

Carola Doerr

List of Publications by Year in descending order

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Version: 2024-02-01

103
papers

1,787
citations

430874

18
h-index

395702

33
g-index

103
all docs

103
docs citations

103
times ranked

330
citing authors

| # | ARTICLE | IF | CITATIONS |
|----|---|-----|-----------|
| 1 | Multiplicative Drift Analysis. <i>Algorithmica</i> , 2012, 64, 673-697. | 1.3 | 187 |
| 2 | From black-box complexity to designing new genetic algorithms. <i>Theoretical Computer Science</i> , 2015, 567, 87-104. | 0.9 | 166 |
| 3 | Optimal Static and Self-Adjusting Parameter Choices for the $(1+(\lambda, \lambda))$ Genetic Algorithm. <i>Algorithmica</i> , 2018, 80, 1658-1709. | 1.3 | 96 |
| 4 | Optimal Parameter Choices Through Self-Adjustment. , 2015, , . | | 60 |
| 5 | Optimal Parameter Choices via Precise Black-Box Analysis. , 2016, , . | | 48 |
| 6 | Finding optimal volume subintervals with k points and calculating the star discrepancy are NP-hard problems. <i>Journal of Complexity</i> , 2009, 25, 115-127. | 1.3 | 44 |
| 7 | Theory of Parameter Control for Discrete Black-Box Optimization: Provable Performance Gains Through Dynamic Parameter Choices. <i>Natural Computing Series</i> , 2020, , 271-321. | 2.2 | 44 |
| 8 | Lessons from the black-box. , 2013, , . | | 41 |
| 9 | Benchmarking discrete optimization heuristics with IOHprofiler. <i>Applied Soft Computing Journal</i> , 2020, 88, 106027. | 7.2 | 41 |
| 10 | Static and Self-Adjusting Mutation Strengths for Multi-valued Decision Variables. <i>Algorithmica</i> , 2018, 80, 1732-1768. | 1.3 | 39 |
| 11 | Discrepancy-based evolutionary diversity optimization. , 2018, , . | | 36 |
| 12 | Ranking-Based Black-Box Complexity. <i>Algorithmica</i> , 2014, 68, 571-609. | 1.3 | 35 |
| 13 | Optimal parameter choices via precise black-box analysis. <i>Theoretical Computer Science</i> , 2020, 801, 1-34. | 0.9 | 35 |
| 14 | Exploratory Landscape Analysis is Strongly Sensitive to the Sampling Strategy. <i>Lecture Notes in Computer Science</i> , 2020, , 139-153. | 1.3 | 35 |
| 15 | k -Bit Mutation with Self-Adjusting k Outperforms Standard Bit Mutation. <i>Lecture Notes in Computer Science</i> , 2016, , 824-834. | 1.3 | 32 |
| 16 | A Tight Runtime Analysis of the $(1+(\lambda, \lambda))$ Genetic Algorithm on OneMax. , 2015, , . | | 30 |
| 17 | Self-adjusting mutation rates with provably optimal success rules. , 2019, , . | | 29 |
| 18 | Simple on-the-fly parameter selection mechanisms for two classical discrete black-box optimization benchmark problems. , 2018, , . | | 26 |

| # | ARTICLE | IF | CITATIONS |
|----|---|-----|-----------|
| 19 | Money for Nothing. , 2015, , . | | 25 |
| 20 | The Impact of Random Initialization on the Runtime of Randomized Search Heuristics. <i>Algorithmica</i> , 2016, 75, 529-553. | 1.3 | 25 |
| 21 | A New Randomized Algorithm to Approximate the Star Discrepancy Based on Threshold Accepting. <i>SIAM Journal on Numerical Analysis</i> , 2012, 50, 781-807. | 2.3 | 22 |
| 22 | Playing Mastermind with Constant-Size Memory. <i>Theory of Computing Systems</i> , 2014, 55, 658-684. | 1.1 | 22 |
| 23 | Calculation of Discrepancy Measures and Applications. <i>Lecture Notes in Mathematics</i> , 2014, , 621-678. | 0.2 | 22 |
| 24 | Rumor spreading in random evolving graphs. <i>Random Structures and Algorithms</i> , 2016, 48, 290-312. | 1.1 | 21 |
| 25 | A Simple Proof for the Usefulness of Crossover in Black-Box Optimization. <i>Lecture Notes in Computer Science</i> , 2018, , 29-41. | 1.3 | 20 |
| 26 | Tuning as a means of assessing the benefits of new ideas in interplay with existing algorithmic modules. , 2021, , . | | 20 |
| 27 | IOHalyzer: Detailed Performance Analyses for Iterative Optimization Heuristics. <i>ACM Transactions on Evolutionary Learning</i> , 2022, 2, 1-29. | 3.5 | 20 |
| 28 | Adaptive landscape analysis. , 2019, , . | | 17 |
| 29 | Towards Feature-Based Performance Regression Using Trajectory Data. <i>Lecture Notes in Computer Science</i> , 2021, , 601-617. | 1.3 | 17 |
| 30 | Playing Mastermind With Many Colors. <i>Journal of the ACM</i> , 2016, 63, 1-23. | 2.2 | 16 |
| 31 | Interpolating Local and Global Search by Controlling the Variance of Standard Bit Mutation. , 2019, , . | | 16 |
| 32 | Expressiveness and robustness of landscape features. , 2019, , . | | 16 |
| 33 | Online selection of CMA-ES variants. , 2019, , . | | 16 |
| 34 | The unbiased black-box complexity of partition is polynomial. <i>Artificial Intelligence</i> , 2014, 216, 275-286. | 5.8 | 15 |
| 35 | Towards a theory-guided benchmarking suite for discrete black-box optimization heuristics. , 2018, , . | | 15 |
| 36 | Bayesian performance analysis for black-box optimization benchmarking. , 2019, , . | | 15 |

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|----|---|------|-----------|
| 37 | The query complexity of a permutation-based variant of Mastermind. <i>Discrete Applied Mathematics</i> , 2019, 260, 28-50. | 0.9 | 15 |
| 38 | Unbiased black-box complexities of jump functions. , 2014, , . | | 14 |
| 39 | Introducing Elitist Black-Box Models: When Does Elitist Behavior Weaken the Performance of Evolutionary Algorithms?. <i>Evolutionary Computation</i> , 2017, 25, 587-606. | 3.0 | 14 |
| 40 | Memory-restricted black-box complexity of OneMax. <i>Information Processing Letters</i> , 2012, 112, 32-34. | 0.6 | 13 |
| 41 | Elitist Black-Box Models. , 2015, , . | | 13 |
| 42 | Black-Box Optimization Revisited: Improving Algorithm Selection Wizards Through Massive Benchmarking. <i>IEEE Transactions on Evolutionary Computation</i> , 2022, 26, 490-500. | 10.0 | 13 |
| 43 | Self-Adjusting Mutation Rates with Provably Optimal Success Rules. <i>Algorithmica</i> , 2021, 83, 3108-3147. | 1.3 | 13 |
| 44 | Towards an Adaptive CMA-ES Configurator. <i>Lecture Notes in Computer Science</i> , 2018, , 54-65. | 1.3 | 13 |
| 45 | The Query Complexity of Finding a Hidden Permutation. <i>Lecture Notes in Computer Science</i> , 2013, , 1-11. | 1.3 | 13 |
| 46 | The Right Mutation Strength for Multi-Valued Decision Variables. , 2016, , . | | 13 |
| 47 | Constructing low star discrepancy point sets with genetic algorithms. , 2013, , . | | 12 |
| 48 | Unbiased Black-Box Complexities of Jump Functions. <i>Evolutionary Computation</i> , 2015, 23, 641-670. | 3.0 | 12 |
| 49 | Hyper-parameter tuning for the $(1 + (\hat{\lambda}_1, \hat{\lambda}_2))$ GA. , 2019, , . | | 12 |
| 50 | Optimization of Chance-Constrained Submodular Functions. <i>Proceedings of the AAAI Conference on Artificial Intelligence</i> , 2020, 34, 1460-1467. | 4.9 | 12 |
| 51 | OneMax in Black-Box Models with Several Restrictions. , 2015, , . | | 10 |
| 52 | Maximizing drift is not optimal for solving OneMax. , 2019, , . | | 10 |
| 53 | Integrated vs. sequential approaches for selecting and tuning CMA-ES variants. , 2020, , . | | 10 |
| 54 | The $(1+1)$ Elitist Black-Box Complexity of LeadingOnes. <i>Algorithmica</i> , 2018, 80, 1579-1603. | 1.3 | 9 |

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|----|---|-----|-----------|
| 55 | Fast re-optimization via structural diversity. , 2019, , . | | 9 |
| 56 | Solving Problems with Unknown Solution Length at Almost No Extra Cost. Algorithmica, 2019, 81, 703-748. | 1.3 | 9 |
| 57 | Nevergrad. ACM SIGEVOlution, 2021, 14, 8-15. | 0.5 | 9 |
| 58 | Benchmarking a $(\mu + \lambda)$ Genetic Algorithm with Configurable Crossover Probability. Lecture Notes in Computer Science, 2020, , 699-713. | 1.3 | 9 |
| 59 | Fixed-target runtime analysis. , 2020, , . | | 9 |
| 60 | The impact of random initialization on the runtime of randomized search heuristics. , 2014, , . | | 8 |
| 61 | OneMax in Black-Box Models with Several Restrictions. Algorithmica, 2017, 78, 610-640. | 1.3 | 8 |
| 62 | Unknown solution length problems with no asymptotically optimal run time. , 2017, , . | | 8 |
| 63 | Offspring population size matters when comparing evolutionary algorithms with self-adjusting mutation rates. , 2019, , . | | 8 |
| 64 | Making a case for (Hyper-)parameter tuning as benchmark problems. , 2019, , . | | 8 |
| 65 | Towards large scale automated algorithm design by integrating modular benchmarking frameworks. , 2021, , . | | 8 |
| 66 | Optimal Mutation Rates for the $(1 + \lambda)$ EA on OneMax. Lecture Notes in Computer Science, 2020, , 574-587. | 1.3 | 8 |
| 67 | Provably Optimal Self-adjusting Step Sizes for Multi-valued Decision Variables. Lecture Notes in Computer Science, 2016, , 782-791. | 1.3 | 8 |
| 68 | Towards dynamic algorithm selection for numerical black-box optimization. , 2020, , . | | 8 |
| 69 | Solving Problems with Unknown Solution Length at (Almost) No Extra Cost. , 2015, , . | | 7 |
| 70 | Sensitivity of Parameter Control Mechanisms with Respect to Their Initialization. Lecture Notes in Computer Science, 2018, , 360-372. | 1.3 | 7 |
| 71 | Benchmarking discrete optimization heuristics with IOHprofiler. , 2019, , . | | 6 |
| 72 | Maximizing Drift is Not Optimal for Solving OneMax. Evolutionary Computation, 2021, 29, 1-20. | 3.0 | 6 |

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|----|--|-----|-----------|
| 73 | Evolving Sampling Strategies for One-Shot Optimization Tasks. Lecture Notes in Computer Science, 2020, , 111-124. | 1.3 | 6 |
| 74 | The (1+1) Elitist Black-Box Complexity of LeadingOnes. , 2016, , . | | 4 |
| 75 | Non-static parameter choices in evolutionary computation. , 2017, , . | | 4 |
| 76 | Coupling the design of benchmark with algorithm in landscape-aware solver design. , 2019, , . | | 4 |
| 77 | Optimal static mutation strength distributions for the $(1 + \hat{\mu})$ evolutionary algorithm on OneMax. , 2021, , . | | 4 |
| 78 | OPTION. , 2021, , . | | 4 |
| 79 | Dynamic control parameter choices in evolutionary computation. , 2020, , . | | 4 |
| 80 | Fixed-Target Runtime Analysis. Algorithmica, 2022, 84, 1762-1793. | 1.3 | 4 |
| 81 | Black-box complexity. , 2014, , . | | 3 |
| 82 | Compiling a benchmarking test-suite for combinatorial black-box optimization. , 2018, , . | | 3 |
| 83 | Leveraging benchmarking data for informed one-shot dynamic algorithm selection. , 2021, , . | | 3 |
| 84 | Variance Reduction for Better Sampling in Continuous Domains. Lecture Notes in Computer Science, 2020, , 154-168. | 1.3 | 3 |
| 85 | Simple and optimal randomized fault-tolerant rumor spreading. Distributed Computing, 2016, 29, 89-104. | 0.8 | 2 |
| 86 | Dynamic parameter choices in evolutionary computation. , 2018, , . | | 2 |
| 87 | Fixed-target runtime analysis of the $(1 + 1)$ EA with resampling. , 2019, , . | | 2 |
| 88 | Blending Dynamic Programming with Monte Carlo Simulation for Bounding the Running Time of Evolutionary Algorithms. , 2021, , . | | 2 |
| 89 | Black-box complexity. , 2013, , . | | 1 |
| 90 | Dynamic parameter choices in evolutionary computation. , 2019, , . | | 1 |

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|-----|---|-----|-----------|
| 91 | Preface to the Special Issue on Theory of Genetic and Evolutionary Computation. <i>Algorithmica</i> , 2019, 81, 589-592. | 1.3 | 1 |
| 92 | Computing Minimum Cycle Bases in Weighted Partial 2-Trees in Linear Time. <i>Journal of Graph Algorithms and Applications</i> , 2014, 18, 325-346. | 0.4 | 1 |
| 93 | Hybridizing the 1/5-th Success Rule with Q-Learning for Controlling the Mutation Rate of an Evolutionary Algorithm. <i>Lecture Notes in Computer Science</i> , 2020, , 485-499. | 1.3 | 1 |
| 94 | Direction-reversing quasi-random rumor spreading with restarts. <i>Information Processing Letters</i> , 2013, 113, 921-926. | 0.6 | 0 |
| 95 | The Price of Anarchy for Selfish Ring Routing is Two. <i>ACM Transactions on Economics and Computation</i> , 2014, 2, 1-24. | 1.1 | 0 |
| 96 | Women@GECCO 2014. , 2014, , . | | 0 |
| 97 | 2014 Women@GECCO workshop. <i>ACM SIGEVolution</i> , 2015, 7, 26-27. | 0.5 | 0 |
| 98 | Women@GECCO 2016 Chairs' Welcome. , 2016, , . | | 0 |
| 99 | Theory for Non-Theoreticians. , 2016, , . | | 0 |
| 100 | Preface to the Special Issue on Theory of Genetic and Evolutionary Computation. <i>Algorithmica</i> , 2017, 78, 558-560. | 1.3 | 0 |
| 101 | Illustrating the trade-off between time, quality, and success probability in heuristic search. , 2019, , . | | 0 |
| 102 | MATE: A Model-Based Algorithm Tuning Engine. <i>Lecture Notes in Computer Science</i> , 2021, , 51-67. | 1.3 | 0 |
| 103 | Mutation Rate Control in the $(1+\lambda)$ Evolutionary Algorithm with a Self-adjusting Lower Bound. <i>Communications in Computer and Information Science</i> , 2020, , 305-319. | 0.5 | 0 |