

A Survey of Optimal Power Flow Analysis of Longitudinal Power System

ABSTRACT

This paper presented a survey of publications on Optimal Power Flow (OPF) analysis of longitudinal power system with emphasis on the Nigerian power grid. It explained the nitty-gritty of optimal power flow analysis. The study revealed that application of heuristic optimization techniques to optimal power flow analysis have obviated the drawbacks of the previously used traditional optimization techniques with better solution quality, convergence time and flexibility. Although, the heuristics techniques were not flawless but well off to that of traditional techniques, a careful hybridization of both techniques were seeming best off. This publication will be found handy for power system operators as well as researchers in an attempt to enhance the operations of the electrical power system.

Keywords: Longitudinal power system; Nigerian power system; optimal power flow; power system optimization.

1. INTRODUCTION

Now a days, electrical power is an indispensable product and continues to grow in importance due to its flexibility and other advantages over the other forms of energy. In a deregulated electricity of developing nations, with longitudinal structure of power grid; radial in operation with several long transmission lines where generation centers are sparse and remote from load centers, like the case of Chilean, Nigerian, Taiwan etc. power system. The continuous increase in power demand is fast outpacing the power system infrastructures, which comprises of the generation, transmission and distribution system as well as other ancillary power system equipment; as such, operational violations, complexities and vagaries become evident on such system.

Technically, construction of a new power infrastructure is not only insufficient as a remedy of combating the menace but also militated against by problem right-of-way, environmental or socio-political issue, as well as energy resources management [1]. More so, construction of a new power infrastructure is rather a futuristic approach; cannot meet the present energy need. Enhancement or optimum utilization of the existing power system become a viable resort. However, the performance indices of the system in terms of security, reliability, stability and economical operation have to be in line with the enhancement. This is the concept of Optimal Power Flow (OPF), the subject of this article.

Optimal Power Flow (OPF) is an optimization process applied to power system, it has been widely used in power system operations, analysis, scheduling, planning and energy management over the years and it is still becoming more relevant because of its several capabilities to deal with various situations of modern power system operations [2]. The

optimization process is applicable to power system analysis based on the possibility of modeling power system parameters in terms of variables, constraints and objective function. In power system parlance, OPF is the process of obtaining the optimal setting of the control or decision variables within the electrical power network by optimizing (minimizing or maximizing) objective function of interest without violating the power flow constraints as well as the equipment operating limits while maintaining acceptable system performance in terms of generator capability limits, line flows and output of the compensating devices [3].

Like the conventional (non-optimal) power flow, OPF is also useful for real-time control, operational planning, scheduling, modern Energy Management Systems (EMSs) and also support deregulation transactions of electrical power system. Though the load flow is bereft of yielding the most economic, secured and optimum power system operation but in most cases, it serves as precursor for OPF. While the economic dispatch, which is a particular case of OPF ignores or sometimes, partly up-keep the security of the system but the OPF has the capability to determine the holistic optimal power system operation [1].

OPF, also helps in determine the marginal cost data which in turn aids the pricing aspect of power system operation. It also furnishes the dispatchers or power system operators with possible tradeoffs between different objectives and also enlightens on which of the objectives will pay off, without violation of constraints.

A typical OPF problem is formulated in cognizance to the power network model, objective function, operating limits, and the intended solution technique. Due to its versatility, different formulations represent each of the possible case of OPF and the quality of the result relies on accurate model formulation as well as the solution techniques. Among the OPF formulations are:

1. Optimal Scheduling: ensuring optimal generation with a saving (proper allocation) of the energy resources (fuel) invariably a saving in operating cost (fuel cost in thermal plants), such is a case of OPF called; classical economic dispatch [3].
2. Security - Constrained Optimal Power Flow (SCOPF): Curtailing outages and contingencies while ensuring optimum system operation. Also is the Security - Constrained with Voltage Stability (SCOPF-VS) another particular case of OPF [4]
3. The scope of OPF can also be extended to accommodate Flexible Alternating Current Transmission System (FACTS) devices as well as renewable energy generation [1]

2. METHODOLOGY

The methodology of OPF is synonymous to that of a typical optimization process, with the appropriate problem formulation in terms of objective function, variables, and constraints such that it captures the desire of the system operators; then, the deployment of solution methodologies or optimization techniques.

2.1 Optimal power flow formulation

Several OPF formulations have been reported in the literature to address several instances of the problem. In recent times, the restructuring and developments in power system are causing increment in electric power system complexity. Also, the advent of Independent Power Producers (IPPs) and the prospect of integrating distributed and renewable generation in the grid, further expand the scope of OPF. Thus, various formulations abound, which goes by many names depending on choice of objective function and the constraints. Regardless of the name, any power systems optimization problem that includes a set of power flow equations in the constraints may be classified as a form of OPF [5].

In spite of the changes in the traditional power system operation and control due to increase in power system size and complexities, with the introduction of modern devices and renewable energy to alleviate the bottleneck and maximize system utility, the general structure of OPF formulation still maintains the classical format. Expressed as follows ([6]; [7]):

$$\text{Optimize } F(x, u) \quad (1)$$

Subject to:

$$G(x, u) = 0 \quad (2)$$

$$H_{\min}(x, u) \leq H(x, u) \leq H_{\max}(x, u) \quad (3)$$

Where: (x, u) , vector of controllable or independent variables and dependent or state variables of the system respectively; $F(x, u)$, the objective function: whose selection is based on the operating philosophy of the system operator; $G(x, u)$ and $H(x, u)$, are vector representing the system equality and inequality constraints respectively.

2.1.1 Variables

Optimal power flow analysis requires certain power system variables to be controlled or modified in order to optimize the operation of electrical power system as well as variables to reflect the effect of the optimization process. The variables are thus classified as the control (decision or independent) variables and the state or dependent variables, accordingly. Generally, the state variables are said to be continuous in nature while the control variables may be continuous or discrete; as in the case of switched devices or lines, they are binary [8] ; [9]. In [9] and [10], the examples of the variables are enumerated as follows:

The control variables which includes:

1. Active power at the generator buses except for the slack bus
2. Voltage magnitudes at the generator buses
3. Position of the transformer taps
4. Position of the phase shifter (quad booster) taps
5. Status of the switched capacitors and reactors
6. Control of power electronics (HVDC, FACTS)
7. Amount of load disconnected, etc.

While that of the state variables includes:

1. Voltage magnitudes at load buses
2. Voltage phase angle at all buses
3. Active power output of the slack bus only.
4. Reactive power of all generator buses.
5. Line flows, etc.

139 **2.1.2 Constraints**

140 Constraints, are generally regard as an integral part of a practical optimization problem and
 141 are sometimes use as the key for the classification of OPF problems, for instance, the
 142 security-constrained OPF, economic dispatch, security-constrained with voltage stability etc.
 143 Besides, the system variables has to be within a permissible range (constrained), which
 144 should not be violated except causing damage to electrical power system equipment or
 145 resulting into a mal-operation. The constraints are generally categorized as equality and
 146 inequality constraints. More so, some of these constraints are easily handled except for the
 147 functional dependent ones of the inequality constraints, which employ the method of penalty
 148 functions, lagrange multiplier or others, in handling such functional constraints.
 149 In OPF, the equality constraints are basically the power flow network equations, which can
 150 either be the steady state power flow or the contingency state power flow, either of which is
 151 non-linear though their level of complexity differs widely [10].

152 On the other hand, is the inequality constraints that specified the limits on the equipment of
 153 electrical power system as well as the limits needed to guarantee system security [11]. The
 154 inequality constraints are subdivided as follows as:

155 a) Control variables limits, which includes:

- 156 • Generator real power

$$157 \quad P_{G_i}^{min} \leq P_{G_i} \leq P_{G_i}^{max} \quad (4)$$

- 158 • Generator bus voltage

$$159 \quad V_{G_i}^{min} \leq V_{G_i} \leq V_{G_i}^{max} \quad (5)$$

- 160 • Volt – Ampere Reactive (VAR) power

$$161 \quad Q_{G_i}^{min} \leq Q_{G_i} \leq Q_{G_i}^{max} \quad (6)$$

- 162 • Transformer tap position

$$163 \quad T_i^{min} \leq T_i \leq T_i^{max} \quad (7)$$

164 b) State variables limits :

- 165 • Voltage magnitude of load bus

$$166 \quad V_{L_i}^{min} \leq V_{L_i} \leq V_{L_i}^{max} \quad (8)$$

- 167 • Line flow limits

$$168 \quad S_{l_i} \leq S_{l_i}^{max} \quad (9)$$

169 Additional inequality constraints include, reactive power of generator, prohibited zones of the
 170 generating units, rotor angle stability, limit on transient voltage electromagnetic field levels,
 171 etc [9].

172 **2.1.3 Objective function**

173 Practical OPF problems have several objective functions to reflect the different possible
 174 operations of power system, the objective function is multi-faceted as no single objective
 175 function fit into all the emerging scenarios of OPF. The selection and consideration of the
 176 objective functions depend on the operating philosophy of the power system operator [1].
 177 The most commonly used objective function is the minimization of generation costs with and
 178 without consideration of system losses, since the issue of cost used to take precedence in
 179 power system operations. This is the classical case of OPF, called economic dispatch.
 180 Classical economic dispatch controls only the generation units to dispatch while OPF
 181 controls all power flow within the electrical power system [3].

182 It is to be noted that the cost, is the operating cost and not the total capital outlay of the
 183 power system, which is known in thermal and nuclear stations as the fuel cost. But for the
 184 case of hydro plants, where water is seeming free, there exist techniques for hydro scheme
 185 coordination as well as for incorporating pumped-storage hydro units into OPF formulation
 186 [12]. The fuel cost is usually equated to the operating cost or generating cost with the
 187 realization that other variable cost like: labour cost, maintenance cost, and fuel
 188 transportation cost, etc which are difficult to express directly as a function of the output of the
 189 thermal generator unit, are expressed as a fixed portion of the fuel cost [3],[10].
 190 Emphatically, fixed costs, such as the capital cost of installing equipment, are not included,
 191 only those costs that are a function of unit power output are considered in the OPF
 192 formulation.

193 Besides minimization of generation costs, other objectives function are the minimization of
 194 system losses, maximization of power quality often through minimization from a given
 195 schedule of a control variable (such as voltage deviation) maximization of voltage stability,
 196 load curtailment and emission of certain gases etc. Sometimes, in a multi-objective
 197 problems, the objective functions are augmented with respect to each other, where
 198 importance is attached to a particular objective using the method of weighted sum, as seen
 199 in [11].

200

201

202 **2.2 Optimization Techniques**

203 The wide varieties of OPF formulations and the nature of the OPF problems, as previously
 204 discussed, brought about wide varieties of optimization techniques. In the past decades,
 205 OPF algorithms or techniques were designed in line with simplified assumptions of the
 206 problem formulation. Such techniques were termed as traditional or deterministic or better
 207 still mathematical optimization technique. The technique have been applied to OPF
 208 problems and were used in power industry. However, they suffer some shortcomings, mainly
 209 as a result of the simplification made in the formulation of the problem, without which the
 210 technique might not converge, making the traditional have minimal applications [13].

211 However, the new dawn in optimization computations are the heuristics or non -
 212 deterministic optimization techniques, which differ conceptually from the traditional
 213 techniques, and are found to outweigh the shortcoming of the previously used traditional
 214 methods [13]. It is however noted that, there are still no known universal or almighty
 215 techniques that fits exactly for all varieties of the OPF problems, although some algorithms
 216 might perform excellently well than others in certain OPF model. A common theorem in this
 217 aspect of study is the no free lunch theorem; which states, no algorithms in all aspect is
 218 better than the other except in certain aspect where one may outweighs the others [14].

219 The heuristic techniques are however, reported with many theoretical advantages and
 220 practically outperform the classical techniques. Though, they are computational intensive,

are not inherently applicable to constrained problems and the development of their software package is burdensome relative to the traditional or deterministic techniques. Some of the performance metrics for discerning between the algorithms as used in OPF researches, were identified by [15] [16] as follows: computational speed, reliability, robustness, versatility or flexibility, scalability, solution quality and time of convergence. Evidently, it is very difficult for a single algorithm to possess all these traits. However, [16] stated that solution quality, robustness, time of convergence, reliability, and scalability should be considered in choosing and rating an OPF optimization techniques.

2.2.1 Traditional or deterministic optimization techniques

These techniques are principally based on the criterion of local search for the optimal solution through the feasible region of the solution, they use single path search methods and follow deterministic transition rules. Also known as derivative-based optimization methods, as its employed gradient and Hessian operators [5]. In these techniques, the criterion for optimality is based on Karush-Kuhn-Tucker (KKT) criterion which is a necessary but not sufficient criterion for optimality. These techniques have been widely used in solving optimization problems and OPF problems in particular, the reason being their efficiency, simplicity, solid mathematical foundation and readily available software tools for their implementation [2].

Common among these techniques as applied to OPF are: Newton method, simplex method, Lambda-iterative techniques, Gradient-based techniques, Linear and non-linear programming, Quadratic and dynamic programming and interior point method etc [13] However, in spite of their application to OPF problem, the techniques suffer from the following drawbacks which make them to have minimal applications in solving practical OPF problems as reported in [13], [2], [5] :

- Local solvers; cannot guarantee global optimality except for the case of convex problem; because the Karush-Kuhn-Tucker (KKT) conditions are not sufficient for a global optimum.
- Uses approximate assumptions (such as linearity, differentiability, convexity etc.) which are unlike practical OPF problem.
- Sensitive to objective function and the initial estimate or starting points.
- The majority are meant to handle continuous variables, whereas the practical power system consist of binary or integer and discrete variables.

2.2.2 Heuristic or Non – deterministic optimization techniques

These techniques employed exhaustive or stochastic search with randomness in moving from one solution to the next in the feasible solution region to obtain the optimal solution, this majorly helps in circumventing being trapped in local minima. Thus, they are versatile in handling various OPF format even with non-convexities and complicating constraints that are typical of practical OPF. These techniques are evolved to overcome the drawbacks of conventional techniques .Most of these techniques imitate certain natural phenomenon in their search for an optimal solution, which brought about their various categories [17].

Thus, each one of them have peculiar philosophy, but their common denominator is the systematic exploration of the search space for the solution. For instance, the philosophy of species evolution, is employed in the case of Genetic Algorithms and Evolutionary

programming; the neural system philosophy, as the case of Artificial Neural Networks; the thermal annealing of heated solids as the case of Simulated Annealing; and the philosophy of social behaviors and foraging of living things, as in the case of Ant Colony Optimization, Particle Swarm Optimization, Fire-fly Algorithm, Teaching – Learning - Based optimization and so on, ([9]). These techniques are called many names, popular among are: heuristic, meta-heuristic, artificial intelligent, modern optimization technique etc.

It is to be emphasized that the application of these techniques requires selection of some algorithm specific parameters for their proper performance. Also, these techniques are inherently designed to handle unconstrained problems but with incorporation of penalty terms except when using the direct method, the constrained problems are easily handled. Most of these techniques are sensitive to the choice of parameter and penalty terms, such that the improper selection either increases the computational effort or yields the local optimal solution, also, a change in the parameters change their effectiveness [18]. The difficulty in the selection of algorithm parameters, and their lack of solid mathematical foundation with their complicated programming, are the major drawbacks of these techniques [9]. However, advancement in research are bringing to limelight some techniques that requires selection of fewer algorithm specific parameters, such techniques is the Teaching - Learning-Based Optimization (TLBO), Jaya algorithm among others [18].

2.2.3 Hybrid optimization techniques

Optimization techniques continues to grow in importance due to its wide range of application and thus becomes an active area of research. In spite of the landmark success of both deterministic and non-deterministic optimization techniques generally and in the aspect of OPF in particular, there are still some inherent shortcomings of each of these techniques. This brought about the quest of having a hybrid optimization algorithm techniques that carefully combine two or more techniques into one, such that the advantages of each can be used to strengthen the others or to surmount its disadvantages. Significant improvements such as computation time, convergence properties, and solution quality or parameter robustness over each of the individual methods are achievable [17]. The hybridization could be:

- i. Deterministic method combined : Instances of this as applicable to OPF are the Sequential Quadratic Programming (SQP) combined with quasi – Newton [19], Interior Point Method (IPMS) combined with Benders Decomposition [4], Interior Point Method (IPMS) combined with lagrangian Relaxation and Newton's method [20] etc.
- ii. Deterministic and non-deterministic combined : Examples of this as applicable to various form of OPF are Newton's method combined with Simulated Annealing (SA) [21] , combined chaotic Particle Swarm Optimization (PSO) with linear Interior Point Method (IPM) [22] Newton's method combined with Particle Swarm Optimization (PSO) [23] etc.
- iii. Non - deterministic Methods Combined: Differential Evolution (DE) combined with other meta-heuristics [24]; Particle Swarm Optimization (PSO) combined with Simulated Annealing (SA) [25]; combined Differential Evolution (DE) and Simulated Annealing (SA) [26], etc.

308 3. PREVIOUS STUDIES

309 Application of the variants of Genetic Algorithm (GA) to the problem of economic dispatch of
 310 generation was the focus of [27]. In this study, both the Conventional Genetic Algorithm
 311 (CGA) and Micro Genetic Algorithm (μ GA) were applied to minimize the generation cost, the
 312 power balance constraints was the equality constraints considered. The authors reported
 313 that the major drawback of the conventional genetic algorithms approach was that it can be
 314 time consuming. Micro genetic algorithms approach was proposed as a better time efficient
 315 alternative. The effectiveness of both techniques to solving economic dispatch problem was
 316 initially verified on a 6-bus IEEE test system and then on the 31-bus Nigerian grid systems. It
 317 was concluded that the results obtained from both approaches were satisfactory. However,
 318 from the view point of economic and computational time, micro genetic algorithms performed
 319 better than the conventional genetic algorithms and overly better to that of Newton-
 320 approach, on both the 6-bus IEEE test system and then on the 31-bus longitudinal Nigerian
 321 grid systems.

322
 323 In [28], voltage profile correction and power loss minimization through reactive power control
 324 using Differential Evolution (DE) and Particle Swarm Optimization (PSO) technique was
 325 investigated. The feasibility, effectiveness and generic nature of both Differential Evolution
 326 (DE) and Particle Swarm Optimization (PSO) approaches were demonstrated on the 31- bus
 327 Nigerian grid system and the 39- bus New England power system with MATLAB application
 328 package. The simulation results revealed that both approaches were able to remove the
 329 voltage limit violations, but Particle Swarm Optimization (PSO) procured in some instances
 330 slightly higher power loss reduction as compared with Differential Evolution (DE). However,
 331 Differential Evolution (DE) was observed to require a considerably lower number of function
 332 evaluations while compared with Particle Swarm Optimization (PSO), if this observation
 333 could be substantiated by further investigation on the longitudinal Nigerian grid system, the
 334 DE approach will be more viable for potential real time application in control centre where
 335 the computation time is very relevant.

336
 337 More so, the Elitist Non-dominated Sorting Genetic Algorithm II (NSGA-II), was applied to
 338 solve the multi-objective optimal dispatch of the Nigerian 24-bus hydrothermal power system
 339 with fuel cost and transmission loss as the objectives, with the consideration of power
 340 balance [29]. The authors established that the solutions obtained by elitist non-dominated
 341 sorting genetic algorithm (NSGA-II) converged better over both conventional genetic
 342 algorithms and micro genetic algorithms approaches used in earlier studies on the Nigerian
 343 power grid. It was observed that as the modification of the algorithm increases, their
 344 performance get better.

345
 346 The optimal dispatch of generation with the minimization of system total generation cost,
 347 subjected to power balance constraint equation using Newton Raphson iterative techniques
 348 was examined in [30]. This iterative techniques was applied to Nigerian grid system to
 349 determine the total cost of generation as well as the total system transmission losses. While
 350 the simulation was done with a MATLAB based program. At certain buses where voltage
 351 drops were noticed, Load Tap-changing Transformer (LTCT) were introduced to adjust the
 352 voltage magnitude, which furthered reduced the losses on the system. It was observed that
 353 the optimality in this study was determined based on Karush-Kuhn-Tucker (KKT) criterion;
 354 being a traditional technique, the result obtained trailed that of previous works ([27],[28]and
 355 [28]) in solution quality.

356
 357 Constrained Elitist Genetic Algorithm (CEGA) was adopted in [31] to solve the economic

load dispatch problem of the 31-bus Nigerian power system, to reduce both the transmission power loss and total cost of generation, while maintaining an acceptable generation output. Simulation results show that CEGA performed better while comparing with the result of the micro genetic algorithm (μ GA) and a Conventional Genetic Algorithm (CGA), previously used with the same data set as reported in [27]. It was observed that the modification of the algorithm brought about a better result for the Nigerian power grid.

The optimal load dispatch in the South / South Zone of Nigeria Power System by means of a Particle Swarm optimization and Lambda-iteration techniques was investigated in [32]. The economic load dispatch problem were solved for two different cases, the Sapele plant with three units in generating stations and the Afam plant, with six units in the generating stations. The analysis was simulated on MATLAB software package. The objective was cost minimization with and without consideration of losses. It was reported that PSO gave a better solution in terms fuel cost and losses when compared to the result obtained by lambda-iteration, for the same test case.

4. CONCLUSION

This paper has dissected and presented the nitty-gritty of OPF analysis of a longitudinal power grid with emphasis on the Nigerian power system. From the reviewed works, the heuristic or non – deterministic optimization techniques demonstrated its effectiveness and superiority over the traditional techniques with a better numerical result and computational time unlike the traditional techniques. Although, the programming aspect or the development of software package of the heuristics techniques might be tedious relative to traditional techniques. Noteworthy also, the performance of the non-deterministic techniques get better as their modification and hybridization increases. These are cue for further works. Subsequent works should leverage on the application of non – deterministic and combinatorial (hybrid) optimization techniques to solving OPF problems.

More so, it was evident from the review that bulk of the studies focused on generation cost and transmission losses minimization; a particular case of OPF called economic dispatch. Extension of the scope of OPF to accommodate other operational constraints and objectives with the consideration of FACTS controllers, hydro-plants, distributed generations, are also recommended; if included in the analysis, it will further enhance the performance and operation of the power system.

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