

TABLE I
PERFORMANCE COMPARISON OF TOTAL POWER GENERATION FOR
DIFFERENT VALUE OF b

b	Mean (MW)	Best (MW)	Worst (MW)	Standard Deviation
1.00	4.7641691656	4.7648415724	4.6976008998	6.7241e+03
0.95	4.7648415724	4.7648415724	4.7648415724	5.6161e-10
0.90	4.7630743588	4.7648415724	4.6764808948	1.2433e+04
0.85	4.7641691656	4.7648415724	4.6976008999	6.7241e+03
0.80	4.7648415724	4.7648415724	4.7648415724	4.3903e-10
0.75	4.7648415724	4.7648415724	4.7648415724	3.7441e-10
0.70	4.7648415724	4.7648415724	4.7648415724	3.7441e-10

TABLE II
THE COMPARISON PERFORMANCE OF TOTAL POWER GENERATION (MW)
BETWEEN MFO, SDA AND SED BASED APPROACH

Statistical Evaluation	MFO	SDA	SED
Average	4.7648415724	4.7648415723	4.7644075485
Highest	4.7648415724	4.7648415723	4.7648415242
Lowest	4.7648415724	4.7648415723	4.7627457259
Standard Deviation	5.6161×10^{-10}	1.1039824×10^{-7}	4.513106×10^2

IV. CONCLUSION

This paper presents a data-driven method based on moth-flame optimization (MFO) algorithm has been investigated. This study aims to propose a MFO for a power generation of wind plant and compare the findings with SDA and SED based approaches. In this simulation results, the MFO based method exhibits a slightly higher total power generation than the SDA and SED based approaches. This proves the potential of MFO based approach for data driven method of wind plant control. In the future, it is necessary to improve the convergence speed of MFO since it will take longer time if the size of wind plant is large. In this case, one might consider a multi-resolution version of MFO to increase the convergence speed.

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REFERENCES

- [1] J. R. Marden, S. D. Ruben, and L. Y. Pao, 'A Model-Free Approach to Wind Farm Control Using Game Theoretic Methods', *IEEE Trans. Control Syst. Technol.*, vol. 21, no. 4, pp. 1207–1214, Jul. 2013.
- [2] J. Marden, S. Ruben, and L. Pao, 'Surveying Game Theoretic Approaches for Wind Farm Optimization', in *50th AIAA Aerospace Sciences Meeting Including the New Horizons Forum and Aerospace Exposition*, 2012.
- [3] J. Park, S. Kwon, and K. H. Law, 'Wind farm power maximization based on a cooperative static game approach', in *Active and Passive Smart Structures and Integrated Systems 2013*, 2013, vol. 8688, p. 86880R.
- [4] P. M. O. Gebraad, F. C. van Dam, and J. W. van Wingerden, 'A model-free distributed approach for wind plant control', in *2013 American Control Conference*, 2013, pp. 628–633.
- [5] P. M. O. Gebraad and J. W. van Wingerden, 'Maximum power-point tracking control for wind farms', *Wind Energy*, vol. 18, no. 3, pp. 429–447, Mar. 2015.
- [6] M. R. Hao, R. M. T. R. Ismail, and M.A. Ahmad, 'Using spiral dynamic algorithm for maximizing power production of wind farm', in *2017 International Conference on Applied System Innovation (ICASI)*, 2017, pp. 1706–1709.
- [7] M. A. Ahmad, S. Azuma, and T. Sugie, 'A Model-Free Approach for Maximizing Power Production of Wind Farm Using Multi-Resolution Simultaneous Perturbation Stochastic Approximation', *Energies*, vol. 7, no. 9, pp. 5624–5646, Aug. 2014.
- [8] V. Aristidis, P. Maria, and L. Christos, 'Particle Swarm Optimization (PSO) algorithm for wind farm optimal design', *Int. J. Manag. Sci. Eng. Manag.*, vol. 5, pp. 53–58, May 2013.
- [9] M.A. Ahmad, M. R. Hao, R. M. T. R. Ismail, and A. N. K. Nasir, 'Model-free wind farm control based on random search', in *2016 IEEE International Conference on Automatic Control and Intelligent Systems (I2CACIS)*, 2016, pp. 131–134.
- [10] S. Mirjalili, 'Moth-flame optimization algorithm: A novel nature-inspired heuristic paradigm', *Knowl.-Based Syst.*, vol. 89, no. Supplement C, pp. 228–249, Nov. 2015.
- [11] B. Bentouati, L. Chaib, and S. Chettih, 'Optimal Power Flow using the Moth Flam Optimizer: A case study of the Algerian power system', *Indones. J. Electr. Eng. Comput. Sci.*, vol. 1, no. 3, pp. 431–445, Feb. 2016.
- [12] O. Ceylan, 'Harmonic elimination of multilevel inverters by moth-flame optimization algorithm', in *2016 International Symposium on Industrial Electronics (INDEL)*, 2016, pp. 1–5.
- [13] H. M. Zawbaa, E. Emary, B. Parv, and M. Sharawi, 'Feature selection approach based on moth-flame optimization algorithm', in *2016 IEEE Congress on Evolutionary Computation (CEC)*, 2016, pp. 4612–4617.
- [14] Vikas and S. J. Nanda, 'Multi-objective Moth Flame Optimization', in *2016 International Conference on Advances in Computing, Communications and Informatics (ICACCI)*, 2016, pp. 2470–2476.
- [15] B.S. Yildiz, A.R. Yildiz, 'Moth-flame optimization algorithm to determine optimal machining parameters in manufacturing processes', *Materials Testing*, 59(5), 2017, pp.425-429.
- [16] M. Wang, H. Chen, B. Yang, X. Zhao, L. Hu, Z. Cai, H. Huang, C. Tong, 'Toward an optimal kernel extreme learning machine using a chaotic moth-flame optimization strategy with applications in medical diagnoses', *Neurocomputing*, 267, 2017, pp.69-84.
- [17] A.E. Hassanien, T. Gaber, U. Mokhtar, H. Hefny, 'An improved moth flame optimization algorithm based on rough sets for tomato diseases detection' *Computers and Electronics in Agriculture*, 136, 2017, pp.86-96.
- [18] A.A. Ewees, A.T. Sahlol, M.A. Amasha, 'A Bio-inspired moth-flame optimization algorithm for Arabic handwritten letter recognition', In *International Conference on Control, Artificial Intelligence, Robotics & Optimization (ICCAIRO)*, 2017, pp. 154-159.
- [19] P. Singh, S. Prakash, 'Optical network unit placement in Fiber-Wireless (FiWi) access network by Moth-Flame optimization algorithm', *Optical Fiber Technology*, 36, 2017, pp.403-411.
- [20] A. K. Scholbrock, 'Optimizing Wind Farm Control Strategies to Minimize Wake Loss Effects', *Univ. Colo. Boulder*, 2011.