدف البحث: الهدف الرئيسي لهذه الدراسة هو تطوير خوارزمية تحسين هجينة تتكون من خوارزمية اليراع المعدلة والخوارزمية الجينية لإيجاد حلول أفضل.

منهجية البحث: أولاً، عدلت هذه الدراسة خوارزمية اليراع الأصلي للحصول على حل أمثل من خلال تحديث سطوع اليراع وتجنب المسافة بين اليراعات الفردية من أن تكون بعيدة جداً. بالإضافة إلى ذلك، تم التحكم في الحركات العشوائية لإنتاج معدل تقارب مرتفع. ثانياً، تم حل مشكلة التحسين باستخدام الخوارزمية الجينية القياسية. أخيراً، تم استخدام الحل الذي تم الحصول عليه من خوارزمية اليراع المعدلة كحلول أولية للخوارزمية الجينية القياسية. تم اختبار خوارزمية اليراع المعدلة والخوارزمية الجينية والخوارزمية المهجنة المقترحة على أنظمة IEEE 3-bus و bus-8 و bus-9 و bus-15.

نتائج البحث: تشير النتائج إلى فعالية وتفوق الخوارزميات المقترحة في تقليل وقت التشغيل الإجمالي للمرحلات الأولية مقارنة بالطرق الأخرى المذكورة في الأدبيات لتنسيق مرحلات زيادة التيار.

استنتاجات البحث: بالمقارنة مع خوارزمية اليراع المعدلة والخوارزمية الجينية القياسية، قلت

الخوارزمية الهجينة المقترحة من وقت فاصل التنسيق بين أزواج التتابع الأولية والنسخ الاحتياطي.

الكلمات المفتاحية: الخوارزميات الهجينة ، خوارزمية اليراع ، الخوارزمية الجينية.