



UNIVERSITI PUTRA MALAYSIA

***HYBRID FIREFLY AND PARTICLE SWARM OPTIMIZATION
ALGORITHM FOR MULTI-OBJECTIVE OPTIMAL POWER FLOW WITH
DISTRIBUTED GENERATION***

ABDULLAH KHAN

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By

ABDULLAH KHAN

**Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia, in
Fulfilment of the Requirements for the Degree of Doctor of Philosophy**

January 2022

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Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfilment of the requirement for the degree of Doctor of Philosophy

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January 2022

**Chairman: Associate Professor Hashim Hizam, PhD
Faculty: Engineering**

An optimal power flow (OPF) solution is an essential approach in electric power system operation. Electric service providers are continually working on power generation planning to improve different factors: electricity demands, deregulation of the markets, increasing utilization of distributed generation (DG), and the increasing role of decision-makers. These factors affect the operation plans, have raised the OPF problems' complexities, and require an unfailing optimization algorithm to solve economic and security concerns in different interconnected power systems.

This thesis proposes and simulates the three novel optimization algorithms to handle DG allocation, different single-objective, and multi-objective OPF problems. A new formulation for the multi-objective optimal power flow (MOOPF) problem and DG unit allocation in the power system is also presented. The suggested approaches have been scrutinized and confirmed based on the IEEE 30-bus and 57-bus test systems.

The requirement of the DG installation in the distributed system is to fulfill the power network operation necessities, generally to improve the total loss rises in the system. The DG units have to be allocated with optimal sizes in the network to reach maximum efficacy.

A new meta-heuristic optimization technique called the Slime Mould Algorithm (SMA) approach has a high convergence rate or a few iterations and superior optimization indices analyzed against other algorithms. It can guarantee to enhance the efficiency of exploitation and exploration, based on a sustained balance between exploitation and exploration, to achieve promising statistical results. Therefore, the SMA method is redesigned for optimal location and sizing based on the total active power loss of the systems. And optimal results for single, two, and three DG allocation cases are obtained.

The simulated results from the proposed DG-based SMA approach are also matched with the calculated solutions of the biogeography-based optimization (BBO) approach. The comparison and graphical analysis showed that the total active power loss, required iterations, percentage of the total loss minimization, and DG installed capacities are relatively improved using the suggested SMA algorithm based on the optimal DG unit sizing and location problem in the power systems.

Secondly, a novel and competent meta-heuristic, population-based Hybrid Firefly Particle Swarm Optimization (HFPSO) algorithm is designed to handle various convex and non-linear, single-objective OPF problems. The HFPSO technique hybridizes the Firefly Optimization (FFO) algorithm and the Particle Swarm Optimization (PSO) method to improve the exploitation and exploration strategies and enhance the convergence rate. Moreover, the achieved results revealed the efficacy of the suggested HFPSO algorithm considering the acceptable convergence rate. The statistical examination demonstrated that the proposed method is a reliable and robust optimization approach to deal with OPF problems. Thus, evaluating the applicability and performance of the HFPSO algorithm, it is evident that the proposed method provides a better tool to solve OPF problems of electric power networks.

Finally, a crowding distance and non-dominated-sorting-based multi-objective hybrid firefly & particle swarm optimization (MOHFPSO) algorithm is designed for MOOPF problems. The proposed algorithm is simulated for simultaneous OPF-based conflicting objectives, respectively. Besides, the approach's acquired optimized results are also compared against the simulated original OPF-based MOPSO method and the optimal values of the present literature work to authenticate its effectiveness. Comparing and analyzing the resultant optimal values indicated the proposed MOHFPSO method's dominance in the optimal solution. Consequently, the proposed algorithm with a non-dominated sorting approach can be efficiently applied for small and large power networks.

Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Doktor Falsafah

ALGORITMA PENGOPTIMUMAN HIBRID KELIP-KELIP DAN PARTIKEL SWARM UNTUK ALIRAN KUASA OPTIMAL PELBAGAI OBJEKTIF DENGAN GENERASI TERAGIH

Oleh

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Penyelesaian aliran daya optimum (OPF) adalah pendekatan penting dalam operasi sistem kuasa elektrik. Penyedia perkhidmatan elektrik terus berusaha untuk merancang penjanaan tenaga untuk menambahbaik pelbagai faktor: permintaan elektrik, deregulasi pasaran, peningkatan pemanfaatan penjanaan agihan (DG), dan peningkatan peranan pembuat keputusan. Faktor-faktor ini mempengaruhi rancangan operasi, telah meningkatkan kerumitan masalah OPF, dan memerlukan algoritma pengoptimuman yang berterusan untuk menyelesaikan masalah ekonomi dan keselamatan dalam sistem kuasa yang saling berkaitan.

Tesis ini mencadangkan dan mensimulasikan tiga algoritma pengoptimuman baharu untuk menangani peruntukan DG, masalah OPF objektif tunggal dan pelbagai objektif yang berbeza. Rumusan baru untuk masalah aliran daya optimum pelbagai objektif (MOOPF) dan peruntukan unit DG dalam sistem kuasa juga ditunjukkan. Pendekatan yang dicadangkan telah diteliti dan disahkan berdasarkan sistem ujian IEEE 30-bus dan IEEE 57-bus.

Keperluan pemasangan DG dalam sistem yang diedarkan adalah untuk memenuhi keperluan operasi jaringan kuasa, secara umum untuk memperbaiki jumlah kerugian yang meningkat dalam sistem. Unit DG harus ditempatkan secara ukuran optimum dalam rangkaian untuk mencapai keberkesanan maksimum.

Teknik pengoptimuman meta-heuristik baru iaitu *Slime Mold Algorithm* (SMA) mempunyai kadar penumpuan tinggi atau beberapa lalaran dan indeks pengoptimuman unggul yang dianalisis terhadap algoritma lain. Ini dapat menjamin untuk meningkatkan kecekapan eksploitasi dan eksplorasi, berdasarkan keseimbangan antara eksploitasi dan eksplorasi, untuk mencapai hasil statistik yang menjanjikan. Oleh itu kaedah SMA direka

bentuk semula untuk lokasi dan ukuran yang optimum berdasarkan kehilangan kuasa aktif keseluruhan sistem. Dan hasil yang optimum untuk kes peruntukan tunggal, dua, dan tiga DG diperoleh. Hasil simulasi dari pendekatan SMA berdasarkan DG yang diusulkan juga dipadankan dengan penyelesaian yang dihitung dari pendekatan pengoptimuman berdasarkan biogeografi (BBO). Perbandingan dan analisis grafik menunjukkan bahawa jumlah kehilangan daya aktif, iterasi yang diperlukan, peratusan pengurangan kerugian total, dan kapasiti pemasangan DG relatif lebih baik menggunakan algoritma SMA yang dicadangkan berdasarkan ukuran unit DG dan masalah lokasi dalam sistem kuasa.

Kedua, algoritma novel dan meta-heuristik, berdasarkan populasi *Hybrid Firefly Particle Swarm Optimization* (HFPSO) berasaskan populasi dirancang untuk menangani pelbagai masalah OPF cembung dan tidak linear. Teknik HFPSO menghidupkan algoritma *Firefly Optimization* (FFO) dan kaedah *Particle Swarm Optimization* (PSO) untuk meningkatkan strategi eksploitasi dan penerokaan dan meningkatkan kadar penumpuan. Lebih-lebih lagi, hasil yang dicapai menunjukkan keberkesanan algoritma HFPSO yang dicadangkan dengan mempertimbangkan kadar penumpuan yang boleh diterima. Pemeriksaan statistik menunjukkan bahawa kaedah yang dicadangkan adalah pendekatan pengoptimuman yang boleh dipercayai dan mantap untuk menangani masalah OPF. Oleh itu, dengan menilai kebolehlaksanaan dan prestasi algoritma HFPSO, terbukti bahawa kaedah yang dicadangkan menyediakan alat yang lebih baik untuk menyelesaikan masalah rangkaian kuasa elektrik OPF.

Akhir sekali, algoritma jarak jauh dan algoritma *multi-objective hybrid firefly dan particle swarm optimization* (MOHFPSO) berasaskan jarak yang tidak didominasi dan disusun dirancang untuk masalah MOOPF. Algoritma yang dicadangkan disimulasikan untuk objektif bertentangan berasaskan OPF serentak. Selain itu, hasil pendekatan yang dioptimumkan juga dibandingkan dengan kaedah MOPSO berasaskan OPF asli yang disimulasikan dan nilai optimum ini berfungsi untuk mengesahkan keberkesanannya. Membandingkan dan menganalisis nilai optimum yang dihasilkan menunjukkan keberkesanan kaedah MOHFPSO yang dicadangkan dalam penyelesaian optimum. Hasilnya, algoritma yang dicadangkan dengan pendekatan penyisihan yang tidak didominasi dapat diterapkan secara efisien untuk rangkaian kuasa kecil dan besar.

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LIST OF ABBREVIATIONS

ABC	artificial bee colony algorithm
a_i, b_i, c_i	fuel cost coefficients of the i -th generator
AFPA	adaptive flower pollination algorithm
BHBO	black-hole-based optimization
DSM	different search method
FFO	fruit fly optimization
FC	total fuel cost as an objective function
F_{ij}	component of complex a matrix
GSA	gravitational search algorithm
G_{ij}, B_{ij}	conductance and susceptance amid the node i and j
HFPSO	hybrid firefly & particle swarm optimization
ICB	improved colliding bodies
IEEE	Institute of Electrical and Electronics Engineers
KHA	krill herd algorithm
L_a	voltage stability enhancement as an objective function
MOHFPSO	multi-objective hybrid firefly and particle swarm optimization
MOPSO	multi-objective particle swarm optimization
MODE	decomposition-based modern meta-heuristic algorithms
MOTLBO	multi-objective teaching learning based optimization
MOMICA	multi-objective modified imperialist competitive algorithm
MOHSA	multi-objective harmony search algorithm
MOALO	multi-objective ant lion optimizer
MJAYA	modified jaya algorithm
MODE	multi-objective differential evolution

N	No of lines
NSGA	non-dominated sorting genetic algorithm
OPF	optimal power flow
P_{gi}, P_{di}	active power of a generator and load of the <i>i-th</i> power line
P_{gi}	real power of the <i>i-th</i> generator
P_{gi}^{mn}, P_{gi}^{mx}	limits of real power based on the <i>i-th</i> generator
PL	active power loss as an objective function
Q_{gi}, Q_{di}	reactive power of a generator and load of the <i>i-th</i> power line
Q_{gi}	reactive power of the <i>i-th</i> generator
Q_{gi}^{mn}, Q_{gi}^{mx}	limits of reactive power based on the <i>i-th</i> generator
Q_{Ci}	VAR injection of the <i>i-th</i> shunt capacitor
Q_{Ci}^{mn}, Q_{Ci}^{mx}	limits of VAR injection based on the <i>i-th</i> shunt capacitor
SSO-PSO	salp swarm optimization with particle swarm optimization algorithm
$S_{linei}, S_{linei}^{mx}$	apparent and maximum apparent power of the <i>i-th</i> power line
TLBO	teaching learning based optimization
T_i	turn-ratio of the <i>i-th</i> transformer
T_i^{mn}, T_i^{mx}	limits of turn-ratio based on the <i>i-th</i> transformer
UPM	Universiti Putra Malaysia
VD	minimization of bus voltage deviation as an objective function
V_{gi}^{mx}, V_{gi}^{mn}	limits of voltage based on the <i>i-th</i> generator
V_{gi}	voltage of the <i>i-th</i> generator
V_{li}	voltage magnitude of the <i>i-th</i> load bus
V_{li}^{mn}, V_{li}^{mx}	limits of voltage based on the <i>i-th</i> load bus

CHAPTER 1

INTRODUCTION

1.1 Background of the study

The electrical power sector has evolved in the previous three decades, demanding the requirement to change the power network operation schemes to handle security and economic interests. The current variations consist of continuous growth of the electrical power loads, increasing involvement of decision-makers in the system operation field, relaxation of the power market, and expanding incoming distributed generation (DG) units. These substantial variations can be handled with a capable approach named optimal power flow (OPF) solution in the power operation field.

The continuous advancement in power demand is a constant challenge in the present power system. Load demand has surpassed the installed structure growth in many countries (Frank *et al.*, 2012b). The continuous shutdown has been noted worldwide over the previous decade (Yamashita *et al.*, 2008). These situations are burdened on power lines in various circumstances (Yamashita *et al.*, 2008).

Increasing demands of decision-makers are an additional growing burden to operation schemes of the present power network. The energy decision-maker makes essential estimations regarding conflicting objectives as various practical system operation schemes involve concurrent optimization of these goals.

Allocation of DG equipment in interconnected energy systems has simultaneously increased (Ghosh, Ghoshal, & Ghosh, 2010). The installation of DG technologies has further increased the complexities of dispatching optimal power problems. The system's present DG allocation deals with technical and economic issues and offers benefits, such as loss reduction, congestion mitigation, and transmission cost reduction (Ghosh *et al.*, 2010; Sheng *et al.*, 2014). These problems can be solved by the optimal location and capacity of DG units. Therefore, a formulation that can optimize all dependent variables and allocate penetrated DG units is required for the OPF problems.

Finally, the relaxation of power business is a significant issue to address. Various countries have introduced essential improvements in their energy markets to stop the monopoly in the last few years (Archana Singh, 2011). Legislation has been passed to allow the power providers to facilitate the users to choose their energy suppliers. Generation cost optimization in this private-enterprise situation is the main priority. The liberalized energy enterprises need intelligent and influential optimization approaches for OPF solutions to handle the different competitors with real-time simulation necessities.

The OPF solution is becoming a basic method over the past thirty years to deal with planning and operation in the electricity market. Various OPF mathematical equations are suggested to improve numerous goals over optimal adjustment of the power systems control variables, simultaneously applying constraints for operational purposes within stated limits. These optimization problems are highly non-convex multi-modal, and non-linear, with many local optima and specific global optima (Abou El Ela *et al.*, 2010). It is essential to state that the intricacy to deal with the OPF raises as the system size increases (Frank *et al.*, 2012b) because of the present growth.

Various energy systems operation problems consist of more than one different competitive goal to optimize at the same time; therefore, the multi-objective optimal power flow (MOOPF) approach has acquired attention in power services (Hazra, J., Sinah, 2011; H. Chen, *et al.*, 2014). This method is broadly used as a fundamental approach in network operation to enhance the modern energy sector's reliability, security, and economy (Khorsandi *et al.*, 2013). There are many growing issues and complexities to deal with the OPF issues.

To summarize, proposing proficient optimization methods to solve the single-objective and multi-objective OPF issues based on penetration of DG units considering the present substantial growth in the power field is a requirement.

1.2 Problem Statement

Different optimization algorithms based on OPF problems and DG units allocation are designed and applied for the electrical power system. A genetic algorithm (GA) has been proposed for the unity power factor network to obtain optimal sizing and location of DG unit, based on losses reduction and power enhancement. The main drawback of the approach is its slow convergence because of unbalanced exploration and exploitation. Another proposed approach, called Ant Colony optimization, is used to locate the site of the DG unit with the network reliability improvement as an objective function. The method cannot find the size of the DG unit simultaneously. Many authors have modeled various methods to find the capacity and location of DG units, one of these approaches is the particle swarm optimization (PSO) algorithm. The meta heuristic methods are based on a randomly generated population with two stages to update the particle position and calculate its velocity. The PSO needs less memory and simulation time. However, one of its drawbacks is partial optimization.

The shuffled frog leaping population-based method is used for optimum sizing and location of DG units. The approach is applied for handling various multi-modal, nonlinear, non-differentiable, and complex problems. However, the main limitations of the method are premature and slow convergence because of unbalancing between exploration and exploitation. Genetic algorithm (GA) is an easy technique widely applied for DG optimal integration. The approach can solve the non-continuous, non-differential, and non-dimensional problems. The optimum values can be achieved in real-time but with disadvantages of low convergence speed and random solutions. The bacterial foraging optimization approach (BFOA) has been applied to handle different power

system optimization cases due to its capacity in searching for the favorable region of the solution space. However, due to the complexity of the technique, the authors are interested in designing a straightforward approach to improve the convergence rate.

The above algorithms have been employed to effectively solve DG sources' optimum allocation problems. However, each approach has its disadvantages in terms of convergence time, unbalance between exploration/exploitation, simulation time, small-scale network effectiveness, and economic aspects.

Many developers designed and used classical gradient methods to solve the OPF problems, such as the newton method, quadratic programming (QP), linear programming (LP), interior-point methods (IPMs), and decomposition algorithms. These techniques get the optimum global point as a solution in particular cases. However, these approaches have some drawbacks, such as failure to handle the non-differentiable objective functions, confining in local optimal points, and being very sensitive to initial search or exploration points. Therefore suggesting substitute approaches to avert the mentioned weaknesses is essential. Subsequently, to remove the shortcoming of the above classical optimization techniques, different natural-inspired optimization algorithms are proposed and used to tackle the OPF problems, such as particle swarm optimization (PSO) method, genetic algorithm (GA) algorithm, harmony search (HS) method, gravitational search approach (GSA), artificial bee colony (ABC), and differential evolution (DE) technique. These stated techniques effectively obtain optimized values of various non-linear OPF objectives. Unfortunately, apart from their benefits, some of these methods are ineffective for global optimization, such as trapping in local optima because of unbalancing between exploration/exploitation and premature convergence for various OPF problems.

Recently, various hybrid optimization algorithms have been suggested to handle several OPF problems. The basic idea behind offering these hybrid approaches is to take advantage of each method, such as a balance between exploration/exploitation and design a more effective method that can deal with the population and classical-based algorithms weaknesses in the OPF field.

To sum up, proposing efficient optimization methods that can remove the shortcomings of present optimization algorithms and successfully solve various SOOPF/MOOPF problems and DG units allocation is required because of the current essential growth in the energy sector.

1.3 Research Aim and Objectives

The primary purpose of this thesis is to design and develop optimization methods to allocate DG units and solve various single and multi-objective OPF problems. The thesis's main objectives are:

1. To utilize an efficient Slime Mould Algorithm (SMA) based on optimal location and sizing of different numbers of DG units.
2. To apply a population-based Hybrid Firefly Particle Swarm Optimization (HFPSO) algorithm for various single-objective OPF problems.
3. To develop and use a technique based on Hybrid Firefly Particle Swarm Optimization (HFPSO) algorithm to deal with multi-objective OPF problems.

1.4 Motivations

Various reasons motivated this research: -

1. The wrong location and size of the DG units cause power loss, reverse power flow, voltage instability, power quality problems, and it could lead to complete system failure in an energy system. Therefore, proposing an approach to deal with DG's optimal size and the suitable location of the system to reach maximum efficiency is required.
2. Optimization approach of the present hybrid population-based optimization techniques designed and applied for solving the OPF problems needs a balance between exploitation and exploration to achieve an optimum solution. Any unbalance results in trapped in the local optima or produces the computational load. So, one of the key motivations for accomplishing this research work is applying a new hybrid meta-heuristic optimization algorithm that balances exploration and exploitation to lessen premature convergence, avoids trapping in local optima, and improves the convergence rate.
3. Designing a practical optimal Pareto front in a single simulation for MOOPF problems is complicated. Unfortunately, various proposed MOOPF based optimization algorithms can't simulate a Pareto optimal set and then global Pareto optimal front in a single execution. The conventional optimization approach converts a MOOPF problem to a single objective optimization problem based on the proper weighting factor approach. But the method mentioned above should be run as many times as the number of Pareto optimal sets to construct a global Pareto optimal front (Sivasubramani & Swarup, 2011). Therefore, a multi-objective hybrid firefly & particle swarm optimization (MOHFPSO) method based on non-dominated sorting and crowding distance is needed to design to deal with the MOOPF problem that can generate a global Pareto optimal set and global Pareto optimal front in a single execution. Euclidean distance approach is essential to select an optimal compromised solution. Also, a balance between exploration and exploitation in the method is necessary. Thus suggestion of a new multi-objective optimization method that can efficiently solve MOOPF problems in a single execution is a requirement.
4. Various heuristic methods may give impractical solutions based on violating the dependent variables constraints for OPF problems (Rezaei *et al.*, 2013; Ananthi *et*

al., 2014; Radosavljević *et al.*, 2015). In general, the optimization techniques based on the stochastic approach can not apply the operational limits. Therefore the method that can deal with all limitations is essential.

1.5 Scope and Limitation of the Study

The scope of this research work is bounded to suggest a new efficient single and multi-objective optimization method to deal with AC optimal power flow (ACOPF) problems without and with the allocation of DG units in the power systems. Five single-objective OPF based optimization cases are considered to solve: fuel cost minimization, voltage profile improvement, voltage stability enhancement, active power loss reduction, reactive power losses reduction. Furthermore, various cases of two and three combinations of the goal functions are considered simultaneously for multi-objective OPF optimization.

To authenticate the suggested methods, the scope of the thesis work also includes a complete comparison with OPF based optimization algorithms stated in the present literature. The OPF is a substantial and widely studied problem of power network operational issues such as security-constrained economic dispatch (SCED), economic dispatch (ED), and optimal reactive power dispatch (ORPD) were not included in this research work. Moreover, SMA based optimization algorithm is used to identify the optimal size and location of the DG units in the system. Meanwhile, DG's active power is used as a control variable to find the DG's size. Due to the limitation of this thesis work, only two standard bus test networks, such as IEEE 30-bus and IEEE 57-bus test systems, are used to analyze the effectiveness of the proposed approaches.

1.6 Contributions of the Study

In this thesis work, considerable contributions are made to the area of the power system to solve the OPF and MOOPF problems with and without the influence of DG units. In this thesis, the significant educational contributions for the scientific group is detailed as follow:-

1. An efficient SMA-based optimization algorithm is proposed and applied to solve the DG unit's optimum sizing and location in a single run. An equilibrium is kept between local and global search areas to find optimal value as a main contribution of the algorithm. Moreover, mathematical formulation and set of the associated constraints are implemented considering DG unit characteristics, enhancing the effectiveness of installing DG units in the power distribution system based on total power loss reduction.
2. A population-based hybrid firefly particle swarm optimization (HFPSO) approach is proposed and used for several single-objective OPF issues based on the DG

allocation for the first time in this thesis. Also, this hybrid approach keeps a balance between exploration and exploitation.

3 Another version of the HFPSO method, namely the multi-objective HFPSO approach, is designed, proposed, and applied to deal with multi-objective OPF problems in this thesis work. The proposed algorithm is based non-dominated sorting approach to obtain Pareto optimal front, while the optimal compromised solution is selected by applying the Euclidean distance concept. The proposed algorithm is used to two and three conflicting objectives simultaneously through ten different cases.

1.7 Thesis Layout

The rest of the thesis is arranged as follows. Chapter 2 discusses the literature review. Firstly, optimal power flow problems are overviewed. Following the previously proposed heuristic, classical and hybrid optimization algorithms used for various OPF problems are closely reviewed, compared, analyzed, and summarized.

Chapter 3 details the proposed techniques. First, the OPF problem's mathematical formulations are introduced. Then, Slime Mould Algorithm (SMA) based optimal location and sizing of DG units are designed and illustrated. Next, an effective population-based Hybrid Firefly Particle Swarm Optimization (HFPSO) method for single-objective and multi-objective OPF considering DG unit allocation is introduced and modeled, respectively.

Chapter 4 presents the simulated results, detailed discussions, and analysis of the proposed OPF based optimization approaches with the method stated in the current literature. In this thesis, two standard test systems, such as IEEE 30-bus and IEEE 57-bus test systems, are used to test, validate and demonstrate the efficacy of the suggested methods. Lastly, conclusions considering the proposed algorithm's implementation illustrated and future research work's recommendations are detailed in chapter 5, correspondingly.

REFERENCES

- Abadie, J., & Carpentier, J. (1969). Generalization of the Wolfe Reduced Gradient Method to the Case of Nonlinear Constraints. *R. Fletcher (Ed.), Optimization, Academic Press*, 37–47.
- Abd-Elazim, S. M., & Ali, E. S. (2013). A hybrid Particle Swarm Optimization and Bacterial Foraging for optimal Power System Stabilizers design. *International Journal of Electrical Power and Energy Systems*, 46(1), 334–341. <https://doi.org/10.1016/j.ijepes.2012.10.047>.
- Abdel, S., Soliman, H., Abdel, & Mantawy, A. H. (2013). Modern Optimization Techniques with Application in Electric Power Systems. *Journal of Chemical Information and Modeling*. 281-292. doi: 10.1007/978-1-4614-1752-1.
- Abdul Kadir, A. F., Mohamed, A., Shareef, H., Wanik, M. Z. C., & Ibrahim, A. A. (2013). Optimal sizing and placement of distributed generation in distribution system considering losses and THDV using gravitational search algorithm. *Przeglad Elektrotechniczny*. 89(4), 132-136.
- Abido, M. A. (2002a). Optimal design of Power System Stabilizers Using Particle Swarm Optimization. *IEEE Power Engineering Review*. 17(3), 406-413. <https://doi.org/10.1109/MPER.2002.4312374>.
- Abido, M. A. (2002b). Optimal power flow using particle swarm optimization. *International Journal of Electrical Power & Energy Systems*, 24, 563–571. doi: [http://dx.doi.org/10.1016/S0142-0615\(01\)00067-9](http://dx.doi.org/10.1016/S0142-0615(01)00067-9).
- Abido, M. A. (2009). Multiobjective particle swarm optimization for environmental/economic dispatch problem. *Electric Power Systems Research*. 79(7), 1105-1113. <https://doi.org/10.1016/j.epsr.2009.02.005>.
- Abou El-Ela, A. A., Allam, S. M., & Shatla, M. M. (2010). Maximal optimal benefits of distributed generation using genetic algorithms. *Electric Power Systems Research*. 80(7), 869-877. <https://doi.org/10.1016/j.epsr.2009.12.021>.
- Abou El Ela, A. A., Abido, M. A., & Spea, S. R. (2010). Optimal power flow using differential evolution algorithm. *Electric Power Systems Research*. 80(7), 878-885. <https://doi.org/10.1016/j.epsr.2009.12.018>.
- Aboubakr Khelifi, B. B. and S. C. (2018). Optimal power flow using Hybrid Particle Swarm Optimization and Moth Flame Optimizer approach. *Journal of Science and Engineering Sciences*, 07(33–41).
- Aguado, J. A., Quintana, V. H., & Conejo, A. J. (1999). Optimal power flows of interconnected power systems. In *1999 IEEE Power Engineering Society Summer Meeting, PES 1999 - Conference Proceedings* (pp. 814–819). <https://doi.org/10.1109/PESS.1999.787421>.
- Al-Kaabi, M., & Al-Bahrani, L. (2020). Modified artificial bee colony optimization

technique with different objective function of constraints optimal power flow. *International Journal of Intelligent Engineering and Systems*. 13(4), 378-388. <https://doi.org/10.22266/IJIES2020.0831.33>.

Almasabi, S., Alharbi, F. T., & Mitra, J. (2016). Opposition-based elitist real genetic algorithm for optimal power flow. In *NAPS 2016 - 48th North American Power Symposium, Proceedings*. <https://doi.org/10.1109/NAPS.2016.7747958>.

Alsac, O., Bright, J., Prais, M., & Stott, B. (1990). Further developments in lp-based optimal power flow. *IEEE Transactions on Power Systems*. 5(3), 697-711. doi:10.1109/59.65896.

Alsac, O., & Stott, B. (1974). Optimal load flow with steady-state security. *IEEE Transactions on Power Apparatus and Systems*. <https://doi.org/10.1109/TPAS.1974.293972>.

Amiri, B., Fathian, M., & Maroosi, A. (2009). Application of shuffled frog-leaping algorithm on clustering. *International Journal of Advanced Manufacturing Technology*. 45(1), 199-209. <https://doi.org/10.1007/s00170-009-1958-2>.

Aoki, K., Nishikori, A., & Yokoyama, R. (1987). Constrained load flow using recursive quadratic programming. *IEEE Transactions on Power Systems*, 2(1), 8-16. doi: 10.1109/TPWRS.1987.4335064.

Archana Singh, P. D. S. C. (2011). Electricity sector restructuring experience of different countries. *International Journal of Scientific & Engineering Research*, 2(4), 1-8.

Armaghani, S., Amjady, N., & Abedinia, O. (2015). Security constrained multi-period optimal power flow by a new enhanced artificial bee colony. *Applied Soft Computing Journal*. 37, 382-395. <https://doi.org/10.1016/j.asoc.2015.08.024>.

Attia, A. F., Al-Turki, Y. A., & Abusorrah, A. M. (2012). Optimal power flow using adapted genetic algorithm with adjusting population size. *Electric Power Components and Systems*. 40(11), 1285-1299 <https://doi.org/10.1080/15325008.2012.689417>.

Ayan, K., & Kiliç, U. (2012). Artificial bee colony algorithm solution for optimal reactive power flow. *Applied Soft Computing Journal*. 12(5), 1477-1482. <https://doi.org/10.1016/j.asoc.2012.01.006>.

Aydilek, İ. B. (2018). A hybrid firefly and particle swarm optimization algorithm for computationally expensive numerical problems. *Applied Soft Computing Journal*. 66, 232-249. <https://doi.org/10.1016/j.asoc.2018.02.025>.

Bachir, Bentouati, Saliha, Chettih, Pradeep, Jangir, ... Trivedi. (n.d.). A solution to the optimal power flow using multi-verse optimizer. *Journal of Electrical Systems*, 12(4), 716-733.

Badar, A. Q. H., Umre, B. S., & Junghare, A. S. (2012). Reactive power control using dynamic Particle Swarm Optimization for real power loss minimization. *International Journal of Electrical Power and Energy Systems*. 41(1), 133-136.

<https://doi.org/10.1016/j.ijepes.2012.03.030>.

- Bai, W., Eke, I., & Lee, K. Y. (2017). An improved artificial bee colony optimization algorithm based on orthogonal learning for optimal power flow problem. *Control Engineering Practice*, 61, 163-172. <https://doi.org/10.1016/j.conengprac.2017.02.010>.
- Bakirtzis, A. G., Biskas, P. N., Zoumas, C. E., & Petridis, V. (2002). Optimal power flow by enhanced genetic algorithm. *IEEE Transactions on Power Systems*, 17(2), 229-236. <https://doi.org/10.1109/TPWRS.2002.1007886>.
- Balasubramanian, S., Gokhale, R. V., & Sekar, A. (2015). A new AC optimal power flow formulation and solution using Genetic Algorithm based on P-Q decomposition. In *2015 North American Power Symposium, NAPS 2015*. <https://doi.org/10.1109/NAPS.2015.7335259>.
- Bansilal, Thukaram, D., & Parthasarathy, K. (1996). Optimal reactive power dispatch algorithm for voltage stability improvement. *International Journal of Electrical Power and Energy Systems*, 18(7), 461-468. [https://doi.org/10.1016/0142-0615\(96\)00004-X](https://doi.org/10.1016/0142-0615(96)00004-X).
- Basu, M. (2015). Modified particle swarm optimization for nonconvex economic dispatch problems. *International Journal of Electrical Power and Energy Systems*. <https://doi.org/10.1016/j.ijepes.2015.01.015>.
- Basu, M. (2016). Multi-objective optimal reactive power dispatch using multi-objective differential evolution. *International Journal of Electrical Power and Energy Systems*, 82, 213-224. <https://doi.org/10.1016/j.ijepes.2016.03.024>.
- Bedrinana, M. F., & C. (2009). Step size optimization based interior point algorithm: Applications and treatment of ill-conditioning in optimal power flow solutions. *Paper Presented at the 2009 IEEE Power & Energy Society General Meeting*.
- Bentouati, B., Javaid, M. S., Boucekara, H. R. E. H., & El-Fergany, A. A. (2020). Optimizing performance attributes of electric power systems using chaotic salp swarm optimizer. *International Journal of Management Science and Engineering Management*, 165-175. <https://doi.org/10.1080/17509653.2019.1677197>.
- Bhattacharya, A., & Chattopadhyay, P. K. (2011). Application of biogeography-based optimisation to solve different optimal power flow problems. *IET Generation, Transmission and Distribution*, 5(1), 70-80. <https://doi.org/10.1049/iet-gtd.2010.0237>.
- Bhattacharya, Aniruddha, & Chattopadhyay, P. K. (2010). Hybrid differential evolution with biogeography-based optimization for solution of economic load dispatch. *IEEE Transactions on Power Systems*, 25(4). <https://doi.org/10.1109/TPWRS.2010.2043270>
- Bhowmik, A. R., & Chakraborty, A. K. (2014a). Optimal reactive power flow using non dominated sorting multi objective gravitational search algorithm. In *2014 11th International Conference on Electrical Engineering/Electronics, Computer,*

Telecommunications and Information Technology, ECTI-CON 2014.
<https://doi.org/10.1109/ECTICon.2014.6839706>.

- Bhowmik, A. R., & Chakraborty, A. K. (2014b). Solution of optimal power flow using nondominated sorting multi objective gravitational search algorithm. *International Journal of Electrical Power and Energy Systems.* , 64, 1237-1250. <https://doi.org/10.1016/j.ijepes.2014.04.053>.
- Bhowmik, A. R., & Chakraborty, A. K. (2018). Non dominated sorting based multi objective GSA for solving optimal power flow problems. In *IEEE International Conference on Power, Control, Signals and Instrumentation Engineering, ICPCSI 2017.* <https://doi.org/10.1109/ICPCSI.2017.8392108>.
- Biswas, P. P., Suganthan, P. N., Mallipeddi, R., & Amaratunga, G. A. J. (2018). Optimal power flow solutions using differential evolution algorithm integrated with effective constraint handling techniques. *Engineering Applications of Artificial Intelligence*, 68, 81-100. <https://doi.org/10.1016/j.engappai.2017.10.019>.
- Boucekara, H. (2020). Solution of the optimal power flow problem considering security constraints using an improved chaotic electromagnetic field optimization algorithm. *Neural Computing and Applications*, 32, 2683-2703. <https://doi.org/10.1007/s00521-019-04298-3>.
- Boucekara, H. R. E. H. (2014). Optimal power flow using black-hole-based optimization approach. *Applied Soft Computing Journal*, 24, 879-888. <https://doi.org/10.1016/j.asoc.2014.08.056>.
- Boucekara, H. R. E. H., Abido, M. A., & Boucherma, M. (2014). Optimal power flow using Teaching-Learning-Based Optimization technique. *Electric Power Systems Research*, 114, 49-59. <https://doi.org/10.1016/j.epsr.2014.03.032>.
- Boucekara, H. R. E. H., Abido, M. A., Chaib, A. E., & Mehasni, R. (2014). Optimal power flow using the league championship algorithm: A case study of the Algerian power system. *Energy Conversion and Management*, 87, 58-70. <https://doi.org/10.1016/j.enconman.2014.06.088>.
- Burchett, R. C., Happ, H. H., & Vierath, D. R. (1984). Quadratically Convergent Optimal Power Flow. *IEEE Transactions on Power Apparatus and Systems*, PAS-103(11), 3267-3275. <https://doi.org/10.1109/TPAS.1984.318568>.
- Burchett, R. C., Happ, H. H., & Wirgau, K. A. (1982). Large Scale Optimal Power Flow. *IEEE Transactions on Power Apparatus and Systems*, 101(10), 3722-3732. <https://doi.org/10.1109/TPAS.1982.317057>.
- Chaouki Khammassia, S. K. (n.d.). *A NSGA2-LRWrapper Approach for Feature Selection in Network Intrusion Detection. research work to get a Ph.D. degree in computer science.* LARODEC Laboratory, Institut Supérieur de Gestion de Tunis, Université de Tunis, 41 Rue de la liberté, Le Bardo 2000, 70, 255-277, Tunisia. <https://doi.org/10.1016/j.cose.2017.06.005>.
- Chen, G., Liu, L., Zhang, Z., & Huang, S. (2017). Optimal reactive power dispatch by

- improved GSA-based algorithm with the novel strategies to handle constraints. *Applied Soft Computing*, 50, 58-70. <https://doi.org/10.1016/j.asoc.2016.11.008>.
- Chen, G., Qiu, S., Zhang, Z., Sun, Z., & Liao, H. (2017). Optimal Power Flow Using Gbest-Guided Cuckoo Search Algorithm with Feedback Control Strategy and Constraint Domination Rule. *Mathematical Problems in Engineering*. <https://doi.org/10.1155/2017/9067520>.
- Chen, H., Bo, M. L., & Zhu, Y. (2014). Multi-hive bee foraging algorithm for multi-objective optimal power flow considering the cost, loss, and emission. *International Journal of Electrical Power and Energy Systems*, 60, 203-220. <https://doi.org/10.1016/j.ijepes.2014.02.017>.
- Chiang, H.-D., Wang, B., & Jiang, Q.-Y. (2009). Applications of TRUST-TECH Methodology in Optimal Power Flow of Power Systems. *Optimization in the Energy Industry, Berlin, Heidelberg: Springer*, 297-318. https://doi.org/10.1007/978-3-540-88965-6_13.
- Christy, A., Raj, P. A. D. V., Padmanaban, S., Selvamuthukumaran, R., & Ertas, A. H. (2016). A bio-inspired novel optimization technique for reactive power flow. *Engineering Science and Technology, an International Journal*, 19(4), 1682-1692. <https://doi.org/10.1016/j.jestch.2016.07.011>.
- Chung, C. Y., Liang, C. H., Wong, K. P., & Duan, X. Z. (2010). Hybrid algorithm of differential evolution and evolutionary programming for optimal reactive power flow. *IET Generation, Transmission and Distribution*, 4(1), 84. <https://doi.org/10.1049/iet-gtd.2009.0007>.
- Coelho, L. dos S., & Mariani, V. C. (2009). An improved harmony search algorithm for power economic load dispatch. *Energy Conversion and Management*, 50(10), 2522-2526. <https://doi.org/10.1016/j.enconman.2009.05.034>.
- Crisan, O., & Mohtadi, M. A. (1992). Efficient identification of binding inequality constraints in optimal power flow Newton approach. *IEE Proceedings C - Generation, Transmission and Distribution*, 139(5), 365-370. doi: 10.1049/ip-c.1992.0053.
- de Carvalho, E. P., dos Santos, A., & Ma, T. F. (2008). Reduced gradient method combined with augmented Lagrangian and barrier for the optimal power flow problem. *Applied Mathematics and Computation*. <https://doi.org/10.1016/j.amc.2007.11.025>.
- Dommel, H. W., & Tinney, W. F. (1968). Optimal Power Flow Solutions. *IEEE Transactions on Power Apparatus and Systems, Volume: PA(10)*, 1866-1876. <https://doi.org/10.1109/TPAS.1968.292150>.
- Duman, S., Güvenç, U., Sönmez, Y., & Yörükeren, N. (2012). Optimal power flow using gravitational search algorithm. *Energy Conversion and Management*, (59), 86-95. <https://doi.org/10.1016/j.enconman.2012.02.024>.
- Duong, M. Q., Pham, T. D., Nguyen, T. T., Doan, A. T., & Van Tran, H. (2019).

- Determination of optimal location and sizing of solar photovoltaic distribution generation units in radial distribution systems. *Energies*, 12(1), 174. <https://doi.org/10.3390/en12010174>.
- E. C. Housos, G. D. I. (1982). A Sparse Variable Metric Optimization Method Applied to the Solution of Power System Problems. *IEEE Transactions on Power Apparatus and Systems, PAS-101*(1), 195–202.
- El-Hawary, M. E. (1993). Optimal Economic Operation of Large Scale Electric Power Systems: A Review. In *Joint International Power Conference "Athens Power Tech": Planning, Operation and Control in Today's Electric Power Systems, APT 1993 - Proceedings*. <https://doi.org/10.1109/APT.1993.686871>.
- El-Kady, M. A., Burchett, R. C., Happ, H. H., & Vie Rath, D. R. (1986). Assessment of real-time optimal voltage control. *IEEE Transactions on Power Systems*. 1(2), 98-105. <https://doi.org/10.1109/TPWRS.1986.4334912>.
- Elsaiah, S., Cai, N., Benidris, M., & Mitra, J. (2015). Fast economic power dispatch method for power system planning studies. *IET Generation, Transmission and Distribution*, 9(5), 417-426. <https://doi.org/10.1049/iet-gtd.2014.0130>.
- Fortenbacher, P., & Demiray, T. (2019). Linear/quadratic programming-based optimal power flow using linear power flow and absolute loss approximations. *International Journal of Electrical Power and Energy Systems*, 107, 680-689. <https://doi.org/10.1016/j.ijepes.2018.12.008>.
- Frank, S., & Rebennack, S. (2016). An introduction to optimal power flow: Theory, formulation, and examples. *IIE Transactions (Institute of Industrial Engineers)*, 48(12), 1172–1197. <https://doi.org/10.1080/0740817X.2016.1189626>.
- Frank, S., Steponavice, I., & Rebennack, S. (2012a). Optimal power flow: a bibliographic survey I. *Energy Systems*, 3(3), 221-258. <https://doi.org/10.1007/s12667-012-0056-y>.
- Frank, S., Steponavice, I., & Rebennack, S. (2012b). Optimal power flow: A bibliographic survey I Formulations and deterministic methods. *Energy Systems*. <https://doi.org/10.1007/s12667-012-0056-y>.
- Frank, S., Steponavice, I., & Rebennack, S. (2012c). Optimal power flow: a bibliographic survey II. *Energy Systems*. 3(3), 259-289. <https://doi.org/10.1007/s12667-012-0057-x>
- G.R.M. da Costa , C.E.U. Costa, A. M. de S. (2000). Comparative studies of optimization methods for the optimal power flow problem. *Electric Power Systems Research*, 3(56), 249–254. doi: [http://dx.doi.org/10.1016/S0378-7796\(00\)00114-0](http://dx.doi.org/10.1016/S0378-7796(00)00114-0).
- Gacem, A., & Benattous, D. (2017). Hybrid genetic algorithm and particle swarm for optimal power flow with non-smooth fuel cost functions. *International Journal of Systems Assurance Engineering and Management*, 1-8. <https://doi.org/10.1007/s13198-014-0312-8>.

- Ghasemi, M., Ghavidel, S., Ghanbarian, M. M., Gharibzadeh, M., & Azizi Vahed, A. (2014). Multi-objective optimal power flow considering the cost, emission, voltage deviation and power losses using multi-objective modified imperialist competitive algorithm. *Energy*, 78, 276-289. <https://doi.org/10.1016/j.energy.2014.10.007>.
- Ghasemi, M., Ghavidel, S., Ghanbarian, M. M., Massrur, H. R., & Gharibzadeh, M. (2014). Application of imperialist competitive algorithm with its modified techniques for multi-objective optimal power flow problem: A comparative study. *Information Sciences*, 281, 225-247. <https://doi.org/10.1016/j.ins.2014.05.040>.
- Ghosh, S., Ghoshal, S. P., & Ghosh, S. (2010). Optimal sizing and placement of distributed generation in a network system. *International Journal of Electrical Power and Energy Systems*, 32(8), 849-856. <https://doi.org/10.1016/j.ijepes.2010.01.029>.
- Granelli, G. P., & Montagna, M. (2000). Security-constrained economic dispatch using dual quadratic programming. *Electric Power Systems Research*, 56(1), 71-80. [https://doi.org/10.1016/S0378-7796\(00\)00097-3](https://doi.org/10.1016/S0378-7796(00)00097-3).
- Hazra, J., Sinah, A. K. (2011). A multi-objective optimal power flow using particle swarm optimization. *Europion Transactions on Electrical Power*, 21(1)(1), 1028–1045. doi: 10.1002/etep.494.
- Herbadji, O., Slimani, L., & Bouktir, T. (2019). Optimal power flow with four conflicting objective functions using multiobjective ant lion algorithm: A case study of the algerian electrical network. *Iranian Journal of Electrical and Electronic Engineering*. <https://doi.org/10.22068/IJEEE.15.1.94>.
- Herbadji, Ouafa, Slimani, L., & Bouktir, T. (2017). Multi-objective optimal power flow considering the fuel cost, emission, voltage deviation and power losses using Multi-Objective Dragonfly algorithm. In: *International Conference on Recent Advances in Electrical Systems*, 191–197.
- Hobson, E. (1978). Power system security control calculations using linear programming, part I. *IEEE Transactions on Power Apparatus and Systems*, PAS-97(5), 1713-1720. <https://doi.org/10.1109/TPAS.1978.354664>.
- J. A., & Zhu, J. Z. (1999). Improved interior point method for OPF problems. *IEEE Transactions on Power Systems*, 14(4), 1114–1120. doi: 10.1109/59.780938.
- Jabr, R. A. (2003). A Primal-Dual Interior-Point Method to Solve the Optimal Power Flow Dispatching Problem. *Optimization and Engineering*, 4(4), 309-336. <https://doi.org/10.1023/b:opte.0000005390.63406.1e>.
- Jadhav, H. T., & Bamane, P. D. (2016). Temperature dependent optimal power flow using g-best guided artificial bee colony algorithm. *International Journal of Electrical Power and Energy Systems*, 77, 77-90. <https://doi.org/10.1016/j.ijepes.2015.11.026>.
- Jadoun, V. K., Gupta, N., Niazi, K. R., & Swarnkar, A. (2015). Modulated particle swarm optimization for economic emission dispatch. *International Journal of Electrical*

- Jeyadevi, S., Baskar, S., Babulal, C. K., & Willjuice Iruthayarajan, M. (2011). Solving multiobjective optimal reactive power dispatch using modified NSGA-II. *International Journal of Electrical Power and Energy Systems*. 33(2), 219-228. <https://doi.org/10.1016/j.ijepes.2010.08.017>.
- Karmarkar, N. (1984). A new polynomial-time algorithm for linear programming. *Combinatorica*, 4(4), 373-395. <https://doi.org/10.1007/BF02579150>
- Kennedy, J., & Eberhart, R. (1995a). Particle Swarm Optimization, Proceedings of IEEE International Conference on Neural Networks Vol. IV: 1942–1948. <https://doi.org/10.1109/ICNN.1995.488968>.
- Kennedy, J., & Eberhart, R. (1995b). Particle Swarm Optimization James. In *Proceedings of the IEEE International Conference on Neural Networks*. <https://doi.org/10.1109/ICNN.1995.488968>.
- Kessler, D. (1982). Plasmodial Structure and Motility. In *Cell Biology of Physarum and Didymium*. <https://doi.org/10.1016/b978-0-12-049601-3.50010-9>.
- Khamidov, S., Normuratov, B., Pulatov, B., & Kilichov, O. (2020). Optimization of power flow through facts in electrical networks. In *IOP Conference Series: Materials Science and Engineering*. <https://doi.org/10.1088/1757-899X/883/1/012128>.
- Khazali, A. H., & Kalantar, M. (2011). Optimal reactive power dispatch based on harmony search algorithm. *International Journal of Electrical Power and Energy Systems*, 33(3), 684-692. <https://doi.org/10.1016/j.ijepes.2010.11.018>.
- Khelifi, A., Bentouati, B., Chettih, S., & El-Sehiemy, R. A. (2019). A hybrid cuckoo search and krill herd technique for solving problem of optimal power flow in power systems. *Journal of Electrical Systems*, 15(3) 375-391.
- Khorsandi, A., Hosseinian, S. H., & Ghazanfari, A. (2013). Modified artificial bee colony algorithm based on fuzzy multi-objective technique for optimal power flow problem. *Electric Power Systems Research*, 95, 206-213. <https://doi.org/10.1016/j.eprsr.2012.09.002>.
- Kim, B. H. (1999). A fast distributed implementation of optimal power flow. *IEEE Transactions on Power Systems*. 14(3), 858-864, <https://doi.org/10.1109/59.780896>.
- Kim, J. Y., Mun, K. J., Kim, H. S., & Park, J. H. (2011). Optimal power system operation using parallel processing system and PSO algorithm. *International Journal of Electrical Power and Energy Systems*, 33(8), 1457-1461. <https://doi.org/10.1016/j.ijepes.2011.06.026>.
- Kirschen, D. S., & Van Meeteren, H. P. (1988). Mw/Voltage control in a linear programming based optimal power flow. *IEEE Transactions on Power Systems*, 3(2), 481-489. <https://doi.org/10.1109/59.192899>.

- Kumari, M. S., & Maheswarapu, S. (2010). Enhanced Genetic Algorithm based computation technique for multi-objective Optimal Power Flow solution. *International Journal of Electrical Power and Energy Systems*, 32(6), 736-742. <https://doi.org/10.1016/j.ijepes.2010.01.010>.
- Lai, L. L., Ma, J. T., Yokoyama, R., & Zhao, M. (1997). Improved genetic algorithms for optimal power flow under both normal and contingency operation States. *International Journal of Electrical Power and Energy Systems*, 19(5), 287-292. [https://doi.org/10.1016/s0142-0615\(96\)00051-8](https://doi.org/10.1016/s0142-0615(96)00051-8).
- Lee, K. Y., Park, Y. M., & Ortiz, J. L. (1985). A United Approach to Optimal Real and Reactive Power Dispatch. *IEEE Power Engineering Review*, 104(5), 1147-1153. <https://doi.org/10.1109/MPER.1985.5526580>.
- Lee, Kwang Y., Bai, X., & Park, Y. M. (1995). Optimization Method for Reactive Power Planning by Using a Modified Simple Genetic Algorithm. *IEEE Transactions on Power Systems*, 10(4), 1843-1850. <https://doi.org/10.1109/59.476049>.
- Li, Q. (2009). Shuffled frog leaping algorithm based optimal reactive power flow. In *Proceedings - 1st International Symposium on Computer Network and Multimedia Technology, CNMT 2009*. <https://doi.org/10.1109/CNMT.2009.5374681>.
- Li, Shimin, Chen, H., Wang, M., Heidari, A. A., & Mirjalili, S. (2020). Slime mould algorithm: A new method for stochastic optimization. *Future Generation Computer Systems*, 111, 300-323. <https://doi.org/10.1016/j.future.2020.03.055>.
- Li, Shuijia, Gong, W., Wang, L., Yan, X., & Hu, C. (2020). Optimal power flow by means of improved adaptive differential evolution. *Energy*. <https://doi.org/10.1016/j.energy.2020.117314>.
- Li, Y., Wang, Y., & Li, B. (2013). A hybrid artificial bee colony assisted differential evolution algorithm for optimal reactive power flow. *International Journal of Electrical Power and Energy Systems*, 52, 25-33. <https://doi.org/10.1016/j.ijepes.2013.03.016>.
- Liu, L., Wang, X., Ding, X., & Chen, H. (2009). A robust approach to optimal power flow with discrete variables. *IEEE Transactions on Power Systems*, 24(3), 1182-1190. <https://doi.org/10.1109/TPWRS.2009.2023258>.
- Mahdad, B., & Srairi, K. (2016). Security constrained optimal power flow solution using new adaptive partitioning flower pollination algorithm. *Applied Soft Computing Journal*, 46(c) 501-522. <https://doi.org/10.1016/j.asoc.2016.05.027>.
- Mahdavi, M., Fesanghary, M., & Damangir, E. (2007). An improved harmony search algorithm for solving optimization problems. *Applied Mathematics and Computation*, 188(2), 1567-1579. <https://doi.org/10.1016/j.amc.2006.11.033>.
- Medina, M. A., Das, S., Coello Coello, C. A., & Ramírez, J. M. (2014). Decomposition-based modern metaheuristic algorithms for multi-objective optimal power flow -

A comparative study. *Engineering Applications of Artificial Intelligence*, 32, 10-20. <https://doi.org/10.1016/j.engappai.2014.01.016>.

Mirzaei, M., Jasni, J., Hizam, H., Abdul Wahab, N. I., & Moazami, E. (2013). Static voltage stability analysis using generalized regression neural network. In *Proceedings of the 2013 IEEE 7th International Power Engineering and Optimization Conference, PEOCO 2013*. <https://doi.org/10.1109/PEOCO.2013.6564579>.

Mohd Ilyas, Syed Mohammad Tanweer, A. R. (2013). Optimal Placement of Distributed Generation on Radial Distribution System for Loss Minimisation & Improvement of Voltage Profile. *International Journal of Modern Engineering Research (IJMER)*, 3(4), 2296–2312.

Montoya, O. D., Escobar, A. F., & Garrido, V. M. (2020). Power flow solution in direct current grids using the linear conjugate gradient approach. In *Journal of Physics: Conference Series*. <https://doi.org/10.1088/1742-6596/1448/1/012016>.

Mota-Palomino, R., & Quintana, V. H. (1984). A penalty function-linear programming method for solving power system constrained economic operation problems. *IEEE Transactions on Power Apparatus and Systems*, PAS-103(6), 1414-1422. <https://doi.org/10.1109/TPAS.1984.318478>.

Naderi, E., Pourakbari-Kasmaei, M., & Abdi, H. (2019). An efficient particle swarm optimization algorithm to solve optimal power flow problem integrated with FACTS devices. *Applied Soft Computing Journal*, 80, 243-262. <https://doi.org/10.1016/j.asoc.2019.04.012>.

Nangia, U., Jain, N. K., & Wadhwa, C. L. (2001). Multiobjective optimal load flow based on ideal distance minimization in 3D space. *International Journal of Electrical Power and Energy Systems*, 23(8), 847-855. [https://doi.org/10.1016/S0142-0615\(00\)00085-5](https://doi.org/10.1016/S0142-0615(00)00085-5).

Narimani, M. R., Azizipanah-Abarghooee, R., Zoghdar-Moghadam-Shahrekohne, B., & Gholami, K. (2013). A novel approach to multi-objective optimal power flow by a new hybrid optimization algorithm considering generator constraints and multi-fuel type. *Energy*, 49, 119-136. <https://doi.org/10.1016/j.energy.2012.09.031>.

Nayak, M. R., Nayak, C. K., & Rout, P. K. (2012). Application of Multi-Objective Teaching Learning based Optimization Algorithm to Optimal Power Flow Problem. *Procedia Technology*, 6, 255-264. <https://doi.org/10.1016/j.protcy.2012.10.031>.

Niknam, T., Narimani, M. R., Azizipanah-Abarghooee, R., & Bahmani-Firouzi, B. (2013). Multiobjective optimal reactive power dispatch and voltage control: A new opposition-based self-adaptive modified gravitational search algorithm. *IEEE Systems Journal*, 7(4), 742-753. <https://doi.org/10.1109/JSYST.2012.2227217>.

Niknam, T., Narimani, M. R., & Jabbari, M. (2013). Dynamic optimal power flow using hybrid particle swarm optimization and simulated annealing. *International Transactions on Electrical Energy Systems*, 23(7), 975-1001.

<https://doi.org/10.1002/etep.1633>.

- Niknam, T., Narimani, M. rasoul, Jabbari, M., & Malekpour, A. R. (2011). A modified shuffle frog leaping algorithm for multi-objective optimal power flow. *Energy*, 36(11),6420-6432. <https://doi.org/10.1016/j.energy.2011.09.027>.
- Nogales, F. J., Prieto, F. J., & Conejo, A. J. (2003). A Decomposition Methodology Applied to the Multi-Area Optimal Power Flow Problem. *Annals of Operations Research*, 120(1), 99-116. <https://doi.org/10.1023/A:1023374312364>.
- Ongsakul, W., & Tantimaporn, T. (2006). Optimal power flow by improved evolutionary programming. *Electric Power Components and Systems*, 34(1), 79-95. <https://doi.org/10.1080/15325000691001458>.
- Pandiarajan, K., & Babulal, C. K. (2016). Fuzzy harmony search algorithm based optimal power flow for power system security enhancement. *International Journal of Electrical Power and Energy Systems*, 78, 72-79. <https://doi.org/10.1016/j.ijepes.2015.11.053>.
- Peschon, J., Bree, D. W., & Hajdu, L. P. (1972). Optimal power-flow solutions for power system planning. *Proceedings of the IEEE*, 60(1), 64-70.
- Power Systems Test Case Archive. (2006). Retrieved from <https://labs.ece.uw.edu/pstca/>
- Pulluri, H., Naresh, R., & Sharma, V. (2017). An enhanced self-adaptive differential evolution based solution methodology for multiobjective optimal power flow. *Applied Soft Computing Journal*, 54, 229-245. <https://doi.org/10.1016/j.asoc.2017.01.030>.
- R.A.Fernandes, H.H.Happ, K. A. W. (1980). Optimal reactive power flow for improved system operations. *International Journal of Electrical Power & Energy Systems*, 2(3), 133-139. doi: [http://dx.doi.org/10.1016/0142-0615\(80\)90022-8](http://dx.doi.org/10.1016/0142-0615(80)90022-8).
- Radosavljević, J., Klimenta, D., Jevtić, M., & Arsić, N. (2015). Optimal Power Flow Using a Hybrid Optimization Algorithm of Particle Swarm Optimization and Gravitational Search Algorithm. *Electric Power Components and Systems*, 43(17), 1958-1970. <https://doi.org/10.1080/15325008.2015.1061620>.
- Ragab A, E. S., F, S., Bachir, B., & M. A, A. (2019). A novel multi-objective hybrid particle swarm and salp optimization algorithm for technical-economicalenvironmental operation in power systems. *Energy*, 193. <https://doi.org/https://doi.org/10.1016/j.energy.2019.116817>.
- Ramesh Kumar, A., & Premalatha, L. (2015). Optimal power flow for a deregulated power system using adaptive real coded biogeography-based optimization. *International Journal of Electrical Power and Energy Systems*, 73, 393-399. <https://doi.org/10.1016/j.ijepes.2015.05.011>.
- Rao, R. S., Ravindra, K., Satish, K., & Narasimham, S. V. L. (2013). Power loss minimization in distribution system using network reconfiguration in the presence of distributed generation. *IEEE Transactions on Power Systems*, 28(1), 317-325. <https://doi.org/10.1109/TPWRS.2012.2197227>.

- Reid, G. F., & Hasdorff, L. (1973). Economic dispatch using quadratic programming. *IEEE Transactions on Power Apparatus and Systems*, Volume: PA(6), 2015–2023. <https://doi.org/10.1109/TPAS.1973.293582>.
- Rezaee Jordehi, A., & Jasni, J. (2012). Particle swarm optimisation for discrete optimisation problems: a review. *Artificial Intelligence Review*, 43(2), 243–258. <https://doi.org/10.1007/s10462-012-9373-8>.
- Rezaei Adaryani, M., & Karami, A. (2013). Artificial bee colony algorithm for solving multi-objective optimal power flow problem. *International Journal of Electrical Power and Energy Systems*, 53, 219-230. <https://doi.org/10.1016/j.ijepes.2013.04.021>.
- Rezaie, H., Kazemi-Rahbar, M. H., Vahidi, B., & Rastegar, H. (2019). Solution of combined economic and emission dispatch problem using a novel chaotic improved harmony search algorithm. *Journal of Computational Design and Engineering*. <https://doi.org/10.1016/j.jcde.2018.08.001>.
- Roa-Sepulveda, C. A., & Pavez-Lazo, B. J. (2003). A solution to the optimal power flow using simulated annealing. *International Journal of Electrical Power and Energy Systems*, 25(1), 47-57. [https://doi.org/10.1016/S0142-0615\(02\)00020-0](https://doi.org/10.1016/S0142-0615(02)00020-0).
- Roy, P. K., & Paul, C. (2015). Optimal power flow using krill herd algorithm. *International Transactions on Electrical Energy Systems*, 25(8), 1397-1419. <https://doi.org/10.1002/etep.1888>.
- Ruhela, & Dinesh Singh. (2014). *A study of computational complexity of algorithms for numerical methods*. University of Rajasthan. Retrieved from <http://hdl.handle.net/10603/148188>.
- Santos, a. . J. (1995). Optimal-power-flow solution by Newton's method applied to an augmented Lagrangian function. *IEE Proceedings - Generation, Transmission and Distribution*, 142(1), 33-36, <https://doi.org/10.1049/ip-gtd:19951586>.
- Sarkheyli, A., Zain, A. M., & Sharif, S. (2015). The role of basic, modified and hybrid shuffled frog leaping algorithm on optimization problems: a review. *Soft Computing*, 19(7), 2011-2038. <https://doi.org/10.1007/s00500-014-1388-4>.
- Satyajit Bhuyan, S. H. (2014). PowerFlow Analysis on IEEE 57 bus System using MATLAB. *International Journal of Engineering Research & Technology (IJERT)*, 3(8), 1116-1171.
- Sayah, S., & Zehar, K. (2008). Modified differential evolution algorithm for optimal power flow with non-smooth cost functions. *Energy Conversion and Management*49(11), 3036-3042, <https://doi.org/10.1016/j.enconman.2008.06.014>
- Secui, D. C. (2015). A new modified artificial bee colony algorithm for the economic dispatch problem. *Energy Conversion and Management*, 89, 43-62. <https://doi.org/10.1016/j.enconman.2014.09.034>.
- Shaheen, A. M., El-Sehiemy, R. A., & Farrag, S. M. (2016). Solving multi-objective

optimal power flow problem via forced initialised differential evolution algorithm. *IET Generation, Transmission and Distribution*, 10(7), 1634-1647. <https://doi.org/10.1049/iet-gtd.2015.0892>.

Shaheen, A. M., Farrag, S. M., & El-Sehiemy, R. A. (2017). MOPF solution methodology. *IET Generation, Transmission and Distribution*, 11(2), 570-581. <https://doi.org/10.1049/iet-gtd.2016.1379>.

Sheng, W., Liu, K. yan, & Cheng, S. (2014). Optimal power flow algorithm and analysis in distribution system considering distributed generation. *IET Generation, Transmission and Distribution*, 8(2), 261-272. <https://doi.org/10.1049/iet-gtd.2013.0389>.

Shi, Y., & Eberhart, R. C. (1999). Empirical study of particle swarm optimization. In *Proceedings of the 1999 Congress on Evolutionary Computation, CEC 1999*. <https://doi.org/10.1109/CEC.1999.785511>.

Shoults, R. R., & Sun, D. T. (1982). Optimal power flow based upon P-Q decomposition. *IEEE Transactions on Power Apparatus and Systems*, PAS- 101(2), 397-405. <https://doi.org/10.1109/TPAS.1982.317120>.

Singh, R. P., Mukherjee, V., & Ghoshal, S. P. (2016). Particle swarm optimization with an aging leader and challengers algorithm for the solution of optimal power flow problem. *Applied Soft Computing Journal*, 40, 161-177. <https://doi.org/10.1016/j.asoc.2015.11.027>.

Sinsuphan, N., Leeton, U., & Kulworawanichpong, T. (2013). Optimal power flow solution using improved harmony search method. *Applied Soft Computing Journal*, 13(5), 2364-2374. <https://doi.org/10.1016/j.asoc.2013.01.024>.

Sivasubramani, S., & Swarup, K. S. (2011). Multi-objective harmony search algorithm for optimal power flow problem. *International Journal of Electrical Power and Energy Systems*, 33(3), 745-75.2 <https://doi.org/10.1016/j.ijepes.2010.12.031>.

Somasundaram, P., & Kuppusamy, K. (2005). Application of evolutionary programming to security constrained economic dispatch. *International Journal of Electrical Power and Energy Systems*. <https://doi.org/10.1016/j.ijepes.2004.12.006>.

Somasundaram, P., Kuppusamy, K., & Kumudini Devi, R. P. (2004). Evolutionary programming based security constrained optimal power flow. *Electric Power Systems Research*, 27(5-6), 343-351. <https://doi.org/10.1016/j.eprs.2004.02.006>.

Sood, Y. R. (2007). Evolutionary programming based optimal power flow and its validation for deregulated power system analysis. *International Journal of Electrical Power and Energy Systems*, 29(1), 65-75. <https://doi.org/10.1016/j.ijepes.2006.03.024>.

Sousa, A. A., Torres, G. L., & Cañizares, C. A. (2011). Robust optimal power flow solution using trust region and interior-point methods. *IEEE Transactions on Power Systems*, 26(2), 487-499. <https://doi.org/10.1109/TPWRS.2010.2068568>.

- Sousa, T., Soares, J., Vale, Z. A., Morais, H., & Faria, P. (2011). Simulated Annealing metaheuristic to solve the optimal power flow. In *IEEE Power and Energy Society General Meeting*. <https://doi.org/10.1109/PES.2011.6039543>.
- Suganthi, S. T., & Devaraj, D. (2013). An improved differential evolution based approach for emission constrained optimal power flow. In *2013 International Conference on Energy Efficient Technologies for Sustainability, ICEETS 2013*. <https://doi.org/10.1109/ICEETS.2013.6533576>.
- Suliman, M. S., Hizam, H., & Othman, M. L. (2020). Determining penetration limit of central PVDG topology considering the stochastic behaviour of PV generation and loads to reduce power losses and improve voltage profiles. *IET Renewable Power Generation*, 14(14), 2629-2638. <https://doi.org/10.1049/iet-rpg.2019.1376>.
- Tangpatiphan, K., & Yokoyama, A. (2009). Optimal power flow with steady-state voltage stability consideration using improved evolutionary programming. In *2009 IEEE Bucharest PowerTech: Innovative Ideas Toward the Electrical Grid of the Future*. <https://doi.org/10.1109/PTC.2009.5282214>.
- The MathWorks, Inc.: Natick, MA, U. (2016). R2016a, MATLAB Release.
- Thong, V. V., Driesen, J., & Belmans, R. (2007). Transmission system operation concerns with high penetration level of distributed generation. In *Proceedings of the Universities Power Engineering Conference*, 867-871. <https://doi.org/10.1109/UPEC.2007.4469063>.
- Torres, G. L., & Quintana, V. H. (2001). On a nonlinear multiple-centrality-corrections interior-point method for optimal power flow. *IEEE Transactions on Power Systems*, 16(2), 222-228. <https://doi.org/10.1109/59.918290>.
- Tu, S., Wachter, A., & Wei, E. (2021). A Two-Stage Decomposition Approach for AC Optimal Power Flow. *IEEE Transactions on Power Systems*, 36(1), 303-312. <https://doi.org/10.1109/TPWRS.2020.3002189>.
- Vaisakh, K., & Srinivas, L. R. (2011). Evolving ant direction differential evolution for OPF with non-smooth cost functions. *Engineering Applications of Artificial Intelligence*, 24(3), 426-436. <https://doi.org/10.1016/j.engappai.2010.10.019>.
- Varadarajan, M., & Swarup, K. S. (2008). Solving multi-objective optimal power flow using differential evolution. *IET Generation, Transmission and Distribution*, 2(5), 720-730. <https://doi.org/10.1049/iet-gtd:20070457>.
- Vargas, L. S., Quintana, V. H., & Vannelli, A. (1993). A tutorial description of an interior point method and its applications to security-constrained economic dispatch. *IEEE Transactions on Power Systems*, 8(3), 1315-1324. <https://doi.org/10.1109/59.260862>.
- Vasebi, A., Fesanghary, M., & Bathaee, S. M. T. (2007). Combined heat and power economic dispatch by harmony search algorithm. *International Journal of Electrical Power and Energy Systems*, 29(10), 713-719. <https://doi.org/10.1016/j.ijepes.2007.06.006>.

- Wang, L., & Singh, C. (2007). Environmental/economic power dispatch using a fuzzified multi-objective particle swarm optimization algorithm. *Electric Power Systems Research*, 77(12), 1654-1664. <https://doi.org/10.1016/j.epsr.2006.11.012>.
- Wang, L., & Singh, C. (2008). Stochastic economic emission load dispatch through a modified particle swarm optimization algorithm. *Electric Power Systems Research*. 78(8), 1466-1476. <https://doi.org/10.1016/j.epsr.2008.01.012>.
- Warid, W., Hizam, H., Mariun, N., & Abdul-Wahab, N. I. (2016). Optimal power flow using the Jaya algorithm. *Energies*, 9(9), 678. <https://doi.org/10.3390/en9090678>.
- Warid, W., Hizam, H., Mariun, N., & Abdul Wahab, N. I. (2018). A novel quasi-oppositional modified Jaya algorithm for multi-objective optimal power flow solution. *Applied Soft Computing Journal*, 65, 360-373. <https://doi.org/10.1016/j.asoc.2018.01.039>.
- Warid, W. S. (2017). *Optimal power flow based on fuzzy linear programming and modified Jaya algorithms*. Universiti Putra Malaysia. Retrieved from https://eng.upm.edu.my/upload/dokumen/20180919161736CV_PM_Dr_Hashim_07062018.pdf.
- Wu, Y. C., Debs, A. S., & Marsten, R. E. (1994). A Direct Nonlinear Predictor-Corrector Primal-Dual Interior Point Algorithm for Optimal Power Flows. *IEEE Transactions on Power Systems*, 9(2), 876-883. <https://doi.org/10.1109/59.317660>
- Yamashita, K., Joo, S. K., Li, J., Zhang, P., & Liu, C. C. (2008). Analysis, control, and economic impact assessment of major blackout events. *European Transactions on Electrical Power*, 18(8), 854-871. <https://doi.org/10.1002/etep.304>.
- Yang, X.-S. (2010). Firefly Algorithm, Stochastic Test Functions and Design Optimisation. *Int. J. Bio-Inspired Computation*, 2(2), 78–84..
- Yang, X. S. (2010). Firefly algorithm, Leavy flights and global optimization. In *Research and Development in Intelligent Systems XXVI: Incorporating Applications and Innovations in Intelligent Systems XVII*. <https://doi.org/10.1007/978-1-84882-983-1-15>.
- Yang, X. S., & He, X. (2013). Firefly algorithm: recent advances and applications. *International Journal of Swarm Intelligence*, 1(1), 36-50. <https://doi.org/10.1504/ijsi.2013.055801>.
- Yao, X., Liu, Y., & Lin, G. (1999). Evolutionary programming made faster. *IEEE Transactions on Evolutionary Computation*. <https://doi.org/10.1109/4235.771163>.
- Younes, M., Khodja, F., & Kherfane, R. L. (2014). Multi-objective economic emission dispatch solution using hybrid FFA (firefly algorithm) and considering wind power penetration. *Energy*, 67, 595-606. <https://doi.org/10.1016/j.energy.2013.12.043>.
- Yuryevich, J. (1999). Evolutionary programming based optimal power flow algorithm. *IEEE Transactions on Power Systems*, 14(4), 1245-1250. <https://doi.org/10.1109/59.801880>.

Zeng, Y., & Sun, Y. (2014). Solving multiobjective optimal reactive power dispatch using improved multiobjective particle swarm optimization. *IEEE. 26th Chinese Control and Decision Conference*, 2014. DOI: 10.1109/CCDC.2014.6852312.

Zhang, Y., Li, Y., Xia, F., & Luo, Z. (2012). Immunity-based gravitational search algorithm. In *Lecture Notes in Computer Science (including subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*. (754-761). https://doi.org/10.1007/978-3-642-34062-8_98.

Ziane, I., Benhamida, F., & Graa, A. (2017). Simulated annealing algorithm for combined economic and emission power dispatch using max/max price penalty factor. *Neural Computing and Applications*, 1-9. <https://doi.org/10.1007/s00521-016-2335-3>



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