

Bibliography

- [1] Nguyen Dang and Patrick De Causmaecker. “Analysis of algorithm components and parameters: some case studies”. In: *International Conference on Learning and Intelligent Optimization*. Springer, 2018, pp. 288–303.
- [2] David S Johnson et al. “Experimental analysis of heuristics for the ATSP”. In: *The traveling salesman problem and its variations*. Springer, 2007, pp. 445–487.
- [3] M. Dorigo, V. Maniezzo, and A. Coloni. “Ant system: optimization by a colony of cooperating agents”. In: *IEEE Transactions on Systems, Man, and Cybernetics, Part B (Cybernetics)* 26.1 (1996), pp. 29–41. DOI: 10.1109/3477.484436.
- [4] Mauro Birattari and Janusz Kacprzyk. *Tuning metaheuristics: a machine learning perspective*. Vol. 197. Springer, 2009.
- [5] S. K. Smit and A. E. Eiben. “Comparing parameter tuning methods for evolutionary algorithms”. In: *2009 IEEE Congress on Evolutionary Computation*. 2009, pp. 399–406. DOI: 10.1109/CEC.2009.4982974.
- [6] Peng Li and Hua Zhu. “Parameter selection for ant colony algorithm based on bacterial foraging algorithm”. In: *Mathematical Problems in Engineering* 2016 (2016).
- [7] Matthias Feurer and Frank Hutter. “Hyperparameter optimization”. In: *Automated Machine Learning*. Springer, Cham, 2019, pp. 3–33.
- [8] Abraham P Punnen. “The traveling salesman problem: Applications, formulations and variations”. In: *The traveling salesman problem and its variations*. Springer, 2007, pp. 1–28.

- [9] C. Huang, Y. Li, and X. Yao. “A Survey of Automatic Parameter Tuning Methods for Metaheuristics”. In: *IEEE Transactions on Evolutionary Computation* 24.2 (2020), pp. 201–216. doi: 10.1109/TEVC.2019.2921598.
- [10] Dorian Gaertner and Keith L Clark. “On Optimal Parameters for Ant Colony Optimization Algorithms.” In: *IC-AI*. Citeseer. 2005, pp. 83–89.
- [11] Pauli Virtanen et al. “SciPy 1.0: Fundamental Algorithms for Scientific Computing in Python”. In: *Nature Methods* 17 (2020), pp. 261–272. doi: 10.1038/s41592-019-0686-2.
- [12] C. E. Rasmussen and C. K. I. Williams. *Gaussian Processes for Machine Learning*. the MIT Press, 2006. ISBN: 026218253X.
- [13] Eric Brochu, Vlad M. Cora, and Nando de Freitas. *A Tutorial on Bayesian Optimization of Expensive Cost Functions, with Application to Active User Modeling and Hierarchical Reinforcement Learning*. 2010. arXiv: 1012.2599 [cs.LG].
- [14] Cristiano Cervellera and Danilo Maccio. “Learning With Kernel Smoothing Models and Low-Discrepancy Sampling”. In: *IEEE Transactions on Neural Networks and Learning Systems* 24.3 (2013), pp. 504–509. doi: 10.1109/TNNLS.2012.2236353.
- [15] C. Schretter, L. Kobbelt, and Paul-Olivier Dehaye. “Golden Ratio Sequences for Low-Discrepancy Sampling”. In: *J. Graph. Tools* 16 (2012), pp. 95–104.
- [16] Frank Hutter, Holger H Hoos, and Thomas Stützle. “Automatic algorithm configuration based on local search”. In: *Aaai*. Vol. 7. 2007, pp. 1152–1157.
- [17] Eduardo Batista de Moraes Barbosa and Edson Luiz França Senne. “Improving the fine-tuning of metaheuristics: an approach combining design of experiments and racing algorithms”. In: *Journal of Optimization* 2017 (2017). doi: <https://doi.org/10.1155/2017/8042436>.
- [18] Tim Head et al. “scikit-optimize/scikit-optimize”. Version 0.8.1. In: (Sept. 2020). doi: 10.5281/zenodo.4014775.
- [19] Gerhard Reinelt. “TSPLIB—A traveling salesman problem library”. In: *ORSA journal on computing* 3.4 (1991), pp. 376–384.

- [20] Thomas Hammerl. *Ant colony optimization for tree and hypertree decompositions*. 2009.

Appendix A

Implementation

The implementation of the optimization procedure can be found at the following github repo: <https://github.com/klaswijk/bayesian-tuning>.

The libaco implementation can be found at the following github repo: <https://github.com/emmyyin/libaco>.