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Welcome

Dear GECCO attendees,

Welcome to the Genetic and Evolutionary Computation Conference (GECCO). After the tremendously successful GECCO 2019 in Prague, Czech Republic, GECCO came back to the Americas: to Cancún, México. Unfortunately, and because of the coronavirus crisis, GECCO became, for the first time, an online conference.

GECCO is the largest peer-reviewed conference in the field of Evolutionary Computation, and the main conference of the Special Interest Group on Genetic and Evolutionary Computation (SIGEVO) of the Association for Computing Machinery (ACM). GECCO implements a rigorous and selective review process to identify the most important and technically sound papers to publish. The technical program is divided into thirteen tracks reflecting all aspects of our field and each of these tracks were chaired by two experts who managed the review process and made the decisions on the papers in coordination with the Editor-in-Chief.

This year, we received 415 papers and accepted 149 of them, which results in a 36% acceptance rate. Those papers will be presented during the online conference using live streaming. We additionally accepted 132 posters that will also be presented in two online posters sessions.

Besides the technical tracks, GECCO includes 42 tutorials selected from among 54 proposals as well as 21 workshops reflecting important topics in our field. They will take place during the first two days of the online conference, together with several competitions.

We are thrilled to welcome as keynote speakers to Leslie Valiant, from Harvard University and Kalyanmoy Deb, from Michigan State University. We are also honored that Darrell Whitley will give the SIGEVO keynote this year.

We would like to thank all authors for submitting their work to GECCO 2020. We also wish to thank all the tutorial speakers as well as to the workshop and competition organizers.

The organization of a conference like GECCO is a tremendous task relying on many people. We sincerely appreciate their work. We would like to thank first to all our chairs: track, tutorials, workshops, publicity, competitions, late breaking abstracts, and hot off the press. We also thank the organizers of the Humies, and of Women@GECCO as well as to the members of our program committee.

Some members of the organization team deserve a special mention and recognition: José A. Lozano, Editor-in-Chief, Josu Ceberio, Proceedings Chair, Arturo Hernández Aguirre, Local Chair and Gregorio Toscano Pulido, who was Publicity Chair but played a key role by developing a system that could make possible an online conference with a more friendly user interface. All of them spent long hours working to make this conference possible. We also thank Ahmed Kheiri for allowing us to use his algorithm to optimize the schedule of the conference. It is also worth mentioning the hard work done by Brenda Ramirez and Roxane Rose who helped us with the registrations and the logistics of the event. Finally, we also thank Franz Rothlauf, Enrique Alba, Darrell Whitley, Jürgen Branke and Peter Bosman from SIGEVO for their valuable advice and guidance.

On behalf GECCO, we would also like to thank Uber, Beacon and CINVESTAV-IPN for their sponsorship and support.

Enjoy the conference!

Carlos A. Coello Coello
GECCO 2020 General Chair
CINVESTAV-IPN
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CS — Complex Systems (Artificial Life/Artificial Immune Systems/Generative and Developmental Systems/Evolvable Robotics)
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DETA — Digital Entertainment Technologies and Arts
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Vanessa Volz, modl.ai

ECOM — Evolutionary Combinatorial Optimization and Metaheuristics
Francisco Chicano, University of Malaga
Luis Paquete, University of Coimbra

EML — Evolutionary Machine Learning
Jaume Bacardit, Newcastle University
Bing Xue, Victoria University of Wellington

EMO — Evolutionary Multiobjective Optimization
Jonathan Edward Fieldsend, University of Exeter
Sanaz Mostaghim, University of Magdeburg

ENUM — Evolutionary Numerical Optimization
Dirk V. Arnold, Dalhousie University
Oliver Schuetze, CINVESTAV-IPN

GA — Genetic Algorithms
Gabriela Ochoa, University of Stirling
Carlos Segura, Centro de Investigación en Matemáticas (CIMAT)

GECH — General Evolutionary Computation and Hybrids
Carlos Cotta, University of Málaga
Michael T. M. Emmerich, LIACS

GP — Genetic Programming
Miguel Nicolau, University College Dublin
Mengjie Zhang, Victoria University of Wellington

RWA — Real World Applications
Robin Purshouse, University of Sheffield
Tapabrata Ray, University of New South Wales

SBSE — Search-Based Software Engineering
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Bilel Derbel, Univ. Lille, Inria Lille - Nord Europe
Travis Desell, Rochester Institute of Technology
Andre Deutz, Leiden University
Grant Dick, University of Otago, Information Science Dept.
Quốc Bảo Diệp, VSB-TU Ostrava
Federico Divina, Pablo de Olavide University
Alexander Dockhorn, Otto-von-Guericke University Magdeburg
Karl F. Doerner, University of Vienna
Carola Doerr, CNRS, Sorbonne University
Benjamin Doerr, Ecole Polytechnique, Laboratoire d’Informatique (LIX)
Carola Doerr, CNRS, Sorbonne University
Emily L. Dolson, Michigan State University
Alan Dorin, Monash University
Bernabe Dorronsoro, University of Cadiz
Erik Dowgan, Jozef Stefan Institute
Jan Drchal, Department of Computer Science and Engineering, Faculty of Electrical Engineering, Czech Technical University
Rafał Drezewski, AGH University of Science and Technology
Madalina Drugan, IT.Learns.Online
Wei Du, East China University of Science and Technology
Juan Durillo, Leibniz Supercomputing Center of the Bavarian Academy of Science and Humanities
Joao Duro, The University of Sheffield
Richard Duro, Universidade da Coruña
Marc Ebner, Ernst-Moritz-Universität Greifswald, Germany
Tome Eftimov, Jožef Stefan Institute, Stanford University
A.E. Eiben, VU University Amsterdam, Vrije Universiteit Amsterdam
Aniko Ekart, Aston University
Mohammed El-Abd, American University of Kuwait
Michael T. M. Emmerich, LIACS
Andries P. Engelbrecht, Stellenbosch University
Michael Epitropakis, The Signal Group
Anton V. Eremeev, Omsk Branch of Sobolev Institute of Mathematics, The Institute of Scientific Information for Social Sciences RAS
Rafael Ertel, University of Exeter
Richard Everson, University of Exeter
Zhun Fan, STU
Liang Feng, Chongqing University
Lavinia Ferariu, “Gheorghe Asachi” Technical University of
Cristian Ramirez, Otto von Guericke Universität
Jeremy Rapin, Facebook
Khaled Rasheed, University of Georgia
Tapabrata Ray, University of New South Wales
Margarita Rebollo Coy, TH Köln, VU University Amsterdam
Patrick M. Reed, Cornell University
Frederik Rehbach, TH Køeln
Gustavo Reis, School of Technology and Management, Polytechnic Institute of Leiria, Portugal
Lewis Rydh, PC Member
Maria Cristina Riff, UTFSM
Sebastian Risi, IT University of Copenhagen
Marcus Ritt, Federal University of Rio Grande do Sul
Peter Rockett, University of Sheffield
Eduardo Rodríguez-Tello, CINVESTAV - Tamaulipas
Nicolas Rojas-Morales, Universidad Técnica Federico Santa María
Andrea Roli, Alma Mater Studiorum Università di Bologna
Alejandro Rosales-Pérez, Centro de Investigación en Matemáticas
Nick Ross, University of Exeter
Franz Rothlauf, University of Mannz
Jonathon Rowe, University of Birmingham
Proteek Chandan Roy, Michigan State University
Guenther Rudolph, TU Dortmund University
Ruben Ruiz, Polytechnic University of Valencia
Thomas Runkler, Siemens AG, Technical University of Munich
Conor Ryan, University of Limerick, Lero - The Irish Software Research Centre
Rubén Saborido, University of Malaga
Ali Safari Khatouni, Dalhousie University
Houari Saahraoui, DIBO, Université de Montréal
Naoki Sakamoto, University of Tsukuba, RIKEN AIP
Carolina Salto, Fac. de Ingeniería - UNLPam - Argentina
Luciano Sanchez, Universidad de Oviedo
Nayat Sanchez-Pi, Inria Chile
Nicholas Sanders, University of Exeter
Roberto Santana, Universidade de Ouro Preto, Brazil
Valentino Santucci, University of Perugia
Yuji Sato, Hosei University
Hiroyuki Sato, The University of Electro-Communications
Fédéric Saubion, University of Angers, France
Saket Saurabh, IMSc, University of Bergen
Umerto Scotar, ICAR-CNR
Robert Schaefer, AWH University of Science and Technology
Sebastian Schmitt, Honda Research Institute Europe GmbH
Manuel Schmitt, University of Erlangen-Nuremberg
Jacob Schrum, Department of Mathematics and Computer Science, Southwestern University
Oliver Schuetze, CINVESTAV-México
Marco Scirea, University of Southern Denmark
Eric Scott, George Mason University, MITRE
Ignacio Segovia-Domínguez, Center for Research in Mathematics
Eduardo Segredo, Universidad de La Laguna, Edinburgh Napier University
Carlos Segura, Centro de Investigación en Matemáticas (CIMAT)
Lukas Sekanina, Brno University of Technology, Czech Republic
Sevil Sen, Hacettepe University, Turkey
Bernhard Sendhoff, Honda Research Institute Europe
Roman Senkerik, Tomas Bata University
Kevin Seppi, Brigham Young University
Cameron Shand, University of Manchester
Ke Shang, Southern University of Science and Technology, Southeast University
Feng Shi, Central South University
Koji Shimoyama, Tohoku University
Shinichi Shirakawa, Yokohama National University
Sara Silva, FCUL
Kevin Sim, Edinburgh Napier University
Anabela Simões, DEIS/ISEC - Coimbra Polytechnic
Lana Sinapayen, Sony Computer Science Laboratories, Inc.; Earth Life Sciences Institute
Hemant K. Singh, University of New South Wales
Moshe Sipper, Ben-Gurion University
Robert Smith, Dalhousie University
Stephen L. Smith, University of York
Joao Soares, GECAD, Polytechnic of Porto
Dominik Sobania, University of Mainz
Christine Solnon, LIRIS, INSA de Lyon
Oliver Somnin, INSA Lyon, Inria
Andy song, RMIT University
Lisa Soros, University of Central Florida
Victor Adrian Sosa Hernandez, ITESM-CEM
Lee Specter, Hampshire College; Amherst College, UMass Amherst
Patrick Spettel, Vorarlberg University of Applied Sciences
Ambuj Kumar Srivastava, The University of Sheffield
Anthony Stein, University of Hohenheim
Thomas Stibor, GSI Helmholtz Centre for Heavy Ion Research
Sebastian Stich, École polytechnique fédérale de Lausanne
Catalin Stoean, University of Craiova, Romania
Daniel H. Stolfi, Instituto de Matemática e Estatística, TH Köln
Jörg Stork, Institute of Data Science, Engineering, and Analytics, TH Köln
Thomas Stützle, Université Libre de Bruxelles
Dirk Sudholt, University of Sheffield
Ponnuthurai Suganthan, NTU
Chaoli Sun, Taiyuan University of Science and Technology
Yanan Sun, Sichuan University, Victoria University of Wellington
Andrew M. Sutton, University of Minnesota Duluth
Reiji Suzuki, Nagoya University
Jerry Swan, INAI-SENSE SA, University of York
Keiki Takadama, The University of Electro-Communications
Ricardo Takahashi, Universidade Federal de Minas Gerais
Ryoji Tanabe, Yokohama National University
Ivan Tanev, Faculty of Engineering, Doshisha University
Ernesto Tarantino, ICAR - CNR
Danesh Tarapore, University of Southampton
Sara Tari, University of Lille
Tomoaki Tatsuoka, Tokyo University of Science
Daniel R. Tauritz, Auburn University
Roberto Tavares, Federal University of Sao Carlos
Jürgen Teich, Friedrich-Alexander-Universität Erlangen-Nürnberg (FAU)
Hugo Terashima-Marín, Tecnológico de Monterrey
German Terrazas Angulo, University of Cambridge
Andrea G. B. Tettamanzi, Université Côte d’Azur
Johannes Textor, University of Utrecht
Olivier Teytaud, TAO, Inria
Dirk Thieren, Utrecht University
Dhananjay Thingray, Deakin University
Spencer Thomas, National Physical Laboratory
Sarah Thomson, University of Stirling
Jie Tian, Taiyuan University of Science and Technology
Renato Tinos, University of São Paulo
Julian Togelius, New York University
Lukas Tomaszek, VSB-TUO
Sven Tomforde, Kiel University
Jim Torresen, University of Oslo
Gregorio Toscano, Cinevest-IPN
Cheikh Toure, École Polytechnique, Inria
Cao Tran, Victoria University of Wellington
Heike Trautmann, University of Münster
Isaac Triguero, University of Nottingham
Krzysztof Trojanowski, Cardinal Stefan Wyszyński University in Warsaw
Leonardo Trujillo, Instituto Tecnológico de Tijuana
Giuseppe A. Trunfio, University of Sassari
Thanh Cong Truong, VSB - Technical University of Ostrava
Tea Tusar, Jozef Stefan Institute
Cem C. Tutum, University of Texas at Austin
Kyungpook University, Kyungpook National University
Paulo Urbano, University of Lisbon
Ryan Urbanowicz, University of Pennsylvania
Neil Urrutia, Edinburgh Napier University
Zita Vale, Politecnico di Porto
Pablo Valledor, ArcelorMittal
Koen van der Blom, Leiden University
Sander van Rijn, LIACS, Leiden University
Bas van Stein, LIACS, Leiden University
Leonardo Vanneschi, ISEGI, Universidade Nova de Lisboa
Konstantinos Varelas, Inria Saclay Ile-de-France, Thales LAS France SAS
Danilo Vasconcellos Vargas, Kyushu University
Zdenek Vasek, Brno University of Technology
Vassilis Vassiliades, Research Centre on Interactive Media, Smart Systems and Emerging Technologies; University of Cyprus
Igor Vatolkin, TU Dortmund
Neil Vaughan, University of Chester, Royal Academy of Engineering (RAEng)
Frank Veenstra, IT University of Copenhagen
Nadarajen Veerapan, Université de Lille
Sebastian Ventura, Universidad de Cordoba
Sebastien Verel, Université du Littoral Côte d’Opale
Silvia Vergilio, Federal University of Paraná
Ana Viana, INESC TEC/Polytechnic of Porto
Petra Vlinderova, Institute of Computer Science of ASCR
Adam Viktorin, Tomas Bata University in Zlín
Marco Villani, University of Modena and Reggio Emilia, Italy
Alijosa Vodopija, Jozef Stefan Institute, Jozef Stefan International Postgraduate School
Thomas Vogel, Humboldt University Berlin
Vanessa Voltz, modl.ai
Fernando J. Von Zuben, Unicamp
Vida Vukasinovic, JSI
Stefan Wagner, University of Applied Sciences Upper Austria
Neal Wagner, Systems & Technology Research
Markus Wagner, School of Computer Science, The University of Adelaide
David Walker, University of Plymouth
Feng Wang, Wuhan University, School of Computer Science
Handing Wang, Xidian University
Hao Wang, Sorbonne Université, CNRS
Rui Wang, Department of Automatic Control and Systems Engineering, University of Sheffield; National University of Defense Technology
Rui Wang, Uber AI Labs
Shuai Wang, Xidian University
Xilu Wang, Miss
Yifei Wang, Georgia Tech
Elizabeth Wanner, Aston University
Jaroslaw Was, AGH University of Science and Technology, Poland
Thomas Weise, University of Science and Technology of China (USTC), School of Computer Science and Technology
Simon Wessing, Technische Universität Dortmund
Peter Alexander Whigham, Univ. of Otago, Information Science Dept.
David White, University of Sheffield
Darrell Whitley, Colorado State University
Pawel Widera, Newcastle University
Josh Wilkerson, NAVAIR
Stewart W. Wilson, Prediction Dynamics
Dennis Wilson, ISAE-SUPAERO
Garnett Wilson, Dalhousie University
Stephan Winkler, University Of Applied Sciences Upper Austria
Carsten Witt, Technical University Of Denmark
Man Leung Wong, Lingnan University, Hong Kong
John Woodward, Queen Mary University of London
Guohua Wu, National University of Defense Technology
Huayang Xie, Oracle New Zealand
Yang Xin-Sheng, Middlesex University
Ning Xiong, Malardalen University
Hang Xu, Victoria University of Wellington, Putian University
Bing Xue, Victoria University of Wellington
Takeshi Yamada, NTT Communication Science Labs.
Cuie Yang, Hong Kong Baptist University
Kai Feng Yang, Leiden University
Georgios N. Yannakakis, Institute of Digital Games, University of Malta, Malta; University of Malta
Donya Yazdani, University of Sheffield
Iryna Yevseyeva, De Montfort University
Shin Yoo, Korea Advanced Institute of Science and Technology
Tian-Li Yu, Department of Electrical Engineering, National Taiwan University
Guo Yu, Department of Computer Science of University of Surrey
Yuan Yuan, Michigan State University
Cai Tong Yue, Zhengzhou University
Shiu Yin Yuen, City University of Hong Kong
Martin Zaefferer, TH Köln
Amelia Zafra, University of Cordoba
Saúl Zapotecas-Martínez, Universidad Autónoma Metropolitana (UAM)
Christine Zarges, Aberystwyth University
Ivan Zelinka, VSB-TU Ostrava
Jan Zenisek, University of Applied Sciences Upper Austria
Zhi-Hui Zhan, South China University of Technology
Qingfu Zhang, City University of Hong Kong, City University of Hong Kong Shenzhen Research Institute
Mengjie Zhang, Victoria University of Wellington
Xingyi Zhang, Anhui University
Weijie Zheng, Southern University of Science and Technology, University of Science and Technology of China
Jinghui Zhong, South China University of Technology
Aimin Zhou, Department of Computer Science and Technology, East China Normal University
Yan Zhou, Northeastern University
Heiner Zille, Otto-von-Guericke University Magdeburg
Nur Zincir-Heywood, Dalhousie University
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Time Zone

The schedule of GECCO 2020 appears in Cancún's time zone (UTC/GMT-5 hours), which, in July, is equivalent to US Central time. For your convenience, we provide here the equivalent time in some selected major cities around the world:

- For New York (USA), please add 1 hour.
- For Toronto (Canada), please add 1 hour.
- For Chicago/Houston (USA), consider the exact same time.
- For Los Angeles/San Francisco (USA), please subtract 2 hours.
- For Vancouver (Canada), please subtract 2 hours.
- For Paris (France)/Madrid (Spain)/Frankfurt (Germany), please add 7 hours.
- For London (UK)/Lisbon (Portugal), please add 6 hours.
- For Sao Paulo (Brazil), please add 2 hours.
- For New Delhi (India), please add 10:30 hours.
- For Shanghai/Beijing (China), please add 13 hours.
- For Taipei (Taiwan), please add 13 hours.
- For Tokyo (Japan), please add 14 hours.
- For Johannesburg/Cape Town (South Africa), please add 7 hours.
- For Wellington (New Zealand), please add 17 hours.
Schedule
## Schedule at a Glance

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<td><strong>Tutorials, Workshops, and Competitions</strong></td>
<td><strong>INVITED KEYNOTE</strong></td>
<td><strong>INVITED KEYNOTE</strong></td>
<td><strong>Paper Sessions</strong></td>
</tr>
<tr>
<td>08:30-10:20</td>
<td>08:30-10:20</td>
<td><strong>Leslie Valiant</strong></td>
<td><strong>Kalyanmoy Deb</strong></td>
<td>09:00-10:40</td>
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<td><strong>Break</strong></td>
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<tr>
<td><strong>Tutorials and Workshops</strong></td>
<td><strong>Tutorials, Workshops, and Competitions</strong></td>
<td><strong>Paper Sessions and HOP</strong></td>
<td><strong>Poster Session 2</strong></td>
<td><strong>SIGEVO Keynote</strong></td>
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<td>10:40-12:30</td>
<td>10:40-12:30</td>
<td>10:30-12:10</td>
<td>10:30-12:10</td>
<td><strong>Darrell Whitley</strong></td>
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<tr>
<td><strong>Lunch break</strong></td>
<td><strong>Women@GECCO</strong></td>
<td><strong>Job Market</strong></td>
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<td><strong>Lunch break</strong></td>
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<tr>
<td>12:30-14:00</td>
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<td>12:10-13:40</td>
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<tr>
<td><strong>Tutorials and Workshops</strong></td>
<td><strong>Tutorials and Workshops</strong></td>
<td><strong>Paper Sessions, HOP and HUMIES</strong></td>
<td><strong>Paper Sessions</strong></td>
<td><strong>Paper Sessions</strong></td>
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<tr>
<td><strong>Tutorials and Workshops</strong></td>
<td><strong>Tutorials, Workshops, and LBA</strong></td>
<td><strong>Paper Sessions</strong></td>
<td><strong>Paper Sessions</strong></td>
<td><strong>Poster Session 1</strong></td>
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<tr>
<td>16:10-18:00</td>
<td>16:10-18:00</td>
<td>15:40-17:20</td>
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<td>17:40-19:20</td>
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<td><strong>Poster Session 1</strong></td>
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<td>17:40-19:20</td>
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### Workshop and Tutorial Sessions, Wednesday, July 8

<table>
<thead>
<tr>
<th>Time</th>
<th>Tutorials</th>
<th>Workshops</th>
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<tbody>
<tr>
<td>08:30-10:20</td>
<td>EAPwU — Evolutionary Algorithms for Problems with Uncertainty</td>
<td>GI: Taking real-world source code and improving it using genetic programming</td>
</tr>
<tr>
<td></td>
<td>Filipic, Tušar</td>
<td>Haraldsson, Woodward, Wagner, Alexander</td>
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<tr>
<td></td>
<td>NEWK — Neuroevolution at work</td>
<td>Evolutionary Computation for Digital Art</td>
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<td>Neumann, Neumann</td>
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<tr>
<td>10:40-12:30</td>
<td>Visualization in Multiobjective Optimization</td>
<td>SAEOpt — Surrogate-Assisted Evolutionary Optimisation</td>
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<tr>
<td>14:00-15:50</td>
<td>Introductory Mathematical Programming for EC</td>
<td>Evolutionary Multi-objective Optimization: Past, Present and Future</td>
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<td></td>
<td>Shir</td>
<td>Deb</td>
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<tr>
<td>16:10-18:00</td>
<td>EvoSoft — Evolutionary Computation Software Systems</td>
<td>Evolutionary Many-Objective Optimization</td>
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<td>Ishibuchi, Sato</td>
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<td></td>
<td>ECPERM — Evolutionary Computation for Permutation Problems</td>
<td>iGECCO — Interactive Methods @ GECCO</td>
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<td>Lehre, Oliveto</td>
<td>Stein, Nakata</td>
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<tr>
<td>14:00-15:50</td>
<td>Introduction to Genetic Programming</td>
<td>Evolution of Neural Networks</td>
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<td></td>
<td>O’Reilly, Hemberg</td>
<td>Miikkulainen</td>
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<td>Evolutionary Computation: A Unified Approach</td>
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<td>De Jong</td>
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<td>Model-Based Evolutionary Algorithms</td>
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<td>Thierens, Bosman</td>
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<td>Evolutionary Computation and Games</td>
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<td></td>
<td>Fitness landscape analysis to understand and predict alg. performance for single- and multi-objective opt.</td>
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<td>Verel, Derbel</td>
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<td>EvoSoft — Evolutionary Computation Software Systems</td>
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<td></td>
<td>PDEIM — Parallel and Distributed Evolutional Inspired Methods</td>
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<td>GreenAI — Evolutionary and machine learning solutions in environment, renewable and ecologically-aware scenarios</td>
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<td>Quality-Diversity Optimization</td>
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<td>Cully, Mouret, Doncieux</td>
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<td>VizGEC 2020 — Visualisation Methods in Genetic and Evolutionary Computation</td>
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<td>IAM — Industrial Application of Metaheuristics</td>
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<td></td>
<td>Representations for Evolutionary Algorithms</td>
<td>A Gentle Introduction to Theory (for Non-Theoreticians)</td>
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<td>Rothlauf</td>
<td>Doerr</td>
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<td>IAM — Industrial Application of Metaheuristics</td>
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# Workshop and Tutorial Sessions, Thursday, July 9

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<tr>
<td>08:30-10:20</td>
<td>Decomposition Multi-Objective Optimisation: Current Developments and Future Opportunities Li, Zhang, Zapotecas</td>
<td>Sequential Experimentation by Evolutionary Algorithms Shir, Bäck</td>
<td>Addressing Ethical Challenges within Evolutionary Computation Applications Torresen</td>
<td>LBA – Late-Breaking Abstracts p. 51</td>
</tr>
<tr>
<td>14:00-15:50</td>
<td>Dynamic Control Parameter Choices in Evolutionary Computation Papa, Doerr</td>
<td>Solving Complex Problems with Coevolutionary Algorithms Krawiec, Heywood</td>
<td>Evol. Comp. and Deep Learning for Image Analysis, Signal Processing and Pattern Recognition Zhang, Cagnoni</td>
<td>Evolutionary Algorithms in Biomedical Data Mining: Challenges, Solutions, and Frontiers Urbanowicz, Sipper</td>
</tr>
<tr>
<td>16:10-18:00</td>
<td>Search Based Software Engineering: challenges, opportunities and recent applications Ouni</td>
<td>Recent Advances in Particle Swarm Optimization Analysis and Understanding Engelbrecht, Cleghorn</td>
<td>SWINGA — Swarm Intelligence Algorithms: Foundations, Perspectives and Challenges p. 57</td>
<td>Benchmarking and Analyzing Iterative Optimization Heuristics with IOHprofiler Wang, Doerr, Shir, Bäck</td>
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<td>Competitions</td>
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<td></td>
<td>Automated Algorithm Configuration and Design López-Ibáñez, Stutzle</td>
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<td></td>
<td>MedGEC — Workshop on Medical Applications of Genetic and Evolutionary Computation p. 48</td>
<td>Design Principles for Matrix Adaptation Evolution Strategies Beyer</td>
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<tr>
<td></td>
<td>Tutorials</td>
<td>Workshops</td>
<td>Competitions</td>
<td>LBA session</td>
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<tr>
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<td>IWLCs — International Workshop on Learning Classifier Systems</td>
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<tr>
<td></td>
<td>Tutorials</td>
<td>Workshops</td>
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<td>LBA session</td>
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## Parallel Sessions, Friday, July 10 through Sunday, July 12

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- **GECH2**
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### Notes

- `☆` Sessions with best paper nominees
- **HUMIES**
- **HOP**

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*Schedule*
Track List and Abbreviations

**ACO-SI:** Ant Colony Optimization and Swarm Intelligence

**CS:** Complex Systems (Artificial Life / Artificial Immune Systems / Generative and Developmental Systems / Evolutionary Robotics / Evolvable Hardware)

**DETA:** Digital Entertainment Technologies and Arts

**ECOM:** Evolutionary Combinatorial Optimization and Metaheuristics

**EML:** Evolutionary Machine Learning

**EMO:** Evolutionary Multiobjective Optimization

**ENUM:** Evolutionary Numerical Optimization

**GA:** Genetic Algorithms

**GECH:** General Evolutionary Computation and Hybrids

**GP:** Genetic Programming

**HUMIES:** Annual “Humies” Awards For Human-Competitive Results

**HOP:** Hot Off the Press

**RWA:** Real World Applications

**SBSE:** Search-Based Software Engineering

**THEORY:** Theory
Keynotes
Living organisms function according to protein circuits. Darwin’s theory of evolution suggests that these circuits have evolved through variation guided by natural selection. However, it is currently unknown what variation mechanisms can give rise to protein circuits of the complexity found in biology, within realistic population sizes and realistic numbers of generations.

We suggest that computational learning theory offers the framework for investigating this question, of how circuits can come into being via a Darwinian process without a designer. We formulate evolution as a form of learning from examples. The targets of the learning process are the protein expression functions that come closest to best behavior in the specific environment. The learning process is constrained so that the feedback from the experiences is Darwinian. We formulate a notion of evolvability that distinguishes function classes that are evolvable with polynomially bounded resources from those that are not. The dilemma is that if the function class that describes the expression levels of proteins in terms of each other, is too restrictive, then it will not support biology, while if it is too expressive then no evolution algorithm will exist to navigate it. We shall review current work in this area.

Biosketch: Leslie Valiant was educated at King’s College, Cambridge; Imperial College, London; and at Warwick University where he received his Ph.D. in computer science in 1974. He is currently T. Jefferson Coolidge Professor of Computer Science and Applied Mathematics in the School of Engineering and Applied Sciences at Harvard University, where he has taught since 1982. Before coming to Harvard he had taught at Carnegie Mellon University, Leeds University, and the University of Edinburgh.

His work has ranged over several areas of theoretical computer science, particularly complexity theory, learning, and parallel computation. He also has interests in computational neuroscience, evolution and artificial intelligence and is the author of two books, Circuits of the Mind, and Probably Approximately Correct.

He received the Nevanlinna Prize at the International Congress of Mathematicians in 1986, the Knuth Award in 1997, the European Association for Theoretical Computer Science EATCS Award in 2008, and the 2010 A. M. Turing Award. He is a Fellow of the Royal Society (London) and a member of the National Academy of Sciences (USA).
Evolutionary Computation's Niche for Solving Multi-Criterion Optimization Problems

Kalyanmoy Deb, Michigan State University, United States

Evolutionary computation (EC) involves a careful collaborative and iterative update of a population of solutions to reach near a desired target. In a single-objective optimization problem, the respective optimal solution is often the single target. In a multi-criterion optimization problem, the target is a set of Pareto-optimal solutions. Although EC field started with solving single-objective problems, EC researchers soon realized that they were ideal for finding a well-diversified set of multiple Pareto-optimal solutions simultaneously for multi-criterion optimization problems, a clear niche of EC compared to their point-based classical counterparts. In this keynote talk, we provide a brief chronology of events on the evolutionary multi-criterion optimization (EMO) field in the past almost three decades, key challenges it faced, and key events and publications which pushed the field forward. Moreover, we shall provide a brief account of the current activities and speaker's view of what lies ahead in this talk.

Biosketch: Kalyanmoy Deb is Koenig Endowed Chair Professor at Department of Electrical and Computer Engineering in Michigan State University. His research interests are in evolutionary optimization and their application in multi-criterion optimization, bilevel optimization, modeling, and machine learning. He was awarded IEEE CIS EC Pioneer award, Infosys Prize, TWAS Prize in Engineering Sciences, CajAstur Mandani Prize, Distinguished Alumni Award from IIT Kharagpur, Edgeworth-Pareto award, Bhatnagar Prize in Engineering Sciences, and Bessel Research award from Germany. He is fellow of IEEE and ASME. He has published over 520 research papers with Google Scholar citation of over 137,000 with h-index 115. More information can be found from http://www.coin-lab.org.

Removing Randomness from Evolutionary Algorithms

Darrell Whitley, Colorado State University, United States

It is natural to think of Evolutionary Algorithms as highly stochastic search methods. This can also make Evolutionary Algorithms, and particularly recombination, quite difficult to analyze. One way to reduce randomness involves the quadratization of functions, which is commonly used by modern optimization methods, and also has applications in quantum computing. After a function is made quadratic, random mutation is obsolete and unnecessary; the location of improving moves can be calculated deterministically, on average in $O(1)$ time. Seemingly impossible problems, such as the Needle-in-a-Haystack, becomes trivial to solve in quadratic form. One can also probably tunnel, or jump, between local optima and quasi-local optima in $O(n)$ time using deterministic genetic recombination. The talk also explores how removing randomness from Evolutionary Algorithms might provide new insights into natural evolution. Finally, a form of evolutionary algorithm is proposed where premature convergence is impossible and the evolutionary potential of the population remains open-ended.

Biosketch: Darrell Whitley is a Professor of Computer Science at Colorado State University. He served as the Chair of the International Society of Genetic Algorithm from 1993 to 1997, and as the Editor-in-Chief of the journal Evolutionary Computation from 1997 to 2003. He was Chair of the Governing Board of ACM SIGEVO from 2007 to 2011. He was named an ACM Fellow in 2019 for his contributions to the field of genetic and evolutionary computation.
Tutorials
**Introductory Tutorials**

**Hyper-heuristics**
John R. Woodward, Queen Mary University of London  
Daniel R. Tauritz, Missouri University of Science and Technology  
Wednesday, July 8, 08:30-10:20

**Evolutionary Computation: A Unified Approach**
Kenneth De Jong, George Mason University  
Wednesday, July 8, 08:30-10:20

**Runtime Analysis of Evolutionary Algorithms: Basic Introduction**
Per Kristian Lehre, University of Birmingham  
Pietro S. Oliveto, University of Sheffield  
Wednesday, July 8, 08:30-10:20

**Introductory Mathematical Programming for EC**
Ofer M. Shir, Tel-Hai College  
Wednesday, July 8, 10:40-12:30

**Model-Based Evolutionary Algorithms**
Dirk Thierens, Utrecht University  
Peter A. N. Bosman, Centrum Wiskunde & Informatica (CWI)  
Wednesday, July 8, 10:40-12:30

**Introduction to Genetic Programming**
Una-May O’Reilly, CSAIL, Massachusetts Institute of Technology  
Erik Hemberg, Massachusetts Institute of Technology  
Wednesday, July 8, 10:40-12:30

**Representations for Evolutionary Algorithms**
Franz Rothlauf, Universität Mainz  
Wednesday, July 8, 10:40-12:30

**Evolutionary Multi-objective Optimization: Past, Present and Future**
Kalyanmoy Deb, Michigan State University  
Wednesday, July 8, 14:00-15:50

**Evolutionary Computation and Games**
Julian Togelius, New York University  
Sebastian Risi, IT University of Copenhagen  
Georgios N. Yannakakis, University of Malta  
Wednesday, July 8, 14:00-15:50

**Evolution of Neural Networks**
Risto Miikkulainen, The University of Texas at Austin  
Wednesday, July 8, 14:00-15:50

**A Gentle Introduction to Theory (for Non-Theoreticians)**
Benjamin Doerr, Ecole Polytechnique  
Wednesday, July 8, 14:00-15:50

**Evolutionary Many-Objective Optimization**
Hisao Ishibuchi, Southern University of Science Technology (SUSTech)  
Hiroyuki Sato, The University of Electro-Communications  
Wednesday, July 8, 16:10-18:00

**Learning Classifier Systems: From Principles to Modern Systems**
Anthony Stein, University of Hohenheim  
Masaya Nakata, Yokohama National University  
Wednesday, July 8, 16:10-18:00

**Neuroevolution for Deep Reinforcement Learning Problems**
David Ha, Google Brain  
Wednesday, July 8, 16:10-18:00
# Advanced Tutorials

**Visualization in Multiobjective Optimization**
Bogdan Filipic, Jozef Stefan Institute  
Tea Tušar, Jozef Stefan Institute  
Wednesday, July 8, 10:40-12:30

**Quality-Diversity Optimization**
Antoine Cully, Imperial College  
Jean-Baptiste Mouret, Inria  
Stéphane Doncieux, Sorbonne Université  
Wednesday, July 8, 10:40-12:30

**Genetic improvement: Taking real-world source code and improving it using genetic programming**
Saemundur O. Haraldsson, University of Stirling  
John R. Woodward, Queen Mary University of London  
Markus Wagner, The University of Adelaide  
Bradley Alexander, The University of Adelaide  
Wednesday, July 8, 14:00-15:50

**Fitness landscape analysis to understand and predict algorithm performance for single- and multi-objective optimization**
Sébastien Verel, Univ. Littoral Côte d’Opale  
Bilel Derbel, Univ. Lille  
Wednesday, July 8, 14:00-15:50

**Theory and Practice of Population Diversity in Evolutionary Computation**
Dirk Sudholt, University of Sheffield  
Giovanni Squillero, Politecnico di Torino  
Wednesday, July 8, 14:00-15:50

**Evolutionary Computation for Digital Art**
Aneta Neumann, The University of Adelaide  
Frank Neumann, The University of Adelaide  
Wednesday, July 8, 16:10-18:00

**Decomposition Multi-Objective Optimisation: Current Developments and Future Opportunities**
Ke Li, University of Exeter  
Qingfu Zhang, City University of Hong Kong  
Saúl Zapotecas-Martínez, Autonomous Metropolitan University  
Thursday, July 9, 08:30-10:20

**Dynamic Control Parameter Choices in Evolutionary Computation**
Gregor Papa, Jozef Stefan Institute  
Carola Doerr, CNRS  
Thursday, July 9, 08:30-10:20

**Semantic Genetic Programming**
Alberto Moraglio, University of Exeter  
Krzysztof Krawiec, Poznan University of Technology  
Thursday, July 9, 08:30-10:20

**Sequential Experimentation by Evolutionary Algorithms**
Ofer M. Shir, Tel-Hai College  
Thomas Bäck, Leiden University  
Thursday, July 9, 10:40-12:30
Solving Complex Problems with Coevolutionary Algorithms
Krzysztof Krawiec, Poznan University of Technology
Malcolm Heywood, Dalhousie University
Thursday, July 9, 10:40-12:30

Recent Advances in Particle Swarm Optimization Analysis and Understanding
Andries Petrus Engelbrecht, Stellenbosch University
Christopher W. Cleghorn, University of Pretoria
Thursday, July 9, 10:40-12:30

Statistical Analyses for Meta-heuristic Stochastic Optimization Algorithms
Tome Eftimov, Jozef Stefan Institute
Peter Korošec, Jozef Stefan Institute
Thursday, July 9, 14:00-15:50

Design Principles for Matrix Adaptation Evolution Strategies
Hans-Georg Beyer, Vorarlberg University of Applied Sciences
Thursday, July 9, 14:00-15:50

A Hands-on Guide to Distributed Computing Paradigms for Evolutionary Computation
Rui Wang, Uber AI
Jiale Zhi, Uber AI
Thursday, July 9, 16:10-18:00

Benchmarking and Analyzing Iterative Optimization Heuristics with IOHprofiler
Hao Wang, Sorbonne Université
Carola Doerr, Sorbonne Université
Ofer M. Shir, Tel-Hai College
Thomas Bäck, Leiden University
Thursday, July 9, 16:10-18:00

Specialized Tutorials

Addressing Ethical Challenges within Evolutionary Computation Applications
Jim Torresen, University of Oslo
Thursday, July 9, 08:30-10:20

Search Based Software Engineering: challenges, opportunities and recent applications
Ali Ouni, Ecole de technologie supérieure
Thursday, July 9, 08:30-10:20

Automated Algorithm Configuration and Design
Manuel López-Ibáñez, University of Manchester
Thomas Stutzle, Université Libre de Bruxelles
Thursday, July 9, 08:30-10:20

Evolutionary Computer Vision
Gustavo Olague, CICESE
Thursday, July 9, 10:40-12:30

Theory of Estimation-of-Distribution Algorithms
Carsten Witt, Technical University Of Denmark
Thursday, July 9, 10:40-12:30
**Evolutionary Algorithms and Machine Learning: synergies and challenges**
Giovanni Squillero, *Politecnico di Torino*
Alberto Tonda, INRAe

**Thursday, July 9, 10:40-12:30**

**Evolutionary Computation and Machine Learning in Cryptology**
Stjepan Picek, *Delft University of Technology*
Domagoj Jakobovic, *University of Zagreb*

**Thursday, July 9, 14:00-15:50**

**Swarm Intelligence in Cybersecurity**
Ivan Zelinka, *VSB-TU Ostrava*
Roman Senkerik, *UTB in Zlin*

**Thursday, July 9, 14:00-15:50**

**Evolutionary Computation and Evolutionary Deep Learning for Image Analysis, Signal Processing and Pattern Recognition**
Mengjie Zhang, *Victoria University of Wellington*
Stefano Cagnoni, *University of Parma*

**Thursday, July 9, 14:00-15:50**

**Evolutionary Algorithms in Biomedical Data Mining: Challenges, Solutions, and Frontiers**
Ryan J. Urbanowicz, *University of Pennsylvania*
Moshe Sipper, *Ben-Gurion University*

**Thursday, July 9, 16:10-18:00**

**Push**
Lee Spector, *Amherst College*

**Thursday, July 9, 16:10-18:00**

**Evolutionary Computation for Feature Selection and Feature Construction**
Bing Xue, *Victoria University of Wellington*
Mengjie Zhang, *Victoria University of Wellington*

**Thursday, July 9, 16:10-18:00**
Workshops,
Late Breaking Abstracts,
and Women@GECCO
EAPwU — Evolutionary Algorithms for Problems with Uncertainty

Organizers: Jonathan E. Fieldsend (University of Exeter); Ozgur Akman (University of Exeter); Khulood Alyahya (University of Exeter); Jürgen Branke (University of Warwick)

Time: Wednesday, July 8, 8:30 - 10:20

Solution Approaches for the Dynamic Stacking Problem
Sebastian Raggl, Andreas Beham, Stefan Wagner, Michael Affenzeller

Invited talk: Genetic Programming Hyperheuristics in Dynamic Scheduling
Domagoj Jakobovic, Marko Durasević

Invited talk: Evolutionary Computation for Dynamic Multi-objective Optimisation Problems
Shengxiang Yang

NEWK — Neuroevolution at work

Organizers: Ivanoe de Falco (ICAR CNR); Antonio Della Cioppa (University of Salerno); Umberto Scafuri (ICAR CNR); Ernesto Tarantino (ICAR CNR)

Time: Wednesday, July 8, 8:30 - 10:20

Neuro Evolutional with Game-Driven Cultural Algorithms
Faisal Waris, Robert G. Reynolds

Learning to Walk - Reward Relevance within an Enhanced Neuroevolution Approach
Ivan Colucci, Giuseppe Pellegrino, Antonio Della Cioppa, Angelo Marcelli

Multi-Objective Evolutionary GAN
Marco Biaioletti, Carlos Artemio Coello Coello, Gabriele Di Bari, Valentina Poggioni

GEVO-ML: Optimizing Machine Learning Codes with Evolutionary Computation
Jhe-Yu Liou, Xiaodong Wang, Stephanie Forrest, Carole-Jean Wu

Mutational Puissance Assisted Neuroevolution
Divya D. Kulkarni, Shivashankar B. Nair

EvoSoft — Evolutionary Computation Software Systems

Organizers: Michael Affenzeller (University of Applied Science Upper Austria); Stefan Wagner (University of Applied Sciences Upper Austria)

Time: Wednesday, July 8, 8:30 - 12:30

Welcome & Opening
Workshop chairs

Open Source Evolutionary Structured Optimization
Jérémy Rapin, Pauline Bennet, Emmanuel Centeno, Daniel Haziza, Antoine Moreau, Olivier Teytaud

GA-lapagos, an Open-Source C Framework including a Python-based System for Data Analysis
Peter Jamieson, José Augusto M. Nacif, Ricardo Ferreira
Implementation matters, also in concurrent evolutionary algorithms  
JJ Merelo, Mario García Valdez, Sergio Rojas Galeano

Break — 10:20 - 10:40

Operon C++: an Efficient Genetic Programming Framework for Symbolic Regression  
Bogdan Burlacu, Gabriel Kronberger, Michael Kommenda

Integrating HeuristicLab with Compilers and Interpreters for Non-Functional Code Optimization  
Daniel Dorfmeister, Oliver Krauss

Library for Evolutionary Algorithms in Python (LEAP)  
Mark Coletti, Eric Scott, Jeffrey Bassett

Closing Remarks  
Workshop chairs

ECPERM — Evolutionary Computation for Permutation Problems

Organizers: Marco Baioletti (University of Perugia); Josu Ceberio (University of Basque Country); John McCall (University Robert Gordon, Aberdeen); Alfredo Milani (University of Perugia)

Time: Wednesday, July 8, 8:30 - 12:30

Opening  
Marco Baioletti, Josu Ceberio, John McCall, Alfredo Milani

Keynote talk: Exploiting the algebraic properties of the permutation space in Evolutionary Computation  
Valentino Santucci

Monte Carlo Tree Search on Perfect Rectangle Packing Problem Instances  
Daan Van den Berg, Igor Pejic

dMFEA-II: An Adaptive Multifactorial Evolutionary Algorithm for Permutation-based Discrete Optimization Problems  
Eneco Osaba, Aritz D. Martinez, Akemi Galvez, Andrés Iglesias, Javier Del Ser

Break — 10:20 - 10:40

Tabu Search and Iterated Greedy for a Flow Shop Scheduling Problem with Worker Assignment  
José Elias Claudio Arroyo, Lucas Batista Fialho, Matheus Araujo

Gradient Search in the Space of Permutations: an application for the Linear Ordering Problem  
Valentino Santucci, Josu Ceberio, Marco Baioletti

On the solvability of routing multiple point-to-point paths in Manhattan meshes  
Yannick Vinkesteijn, Daan Van den Berg, Reitze Jansen

An experimental evaluation of the Algebraic Differential Evolution algorithm on the Single Row Facility Layout Problem  
Gabriele Di Bari, Marco Baioletti, Valentino Santucci

The $((1 + \lambda, \lambda))$ Genetic Algorithm for Permutations  
Anton Bassin, Maxim Buzdalov
Closing
Marco Baioletti, Josu Ceberio, John McCall, Alfredo Milani

GreenAI — Green AI: Evolutionary and machine learning solutions in environment, renewable and ecologically-aware scenarios

Organizers: Nayat Sánchez-Pi (Rio de Janeiro State University); Luis Martí (Inria Chile)
Time: Wednesday, July 8, 8:30 - 10:20

Welcome and introduction
Workshop chairs

Invited talk: The Convergence of Artificial Intelligence and High Performance Computing Towards Green AI Solutions
Mariza Ferro

Invited talk: Tackling Climate Change with Machine Learning
Priya L. Donti, David Rolnik

Invited talk: Why GPUs are Green?
Esteban Clua

Invited talk: AI for Green Computing Vs. Greening AI
Romain Rouvoy

Invited talk: Towards a Green AI: Addressing the ecological footprint of machine learning
Nayat Sanchez-Pi, Luis Martí

Open discussion, questions and final remarks

IAM — Industrial Application of Metaheuristics

Organizers: Silvino Fernandez Alzueta (ArcelorMittal); Thomas Stützle (Université Libre de Bruxelles); Pablo Valledor (ArcelorMittal)
Time: Wednesday, July 8, 8:30 - 10:20

Keynote: Industrial Applications of Metaheuristics: What is needed to make it a WIN-WIN situation
Boris Naujoks

A Pareto Front-based Metric to Identify Major Bitcoin Network Influencers
Jonathan Gillett, Shahryar Rahnamayan, Masoud Makrehchi, Azam Asilian Bidgoli

A Benchmark of Recent Population-Based Metaheuristic Algorithms for Multi-Layer Neural Network Training
Seyed Jalaleddin Mousavirad, Gerald Schaefer, Seyed Mohammad Jafar Jalali, Iakov Korovin

Scheduling unrelated parallel machines with family setups and resource constraints to minimize total tardiness
Júlio C. S. N. Pinheiro, José Elias Claudio Arroyo, Lucas Batista Fialho
PDEIM — Parallel and Distributed Evolutionary Inspired Methods

**Organizers:** Ernesto Tarantino (ICAR CNR); Ivanoe de Falco (ICAR CNR); Antonio Della Cioppa (University of Salerno); Umberto Scafuri (ICAR CNR)

**Time:** Wednesday, July 8, 10:40 - 12:30

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An Evolutionary Approach for Constructing Multi-Stage Classifiers

Nolan Hamilton, Errin Wesley Fulp

A Parallel Two-Stage Genetic Algorithm for Route Planning

H. David Mathias, Samantha S. Foley

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ExaEvo: Topological Optimization and Scalability of Evolutionary Algorithms

Ian Bradley Morgan, Daniel R. Tauritz

Exploiting Multi-Objective Parallel Extremal Optimization Features in Dynamic Load Balancing

Ivanoe De Falco, Eryk Laskowski, Richard Olejnik, Umberto Scafuri, Ernesto Tarantino, Marek Truduj

cMOGA/D: a Novel Cellular GA based on Decomposition to Tackle Constrained Multiobjective Problems

Cosijopii Garcia-Garcia, Alicia Morales-Reyes, Maria-Guadalupe Martínez-Peñaloza

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SAEOpt — Surrogate-Assisted Evolutionary Optimisation

**Organizers:** Alma Rahat (University of Plymouth); Richard M. Everson (University of Exeter); Jonathan Fieldsend (University of Exeter); Hao Wang (University of Leiden); Yaochu Yin (University of Surrey)

**Time:** Wednesday, July 8, 14:00 - 15:50

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Keynote: A Tour D’Horizon in Multi-Objective Optimization with Expensive Black-Box Evaluations

Michael Emmerich

A Surrogate-Assisted GA Enabling High-throughput ML by Optimal Feature and Discretization Selection

Johan Garcia

What Do you Mean? The Role of the Mean Function in Bayesian Optimisation

George De Ath, Jonathan E. Fieldsend, Richard M. Everson

Bayesian Methods for Multi-objective Optimization of a Supersonic Wing Platform

Timothy MS Jim, Ghifari Adam Faza, Pramudita Satria Palar, Koji Shimoyama

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iGECCO — Interactive Methods @ GECCO

**Organizers:** Matthew Johns (University of Exeter); Nick Ross (University of Exeter); Ed Keedwell (University of Exeter); Herman Mahmoud (University of Exeter); David Walker (University of Plymouth)

**Time:** Wednesday, July 8, 14:00 - 18:00

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Keynote: Interactive evolutionary computation: from interactive to participative

Evelyne Lutton

**Break — 15:50 - 16:10**
Characteristic Analysis of Auditory Perception and Aesthetics in Sound Composition Optimization Using Revised Interactive Differential Evolution

Hayato Shindo, Yan Pei

Introducing Counterexamples to a Robotic Agent Using Clicker Training

Travis DeVault, Robert B. Heckendorn, Terence Soule

HOWS Interactive System Demo

Herman Mahmoud, Matthew Johns

Panel Discussion

Matthew Johns, Nick Ross, Ed Keedwell, Herman Mahmoud, David Walker, Evelyne Lutton

VizGEC 2020 — Visualisation Methods in Genetic and Evolutionary Computation

Organizers: David Walker (University of Plymouth); Richard Everson (University of Exeter); Rui Wang (Uber); Neil Vaughan (University of Chester)

Time: Wednesday, July 8, 14:00 - 15:50

Keynote: Harmonising Many-objective Visualisation

Richard M. Everson

Visual Mapping of Multi-objective Optimization Problems and Evolutionary Algorithms

Kohei Yamamoto, Tomoaki Takagi, Keiki Takadama, Hiroyuki Sato

Using PageRank to Uncover Patterns in Search Behavior Induced by the Bit Flip Operator

Thomas Green, Timothy Andersen

BENCHMARK — Good Benchmarking Practices for Evolutionary Computation

Organizers: Tome Eftimov (Institute Jozef Stefan); William La Cava (University of Pensilvania); Boris Naujoks (Cologne University of Applied Science); Pietro S. Oliveto (University of Sheffield); Carola Doerr (CNRS, Sorbonne University); Vanessa Volz (modl.ai); Thomas Weise (Hefei University)

Time: Wednesday, July 8, 16:10 - 18:00

Discussion: What is the purpose of benchmarking?

Discussion: What are good benchmarking practices / how to do benchmarking / what not to do?

A Benchmark with Facile Adjustment of Difficulty for Many-Objective Genetic Programming and Its Reference Set

Makoto Ohki

Discussion: Open questions/issues in benchmarking / What to do next?
RWACMO — Real-World Applications of Continuous and Mixed-Integer Optimization

Organizers: Akira Oyama (Japan Aerospace Exploration Agency); Koji Shimoyama (Tohoku University); Hemant Kumar Singh (University of New South Wales), Kazuhisa Chiba (University of Electro-Communications), Pramudita Satria Palar (Bandung Institute of Technology)

Time: Thursday, July 9, 8:30 - 10:20

Invited Talk: Multidisciplinary Design Optimization: Some Developments and Challenges
Tapabrata Ray

Uncertainty Quantification Methods for Evolutionary Optimization under Uncertainty
Pramudita Satria Palar, Koji Shimoyama, Lavi Rizki Zuhal

High-Dimensional Multi-Level Maximum Variance Threshold Selection for Image Segmentation: A Benchmark of Recent Population-based Metaheuristic Algorithms
Seyed Jalaleddin Mousavirad, Gerald Schaefer, Zahra Movahedi, Iakov Korovin

SWINGA — Swarm Intelligence Algorithms: Foundations, Perspectives and Challenges

Organizers: Ivan Zelinka (Technical University of Ostrava); Swagatam Das (Indian Statistical Institute); Ponnuthural Nagaratnam Suganthan (Nanyang Technological University); Roman Senkerik (Tomas Bata University)

Time: Thursday, July 9, 8:30 - 12:30

Discrete Self Organizing Algorithm for Pollution Vehicle Routing Problem
Donald Davendra, Magdalena Bialic-Davendera

gBeam-ACO: a greedy and faster variant of Beam-ACO
Jeff Hajewski, Suely Oliveira, David Stewart, Laura Weiler

Archive-Based Swarms
Nishant Rodrigues, Chilukuri Mohan

A Modular Hybridization of Particle Swarm Optimization and Differential Evolution
Rick Boks, Hao Wang, Thomas Bäck

Break — 10:20 - 10:40

Colour Quantisation using Self-Organizing Migrating Algorithm
Seyed Jalaleddin Mousavirad, Gerald Schaefer, Iakov Korovin

High-Dimensional Multi-Level Image Thresholding using Self-Organizing Migrating Algorithm
Seyed Jalaleddin Mousavirad, Gerald Schaefer, Iakov Korovin

Self-organizing Migrating Algorithm with Clustering-aided Migration
Tomas Kadavy, Michal Pluhacek, Adam Viktorin, Roman Senkerik

Ensemble of Strategies and Perturbation Parameter Based SOMA for Optimal Stabilization of Chaotic Oscillations
Roman Senkerik, Tomas Kadavy, Adam Viktorin, Michal Pluhacek
Applications of swarm intelligence algorithms countering the cyber threats
Thanh Cong Truong, Tan-Phu Huynh, Ivan Zelinka

MedGEC — Medical Applications of Genetic and Evolutionary Computation

Organizers: Neil Vaughan (University of Chester); Stephen Smith (University of York); Stefano Cagnoni (University of Parma); Robert M. Patton (Oak Ridge National Laboratory)

Time: Thursday, July 9, 8:30 - 10:20

Introduction
Stephen L. Smith, Stefano Cagnoni, Neil Vaughan, Robert M. Patton

Invited talk: Negative emotion classification using EEG signals
Arpit Bhardwaj

Invited talk: Co-evolution and the opportunities for patient assessment
Michaela Drahosova

Invited talk: Data-clouds, deep learning and white box algorithms for patient monitoring
Marta Vallejo

Invited talk: What do doctors expect from computer scientists to fight Covid-19 or possible future pandemics?
Tito Poli, Nicola Sverzellati

Invited talk: EXPOSED – An occupant exposure model for confined spaces to retrofit crowd models during a pandemic
Ruggiero Lovreglio

Invited talk: Modeling the peripheral circulation
Leonardo Bocchi, Romeo Martini

Invited talk: Evolving a model of the retina for eyesight loss
Neil Vaughan

Workshop close
Stephen L. Smith, Stefano Cagnoni, Neil Vaughan, Robert M. Patton

ECADA — 10th Workshop on Evolutionary Computation for the Automated Design of Algorithms

Organizers: John R. Woodward (Queen Mary University-London); Daniel R. Tauritz (Auburn University); Emma Hart (Edinburgh-Napier University)

Time: Thursday, July 9, 10:40 - 12:30

Introduction
John R. Woodward, Daniel R. Tauritz, Emma Hart

Time-Dependent Automatic Parameter Configuration of a Local Search Algorithm
Weerapan Sae-Dan, Marie-Eléonore Kessaci, Nadarajen Veerapen, Laetitia Jourdan
Dynamic Primitive Granularity Control: An Exploration of Unique Design Considerations  
Braden N. Tisdale, Aaron Scott Pope, Daniel R. Tauritz

On the Sensitivity Analysis of Cartesian Genetic Programming Hyper-Heuristic  
Luis Filipe Araujo Pessoa, Bernd Hellingrath, Fernando Buarque de Lima Neto

The Automated Design of Local Optimizers for Memetic Algorithms Employing Supportive Coevolution  
Nathaniel R. Kamrath, Aaron Scott Pope, Daniel R. Tauritz

IWLCS 2020 — 23rd International Workshop on Learning Classifier Systems

Organizers: Anthony Stein (University of Augsburg); Masaya Nakata (Yokohama National University); David Pätzel (University of Augsburg)

Time: Thursday, July 9, 10:40 - 15:50

Welcome Note  
Anthony Stein, Masaya Nakata, David Pätzel

Keynote: Interpretability challenges and opportunities in rule-based machine learning  
Ryan J. Urbanowicz

An Overview of LCS Research from IWLS 2019 to 2020  
David Pätz, Anthony Stein, Masaya Nakata

A Scikit-learn Compatible Learning Classifier System  
Robert F. Zhang, Ryan J. Urbanowicz

XCS as a Reinforcement Learning Approach to Automatic Test Case Prioritization  
Lukas Rosenbauer, Anthony Stein, Roland Maier, David Pätz, Jörg Hähner

Lunch break — 12:30 - 14:00

Generic Approaches for Parallel Rule Matching in Learning Classifier Systems  
Lukas Rosenbauer, Anthony Stein, Roland Maier, Jörg Hähner

An Adaption Mechanism for the Error Threshold of XCSF  
Tim Hansmeier, Paul Kaufmann, Marco Platzner

Learning Classifier Systems: Appreciating the Lateralized Approach  
Abubakar Siddique, Will Neil Browne, Gina M. Grimshaw

PEPACS: Integrating Probability-Enhanced Predictions to ACS2  
Romain Orhand, Anne Jeannin-Girardon, Pierre Parrend, Pierre Collet

Investigating Exploration Techniques for ACS in Discretized Real-Valued Environments  
Norbert Kozlowski, Olgierd Unold

General discussion
**Women@GECCO**

**Organizers:** Elizabeth Wanner (Aston University); Leslie Angélica Pérez Cáceres (Pontificia Universidad Católica de Valparaíso)

**Time:** Thursday, July 9, 12:30-14:00

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**A psychologist’s Pandemic Advice:**

**Mental Health Matters**  
Suzanne Skeete, *Inspirational Woman of the Year 2018, Managing Director of Tappy Twins, Teenline, TT Training Academy and Founder of the Courageous Kids Awards*

**COVID-19 and Posttraumatic Growth - the creative power of destruction**  
Katarzyna Wawer Dziedzia, *Clinical Psychologist, Birmingham, UK*

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**Invited Talks:**

**Mathematical Programming strengthening Multi-objective Evolutionary Algorithms**  
Adriana Lara-López, *Instituto Politécnico Nacional, Mexico*

**My path to supporting decision makers in finding the best compromise**  
Kaisa Miettinen, *University of Jyväskylä, Finland*

**Ask me anything**  
Una-May O’Reilly, *Computer Science & Artificial Intelligence Lab, MIT*

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**Women@GECCO meets pub quiz**

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**ADRR — Automated Design of Robots for the Real-world**

**Organizers:** David Howard (CSIRO); Emma Hart (Edinburgh Napier University); Gusz Eiben (VU Amsterdam)

**Time:** Thursday, July 9, 14:00 - 15:50

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**Introduction to the workshop**  
Workshop chair

**Keynote: AutoS2R: Scaling up the automated design and manufacture of soft- and biological robots**  
Josh Bongard

**Real World Morphological Evolution is Feasible**  
Tønnes F. Nygaard, Gerard Howard, Kyrre Glette

**Reusability vs Morphological Space in Physical Robot Evolution**  
Rodrigo Moreno, Andres Faina

**If It Evolves It Needs To Learn**  
Agoston E. Eiben, Emma Hart

**Keynote: The information return on investment (IROI) condition in self-replicating machines**  
Alex Ellery

**Automatically designing the behaviours of falling paper**  
Toby Howison, Fumiya Lida
Evolutionary Stress Factors for Adaptable Robot 'Personalities'
   Edmund R. Hunt

Path Towards Multilevel Evolution of Robots
   Shelvin Chand, Gerard Howard

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**SecDef 2020 — Genetic and Evolutionary Computation in Defense, Security, and Risk Management**

**Organizers:** Erik Hemberg (MIT CSAIL); Riyad Alshammari (King Saud bin Abdulaziz University, Saudi Arabia); Tokunbo Makanju (New York Institute of Technology)

**Time:** Thursday, July 9, 14:00 - 15:50

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**Welcome and introduction**
   Workshop chair

- **Delivering diverse web server configuration in a moving target defense using evolutionary algorithms**
  Ernesto Serrano Collado, Mario García Valdez, JJ Merelo

- **Securing the Software Defined Perimeter with Co-optimization**
  Michal Shlapentokh-Rothman, Erik Hemberg, Una-May O’Reilly

- **Using Evolutionary Algorithms and Pareto Ranking to Identify Secure Virtual Local Area Networks**
  Alina Pacheco Rodríguez, Errin Wesley Fulp, David John, Jinku Cui

- **Exploring an Artificial Arms Race for Malware Detection**
  Zachary Wilkins, Ibrahim Zincir, Nur Zincir-Heywood

- **Adversarial Threats to Large Satellite Networks (ATLAS-N): A Coevolutionary Approach Based on FlipIt**
  Jay Patel, Dhathri H. Somavarapu, Deacon Seals, Daniel R. Tauritz, Davide Guzzetti

**Summary and goodbye**
   Workshop chair

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**LBA — Late Breaking Abstracts**

**Organizers:** Ke Tang (Southern University of Science and Technology)

**Time:** Saturday, July 9, 16:10-18:00

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- **A Genetic Algorithm to Optimize SMOTE and GAN Ratios in Class Imbalanced Datasets**
  Hwi-Yeon Cho, Yong-Hyuk Kim

- **A Daily Stock Index Predictor Using Feature Selection Based on a Genetic Wrapper**
  Dong-Hee Cho, Seung-Hyun Moon, Yong-Hyuk Kim

- **An Effective Variable Transfer Strategy In Multitasking Optimization**
  Qingzheng Xu, Balin Tian, Lei Wang, Feng Zou, Qian Sun

- **Antimander: Open Source Detection of Gerrymandering though Multi-Objective Evolutionary Algorithms**
  Joel Simon, Joel Lehman
On the Co-authorship Network in Evolutionary Computation
   Dong-Pil Yu, Yong-Hyuk Kim

A Surrogate Model Using Deep Neural Networks for Optimal Oil Skimmer Assignment
   Kim Hye-Jin, Kim Yong-Hyuk

Usage of a Genetic Algorithm for optimizing stock usage
   Mateus Interciso, Plinio Garcia

On the order of variables for multitasking optimization
   Lei Wang, Qian Sun, Qingzheng Xu, Balin Tian, Wei Li

Data driven building of realistic neuron model using IBEA and CMA evolution strategies
   Tanguy Damart, Werner Van Geit, Henry Markram

A Bio-inspired Approach for the Spectrum Allocation Problem in IoT Networks
   Jesús Gómez-Avilés, Ángel G. Andrade, Anabel Martínez-Vargas

Combined Kriging Surrogate Model for Efficient Global Optimization Using the Optimal Weighting Method
   Tanguy Appriou, Koji Shimoyama

Towards improvement of SUNA in Multiplexers with preliminary results of simple Logic Gate neuron variation
   Anh Duc Ta, Danilo Vasconcellos Vargas

Constraint Handling in Genotype to Phenotype Mapping and Genetic Operators for Project Staffing
   Soo Ling Lim, Yi Kuo, Peter John Bentley

An RTS-Based Algorithm for Noisy Optimization by Strategic Sample Accumulation
   Jeongmin Kim, Kwang Ryel Ryu

Automatic Evolutionary Learning of Composite Models With Knowledge Enrichment
   Anna Kalyuzhnaya, Nikolay Nikitin, Pavel Vychuzhanin, Alexander Hvató, Alexander Boukhansovsky

Quality enhancement of stochastic feature subset selection via genetic algorithms. Assessment in
   Bioinformatics data sets
   Antonio J. Tallón-Ballesteros, Tengyue Li, Simon Fong

Teng-Yue Algorithm: A Novel Metaheuristic Search Method for Fast Cancer Classification
   Tengyue Li, Simon Fong, Antonio J. Tallón-Ballesteros

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GI@GECCO — The Ninth Genetic Improvement Workshop (2020)

Organizers: Bradley Alexander (University of Adelaide); Alexander (Sandy) Brownlee (University of Stirling); Saemundur O. Haraldsson (University of Stirling); Markus Wagner (University of Adelaide); John R. Woodward (Queen Mary University-London)

Time: Thursday, July 9, 16:10 - 18:00

Welcome
   Workshop chairs

An Annotated Dataset of Stack Overflow Post Edits
   Sebastian Baltes, Markus Wagner
Optimising the Fit of Stack Overflow Code Snippets into Existing Code
Brittany Reid, Christoph Treude, Markus Wagner

Genetic Improvement of Software Efficiency: The Curse of Fitness Estimation
Mamoud Bokhari, Markus Wagner, Brad Alexander

Evolving sqrt into 1/x via software data maintenance
William Langdon, Oliver Krauss

Tuning Genetic Algorithm Parameters using Design of Experiments
Mohsen Mosayebi, Manbir Sodhi

Discussions: What's next?

E-MaOP + DTEO — Evolutionary Many-objective Optimization + Decomposition Techniques in Evolutionary Optimization

Organizers: Rui Wang (National University of Defense Technology); Ran Cheng (Southern University of Science and Technology); Guohua Wu (Central South University); Miqing Li (University of Birmingham); Hisao Ishibuchi (Southern University of Science and Technology); Bilel Derbel (Univ. Lille, Inria Lille - Nord Europe); Ke Li (University of Exeter); Xiaodong Li (RMIT University); Saúl Zapotecas-Martínez (SHINSHU University), Qingfu Zhang (City University of Hong Kong)

Time: Thursday, July 9, 16:10 - 18:00

Preliminary Study of Adaptive Grid-based Decomposition on Many-objective Evolutionary Optimization
Kensuke Kano, Tomoaki Takagi, Keiki Takadama, Hiroyuki Sato

Incremental Lattice Design of Weight Vector Set
Tomoaki Takagi, Keiki Takadama, Hiroyuki Sato

Multi-objective Optimization in the Agile Software Project Scheduling using Decomposition
Saúl Zapotecas-Martínez, Abel García-Nájera, Humberto Humberto
Humies, Competitions, Hot Off the Press, and Job Market
17th Annual Humies Awards for Human Competitive Results

Presentations: Friday, July 10, 13:40-15:20

Announcement of Awards: Sunday, July 12, 12:20-13:50

On-location chair: Erik D. Goodman


Publicity Chair: William Langdon

Prizes: prizes totaling $10,000 to be awarded

Detailed Information: www.human-competitive.org

Techniques of genetic and evolutionary computation are being increasingly applied to difficult real-world problems—often yielding results that are not merely academically interesting, but competitive with the work done by creative and inventive humans. Starting at the Genetic and Evolutionary Computation Conference (GECCO) in 2004, cash prizes have been awarded for human competitive results that had been produced by some form of genetic and evolutionary computation in the previous year.

The total prize money for the Humies awards is $10,000 US dollars. As a result of detailed consideration of the entries in this year’s Humies competition, the selected finalists will each be invited to give a short prerecorded video presentation during GECCO. Each presentation will be 10 minutes. The presentations are open to all GECCO participants. After the session the judges will confer and select winners for Bronze (either one prize of $2,000 or two prizes of $1,000) Silver ($3,000) and Gold ($5,000) awards. The awards will be announced during the GECCO closing ceremony.
Competition on Single Objective Bound Constrained Numerical Optimization

Organizers: Ponnuthurai Nagaratnam Suganthan, Mostafa Ali, J. J. Liang, B. Y. Qu, Cai Tong Yue, Kenneth Price

Time: Thursday, July 9, 8:30-8:45

This competition challenges participants to optimize a test-bed consisting of 10 bound constrained functions that range from easy to hard. Each function is sampled at multiple dimensions to provide insights into algorithmic scaling performance. This year’s competition extends the limits for the maximum allowed number of function evaluations beyond those chosen for prior competitions with the goal of learning if the additional time translates into significantly improved final function values.

Competition on Single Objective Constrained Numerical Optimization

Organizers: Ponnuthurai Nagaratnam Suganthan, Guohua Wu, Mostafa Ali, Rammohan Mallipeddi, Abhishek Kumar, Swagatam Das

Time: Thursday, July 9, 8:45-9:00

The goals of this competition are to evaluate the current state of the art in single objective numerical optimization with general constraints and to propose novel benchmark problems with diverse characteristics. Under the above scenarios, for the first time a set of real-world constrained optimization benchmark will be released for this competition.

Competition on the optimal camera placement problem (OCP) and the unicost set covering problem (USCP)

Organizers: Mathieu Brévilliers, Lhassane Idoumghar, Julien Kritter, Julien Lepagnot

Time: Thursday, July 9, 9:00-9:15

The use of camera networks is now common to perform various surveillance tasks. These networks can be implemented together with intelligent systems that analyze video footage, for instance, to detect events of interest, or to identify and track objects or persons. Firstly, the main goal of this competition is to encourage innovative research works in this direction, by proposing to solve optimal camera placement (OCP) problem instances. Secondly, this competition is an opportunity to propose a benchmark testbed for the OCP. Thirdly, this competition is a way of attracting the interest of the scientific community.

Dota 2 1-on-1 Shadow Fiend Laning Competition

Organizers: Robert Smith, Malcolm Heywood

Time: Thursday, July 9, 9:15-9:30

The Dota 2 game represents an example of a multiplayer online battle arena video game. The underlying goal of the game is to control the behaviour/strategy for a ‘hero’ character. Each hero possesses unique abilities, thus resulting in different performance tradeoffs. This competition will assume the 1-on-1 mid lane configuration of Dota 2 using the Shadow Fiend hero. Such a configuration still includes many of the properties that have turned the game into an ‘e-sport’, but without the computational overhead of solving the task for all heroes under multi-lane settings.

Dynamic Stacking Optimization in Uncertain Environments

Organizers: Andreas Beham, Stefan Wagner, Sebastian Raggl

Time: Thursday, July 9, 9:30-9:45
Stacking problems are central to multiple billion-dollar industries. The container shipping industry needs to stack millions of containers every year. In the steel industry the stacking of steel slabs, blooms, and coils needs to be carried out efficiently, affecting the quality of the final product. The immediate availability of data – thanks to the continuing digitalization of industrial production processes – makes the optimization of stacking problems in highly dynamic environments feasible. In this competition your goal is to write a policy that optimizes crane actions in simulated environments.

**Evolutionary Computation in Uncertain Environments: A Smart Grid Application**

**Organizers:** Joao Soares, Fernando Lezama, Bruno Canizes, Zita Vale  
**Time:** Thursday, July 9, 9:45-10:00

This GECCO 2020 competition proposes two test beds in the energy domain: Testbed 1) optimization of a centralized day-ahead energy resource management problem in smart grids under environments with uncertainty. This test bed is similar to the past challenge using a challenging 500-scenario case study with high degree of uncertainty. Testbed 2) bi-level optimization of end-users' bidding strategies in local energy markets. This test bed represents a complex bi-level problem in which competitive agents in the upper-level try to maximize their profits, modifying and depending on the price determined in the lower-level problem, thus resulting in a strong interdependence.

**GECCO2020 Competition on Niching Methods for Multimodal Optimization**

**Organizers:** Mike Preuss, Michael Epitropakis, Xiaodong Li  
**Time:** Thursday, July 9, 10:40-10:55

The aim of the competition is to provide a common platform that encourages fair and easy comparisons across different niching algorithms. The competition allows participants to run their own niching algorithms on 20 benchmark multimodal functions with different characteristics and levels of difficulty. Researchers are welcome to evaluate their niching algorithms using this benchmark suite, and report the results by submitting a paper to the main tracks of GECCO 2020 (i.e., submitting via the online submission system of GECCO 2020).

**Industrial Challenge**

**Organizers:** Frederik Rehbach, Thomas Bartz-Beielstein  
**Time:** Thursday, July 9, 10:55-11:10

As usual, the industrial challenge is posed in cooperation with an industry partner of the institute. This year’s competition partially relies on a test suite that provides expensive computer simulation-based optimization problems and provides an easy evaluation interface that will be used for the setup of our challenge. Our industry partner is willing to publish one of their CFD simulations - the optimization of a gas distribution system (GDS) in a large-scale electrostatic precipitator - as a public challenge. The task in the first track is to find an optimal configuration for the gas distribution system.

**Open Optimization Competition 2020**

**Organizers:** Carola Doerr, Olivier Teytaud, Jérémy Rapin, Thomas Baek  
**Time:** Thursday, July 9, 11:10-11:25

The Open Optimization Competition aims at fostering a collective, community-driven effort towards reproducible, open access, and conveniently interpretable comparisons of different optimization techniques, with the goal to support users in choosing the best algorithms and the best configurations for their problem at hand. The competition has two tracks:

1. A performance-oriented track, which welcomes contributions of efficient optimization algorithms.
2. Contributions towards "better" benchmarking.

While the performance track is hosted within Nevergrad, the contributions track welcomes contributions to both Nevergrad and IOHanalyzer, the analytical and visualization module of IOHprofiler.

**Evolutionary Multi-task Optimization**

**Organizers:** Feng Liang, Kai Qin, Abhishek Gupta, Yuan Yuan, Eric O Scott, Yew Soon Ong  
**Time:** Thursday, July 9, 11:25-11:40

One of the long-standing goals of AI has been to effectively multitask; i.e., learning to solve many tasks at the same time. It is worth noting that although humans are generally unable to tackle multiple problems simultaneously, or within short timespans – as interleaving more than one task usually entails a considerable switching cost during which the brain must re-adjust from one to the other – machines are largely free from such computational bottlenecks. Evolutionary multitasking is an emerging concept in computational intelligence that realizes the theme of efficient multi-task problem-solving in the domain of numerical optimization.
Hot off the Press

**Organizer:** Heike Trautmann, *University of Münster*

**Time:**
- **HOP1:** Friday, July 10, 10:30-12:10
- **HOP2:** Friday, July 10, 13:40-15:20

The Hot Off the Press (HOP) track offers authors of recent papers the opportunity to present their work to the GECCO community, both by giving a talk on one of the three main days of the conference and by having a 2-page abstract appear in the Proceedings Companion, in which the workshop papers, late-breaking abstracts, and tutorials also appear. We invite researchers to submit summaries of their own work recently published in top-tier conferences and journals. Contributions are selected based on their scientific quality and their relevance to the GECCO community. Typical contributions include (but are not limited to) evolutionary computation papers that have appeared at venues different from GECCO, papers comparing different heuristics and optimization methods that have appeared at a general heuristics or optimization venue, papers describing applications of evolutionary methods that have appeared at venues of this application domain, or papers describing methods with relevance to the GECCO community that have appeared at a venue centred around this method’s domain. In any case, it is the authors’ responsibility to make clear why this work is relevant for the GECCO community, and to present the results in a language accessible to the GECCO community.
Job Market

Organizers: Boris Naujoks, TH Köln - Cologne University of Applied Sciences
Tea Tušar, Jozef Stefan Institute

Time: Friday, July 10, 12:10-13:40

At the GECCO Job Market people offering jobs in Evolutionary Computation can advertise open positions and meet with potential candidates. Any kinds of positions are of interest (PhD, Postdoc, Professor, Engineer, etc.) from academia as well as industry. After brief presentations of the available jobs, participants have the possibility to set up face-to-face meetings for further discussions.

Like the whole conference, the GECCO Job Market this year will be organized using Zoom. For face-to-face meetings, Zoom offers so called breakout rooms. We will make one room available for each of the positions on offer and organize the meetings in these rooms accordingly.

The collection of positions on offer can be found at the SIGEVO web site:
Best Paper Nominations
Voting Instructions

**Beware:** Each GECCO attendee has only one vote and can only vote for a single best paper session. The best paper session for which to vote can be decided after attending several best paper sessions (see below).

**Procedure:** Each track has nominated one or more papers for a best paper award (see the list below). There will be one award per “Best Paper” session. Papers competing for the same award are presented in the same “Best Paper” session. Nominees from small tracks are grouped together into the same session. The votes are nominative and cannot be delegated to another attendee.

More detailed instructions of the procedure to vote during the online conference will be provided to all the attendees.

**Best Paper Nominations**

**Ant Colony Optimization and Swarm Intelligence (ACO-SI)**

- **A Multiobjective Optimization Approach for Market Timing**
  Ismail Mohamed, Fernando E.B. Otero
  
  Friday, July 10, 13:40-15:20

**Complex Systems (CS)**

- **Diversity Preservation in Minimal Criterion Coevolution through Resource Limitation**
  Jonathan C. Brant, Kenneth O. Stanley
  
  Friday, July 10, 13:40-15:20

**Evolutionary Combinatorial Optimization and Metaheuristics (ECOM)**

- **Golden Parameter Search: Exploiting Structure to Quickly Configure Parameters in Parallel**
  Yasha Pushak, Holger H. Hoos

  Friday, July 10, 15:40-17:20

- **Automatic Decomposition of Mixed Integer Programs for Lagrangian Relaxation Using a Multiobjective Approach**
  Jake Matt Weiner, Andreas Ernst, Xiaodong Li, Yuan Sun

  Friday, July 10, 15:40-17:20

**Evolutionary Machine Learning (EML)**

- **Improving Constrained Clustering Via Decomposition-based Multiobjective Optimization with Memetic Elitism**
  Germán González-Almagro, Alejandro Rosáles-Pérez, Julián Luengo, José-Ramón Cano, Salvador García

  Saturday, July 11, 17:40-19:20

- **Neuroevolution of Self-Interpretable Agents**
  Yujin Tang, Duong Nguyen, David Ha

  Saturday, July 11, 17:40-19:20

- **Improving Neuroevolutionary Transfer Learning of Deep Recurrent Neural Networks through Network-Aware Adaptation**
  AbdElRahman ElSaid, Joshua Karns, Zimeng Lyu, Daniel Krutz, Alexander Ororbia II, Travis Desell

  Saturday, July 11, 17:40-19:20
Evolutionary Multiobjective Optimization (EMO)

On the elicitation of indirect preferences in Interactive evolutionary multiple objective optimization
Michał K. Tomczyk, Miłosz Kadziński
Friday, July 10, 10:30-12:10

Multi-Objective Hyperparameter Tuning and Feature Selection using Filter Ensembles
Martin Binder, Julia Moosbauer, Janek Thomas, Bernd Bischl
Friday, July 10, 10:30-12:10

Another Difficulty of Inverted Triangular Pareto Fronts for Decomposition-Based Multi-Objective Algorithms
Linjun He, Auraham Camacho, Hisao Ishibuchi
Friday, July 10, 10:30-12:10

Evolutionary Numerical Optimization (ENUM)

Leveraging Conditional Linkage Models in Gray-box Optimization with the Real-Valued Gene-pool Optimal Mixing Evolutionary Algorithm
Anton Bouter, Stefanus C. Maree, Tanja Alderliesten, Peter A. N. Bosman
Friday, July 11, 13:40-15:20

Genetic Algorithms (GA)

Understanding Transforms of Pseudo-Boolean Functions
Darrell Whitley, Hernan Aguirre, Andrew Sutton
Saturday, July 11, 17:40-19:20

Adaptively Preserving Solutions in Both Feasible and Infeasible Regions on Generalized Multiple Constraint Ranking
Yohanes Bimo Dwianto, Hiroaki Fukumoto, Akira Oyama
Saturday, July 11, 17:40-19:20

General Evolutionary Computation and Hybrids (GECH)

Effective Reinforcement Learning through Evolutionary Surrogate-Assisted Prescription
Olivier Francon, Santiago Gonzalez, Babak Hodjat, Elliot Meyerson, Risto Miikkulainen, Xin Qiu, Hormoz Shahrzad
Saturday, July 11, 15:40-17:20

c-shotgun: c-greedy Batch Bayesian Optimisation
George De Ath, Richard M. Everson, Jonathan E. Fieldsend, Alma A. M. Rahat
Saturday, July 11, 15:40-17:20

Analysis of the Performance of Algorithm Configurators for Search Heuristics with Global Mutation Operators
George T. Hall, Pietro S. Oliveto, Dirk Sudholt
Saturday, July 11, 15:40-17:20

Genetic Programming (GP)

Genetic programming approaches to learning fair classifiers
William La Cava, Jason H. Moore
Saturday, July 11, 15:40-17:20

A Modular Memory Framework for Time Series Prediction
Stephen Paul Kelly, Jacob Newsted, Wolfgang Banzhaf, Cedric Gondro
Saturday, July 11, 15:40-17:20

Semantically-Oriented Mutation Operator in Cartesian Genetic Programming for Evolutionary Circuit Design
David Hodan, Vojtech Mrazek, Zdenek Vasicek
Saturday, July 11, 15:40-17:20
**Real World Applications (RWA)**

**Towards Sustainable Forest Management Strategies with MOEAs**
Philipp Back, Antti Suominen, Pekka Malo, Olli Tahvonen, Julian Blank, Kalyanmoy Deb

**Optimisation of Large Wave Farms using a Multi-strategy Evolutionary Framework**
Mehdi Neshat, Bradley Alexander, Eli O. David, Nataliia Y. Sergiienko, Markus Wagner

**Search-Based Software Engineering (SBSE)**

**Seeding Strategies for Multi-Objective Test Case Selection: An Application on Simulation-based Testing**
Aitor Arrieta, Joseba Andoni Agirre, Goiuria Sagardui

**Theory (THEORY)**

**Fast Mutation in Crossover-based Algorithms**
Denis Antipov, Maxim Buzdalov, Benjamin Doerr
Papers and Posters
Another Difficulty of Inverted Triangular Pareto Fronts for Decomposition-Based Multi-Objective Algorithms ★
Linjun He, Auraham Camacho, Hisao Ishibuchi

On the elicitation of indirect preferences in interactive evolutionary multiple objective optimization ★
Michał K. Tomczyk, Miłosz Kadziński

Multi-Objective Hyperparameter Tuning and Feature Selection using Filter Ensembles ★
Martin Binder, Julia Moosbauer, Janek Thomas, Bernd Bischl

Constraint Handling within MOEA/D Through an Additional Scalarizing Function
Saúl Zapotecas-Martínez, Antonin Sebastien Ponsich,

Using Exploratory Landscape Analysis to Visualize Single-objective Problems
Urban Škvorc, Tome Eftimov, Peter Korošec

Model selection for metabolomics: predicting diagnosis of coronary artery disease using automated machine learning
Alena Orlenko, Jason H. Moore

Sharp Bounds for Genetic Drift in Estimation of Distribution Algorithms
Benjamin Doerr, Weijie Zheng

Empirical Linkage Learning
Michał Witold Przewozniczek, Marcin Michal Komarnicki

Evolution of Distributed Neural Controllers for Voxel-based Soft Robots
Eric Medvet, Alberto Bartoli, Andrea De Lorenzo, Giulio Fidel

Quality Diversity for Multi-task Optimization
Jean-Baptiste Mouret, Glenn Maguire

Scaling MAP-Elites to Deep Neuroevolution
Cédric Colas, Joost Huizinga, Vashisht Madhavan, Jeff Clune

Learning behaviour-performance maps with meta-evolution
David M. Bossens, Jean-Baptiste Mouret, Danesh Tarapore
GA1  
Friday, July 10, 10:30-12:10  
Chair: Gabriela Ochoa (University of Stirling)

**An Improved GPU-Accelerated Heuristic Technique Applied to the Capacitated Vehicle Routing Problem**  
Marwan Fouad Abdelatti, Manbir Singh Sodhi  
10:30-10:55

**On measuring and improving the quality of linkage learning in modern evolutionary algorithms applied to solve partially additively separable problems**  
Michal Witold Przewozniczek, Bartosz Frej, Marcin Michal Komarnicki  
10:55-11:20

**Multi-layer Heterogeneous Ensemble with Classifier and Feature Selection**  
Thanh Tien Nguyen, Van Nang Pham, Truong Manh Dang, Vu Anh Luong, John McCall, Alan Wee-Chung Liew  
11:20-11:45

**Genetic Algorithm for the Weight Maximization Problem on Weighted Automata**  
Elena Gutiérrez, Takamasa Okudono, Masaki Waga, Ichiro Hasuo  
11:45-12:10

GECH1  
Friday, July 10, 10:30-12:10  
Chair: Jürgen Branke (University of Warwick)

**Landscape-Aware Fixed-Budget Performance Regression and Algorithm Selection for Modular CMA-ES Variants**  
Anja Jankovic, Carola Doerr  
10:30-10:55

**Integrated vs. Sequential Approaches for Selecting and Tuning CMA-ES Variants**  
Diederick L. Vermetten, Hao Wang, Carola Doerr, Thomas Bäck  
10:55-11:20

**CMA-ES for One-Class Constraint Synthesis**  
Marcin Karmelita, Tomasz P. Pawlak  
11:20-11:45

**Bivariate Estimation-of-Distribution Algorithms Can Find an Exponential Number of Optima**  
Benjamin Doerr, Martin S. Krejca  
11:45-12:10
A Multiobjective Optimization Approach for Market Timing ★
Ismail Mohamed, Fernando E.B. Otero
13:40-14:05

Diversity Preservation in Minimal Criterion Coevolution through Resource Limitation ★
Jonathan C. Brant, Kenneth O. Stanley
14:05-14:30

Evolving ab initio trading strategies in heterogeneous environments
David Rushing Dewhurst, Yi Li, Alex Bogdan, Jasmine Geng
14:30-14:55

SGP-DT: Towards Effective Symbolic Regression with a Semantic GP Approach Based on Dynamic Targets
Stefano Ruberto, Valerio Terragni, Jason H. Moore
13:40-14:05

The Univariate Marginal Distribution Algorithm Copes Well With Deception and Epistasis
Benjamin Doerr, Martin S. Krejca
14:05-14:30

Is the statistical significance between stochastic optimization algorithms' performances also significant in practice?
Tome Eftimov, Peter Korošec
14:30-14:55

Large Scale Biomedical Data Analysis with Tree-based Automated Machine Learning
Trang T. Le, Weixuan Fu, Jason H. Moore
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Novelty Search for Automatic Bug Repair
Omar Villanueva, Leonardo Trujillo, Daniel Hernández
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A Study on Graph Representations for Genetic Programming
Léo Françoso Sotto, Paul Kaufmann, Timothy Atkinson, Roman Kalkreuth, Márcio Porto Basgalupp
14:05-14:30

DAE-GP: Denoising Autoencoder LSTM Networks as Probabilistic Models in Estimation of Distribution Genetic Programming
David Wittenberg, Franz Rothlauf, Dirk Schweim
14:30-14:55

Code Building Genetic Programming
Edward Pantridge, Lee Spector
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AutoLR: An Evolutionary Approach to Learning Rate Policies
Pedro Carvalho, Nuno Lourenço, Filipe Assunção, Penousal Machado
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New Search Operators for Node-Depth Based Encoding
Gustavo Post Sabin, Telma Woerle de Lima, Anderson da Silva Soares

Comparative Mixing for DSMGA-II
Marcin Michal Komarnicki, Michal Witold Przewozniczek, Tomasz Maciej Durda

A Biased Random Key Genetic Algorithm applied to the VRPTW with Skill Requirements and Synchronization Constraints
Alberto Francisco Kummer, Luciana Salete Buriol, Olinto César Bassi de Araújo

Initial Design Strategies and their Effects on Sequential Model-Based Optimization: An Exploratory Case Study Based on BBOB
Jakob Bossek, Carola Doerr, Pascal Kerschke

Using Implicit Multi-Objectives Properties to Mitigate Against Forgetfulness in Coevolutionary Algorithms
Adefunke Akinola, Mark Wineberg

Model-based Optimization with Concept Drifts
Jakob Richter, Junjie Shi, Jian-Jia Chen, Jörg Rahnenführer, Michel Lang

On the Choice of the Parameter Control Mechanism in the \((1 + (\lambda, \lambda))\) Genetic Algorithm
Mario Alejandro Hevia Fajardo, Dirk Sudholt
Towards Dynamic Algorithm Selection for Numerical Black-Box Optimization: Investigating BBOB as a Use Case
Diederick L. Vermetten, Hao Wang, Thomas Bäck, Carola Doerr

Versatile Black-Box Optimization
Jialin Liu, Antoine Moreau, Mike Preuss, Jérémy Rapin, Baptiste Roziere, Fabien Teytaud, Olivier Teytaud

Distributed Random Walks for Fitness Landscape Analysis
Ryan Dieter Lang, Andries Petrus Engelbrecht

A Surrogate-Assisted Metaheuristic for Bilevel Optimization
Jesus-Adolfo Mejia-de-Dios, Efren Mezura-Montes

Golden Parameter Search: Exploiting Structure to Quickly Configure Parameters in Parallel
Yasha Pushak, Holger H. Hoos

Automatic Decomposition of Mixed Integer Programs for Lagrangian Relaxation Using a Multiobjective Approach
Jake Matt Weiner, Andreas Ernst, Xiaodong Li, Yuan Sun

Just-in-Time Batch Scheduling Subject to Batch Size
Sergey Polyakovskiy, Dhananjay Thiruvady, Rym M’Hallah

Advanced Statistical Analysis of Empirical Performance Scaling
Yasha Pushak, Holger H. Hoos

Automated Design of Multi-Level Network Partitioning Heuristics Employing Self-Adaptive Primitive Granularity Control
Aaron Scott Pope, Daniel R. Tauritz

A Memetic Level-based Learning Swarm Optimizer for Large-scale Water Distribution Network Optimization
Ya-Hui Jia, Yi Mei, Mengjie Zhang

Adaptive Augmented Evolutionary Intelligence for the Design of Water Distribution Networks
Matthew Barrie Johns, Herman Mahmoud, Edward Keedwell, Dragan Savic

Evolving Energy Demand Estimation Models over Macroeconomic Indicators
Nuno Lourenço, José Manuel Colmenar, J. Ignacio Hidalgo, Sancho Salcedo Sanz
**EML1**

Chair: Una-May O’Reilly (CSAIL, Massachusetts Institute of Technology; ALFA Group)

15:40-16:05

**Segmented Initialization and Offspring Modification in Evolutionary Algorithms for Bi-objective Feature Selection**
Hang Xu, Bing Xue, Mengjie Zhang

16:05-16:30

**Relatedness Measures to Aid the Transfer of Building Blocks among Multiple Tasks**
Trung Nguyen, Will Neil Browne, Mengjie Zhang

16:30-16:55

**Feature Standardisation and Coefficient Optimisation for Effective Symbolic Regression**
Grant Dick, Caitlin A. Owen, Peter A. Whigham

16:55-17:20

**Re-purposing Heterogeneous Generative Ensembles with Evolutionary Computation**
Jamal Toutouh, Erik Hemberg, Una-May O’Reilly

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**GP2**

Chair: Mengjie Zhang (Victoria University of Wellington)

15:40-16:05

**Adaptive Weighted Splines - A New Representation to Genetic Programming for Symbolic Regression**
Christian Raymand, Qi Chen, Bing Xue, Mengjie Zhang

16:05-16:30

**Constructing Efficient Multigrid Solvers with Genetic Programming**
Jonas Schmitt, Sebastian Kuckuk, Harald Köstler

16:30-16:55

**Symbolic Regression Driven by Training Data and Prior Knowledge**
Jiri Kubalik, Erik Derner, Robert Babuska

16:55-17:20

**Improving Symbolic Regression based on Correlation Between Residuals and Variables**
Qi Chen, Bing Xue, Mengjie Zhang
**SBSE1+ENUM2+THEORY1**
Chair: Justyna Petke (University College London)
(Best Paper nominees are marked with a star)

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<td>Leveraging Conditional Linkage Models in Gray-box Optimization with the Real-Valued Gene-pool Optimal Mixing Evolutionary Algorithm</td>
<td>Anton Bouter, Stefanus C. Maree, Tanja Alderliesten, Peter A. N. Bosman</td>
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<td>14:05-14:30</td>
<td>Seeding Strategies for Multi-Objective Test Case Selection: An Application on Simulation-based testing</td>
<td>Aitor Arrieta, Joseba Andoni Agirre, Goiuria Sagardui</td>
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<td>Denis Antipov, Maxim Buzdalov, Benjamin Doerr</td>
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<td>Causes and Effects of Fitness Landscapes in Unit Test Generation</td>
<td>Nasser Albunian, Gordon Fraser, Dirk Sudholt</td>
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**ACO-SI2**
Chair: Jürgen Branke (University of Warwick)

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<td>An Approach to Assess Swarm Intelligence Algorithms Based on Complex Networks</td>
<td>Clodomir Joaquim Santana, Edward Keedwell, Ronaldo Menezes</td>
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<td>Exploratory Path Planning for Mobile Robots in Dynamic Environments with Ant Colony Optimization</td>
<td>Valéria de Carvalho Santos, Fernando E. B. Otero, Colin Johnson, Fernando Santos Osório, Cláudio Toledo</td>
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<td>A Weighted Population Update Rule for PACO Applied to the Single Machine Total Weighted Tardiness Problem</td>
<td>Daniel Abitz, Tom Hartmann, Martin Middendorf</td>
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**RWA2**
Chair: Martin Zaefferer (TH Köln)

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<td>COUGAR: Clustering Of Unknown malware using Genetic Algorithm Routines</td>
<td>Zachary Wilkins, Nur Zincir-Heywood</td>
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<td>Variable Reduction for Surrogate-Based Optimization</td>
<td>Frederik Rehbach, Lorenzo Gentile, Thomas Bartz-Beielstein</td>
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EML2  Saturday, July 11, 13:40-15:20
Chair: Krzysztof Krawiec (Poznan University of Technology, Center for Artificial Intelligence and Machine Learning)

Safe Crossover of Neural Networks Through Neuron Alignment  13:40-14:05
Thomas Uriot, Dario Izzo

Exploring the Evolution of GANs through Quality Diversity  14:05-14:30
Victor Costa, Nuno Lourenço, João Correia, Penousal Machado

XCS Classifier System with Experience Replay  14:30-14:55
Anthony Stein, Roland Maier, Lukas Rosenbauer, Jörg Hähner

Evolving Inborn Knowledge For Fast Adaptation in Dynamic POMDP Problems  14:55-15:20
Eseoghene Ben-Iwhiwhu, Pawel Ladosz, Jeffery Dick, Wen-Hua Chen, Praveen Pilly, Andrea Soltoggio

GP3  Saturday, July 11, 13:40-15:20
Chair: Lee Spector (Amherst College, Hampshire College)

Unlabeled Multi-Target Regression with Genetic Programming  13:40-14:05
Uriel López, Leonardo Trujillo, Sara Silva, Pierrick Legrand, Leonardo Vanneschi

Synthesis through Unification Genetic Programming  14:05-14:30
Thomas Welsch, Vitaliy Kurlin

Towards an evolutionary-based approach for natural language processing  14:30-14:55
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ECOM2  Saturday, July 11, 13:40-15:20
Chair: Manuel López-Ibáñez (University of Manchester)

Michael Foster, Matthew Hughes, George O’Brien, Pietro S. Oliveto, James Pyle, Dirk Sudholt, James Williams

A Robust Experimental Evaluation of Automated Multi-Label Classification Methods  14:05-14:30
Alex G. C. de Sá, Cristiano G. Pimenta, Gisele L. Pappa, Alex A. Freitas

Solving Constrained Combinatorial Reverse Auctions Using MOEAs: A Comparative Study.  14:30-14:55
Elaine Guerrero-Peña, Fernanda Nakano Kazama, Paulo de Barros Correia, Aluizio F.R. Araújo

Journey to the Center of the Linear Ordering Problem  14:55-15:20
Leticia Hernando, Alexander Mendiburu, Jose A. Lozano
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**Analysis of the Performance of Algorithm Configurators for Search Heuristics with Global Mutation Operators** ★
George T. Hall, Pietro S. Oliveto, Dirk Sudholt
15:40-16:05

**ε-shotgun: ε-greedy Batch Bayesian Optimisation** ★
George De Ath, Richard M. Everson, Jonathan E. Fieldsend, Alma A. M. Rahat
16:05-16:30

**Effective Reinforcement Learning through Evolutionary Surrogate-Assisted Prescription** ★
Olivier Francon, Santiago Gonzalez, Babak Hodjat, Elliot Meyerson, Risto Miikkulainen, Xin Qiu, Hormoz Shahrzad
16:30-16:55

**An Evolutionary Optimization Algorithm for Gradually Saturating Objective Functions**
Dolly Sapra, Andy D. Pimentel
16:55-17:20

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**Genetic programming approaches to learning fair classifiers** ★
William La Cava, Jason H. Moore
15:40-16:05

**Semantically-Oriented Mutation Operator in Cartesian Genetic Programming for Evolutionary Circuit Design** ★
David Hodan, Vojtech Mrazek, Zdenek Vasicek
16:05-16:30

**A Modular Memory Framework for Time Series Prediction** ★
Stephen Paul Kelly, Jacob Newsted, Wolfgang Banzhaf, Cedric Gondro
16:30-16:55

**Multi-Tree Genetic Programming for Feature Construction-Based Domain Adaptation in Symbolic Regression with Incomplete Data**
Baligh Al-Helali, Qi Chen, Bing Xue, Mengjie Zhang
16:55-17:20

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**Impact of NSGA-II Objectives on EEG Feature Selection Related to Motor Imagery**
Miguel Leon, Christoffer Parkkila, Jonatan Tidare, Ning Xiong, Elaine Astrand
15:40-16:05

**Non-deterministic Journey Planning in Multi-modal Transportation Networks: A Meta-heuristic Approach**
Mohammad Haqqani, Xiaodong Li, Xinghuo Yu
16:05-16:30

**A Genetic Programming Approach to Feature Construction for Ensemble Learning in Skin Cancer Detection**
Qurrat Ul Ain, Harith Al-sahaf, Bing Xue, Mengjie Zhang
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**Surrogate-Assisted Asynchronous Multiobjective Algorithm for Nuclear Power Plant Operations**
Valentin Drouet, Sébastien Verel, Jean-Michel Do
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<td><strong>Neural Architecture Search for Sparse DenseNets with Dynamic Compression</strong></td>
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<td><strong>CPPN2GAN: Combining Compositional Pattern Producing Networks and GANs for Large-scale Pattern Generation</strong></td>
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Pavel Krömer, Jan Platos, Vaclav Snasel
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A Deep Learning Approach to Predicting Solutions in Streaming Optimisation Domains
Mohamad Alissa, Kevin Sim, Emma Hart
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Why Many Traveling Salesman Problem Instances Are Easier Than You Think
Swetha Varadarajan, Darrell Whitley, Gabriela Ochoa
16:55-17:20
Neuroevolution of Self-Interpretable Agents ★
Yujin Tang, Duong Nguyen, David Ha 17:40-18:05

Improving Neuroevolutionary Transfer Learning of Deep Recurrent Neural Networks through Network-Aware Adaptation ★
AbdElRahman ElSaid, Joshua Karns, Zimeng Lyu, Travis Desell, Daniel Krutz, Alexander Ororbia II 18:05-18:30

Improving Constrained Clustering Via Decomposition-based Multiobjective Optimization with Memetic Elitism ★
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GeneCAI: Genetic Evolution for Acquiring Compact AI
Mojan Javaheripi, Mohammad Samragh, Tara Javidi, Farinaz Koushanfar 18:55-19:20

Towards Sustainable Forest Management Strategies with MOEAs ★
Philipp Back, Antti Suominen, Pekka Malo, Olli Tahvonen, Julian Blank, Kalyanmoy Deb 17:40-18:05

Optimisation of Large Wave Farms using a Multi-strategy Evolutionary Framework ★
Mehdi Neshat, Bradley Alexander, Nataliia Y. Sergiienko, Markus Wagner 18:05-18:30

Evolutionary Bin Packing for Memory-Efficient Dataflow Inference Acceleration on FPGA
Mairin Kroes, Lucian Petrica, Sorin Cotofana, Michaela Blott 18:30-18:55

Simultaneously Searching and Solving Multiple Avoidable Collisions for Testing Autonomous Driving Systems

Adaptively Preserving Solutions in Both Feasible and Infeasible Regions on Generalized Multiple Constraint Ranking ★
Yohanes Bimo Dwianto, Hiroaki Fukumoto, Akira Oyama 17:40-18:05

Understanding Transforms of Pseudo-Boolean Functions ★
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Evolving Diverse Sets of Tours for the Travelling Salesperson Problem
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Modelling Parameter Configuration Spaces with Local Optima Networks
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Adam Gaier, Alexander Asteroth, Jean-Baptiste Mouret  
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**Covariance Matrix Adaptation for the Rapid Illumination of Behavior Space**  
Matthew C. Fontaine, Julian Togelius, Stefanos Nikolaidis, Amy K. Hoover  
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**Towards Crossing the Reality Gap with Evolved Plastic Neurocontrollers**  
Huanneng Qiu, Matthew Garratt, David Howard, Sreenatha Anavatti  
18:30-18:55

**Novelty Search makes Evolvability Inevitable**  
Stéphane Doncieux, Giuseppe Paolo, Alban Laflaquière, Alexandre Coninx  
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**The Node Weight Dependent Traveling Salesperson Problem: Approximation Algorithms and Randomized Search Heuristics**  
Jakob Bossek, Katrin Casel, Pascal Kerschke, Frank Neumann  
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**The (1 + (λ, λ)) GA Is Even Faster on Multimodal Problems**  
Denis Antipov, Benjamin Doerr, Vitalii Karavaev  
18:05-18:30

**More Effective Randomized Search Heuristics for Graph Coloring Through Dynamic Optimization**  
Jakob Bossek, Frank Neumann, Pan Peng, Dirk Sudholt  
18:30-18:55

**Fixed-Target Runtime Analysis**  
Maxim Buzdalov, Benjamin Doerr, Carola Doerr, Dmitri Vinokurov  
18:55-19:20

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**Multiobjective Tree-structured Parzen Estimator for Computationally Expensive Optimization Problems**  
Yoshihiko Ozaki, Yuki Tanigaki, Shuhei Watanabe, Masaki Onishi  
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**Surrogate Assisted Evolutionary Algorithm for Medium Scale Multi-Objective Optimisation Problems**  
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**Effects of Dominance Resistant Solutions on the Performance of Evolutionary Multi-Objective and Many-Objective Algorithms**  
Hisao Ishibuchi, Takashi Matsumoto, Naoki Masuyama, Yusuke Nojima  
18:30-18:55

**Runtime Analysis of Evolutionary Algorithms with Biased Mutation for the Multi-Objective Minimum Spanning Tree Problem**  
Vahid Roostapour, Jakob Bossek, Frank Neumann  
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<td>Jakob Bossek, Christian Grimme, Heike Trautmann</td>
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<td>Muhammad Sheraz Anjum, Conor Ryan</td>
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<td>Enhancing Search-Based Product Line Design with Crossover Operators</td>
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<td>Multi-Objective Optimal Distribution of Materials in Hybrid Components</td>
<td>Thomas Gossuin, Didier Garray, Vincent Kelner</td>
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<td>Hybrid Genetic Algorithm for Ridesharing with Timing Constraints: Efficiency analysis with Real-World Data</td>
<td>Nirav Patel, N. S. Narayanaswamy, Alok Joshi</td>
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<td>EML5</td>
<td>Evolutionary Optimization of Deep Learning Activation Functions</td>
<td>Garrett Bingham, William Macke, Risto Miikkulainen</td>
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<td>Multi-Fitness Learning for Behavior-Driven Cooperation</td>
<td>Connor Yates, Reid Christopher, Kagan Tumer</td>
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<td>Program Synthesis as Latent Continuous Optimization: Evolutionary Search in Neural Embeddings</td>
<td>Pawel Liskowski, Krzysztof Krawiec, Nihat Engin Toklu, Jerry Swan</td>
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<td>Self-adaptation of XCS learning parameters based on Learning theory</td>
<td>Motoki Horiuchi, Masaya Nakata</td>
<td>10:15-10:40</td>
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<td>THEORY3</td>
<td>A Tight Lower Bound on the Expected Runtime of Standard Steady State Genetic Algorithms</td>
<td>Carsten Witt, Pietro S. Oliveto, Dirk Sudholt</td>
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<td>Does Comma Selection Help To Cope With Local Optima?</td>
<td>Benjamin Doerr</td>
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<td>Self-Adjusting Evolutionary Algorithms for Multimodal Optimization</td>
<td>Amirhossein Rajabi, Carsten Witt</td>
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<td><strong>EMO4</strong></td>
<td><strong>Gap Finding and Validation in Evolutionary Multi- and Many-Objective Optimization</strong></td>
<td>Pablo Valledor Pellicer, Miguel Iglesias Escudero, Silvino Fernández Alzueta, Kalyanmoy Deb</td>
<td>09:00-09:25</td>
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<td><strong>If Unsure, Shuffle: Deductive Sort is Θ(MN^3), but O(MN^2) in Expectation over Input Permutations</strong></td>
<td>Sumit Mishra, Maxim Buzdalov</td>
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<td><strong>What is a Good Direction Vector Set for The R2-based Hypervolume Contribution Approximation</strong></td>
<td>Yang Nan, Ke Shang, Hisao Ishibuchi</td>
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<td><strong>Surrogate-assisted Multi-objective Combinatorial Optimization based on Decomposition and Walsh Basis</strong></td>
<td>Geoffrey Pruvost, Bilel Derbel, Arnaud Liefooghe, Sébastien Verel, Qingfu Zhang</td>
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<td><strong>GECH4</strong></td>
<td><strong>Sensitivity Analysis in Constrained Evolutionary Optimization</strong></td>
<td>Julian Schulte, Volker Nissen</td>
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<td><strong>Expected Improvement versus Predicted Value in Surrogate-Based Optimization</strong></td>
<td>Frederik Rehbach, Martin Zaefferer, Boris Naujoks, Thomas Bartz-Beielstein</td>
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<td><strong>Algorithm Selection of Anytime Algorithms</strong></td>
<td>Alexandre D. Jesus, Arnaud Liefooghe, Bilel Derbel, Luís Paquete</td>
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<td><strong>From Understanding Genetic Drift to a Smart-Restart Parameter-less Compact Genetic Algorithm</strong></td>
<td>Benjamin Doerr, Weijie Zheng</td>
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A preliminary study towards an improved shepherding model
Heba El-Fiqi, Benjamin Campbell, Saber Elsayed, Anthony Perry, Hemant Kumar Singh, Robert Hunjet, Hussein Abbass

Wind-turbine design optimization using a many-objective evolutionary algorithm
Ahsanul Habib, Kamrul Hasan Rahi, Hemant Kumar Singh, Tapabrata Ray

Expediting the convergence of evolutionary algorithms by identifying promising regions of the search space
Kamrul Hasan Rahi, Ahsanul Habib, Hemant Kumar Singh, Tapabrata Ray

Evolving an Artificial Creole
Geoff Nitschke, Gregory Furman

The Expense of Neuro-Morpho Functional Machines
Geoff Nitschke, Scott Hallauer

Diversity-based Design Assist for Large Legged Robots
Gerard Howard, Thomas Lowe, Wade Geles

Divergent Search for Image Classification Behaviors
Jeremy Tan, Bernhard Kainz

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Exploring the BipedalWalker benchmark with MAP-Elites and Curiosity-driven A3C
Vikas Gupta, Nathanael Aubert-Kato, Leo Cazenille

Evolving Neural Network Agents to Play Atari Games with Compact State Representations
Adam Tupper, Kourosh Neshatian,

QoS-Constrained Multi-Objective Distributed Data-intensive Web Service Composition - NSGA-II with Repair Method
Soheila Sadeghiram, Hui Ma, Gang Chen

Divide and Conquer: Seeding Strategies for Multi-Objective Multi-Cloud Composite Applications Deployment
Tao Shi, Hui Ma, Aaron Chen

A Preliminary Approach to Evolutionary Multitasking for Dynamic Flexible Job Shop Scheduling via Genetic Programming
Fangfang Zhang, Yi Mei, Su Nguyen, Mengjie Zhang

Improving an Evolutionary Wrapper for Attack Detection by Including Feature Importance Information
Javier Alexis Maldonado, Maria Cristina Riff

Towards Evolving Robust Neural Architectures to Defend from Adversarial Attacks
Shashank Kotyan, Danilo Vasconcellos Vargas,
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A Genetic Programming Method for Classifier Construction and Cost Learning in High-dimensional Unbalanced Classification
   Wenbin Pei, Bing Xue, Lin Shang, Mengjie Zhang

Evolutionary Neural Network Structure Search for DNN Pruning and Features Separation
   Zhaoyang Wu, Lin Lin, Guoliang Gong, Rui Xu, Mitsuo Gen, Yong Zhou

Evolving Network Structures for Text Classification using Genetic Algorithm
   Hayden Andersen, Xiaoying Gao, Bing Xue, Mengjie Zhang

Evolving Deep Autoencoders
   Jeff Hajewski, Suely Oliveira, Xiaoyu Xing

A Multi-objective Architecture Search for Generative Adversarial Networks
   Masayuki Kobayashi, Tomoharu Nagao

An Evolution-based Approach for Efficient Differentiable Architecture Search
   Masayuki Kobayashi, Tomoharu Nagao

Towards a Pittsburgh-style LCS for Learning Manufacturing Machinery Parametrizations
   Michael Heider, David Pätzelt, Jörg Hähner

Poster Group 4

BACS: Integrating Behavioral Sequences to ACS2
   Romain Orhand, Anne Jeannin-Girardon, Pierre Parrend, Pierre Collet

Automatically Extracting Features for Face Classification Using Multi-Objective Genetic Programming
   Ying Bi, Bing Xue, Mengjie Zhang

Population-based Evolutionary Distributed SGD
   Amna Shahab, Boris Grot

Performance evaluation of the MOEA/D algorithm for the solution of a microgrid planning problem
   Miguel A. Rodriguez, Dario F. Lopez, Sergio F. Contreras, Camilo A. Cortés, Johanna M.A Myrzik

Dynamic Multi-objective Optimisation Problems with Intertemporal Dependencies
   Bernard Van Tonder, Marde Helbig

Improving NSGA-III for Flexible Job Shop Scheduling using Automatic Configuration, Smart Initialization and Local Search
   Yali Wang, Bas van Stein, Michael Emmerich, Thomas Bäck

Using Gradient-free Local Search within MOEAs for the Treatment of Constrained MOPs
   Lourdes Uribe, Adriana Lara, Kalyanmoy Deb, Oliver Schütze

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An Improved Multiobjective Optimization Evolutionary Algorithm Based on Decomposition with Hybrid Penalty Scheme
   Jinglei Guo, Miaomiao Shao, Shouyong Jiang, Shengxiang Yang

Parametrized Benchmarking: an Outline of the Idea and a Feasibility Study
   Karol R. Opara, Anas A. Hadi, Ali W. Mohamed
Multiobjective Ant Colony Optimization for Task Allocation in Vehicle-Based Crowdsourcing  
Jian Shi, Wei-Neng Chen

Look-Ahead Natural Evolutionary Strategies  
Jin Liang Jia, Alfredo Alan Flores Saldivar, Lin Li, Yun Li

Solving Min-Max Optimisation Problems by Means of Bilevel Evolutionary Algorithms: A Preliminary Study  
Margarita Antoniou, Gregor Papa

Continuous Optimization by Hierarchical Gaussian Mixture with Clustering Embedded Resource Allocation  
Jhih-Wei Chen, Ming-Chun Lu, Tian-Li Yu

Performance2vec: A step further in explainable stochastic optimization algorithms' performance  
Tome Eftimov, Gorjan Popovski, Dragi Kocev, Peter Korošec

Poster Group 6

A view of Estimation of Distribution Algorithms through the lens of Expectation-Maximization  
David H. Brookes, Akosua Busia, Clara Wong-Fannjiang, Kevin Murphy, Jennifer Listgarten

A Fast GA for Automatically Evolving CNN Architectures  
Zhou Hong, Wei Fang, Jun Sun, Xiaojun Wu

SHX: Search History Driven Crossover for Real-Coded Genetic Algorithm  
Takumi Nakane, Xuequan Lu, Chao Zhang

Cooperative Coevolutionary Genetic Algorithm Using Hierarchical Clustering of Linkage Tree  
Takatoshi Niwa, Koya Ihara, Shohei Kato

Impact of additional hardware resources on a parallel genetic algorithm  
Glen Cancian, Wayne Pullan

A Test Problem with Difficulty in Decomposing into Sub-problems for Model-based Genetic Algorithms  
Kei Ohnishi, Shota Ikeda, Tian-Li Yu

Poster Group 7

Evolving Genetic Programming Trees in a Rule-Based Learning Framework  
Siddharth Verma, Piyush Borole, Ryan J. Urbanowicz

General Controllers evolved through Grammatical Evolution with a Divergent Search  
Enrique Naredo, Conor Ryan, Ivan Hugo Guevara, Tiziana Margaria, Paulo Urbano, Leonardo Trujillo

Batch Correction of Genomic Data in Chronic Fatigue Syndrome Using CMA-ES  
Alejandro Lopez Rincon, Aletta D. Kraneveld, Alberto Tonda

Stochastic Simulation Optimization Benchmarking Method in Consideration of Finite Period of Service  
Takufumi Yoshida, Daisuke Yamaguchi, Keiichi Handa

An improved brain storm optimization algorithm for fuzzy distributed hybrid flowshop scheduling with setup time  
Junqing Li, Hui Yu, Xiaolong Chen, Wenhan Li, Yu Du, Yuyan Han

Towards Realistic Optimization Benchmarks: A Questionnaire on the Properties of Real-World Problems  
Koen van der Blom, Timo M. Deist, Tea Tušar, Mariapia Marchi, Yusuke Nojima, Akira Oyama, Vanessa Volz, Boris Naujoks
Moving target defense through evolutionary algorithms
Ernesto Serrano Collado, Mario García Valdez, JJ Merelo

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Optimising Word Embeddings With Search-Based Approaches
Max Hort, Federica Sarro

A Realistic Scooter Rebalancing System via Metaheuristics
Guilherme Fernandes, Nuno Oliveira, Paulo Cortez, Rui Mendes

Evolving Multi-Objective Ranking Models For GMV Optimization in E-Commerce
Andrew Stanton, Akhila Ananthram, Pablo Crespo

Multi-Objective Optimization for Worker Cross-Training - The Tri-Objective Case
Andreas Beham, Viktoria A. Hauder, Johannes Karder, Klaus Altendorfer

Preliminary study of applied Binary Neural Networks for Neural Cryptography
Raul Horacio Valencia Tenorio, Chiu Wing Sham, Danilo Vasconcellos Vargas

A GA for non-uniform sampling harmonic analysis
Ulviya Abdulkarimova, Igor Santos Peretta, Anna Ouskova Leonteva, Younes Monjid, Rabih Amhaz, Pierre Collet, Christian Rolando, Marc Haegelin

Multi-objective Exploration of a Granular Matter Design Space
Gary W. Delaney, Gerard Howard

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An Artificial Sequential Immune Responses Model for Anomaly Detection
Wen Zhou, Yiwen Liang

Beer Organoleptic Optimisation: Utilising Swarm Intelligence and Evolutionary Computation Methods
Mohammad Majid Al-Rifaie, Marc Cavazza

Multiobjective Direction Driven Local Search for Constrained Supply Chain Configuration Problem
Xin Zhang, Zhi-Hui Zhan, Jun Zhang

Search-Based Many-Criteria Identification of Microservices from Legacy Systems
Luiz Carvalho Carlos Felix, Alessandro García, Thelma Elita Colanzi, Wesley K. G. Assunção, Maria Julia Lima, Baldoino Fonseca, Márcio Ribeiro, Carlos Lucena

Safer Reinforcement Learning through Evolved Instincts
Djordje Grbic, Sebastian Risi

Combining Sequential Model-Based Algorithm Configuration with Default-Guided Probabilistic Sampling
Marie Anastacio, Holger H. Hoos

 Recommending Peer Reviewers in Modern Code Review : A Multi-Objective Search-based Approach
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Crash Reproduction Using Helper Objectives
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Population Control meets Doob's Martingale Theorems: the Noise-free Multimodal Case
Marie-Liesse Cauwet, Olivier Teytaud

Redundant Binary Representations with Rigorous Tradeoff Between Connectivity and Locality
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Delia-Elena Dumitru, Anca Andreica, Laura Diosan, Zoltan Balint

The effect of differential quality and differential zealotry in the best-of-n problem
Judhi Prasetyo, Giulia De Masi, Elio Tuci, Eliseo Ferrante

Improving BPSO-based feature selection applied to offline WI handwritten signature verification through overfitting control
Victor L. F. Souza, Adriano L. I. Oliveira, Rafael M. O. Cruz, Robert Sabourin

Adaptive Reinforcement Learning through Evolving Self-Modifying Neural Networks
Samuel Robert Schmidgall

Evolving HyperNetworks for Game-Playing Agents
Christian Carvelli, Djordje Grbic, Sebastian Risi

Novelty Producing Synaptic Plasticity
Anil Yaman, Giovanni Iacca, Decebal Constantin Mocanu, George Fletcher, Mykola Pechenizkiy

Comparing Indirect Encodings by Evolutionary Attractor Analysis in the Trait Space of Modular Robots
Matteo De Carlo, Eliseo Ferrante, Agoston E. Eiben

Poster Group 2

Diversity in Swarm Robotics with Task-independent Behavior Characterization
Tanja Katharina Kaiser, Heiko Hamann

Ádea — Evolving glyphs for aiding creativity in typeface design
Daniel Lopes, Penousal Machado, João Correia

Evolving Search Trajectories
Vincent Hénaux, Adrien Goëffon, Frédéric Saubion

An Efficient Evolutionary Solution to the joint Order Batching - Order Picking Planning Problem
Riccardo Lucato, Jonas Falkner, Lars Schmidt-Thieme

Equilibrium in Classification. A New Game Theoretic Approach to Supervized Learning
Rodica Ioana Lung, Mihai Suciu

Evolutionary Super Resolution
Baptiste Roziere, Nathanaël Carraz Rakotonirina, Vlad Hosu, Hanhe Lin, Andry Rasoanaivo, Olivier Teytaud, Camille Couprie

Turing Learning with Hybrid Discriminators: Combining the Best of Active and Passive Learning
Yue Gu, Wei Li, Roderich Gross
**Poster Group 3**

Generation of Consistent Sets of Multi-Label Classification Rules with a Multi-Objective Evolutionary Algorithm  
Thiago Zafalon Miranda, Diorge Brognara Sardinha, Márcio Porto Basgalupp, Yaochu Jin, Ricardo Cerri

Diversity-Driven Wide Learning for Training Distributed Classification Models  
Rui Cardoso, Emma Hart, Jeremy Pitt

HyperFDA: A Bi-level optimization approach to Neural Architecture Search and Hyperparameters’ Optimization via Fractal Decomposition-based Algorithm  
Leo Souquet, Nadiya Shvai, Arcadi Llanza, Amir Nakib

Searching for Activation Functions Using a Self-Adaptive Evolutionary Algorithm  
Andrew Nader, Danielle Azar

DarwiNN: Efficient Distributed Neuroevolution under Communication Constraints  
Gurshaant Malik, Lucian Petrica, Nachiket Kapre, Michaela Blott

An adaptive neuroevolution-based hyperheuristic  
Etor Arza, Josu Ceberio, Aritz Pérez, Ekhiñe Irurozki

The data-driven physical-based equations discovery using evolutionary approach  
Alexander Hvatov, Mikhail Maslyaev

**Poster Group 4**

Multi-objective Data Stream Clustering  
Mohammed Oualid Attaoui, Hanene Azzag, Mustapha Lebbah, Nabil Keskes

A Generic and Computationally Efficient Automated Innovization Method for Power-Law Design Rules  
Kanish Garg, Anish Mukherjee, Sukrit Mittal, Dhish Kumar Saxena, Kalyanmoy Deb

Dynamic Vessel-to-Vessel Routing Using Level-wise Evolutionary Optimization  
Yash Vesikar, Julian Blank, Kalyanmoy Deb, Markku Kallio, Alaleh Maskooki

Learning-based Multi-objective Optimization Through ANN-Assisted Online Innovization  
Sukrit Mittal, Dhish Kumar Saxena, Kalyanmoy Deb

PaletteStarViz: A Visualization Method for Multi-criteria Decision Making from High-dimensional Pareto-optimal Front  
Khaled Talukder, Kalyanmoy Deb

Worst-case Conditional Hardness and Fast Algorithms with Random Inputs for Non-dominated Sorting  
Sorrachai Yingchareonthawornchai, Proteek Chandan Roy, Bundit Laekhanukit, Eric Torng, Kalyanmoy Deb

Time Complexity Analysis of the Dominance Degree Approach for Non-Dominated Sorting  
Sumit Mishra, Maxim Buzdalov, Rakesh Senwar

**Poster Group 5**

A First Step Towards Incremental Evolution of Convolutional Neural Networks  
Dustin K. Barnes, Sara R. Davis, Emily M. Hand, Sushil J. Louis
Framework to select an improved radiographic image using Speed-constrained Modified Particle Swarm Optimization
Luhana Aracelli González Zarza, Alan Mathías Ruiz Díaz Nodari, José Luis Vázquez Noguera, Diego Pedro Pinto Roa

An EDA with Swarm Intelligence for the multi-objective Flexible Job-Shop Problem
Luiz Carlos Felix Carvalho, Márzia Aparecida Fernandes

A Local Hypervolume Contribution Schema to Improve Spread of the Pareto Front and Computational Time
Edgar Manoatl López, Carlos Ignacio Hernández Castellanos

A Many-Objective Route Planning Benchmark Problem for Navigation
Jens Weise, Sanaz Mostaghim

A Quantum Simulation Algorithm For Continuous Optimization
Anna Ouskova Leonteva, Ulviya Abdulkarimova, Anne Jeannin-Girardon, Pierre Parrend, Tobias Martin Wintermantel, Pierre Collet

Differential Evolution with Reversible Linear Transformations
Jakub Mikolaj Tomczak, Ewelina Marta Weglarz-Tomczak, Agoston E. Eiben

Poster Group 6

Differential Evolution with Explicit Control of Diversity for Constrained Optimization
Gabriel Tadeo Vázquez Ballesteros, Carlos Segura

Illuminating Super Mario Bros - Quality-Diversity Within Platformer Level Generation
Oliver Henry Withington

A Parallel and Distributed Multi-population GA with Asynchronous Migrations: Energy-time Analysis for Heterogeneous Systems
Juan José Escobar, Julio Ortega, Antonio Francisco Díaz, Jesús González, Miguel Damas

Efficient Machine Learning through Evolving Combined Deep Neural Networks
Rune Krauss, Marcel Merten, Mirco Bockholt, Saman Froehlich, Rolf Drechsler

Automated Design of Efficient Swarming Behaviours: a Q-Learning Hyper-Heuristic Approach
Gabriel Valentin Dufló, Grégoire Danoy, El-Ghazali Talbi, Pascal Bouvry

Hybrid Bayesian Evolutionary Optimization for Hyperparameter Tuning
Lukas Atkinson, Robin Müller-Bady, Martin Kappes

On the Effect of Walsh/Fourier Transform in Surrogate-assisted Genetic Algorithms
Dong-Pil Yu, Yong-Hyuk Kim

Poster Group 7

Finding a Better Basis on Binary Representation through DNN-based Epistasis Estimation
Yong-Hoon Kim, Yong-Hyuk Kim

Parallelized Bayesian Optimization for Problems with Expensive Evaluation Functions
Margarita Alejandra Rebolledo Coy, Frederik Rehbach, Agoston E. Eiben, Thomas Bartz-Beielstein

Feature Engineering for Improving Robustness of Crossover in Symbolic Regression
Aliyu Sani Sambo, R. Muhammad Atif Azad, Yevgeniya Kovalchuk, Vivek Padmanaabhan Indramohan, Hanifa Shah
Image Feature Learning with a Genetic Programming Autoencoder
Stefano Ruberto, Valerio Terragni, Jason H. Moore

Benchmarking Parent Selection for Program Synthesis by Genetic Programming
Thomas Helmuth, Amr Abdelhady

Transfer Learning of Genetic Programming Instruction Sets
Thomas Helmuth, Edward Pantridge, Grace Woolson, Lee Spector

Counterexample-Driven Genetic Programming without Formal Specifications
Thomas Helmuth, Lee Spector, Edward Pantridge

Poster Group 8

Why and When are Loops Useful in Genetic Programming?
Anil Kumar Saini, Lee Spector

Refined Typed Genetic Programming as a user interface for Genetic Programming
Paulo Santos, Sara Silva, Alcides Fonseca

Enabling XCSF to Cope with Dynamic Environments via an Adaptive Error Threshold
Tim Hansmeier, Paul Kaufmann, Marco Platzner

Effective Image Clustering using Self-Organizing Migrating Algorithm
Seyed Jalaleddin Mousavirad, Gerald Schaefer, Iakov Korovin

Evolved Ensemble of Detectors for Gross Error Detection
Thanh Tien Nguyen, John McCall, Allan Wilson, Lau Ochei, Helen Corbett, Phil Stockton

A new objective function for super-resolution deconvolution of microscopy images by means of a Genetic Algorithm
Axel Mauro Lacapmesure, Sandra Martínez, Oscar Martínez

A Genetic Algorithm for Matching Oil Spill Particles
Hyeon-Chang Lee, Hwi-Yeon Cho, Yong-Hyuk Kim

Poster Group 9

An application of GA and EDA for Passive In-Building Distributed Antenna Systems
Siddhartha Shakya, Kin Poon, Khawla AlShanqiti, Anis Ouali, Andrei Sleptchenko

Real-time Genetic Optimization of Large File Transfers
Hemanta Sapkota, Engin Arslan, Sushil J. Louis

Cable-Stayed Bridge Optimization Solution Space Exploration
João Correia, Fernando Ferreira, Catarina Maçãs

NGAP: A Novel Hybrid Metaheuristic for Round-trip Carsharing Fleet Planning
Boonyarit Changaival, Grégoire Danoy, Dzmitry Kliazovich, Frédéric Guinand, Matthias R. Brust, Jedrzej Musial, Kittichai Lavangnananda, Pascal Bouvry

Exploiting Fault Localisation for Efficient Program Repair
Vesna Nowack, David Bowes, Steve Counsell, Tracy Hall, Saemundur O. Haraldsson, John R. Woodward, Emily Winter
A New Approach to Distribute MOEA Pareto Front Computation
Federica Sarro, Alessio Petrozziello, Dan-Qi He, Shin Yoo

On the Prediction of Continuous Integration Build Failures Using Search-Based Software Engineering
Islem Saidani, Ali Ouni, Moataz Chouchen, Mohamed Wiem Mkaouer

Poster Group 10

Handling Uncertainty in Code Smells Detection using a Possibilistic SBSE Approach
Sofien Boutaib, Slim Bechikh, Carlos Artemio Coello Coello, Chih-Cheng Hung, Lamjed Ben Said

Critical evaluation of Sine Cosine Algorithm and a few recommendations
Qamar Askari, Irfan Younas, Mehreen Saeed

Radial Model of Differential Evolution Dynamics
Karol R. Opara

Bayesian CMA-ES
David Saltiel, Eric Benhamou, Sébastien Verel
Abstracts by Track
Ant Colony Optimization and Swarm Intelligence

**CS2+ACO-SI1**
Friday, July 10, 13:40-15:20

**A Multiobjective Optimization Approach for Market Timing ★**
Ismail Mohamed, *University of Kent*, Fernando Esteban Barril Otero, *University of Kent*

The introduction of electronic exchanges was a crucial point in history as it heralded the arrival of algorithmic trading. Designers of such systems face a number of issues, one of which is deciding when to buy or sell a given security on a financial market. Although Genetic Algorithms (GA) have been the most widely used to tackle this issue, Particle Swarm Optimization (PSO) has seen much lower adoption within the domain. In two previous works, the authors adapted PSO algorithms to tackle market timing and address the shortcomings of the previous approaches both with GA and PSO. The majority of work done to date on market timing tackled it as a single objective optimization problem, which limits its suitability to live trading as designers of such strategies will realistically pursue multiple objectives such as maximizing profits, minimizing exposure to risk and using the shortest strategies to improve execution speed. In this paper, we adapt both a GA and PSO to tackle market timing as a multiobjective optimization problem and provide an in depth discussion of our results and avenues of future research.

**ACO-SI2**
Saturday, July 11, 13:40-15:20

**ACO with automatic parameter selection for a scheduling problem with a job group cumulative constraint**

We consider a RCPSP (resource constrained project scheduling problem), the goal of which is to schedule jobs on machines in order to minimise job tardiness. This problem comes from a real industrial application, and it requires an additional constraint which is a generalisation of the classical cumulative constraint: jobs are partitioned into groups, and the number of active groups must never exceed a given capacity (where a group is active when some of its jobs have started while some others are not yet completed). We first study the complexity of this new constraint. Then, we describe an Ant Colony Optimisation algorithm to solve our problem, and we introduce three different pheromone structures for it. We study the influence of parameters on the solving process, and show that it varies from an instance to another. Hence, we identify a subset of parameters settings with complementary strengths and weaknesses, and we use machine learning to automatically select the best setting for each new instance to solve. We experimentally compare our approach with a tabu search approach and an exact approach on a data set coming from our industrial application.

**An Approach to Assess Swarm Intelligence Algorithms Based on Complex Networks**

The growing number of novel swarm-based meta-heuristics has been raising debates regarding their novelty. These algorithms often claim to be inspired by different concepts from nature but the proponents of these seldom demonstrate whether the novelty goes beyond the nature inspiration. In this work, we employed the concept of interaction networks to capture the interaction patterns that take place in algorithms during the optimisation process. The analyses of these networks reveal aspects of the algorithm such as the tendency to achieve premature convergence, population diversity, and stability. Furthermore, we make use of portrait divergence, a newly-proposed state-of-the-art metric, to assess structural similarities between our interaction networks. Using this approach to analyse the cat swarm optimization (CSO) algorithm, we were able to identify some of the algorithm’s characteristics, assess the impact of one of the CSO’s parameters, and compare this algorithm to two other well-known methods (particle swarm optimization and artificial bee colony). Lastly, we discuss the relationship between the interaction network and the performance of the algorithms assessed.

**Exploratory Path Planning for Mobile Robots in Dynamic Environments with Ant Colony Optimization**

In the path planning task for autonomous mobile robots, robots should be able to plan their trajectory to leave the start position and reach the goal, safely. There are several path planning approaches for mobile robots in the literature. Ant Colony Optimization algorithms have been investigated for this problem, giving promising results. In this paper, we propose the Max-Min Ant System for Dynamic Path Planning algorithm for the exploratory path planning task for autonomous mobile robots based on topological maps. A topological map is an environment representation whose focus is in the main reference points of the environment and their connections. Based on this representation, the path can be composed of a sequence of state/actions pairs, which facilitates the navigability of the path, without the need to have the information of the complete map.
We show experimentally for the locomotion task that our con-
vention that our approach can solve multiple tasks at once. For instance, they can find
algorithms that search for a large set of diverse but high-performing solutions. In some specific situations, they can solve multiple tasks at once. For instance, they can find
algorithm. The new update rule allows to weight different parts of the solutions. PACO with the new update rule is evaluated for the example of the single machine total weighted tardiness problem (SMTWTP). This is an NP-hard optimization problem where the aim is to schedule jobs on a single machine such that their total weighted tardiness is minimized. PACO with the new population update rule is evaluated with several benchmark instances from the OR-Library. Moreover, the impact of the weights of the jobs on the solutions in the population and on the convergence of the algorithm are analyzed experimentally.

The results show that PACO with the new update rule has on average better solution quality than PACO with the standard update rule.

Complex Systems (Artificial Life/Artificial Immune Systems/Generative and Developmental-Systems/Evolutionary Robotics/Evolvable Hardware)

Evolution of Distributed Neural Controllers for Voxel-based Soft Robots
Eric Medvet, DIA, Alberto Bartoli, DIA, Andrea De Lorenzo, DIA, Giulio Fidel, DIA
Voxel-based soft robots (VSRs) are aggregations of elastic, cubic blocks that have sparked the interest of Robotics and Artificial Life researchers. VSRs can move by varying the volume of individual blocks, according to control signals dictated by a controller, possibly based on inputs coming from sensors embedded in the blocks. Neural networks (NNs) have been used as centralized processing units for those sensing controllers, with weights optimized using evolutionary computation. This structuring breaks the intrinsic modularity of VSRs: decomposing a VSR into modules to be assembled in a different way is very hard. In this work we propose an alternative approach that enables full modularity and is based on a distributed neural controller. Each block contains a small NN that outputs signals to adjacent blocks and controls the local volume, based on signals from adjacent blocks and on local sensor readings. We show experimentally for the locomotion task that our controller is as effective as the centralized one. Our experiments also suggest that the proposed framework indeed allows exploiting modularity: VSRs composed of pre-trained parts (body and controller) can be evolved more efficiently than starting from scratch.

Quality Diversity for Multi-task Optimization
Jean-Baptiste Mouret, Inria, Glenn Maguire, Inria
Quality Diversity (QD) algorithms are a recent family of optimization algorithms that search for a large set of diverse but high-performing solutions. In some specific situations, they can solve multiple tasks at once. For instance, they can find
the joint positions required for a robotic arm to reach a set of points, which can also be solved by running a classic optimizer for each target point. However, they cannot solve multiple tasks when the fitness needs to be evaluated independently for each task (e.g., optimizing policies to grasp many different objects). In this paper, we propose an extension of the MAP-Elites algorithm, called Multi-task MAP-Elites, that solves multiple tasks when the fitness function depends on the task. We evaluate it on a simulated parametrized planar arm (10-dimensional search space; 5000 tasks) and on a simulated 6-legged robot with legs of different lengths (36-dimensional search space; 2000 tasks). The results show that in both cases our algorithm outperforms the optimization of each task separately with the CMA-ES algorithm.

Scaling MAP-Elites to Deep Neuroevolution
Cédric Colas, Inria, Joost Huizinga, Uber AI Labs, Vashisht Madhavan, Element Inc., Jeff Clune, OpenAI
Quality-Diversity (QD) algorithms, and MAP-Elites (ME) in particular, have proven very useful for a broad range of applications including enabling real robots to recover quickly from joint damage, solving strongly deceptive maze tasks or evolving robot morphologies to discover new gait configurations. However, present implementations of ME and other QD algorithms seem to be limited to low-dimensional controllers with far fewer parameters than modern deep neural network models. In this paper, we propose to leverage the efficiency of Evolution Strategies (ES) to scale MAP-Elites to high-dimensional controllers parameterized by large neural networks. We design and evaluate a new hybrid algorithm called MAP-Elites with Evolution Strategies (ME-ES) for post-damage recovery in a difficult high-dimensional control task where traditional ME fails. Additionally, we show that ME-ES performs efficient exploration, on par with state-of-the-art exploration algorithms in high-dimensional control tasks with strongly deceptive rewards.

Learning behaviour-performance maps with
meta-evolution
David M. Bossens, University of Southampton, Jean-Baptiste Mouret, Inria, Danesh Tarapore, University of Southampton
The MAP-Elites quality-diversity algorithm has been successful in robotics because it can create a behaviorally diverse set of solutions that later can be used for adaptation, for instance to unanticipated damages. In MAP-Elites, the choice of the behavior space is essential for adaptation, the recovery of performance in unseen environments, since it defines the diversity of the solutions. Current practice is to hand-code a set of behavioural features, however, given the large space of possible behaviour-performance maps, the designer does not know a priori which behavioural features maximise a map's adaptation potential. We introduce a new meta-evolution algorithm that discovers those behavioural features that maximise future adaptations. The proposed method applies Covariance Matrix Adaptation Evolution Strategy to evolve a population of behaviour-performance maps to maximise a meta-fitness function that rewards adaptation. The method stores solutions found by MAP-Elites in a database which allows to rapidly construct new behaviour-performance maps on-the-fly. To evaluate this system, we study the gait of the RHex robot as it adapts to a range of damages sustained on its legs. When compared to MAP-Elites with user-defined behaviour spaces, we demonstrate that the meta-evolution system learns high-performing gaits with or without damages injected to the robot.

Diversity Preservation in Minimal Criterion Coevolution through Resource Limitation
Jonathan C. Brant, University of Central Florida, Kenneth O. Stanley, University of Central Florida
Minimal Criterion Coevolution (MCC) is a recently-introduced algorithm that demonstrates how interactions between two populations, each subject to a simple reproductive constraint, can produce an open-ended search process. Unlike conventional quality diversity (QD) algorithms, which also promote divergence, MCC does not require an explicit characterization of behavior or a comparison of performance, thereby addressing bottlenecks introduced by an intrinsically-finite behavior descriptor and by an assessment of comparative quality. Genetic speciation, a common method of diversity preservation, maintains population diversity in MCC; however, it requires an unnatural explicit comparison of genetic similarity. In nature, organisms are implicitly segregated into niches that each have a carrying capacity dictated by the amount of available resources. To show that MCC can be simpler and more natural while still working effectively, this paper introduces a method of diversity preservation through resource limitation, thereby alleviating the need to formalize and compare genetic distance. Experimental results in a maze navigation domain demonstrate that resource limitation not only maintains higher population diversity in both the maze and agent populations, but also accelerates evolution by forcing individuals to explore new niches, thereby suggesting that resource limitation is an effective, simpler, and more natural alternative for diversity preservation in MCC.

Evolving ab initio trading strategies in heterogeneous environments
David Rushing Dewhurst, University of Vermont, Yi Li, MassMutual Financial Group, Alex Bogdan, MassMutual Financial Group, Jasmine Geng, MassMutual Financial Group
Securities markets are quintessential complex adaptive systems in which heterogeneous agents compete in an attempt to maximize returns. Species of trading agents are also subject to evolutionary pressure as entire classes of strategies become obsolete and new classes emerge. Using an agent-based model of interacting heterogeneous agents as a flexible environment that can endogenously model many diverse market conditions, we subject deep neural networks to evolutionary pressure to create dominant trading agents. After analyzing the performance of these agents and noting the emergence of anomalous superdiffusion through the evolutionary process, we construct a method to turn high-fitness agents into trading algorithms. We backtest these trading algorithms on real high-frequency foreign exchange data, demonstrating that elite trading algorithms are consistently profitable in a variety of market conditions—even though these algorithms had never before been exposed to real financial data. These results provide evidence to suggest that developing trading strategies through repeated simulation and evolution in a mechanistic market model may be a practical and profitable alternative to explicitly training models with past observed market data.

Discovering Representations for Black-box Optimization
Adam Gaier, Hochschule Bonn-Rhein-Sie, Alexander Asteroth, Hochschule Bonn-Rhein-Sieg, Jean-Baptiste Mouret, Inria
The encoding of solutions in black-box optimization is a delicate, handcrafted balance between expressiveness and domain knowledge — between exploring a wide variety of solutions, and ensuring that those solutions are useful. Our main insight is that this process can be automated by generating a dataset of high-performing solutions with a quality diversity algorithm (here, MAP-Elites), then learning a representation with a generative model (here, a Variational Autoencoder) from that dataset. Our second insight is that this representation can be used to scale quality diversity optimization to higher dimensions — but only if we carefully mix solutions generated with the learned representation and those generated with traditional variation operators. We demonstrate these capabilities by learning an
low-dimensional encoding for the inverse kinematics of a thousand joint planar arm. The results show that learned representations make it possible to solve high-dimensional problems with orders of magnitude fewer evaluations than the standard MAP-Elites, and that, once solved, the produced encoding can be used for rapid optimization of novel, but similar, tasks. The presented techniques not only scale up quality diversity algorithms to high dimensions, but show that black-box optimization encodings can be automatically learned, rather than hand designed.

Covariance Matrix Adaptation for the Rapid Illumination of Behavior Space

Matthew C. Fontaine, University of Southern California, Julian Togelius, New York University, Stefanos Nikolaidis, University of Southern California, Amy K. Hoover, New Jersey Institute of Technology

We focus on the challenge of finding a diverse collection of quality solutions on complex continuous domains. While quality diversity (QD) algorithms like Novelty Search with Local Competition (NSLC) and MAP-Elites are designed to generate a diverse range of solutions, these algorithms require a large number of evaluations for exploration of continuous spaces. Meanwhile, variants of the Covariance Matrix Adaptation Evolution Strategy (CMA-ES) are among the best-performing derivative-free optimizers in single-objective continuous domains. This paper proposes a new QD algorithm called Covariance Matrix Adaptation MAP-Elites (CMA-ME). Our new algorithm combines the self-adaptation techniques of CMA-ES with archiving and mapping techniques for maintaining diversity in QD. Results from experiments based on standard continuous optimization benchmarks show that CMA-ME finds better-quality solutions than MAP-Elites; similarly, results on the strategic game Hearthstone show that CMA-ME finds both a higher overall quality and broader diversity of strategies than both CMA-ES and MAP-Elites. Overall, CMA-ME more than doubles the performance of MAP-Elites using standard QD performance metrics. These results suggest that QD algorithms augmented by operators from state-of-the-art optimization algorithms can yield high-performing methods for simultaneously exploring and optimizing continuous search spaces, with significant applications to design, testing, and reinforcement learning among other domains.

Towards Crossing the Reality Gap with Evolved Plastic Neurocontrollers

Huanneng Qiu, The University of New South Wales Canberra, Matthew Garratt, The University of New South Wales Canberra, David Howard, Robotics and Autonomous Systems Group, Sreenatha Anavatti, The University of New South Wales Canberra

A critical issue in evolutionary robotics is the transfer of controllers learned in simulation to reality. This is especially the case for small Unmanned Aerial Vehicles (UAVs), as the platforms are highly dynamic and susceptible to breakage. Previous approaches often require simulation models with a high level of accuracy, otherwise significant errors may arise when the well-designed controller is being deployed onto the targeted platform. Here we try to overcome the transfer problem from a different perspective, by designing a spiking neurocontroller which uses synaptic plasticity to cross the reality gap via online adaptation. Through a set of experiments we show that the evolved plastic spiking controller can maintain its functionality by self-adapting to model changes that take place after evolutionary training, and consequently exhibit better performance than its non-plastic counterpart.

Novelty Search makes Evolvability Inevitable

Stéphane Doncieux, Sorbonne Université, Giuseppe Paolo, Softbank Robotics Europe, Alban Lallaquière, Softbank Robotics Europe, Alexandre Coninx, Sorbonne Université; CNRS

Evolvability is an important feature that impacts the ability of evolutionary processes to find interesting novel solutions and to deal with changing conditions of the problem to solve. The estimation of evolvability is not straightforward and is generally too expensive to be directly used as selective pressure in the evolutionary process. Indirectly promoting evolvability as a side effect of other easier and faster to compute selection pressures would thus be advantageous. In an unbounded behavior space, it has already been shown that evolvable individuals naturally appear and tend to be selected as they are more likely to invade empty behavior niches. Evolvability is thus a natural byproduct of the search in this context. However, practical agents and environments often impose limits on the reachable behavior space. How do these boundaries impact evolvability? In this context, can evolvability still be promoted without explicitly rewarding it? We show that Novelty Search implicitly creates a pressure for high evolvability even in bounded behavior spaces, and explore the reasons for such a behavior. More precisely we show that, throughout the search, the dynamic evaluation of novelty rewards individuals which are very mobile in the behavior space, which in turn promotes evolvability.

Digital Entertainment Technologies and Arts

Interactive Evolution and Exploration Within Latent Level-Design Space of Generative Adversarial Networks

Jacob Schrum, Southwestern University, Jake Gutierrez, South-
well as dozens of CPU cores. Most general-purpose algorithm configuration procedures such as SMAC, Automated algorithm configuration procedures such as SMAC, GGA++, and irace can often find parameter configurations that substantially improve the performance of state-of-the-art algorithms for difficult problems – e.g., a three-fold speedup in the running time required by EAX, a genetic algorithm, to find optimal solutions to a set of widely studied TSP instances. However, it is usually recommended to provide these methods with running time budgets of one or two days of wall clock time as well as dozens of CPU cores. Most general-purpose algorithm configuration methods are based on powerful meta-heuristics that are designed for challenging and complex search landscapes; however, recent work has shown that many algorithms appear to have parameter configuration landscapes with a relatively simple structure. We introduce the golden parameter search (GPS) algorithm, an automatic configuration procedure designed to exploit this structure while optimizing each parameter semi-independently in parallel. We compare GPS to several state-of-the-art algorithm configurators and show that it often finds similar or better parameter configurations using a fraction of the computing time budget across a broad range of scenarios spanning TSP, SAT and MIP.

Evolutionary Combinatorial Optimization and Metaheuristics

Golden Parameter Search: Exploiting Structure to Quickly Configure Parameters in Parallel

Yasha Pushak, University of British Columbia, Holger H. Hoos, Leiden University

Automated algorithm configuration procedures such as SMAC, GGA++, and irace can often find parameter configurations that substantially improve the performance of state-of-the-art algorithms for difficult problems - e.g., a three-fold speedup in the running time required by EAX, a genetic algorithm, to find optimal solutions to a set of widely studied TSP instances. However, it is usually recommended to provide these methods with running time budgets of one or two days of wall clock time as well as dozens of CPU cores. Most general-purpose algorithm configuration methods are based on powerful meta-heuristics that are designed for challenging and complex search landscapes; however, recent work has shown that many algorithms appear to have parameter configuration landscapes with a relatively simple structure. We introduce the golden parameter search (GPS) algorithm, an automatic configuration procedure designed to exploit this structure while optimizing each parameter semi-independently in parallel. We compare GPS to several state-of-the-art algorithm configurators and show that it often finds similar or better parameter configurations using a fraction of the computing time budget across a broad range of scenarios spanning TSP, SAT and MIP.

CPPN2GAN: Combining Compositional Pattern Producing Networks and GANs for Large-scale Pattern Generation

Jacob Schrum, Western University, Vanessa Volz, monl.ai, Sebastian Risi, IT University of Copenhagen

Generative Adversarial Networks (GANs) are an emerging form of indirect encoding. The GAN is trained to induce a latent space on training data, and a real-valued evolutionary algorithm can search that latent space. Such Latent Variable Evolution (LVE) has recently been applied to game levels. However, it is hard for objective scores to capture level features that are appealing to players. Therefore, this paper introduces a tool for interactive LVE of tile-based levels for games. The tool also allows for direct exploration of the latent dimensions, and allows users to play discovered levels. The tool works for a variety of GAN models trained for both Super Mario Bros. and The Legend of Zelda, and is easily generalizable to other games. A user study shows that both the evolution and latent space exploration features are appreciated, with a slight preference for direct exploration, but combining these features allows users to discover even better levels. User feedback also indicates how this system could eventually grow into a commercial design tool, with the addition of a few enhancements.

Evolutionary Combinatorial Optimization and Metaheuristics

Automatic Decomposition of Mixed Integer Programs for Lagrangian Relaxation Using a Multiobjective Approach

Jake Matt Weiner, RMIT University, Andreas Ernst, Monash University, Xiaodong Li, RMIT University, Yuan Sun, RMIT University

This paper presents a new method to automatically decompose general Mixed Integer Programs (MIPs). To do so, we represent the constraint matrix for a general MIP problem as a hypergraph and relax constraints by removing hyperedges from the hypergraph. A Breadth First Search algorithm is used to identify the individual partitions that now exist and the resulting decomposed problem. We propose that good decompositions have both a small number of constraints relaxed and small subproblems in terms of the number of variables. We use the multiobjective Nondominated Sorting Genetic Algorithm II (NSGA-II) to create decompositions which minimize both the size of the subproblems and the number of constraints relaxed. We show through our experiments the types of decompositions our approach generates and test empirically the effectiveness of these decompositions in producing bounds when used in a Lagrangian Relaxation framework. The results demonstrate that the bounds generated by our decompositions are significantly better than those attained by solving the Linear Programming relaxation, as well as the bounds found via random and greedy constraint relaxation and decomposition generation.

Just-in-Time Batch Scheduling Subject to Batch Size

Sergey Polyakovskiy, Deakin University, Dhananjay Thiruvady,
Deakin University, Rym M’Hallah, Kuwait University
This paper considers single-machine just-in-time scheduling of jobs that may be grouped into batches subject to a constraint on batches’ weights. A job has a weight, due date, and earliness and tardiness penalties per unit time. A batch’s processing time is determined by its jobs. Each job inherits its batch’s completion time. The objective is to minimize the weighted sum of earliness and tardiness penalties of all jobs. This problem is challenging; jobs-to-batch assignment changes the batch’s processing time; thus, affects the structure of the entire solution and most importantly of its cost components. This problem is an excellent benchmark for testing linkage learning techniques, which exploit a problem’s structure. We propose a matheuristic LTGA, which integrates evolutionary algorithms with mixed-integer programming (MIP). It builds a linkage tree that extracts the dependency among decision variables of MIP solutions. We compare its performance to a state of the art matheuristic CMSA, which combines the learning component of an ant colony system (ACS) with MIP within a construct, solve, merge, and adapt framework. It uses knowledge extracted from ACS’ ants to construct a restricted MIP solves it efficiently, and feeds it back to ACS. Computational tests indicate that LTGA outperforms MIP solvers and CMSA.

Advanced Statistical Analysis of Empirical Performance Scaling
Yasha Pushak, University of British Columbia, Holger H. Hoos, Leiden University
Theoretical running time complexity analysis is a widely adopted method for studying the scaling behaviour of algorithms. However, theoretical analysis remains intractable for many high-performance, heuristic algorithms. Recent advances in statistical methods for empirical running time scaling analysis have shown that many state-of-the-art algorithms can achieve significantly better scaling in practice than expected. However, current techniques have only been successfully applied to study algorithms on randomly generated instance sets, since they require instances that can be grouped into “bins”, where each instance in a bin has the same size. In practice, real-world instance sets with this property are rarely available. We introduce a novel method that overcomes this limitation. We apply our method to a broad range of scenarios and demonstrate its effectiveness by revealing new insights into the scaling of several prominent algorithms; e.g., the SAT solver lingeling often appears to achieve sub-polynomial scaling on prominent bounded model checking instances, and the training times of scikit-learn’s implementation of SVMs scale as a lower-degree polynomial than expected (approximately 1.51 instead of 2).

Do Sophisticated Evolutionary Algorithms Perform Better than Simple Ones?
Michael Foster, University of Sheffield, Matthew Hughes, University of Sheffield, George O’Brien, University of Sheffield, Pietro S. Oliveto, University of Sheffield, James Pyle, University of Sheffield, Dirk Sudholt, The University of Sheffield, James Williams, University of Sheffield
Evolutionary algorithms (EAs) come in all shapes and sizes. Theoretical investigations focus on simple, bare-bones EAs while applications often use more sophisticated EAs that perform well on the problem at hand. What is often unclear is whether a large degree of algorithm sophistication is necessary, and if so, how much performance is gained by adding complexity to an EA. We address this question by comparing the performance of a wide range of theory-driven EAs, from bare-bones algorithms like the \((1 + 1)\) EA, \(a(2 + 1)\) GA and simple population-based algorithms to more sophisticated ones like the \((1 + (\lambda, \lambda))\) GA and algorithms using fast (heavy-tailed) mutation operators, against sophisticated and highly effective EAs from specific applications. This includes a famous and highly cited Genetic Algorithm for the Multidimensional Knapsack Problem and the Parameterless Population Pyramid for Ising Spin Glasses and MaxSat. While for the Multidimensional Knapsack Problem the sophisticated algorithm performs best, surprisingly, for large Ising and MaxSat instances the simplest algorithm performs best. We also derive conclusions about the usefulness of populations, crossover and fast mutation operators. Empirical results are supported by statistical tests and contrasted against theoretical work in an attempt to link theoretical and empirical results on EAs.

A Robust Experimental Evaluation of Automated Multi-Label Classification Methods
Alex G. C. de Sá, UFMG, Cristiano G. Pimenta, UFMG, Gisele L. Pappa, UFMG, Alex A. Freitas, University of Kent
Automated Machine Learning (AutoML) has emerged to deal with the selection and configuration of algorithms for a given learning task. With the progression of AutoML, several effective methods were introduced, especially for traditional classification and regression problems. Apart from the AutoML success, several issues remain open. One issue, in particular, is the lack of ability of AutoML methods to deal with different types of data. Based on this scenario, this paper approaches AutoML for multi-label classification (MLC) problems. In MLC, each example can be simultaneously associated to several class labels, unlike the standard classification task, where an example is associated to just one class label. In this work, we provide a general comparison of five automated multi-label classification methods – two evolutionary methods, one Bayesian optimization method, one random search and one greedy search – on 14 datasets and three designed search spaces. Overall, we observe that the most prominent method is the one based on a canonical grammar-based genetic programming (GGP) search method, namely Auto-MEKA_GGP Auto-MEKA_GGP
presented the best average results in our comparison and was statistically better than all the other methods in different search spaces and evaluated measures, except when compared to the greedy search method.

**Solving Constrained Combinatorial Reverse Auctions Using MOEAs: A Comparative Study.**

Elaine Guerrero-Peña, Universidad Federal de Pernambuco, Fernanda Nakano Kazama, University of Campinas, Paulo de Barros Correia, University of Campinas, Aluizio E.R. Aratijo, Universidade Federal de Pernambuco

Traditional Combinatorial Reverse Auctions (CRAs) (multiple items and single or multiple attributes) have been effectively adopted in several real-world applications. However, looking for the best solution (a set of winning sellers) is difficult to solve due to CRAs complexity. The use of exact algorithms is quite unsuitable in some real-life auctions due to the exponential time cost of these algorithms. Hence, we have opted for multi-objective evolutionary optimization methods (MOEAs) to find the best compromising solutions. This paper makes a comparative study between different MOEAs to solve the Problem of Determination of Winners (WDP) in a Multi-Attribute Combinatorial Reverse Auctions (MACRA) of several items with several attributes, establishing a model inspired by a WDP found in the literature. Six real problems of purchasing electronic products of different complexities are considered. The algorithms were assessed according to MOEA evaluation metrics and according to the best auction solution.

**Journey to the Center of the Linear Ordering Problem**

Leticia Hernando, University of the Basque Country (UPV/EHU), Alexander Mendiburu, University of the Basque Country (UPV/EHU), Jose A. Lozano, University of the Basque Country (UPV/EHU)

A number of local search based algorithms have been designed to escape from the local optima, such as, iterated local search or variable neighborhood search. The neighborhood chosen for the local search as well as the escape technique play a key role in the performance of these algorithms. Of course, a specific strategy has a different effect on distinct problems or instances. In this paper, we focus on a permutation-based combinatorial optimization problem: the linear ordering problem. We provide a theoretical landscape analysis for the adjacent swap, the swap and the insert neighborhoods. By making connections to other different problems found in the Combinatorics field, we prove that there are some moves in the local optima that will necessarily return a worse or equal solution. The number of these non-better solutions that could be avoided by the escape techniques is considerably large with respect to the number of neighbors. This is a valuable information that can be included in any of those algorithms designed to escape from the local optima, increasing their efficiency.

**Multi-layer local optima networks for the analysis of advanced local search-based algorithms**

Marcella Scoczyńska Ribeiro Martins, The Federal University of Technology - Paraná, Mohamed El Yafrani, Aalborg University, Myriam Delgado, The Federal University of Technology - Paraná, Ricardo Lüders, The Federal University of Technology - Paraná

A Local Optima Network (LON) is a graph model that compresses the fitness landscape of a particular combinatorial optimization problem based on a specific neighborhood operator and a local search algorithm. Determining which and how landscape features affect the effectiveness of search algorithms is relevant for both predicting their performance and improving the design process. This paper proposes the concept of multi-layer LONs as well as a methodology to explore these models aiming at extracting metrics for fitness landscape analysis. Constructing such models, extracting and analyzing their metrics are the preliminary steps into the direction of extending the study on single neighborhood operator heuristics to more sophisticated ones that use multiple operators. Therefore, in the present paper we investigate a two-layer LON obtained from instances of a combinatorial problem using bit-flip and swap operators. First, we enumerate instances of NKlandscape model and use the hill climbing heuristic to build the corresponding LONs. Then, using LON metrics, we analyze how efficiently the search might be when combining both strategies. The experiments show promising results and demonstrate the ability of multi-layer LONs to provide useful information that could be used for in metaheuristics based on multiple operators such as Variable Neighborhood Search.

**Solving the Single Row Facility Layout Problem by Differential Evolution**

Pavel Krömer, VSB-TU Ostrava, Jan Platos, VSB-TU Ostrava, Vaclav Snasel, VSB-TU Ostrava

Differential evolution is an efficient evolutionary optimization paradigm that has shown a good ability to solve a variety of practical problems, including combinatorial optimization ones. Single row facility layout problem is an NP-hard permutation problem often found in facility design, factory construction, production optimization, and other areas. Real-world problems can be cast as large single row facility location problem instances with different high-level properties and efficient algorithms that can solve them efficiently are needed. In this work, the differential evolution is used to solve the single row facility location problem and the ability of three different variants of the algorithm to evolve solutions to various problem instances is studied.

**A Deep Learning Approach to Predicting Solutions in**
Streaming Optimisation Domains
Mohamad Alissa, Edinburgh Napier University, Kevin Sim, Edinburgh Napier University, Emma Hart, Edinburgh Napier University

In the field of combinatorial optimisation, per-instance algorithm selection still remains a challenging problem, particularly with respect to streaming problems such as packing or scheduling. Typical approaches involve training a model to predict the best algorithm based on features extracted from the data, which is well known to be a difficult task and even more challenging with streaming data. We propose a radical approach that bypasses algorithm-selection altogether by training a Deep-Learning model using solutions obtained from a set of heuristic algorithms to directly predict a solution from the instance-data. To validate the concept, we conduct experiments using a packing problem in which items arrive in batches. Experiments conducted on six large datasets using batches of varying size show the model is able to accurately predict solutions, particularly with small batch sizes, and surprisingly in a small number of cases produces better solutions than any of the algorithms used to train the model.

Why Many Traveling Salesman Problem Instances Are Easier Than You Think
Swetha Varadarajan, Colorado State University, Darrell Whitley, Colorado State University, Gabriela Ochoa, Stirling University

While there are many inexact heuristics for generating high quality solutions to the Travelling Salesman Problem, our understanding of why these methods are effective and efficient is still limited. This paper looks at two population based heuristics: the EAX algorithm and the Mixing GA using partition crossover. We show that the local optima used to construct the initial population are also sampling edges found in the global optimum at an extremely high rate: in the majority of TSP instances, the number of global edges in the initial population is more than 73%. Next, we look at how recombination operators increase the representation of edges from the global optimum in the population, or increase the number of global edges in the best solutions in the population. We also look at TSP instances that are more difficult to solve, and again we find that edge frequency information can help to explain algorithm performance. Finally we use these results to suggest new strategies for generating high quality solutions for Travelling Salesman Problems.

Dynamic Bi-Objective Routing of Multiple Vehicles
Jakob Bossek, The University of Adelaide, Christian Grimme, University of Münster, Heike Trautmann, University of Münster

In practice, e.g. in delivery and service scenarios, Vehicle-Routing-Problems (VRPs) often imply repeated decision making on dynamic customer requests. As in classical VRPs, tours have to be planned short while the number of serviced customers has to be maximized at the same time resulting in a multi-objective problem. Beyond that, however, dynamic requests lead to the need for re-planning of not yet realized tour parts, while already realized tour parts are irreversible. In this paper we study this type of bi-objective dynamic VRP including sequential decision making and concurrent realization of decisions. We adopt a recently proposed Dynamic Evolutionary Multi-Objective Algorithm (DEMOA) for a related VRP problem and extend it to the more realistic (here considered) scenario of multiple vehicles. We empirically show that our DEMOA is competitive with a multi-vehicle offline and clairvoyant variant of the proposed DEMOA as well as with the dynamic single-vehicle approach proposed earlier.

Specific Single- and Multi-Objective Evolutionary Algorithms for the Chance-Constrained Knapsack Problem
Yue Xie, The University of Adelaide, Aneta Neumann, The University of Adelaide, Frank Neumann, The University of Adelaide

The chance-constrained knapsack problem is a variant of the classical knapsack problem where each item has a weight distribution instead of a deterministic weight. The objective is to maximize the total profit of the selected items under the condition that the weight of the selected items only exceeds the given weight bound with a small probability \( \alpha \). In this paper, we consider problem-specific single-objective and multi-objective approaches for the problem. We examine the use of heavy-tail mutations and introduce a problem-specific crossover operator to deal with the chance-constrained knapsack problem. Empirical results for single-objective evolutionary algorithms show the effectiveness of our operators compared to the use of classical operators. Moreover, we introduce a new effective multi-objective model for the chance-constrained knapsack problem. We use this model in combination with the problem-specific crossover operator in multi-objective evolutionary algorithms to solve the problem. Our experimental results show that this leads to significant performance improvements when using the approach in evolutionary multi-objective algorithms such as GSEMO and NSGA-II.
**Evolutionary Machine Learning**

**EML1**
Friday, July 10, 15:40–17:20

**Segmented Initialization and Offspring Modification in Evolutionary Algorithms for Bi-objective Feature Selection**

Hang Xu, Victoria University of Wellington, Bing Xue, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

In classification, feature selection mainly aims at reducing the dataset dimensionality and increasing the classification accuracy, which also results in higher computational efficiency than using the original full set of features. Population-based metaheuristic, evolutionary algorithms have been widely used to solve the bi-objective feature selection problem, which minimizes the number of selected features and the error of classification model. However, most of them are not specifically designed for feature selection, and disregard many of its complex characteristics. In this paper, we propose a generic approach that focuses on improving the initialization effectiveness and offspring quality, in order to boost the performance of existing evolutionary algorithms for bi-objective feature selection. To be more specific, a segmented initialization mechanism is used to enhance the exploration width, while an offspring modification mechanism is proposed to ensure the exploitation depth. Combining them together will make a good trade-off between the diversity and convergence. In the experiments, we plug the proposed approach into three different types of multi-objective evolutionary algorithms, and test them on 18 classification datasets with two widely-used performance metrics. The empirical results prove the significant contribution of the proposed approach on the optimization and classification performance.

**Relatedness Measures to Aid the Transfer of Building Blocks among Multiple Tasks**

Trung Nguyen, Victoria University of Wellington, Will Neil Browne, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

Multitask Learning (MTL) is a learning paradigm that deals with multiple different tasks in parallel and transfers knowledge among them. XOF, a Learning Classifier System using tree-based programs to encode building blocks (meta-features), constructs and collects features with rich discriminative information for classification tasks in an Observed List (OL). This paper seeks to facilitate the automation of feature transferring in MTL by utilising the OL. We hypothesise that the best discriminative features of a classification task carry its characteristics. Therefore, the relatedness between any two tasks can be estimated by comparing their most appropriate patterns. We propose a multiple-XOF system, called mXOF that can dynamically adapt feature transfer among XOFs. This system utilises the OL to estimate the task relatedness. This method enables the automation of transferring features. In terms of knowledge discovery, the resemblance estimation provides insightful relations among multiple data. We experimented mXOF on various scenarios, e.g. representative Hierarchical Boolean problems, classification of distinct classes in the UCI Zoo dataset, and unrelated tasks, to validate its abilities of automatic knowledge-transfer and estimating task relatedness. Results show that mXOF can estimate the relatedness reasonably between multiple tasks to aid the learning performance with the dynamic feature transferring.

**Feature Standardisation and Coefficient Optimisation for Effective Symbolic Regression**

Grant Dick, University of Otago, Caitlin A. Owen, University of Otago, Peter A. Whigham, University of Otago

Symbolic regression is a common application of genetic programming where model structure and corresponding parameters are evolved in unison. In the majority of work exploring symbolic regression, features are used directly without acknowledgement of their relative scale or unit. This paper extends recent work on the importance of standardisation of features when conducting symbolic regression. Specifically, z-score standardisation of input features is applied to both inputs and response to ensure that evolution explores a model space with zero mean and unit variance. This paper demonstrates that standardisation allows a simpler function set to be used without increasing bias. Additionally, it is demonstrated that standardisation can significantly improve the performance of coefficient optimisation through gradient descent to produce accurate models. Through analysis of several benchmark data sets, we demonstrate that feature standardisation enables simple but effective approaches that are comparable in performance to the state-of-the-art in symbolic regression.

**Re-purposing Heterogeneous Generative Ensembles with Evolutionary Computation**

Jamal Toutouh, Massachusetts Institute of Technology, Erik Hemberg, Massachusetts Institute of Technology, Una-May O’Reilly, Massachusetts Institute of Technology

Generative Adversarial Networks (GANs) are popular tools for generative modeling. The dynamics of their adversarial learning give rise to convergence pathologies during training such as mode and discriminator collapse. In machine learning, ensembles of predictors demonstrate better results than a single predictor for many tasks. In this study, we apply two evolutionary algorithms (EAs) to create ensembles to re-purpose generative models, i.e., given a set of heterogeneous generators that were optimized for one objective (e.g., minimize Fréchet Inception Distance), create ensembles of them for optimizing a different objective (e.g., maximize the diversity of the generated samples). The first method is restricted by the exact size
of the ensemble and the second method only restricts the upper bound of the ensemble size. Experimental analysis on the MNIST image benchmark demonstrates that both EA ensembles creation methods can re-purpose the models, without reducing their original functionality. The EA-based demonstrate significantly better performance compared to other heuristic-based methods. When comparing both evolutionary, the one with only an upper size bound on the ensemble size is the best.

Safe Crossover of Neural Networks Through Neuron Alignment

Thomas Uriot, European Space Agency, Dario Izzo, European Space Agency

One of the main and largely unexplored challenges in evolving the weights of neural networks using genetic algorithms is to find a sensible crossover operation between parent networks. Indeed, naive crossover leads to functionally damaged offspring that do not retain information from the parents. This is because neural networks are invariant to permutations of neurons, giving rise to multiple ways of representing the same solution. This is often referred to as the competing conventions problem. In this paper, we propose a two-step safe crossover (SC) operator. First, the neurons of the parents are functionally aligned by computing how well they correlate, and only then are the parents recombined. We compare two ways of measuring relationships between neurons: Pairwise Correlation (PwC) and Canonical Correlation Analysis (CCA). We test our safe crossover operators (SC-PwC and SC-CCA) on MNIST and CIFAR-10 by performing arithmetic crossover on the weights of feed-forward neural network pairs. We show that it effectively transmits information from parents to offspring and significantly improves upon naive crossover. Our method is computationally fast, can serve as a way to explore the fitness landscape more efficiently and makes safe crossover a potentially promising operator in future neuroevolution research and applications.

Exploring the Evolution of GANs through Quality Diversity

Victor Costa, University of Coimbra, Nuno Lourenço, University of Coimbra, João Correia, Center for Informatics and Systems of the University of Coimbra, Penousal Machado, CISUC

Generative adversarial networks (GANs) achieved relevant advances in the field of generative algorithms, presenting high-quality results mainly in the context of images. However, GANs are hard to train, and several aspects of the model should be previously designed by hand to ensure training success. In this context, evolutionary algorithms such as COEGAN were proposed to solve the challenges in GAN training. Nevertheless, the lack of diversity and premature optimization can be found in some of these solutions. We propose in this paper the application of a quality-diversity algorithm in the evolution of GANs. The solution is based on the Novelty Search with Local Competition (NSLC) algorithm, adapting the concepts used in COEGAN to this new proposal. We compare our proposal with the original COEGAN model and with an alternative version using a global competition approach. The experimental results evidenced that our proposal increases the diversity of the discovered solutions and leverage the performance of the models found by the algorithm. Furthermore, the global competition approach was able to consistently find better models for GANs.

XCS Classifier System with Experience Replay

Anthony Stein, University of Hohenheim, Roland Maier, University of Augsburg, Lukas Rosenbauer, BSH Home Appliances, Jörg Hähner, University of Augsburg

XCS constitutes the most deeply investigated classifier system today. It offers strong potentials and comes with inherent capabilities for mastering a variety of different learning tasks. Besides outstanding successes in various classification and regression tasks, XCS also proved very effective in certain multi-step environments from the domain of reinforcement learning. Especially in the latter domain, recent advances have been mainly driven by algorithms which model their policies based on deep neural networks, among which the Deep-Q-Network (DQN) being a prominent representative. Experience Replay (ER) constitutes one of the crucial factors for the DQN’s successes, since it facilitates stabilized training of the neural network-based Q-function approximators. Surprisingly, XCS barely takes advantage of similar mechanisms that leverage remembered raw experiences. To bridge this gap, this paper investigates the benefits of extending XCS with ER. We demonstrate that for single-step tasks ER yields strong improvements in terms of sample efficiency. On the downside, however, we reveal that ER might further aggravate well-studied issues not yet solved for XCS when applied to sequential decision problems demanding for long-action-chains.

Evolving Inborn Knowledge For Fast Adaptation in Dynamic POMDP Problems

Eseoghene Ben-Ihiwihu, Loughborough University, Pawel Ladosz, Loughborough University, Jeffery Dick, Loughborough University, Wen-Hua Chen, Loughborough University, Praveen Pilly, HRL Laboratories, Andrea Soltoggio, Loughborough University

Rapid online adaptation to changing tasks is an important problem in machine learning and, recently, a focus of meta-reinforcement learning. However, reinforcement learning (RL) algorithms struggle in POMDP environments because the state of the system, essential in a RL framework, is not always visible. Additionally, hand-designed meta-RL architectures may not include suitable computational structures for specific learning problems. The evolution of online learning mechanisms, on the contrary, has the ability to incorporate learning strategies
into an agent that can (i) evolve memory when required and (ii) optimize adaptation speed to specific online learning problems. In this paper, we exploit the highly adaptive nature of neuromodulated neural networks to evolve a controller that uses the latent space of an autoencoder in a POMDP. The analysis of the evolved networks reveals the ability of the proposed algorithm to acquire inborn knowledge in a variety of aspects such as the detection of cues that reveal implicit rewards, and the ability to evolve location neurons that help with navigation. The integration of inborn knowledge and online plasticity enabled fast adaptation and better performance in comparison to some non-evolutionary meta-reinforcement learning algorithms. The algorithm proved also to succeed in the 3D gaming environment Malmo Minecraft.

Neural Architecture Search (NAS) algorithms have discovered highly novel state-of-the-art Convolutional Neural Networks (CNNs) for image classification, and are beginning to improve our understanding of CNN architectures. However, within NAS research, there are limited studies focussing on the role of skip-connections, and how the configurations of connections between layers can be optimised to improve CNN performance. Our work focusses on developing a new evolutionary NAS algorithm based on adjacency matrices to optimise skip-connection structures, creating more specialised and powerful skip-connection structures within a DenseNet-BC network than previously seen in the literature. Our work further demonstrates how simple adjacency matrices can be interpreted in a way which allows for a more dynamic variant of DenseNet-BC. The final algorithm, using this novel interpretation of adjacency matrices for architecture design and evolved on the CIFAR100 dataset, finds networks with improved performance relative to a baseline DenseNet-BC network on both the CIFAR10 and CIFAR100 datasets, being the first, to our knowledge, NAS algorithm for skip-connection optimisation to do so. Finally, skip-connection structures discovered by our algorithm are analysed, and some important skip-connection patterns are highlighted.

**Abstracts**

**EML3**
Saturday, July 11, 15:40-17:20

**Absumption and Subsumption based Learning Classifier Systems**
Yi Liu, Victoria University of Wellington, Will Neil Browne, Victoria University of Wellington, Bing Xue, Victoria University of Wellington

Learning Classifier Systems (LCSs) are a group of rule-based evolutionary computation techniques, which have been frequently applied to data-mining tasks. Recent experiments have evidenced that, for Boolean domains, LCSs can produce models containing human-discriminable patterns. But, traditional LCSs cannot efficiently discover consistent, general rules - especially in domains that have unbalanced class distribution. The reason is that traditional search methods, e.g. crossover, mutation, and roulette wheel deletion, rely on luck to find and keep optimum rules. Recently, absumption has been introduced to deterministically remove over-general rules, which is complementary to subsumption that deterministically removes over-specific rules. It is hypothesized that utilizing just absumption & subsumption transforms the search process from stochastic to deterministic, which benefits LCSs in evolving interpretable models and removing the need to tune search parameters to the problem. Interrogatable artificial Boolean domains with varying numbers of attributes are considered as benchmarks. The new LCS, termed Absumption Subsumption Classifier System (ASCS), successfully produces interpretable models for all the complex domains tested, whereas the non-optimal rules in existing techniques obscure the patterns. ASCS’s ability to handle complex search spaces is observed for the 14-bits Majority-On problem, where the required 6435 different cooperating rules were discovered for correct pattern visualization.

**Neural Architecture Search for Sparse DenseNets with Dynamic Compression**
Damien O’Neill, Victoria University of Wellington, Bing Xue, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

Deep learning is an important field of machine learning. It is playing a critical role in a variety of applications ranging from self-driving cars to security and surveillance. However, deep networks have deep flaws. For example, they are highly vulnerable to adversarial attacks. One reason may be the homogeneous nature of their knowledge representation, which allows a single disruptive pattern to cause mis-classification. Biological intelligence has lateral asymmetry, which allows heterogeneous, modular learning at different levels of abstraction, enabling different representations of the same object. This work aims to incorporate lateralization and modular learning at different levels of abstraction in an evolutionary machine learning system. The results of image classification tasks show that the lateralized system efficiently learns hierarchical distributions of knowledge, demonstrating performance that is similar to (or better than) other state-of-the-art deep systems as it reasons using multiple representations. Crucially, the novel system outperformed all the state-of-the-art deep models for the classification of normal and adversarial images by 0.43%-2.56% and 2.15%-25.84%, respectively. Lateralisation enabled the system to exhibit robustness beyond previous work, which advocates for the creation of data sets that enable components of objects and the objects themselves to be learned specifically or in an end-to-end manner.

**Lateralized Learning for Robustness Against Adversarial Attacks in a Visual Classification System**
Abubakar Siddique, Victoria University of Wellington, Will Neil Browne, Victoria University of Wellington, Gina M. Grimshaw, Victoria University of Wellington

Deep learning is an important field of machine learning. It is playing a critical role in a variety of applications ranging from self-driving cars to security and surveillance. However, deep networks have deep flaws. For example, they are highly vulnerable to adversarial attacks. One reason may be the homogeneous nature of their knowledge representation, which allows a single disruptive pattern to cause mis-classification. Biological intelligence has lateral asymmetry, which allows heterogeneous, modular learning at different levels of abstraction, enabling different representations of the same object. This work aims to incorporate lateralization and modular learning at different levels of abstraction in an evolutionary machine learning system. The results of image classification tasks show that the lateralized system efficiently learns hierarchical distributions of knowledge, demonstrating performance that is similar to (or better than) other state-of-the-art deep systems as it reasons using multiple representations. Crucially, the novel system outperformed all the state-of-the-art deep models for the classification of normal and adversarial images by 0.43%-2.56% and 2.15%-25.84%, respectively. Lateralisation enabled the system to exhibit robustness beyond previous work, which advocates for the creation of data sets that enable components of objects and the objects themselves to be learned specifically or in an end-to-end manner.
fitness evaluations

Benjamin Patrick Evans, Victoria University of Wellington, Bing Xue, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

A common problem machine learning developers are faced with is overfitting, that is, fitting a pipeline too closely to the training data that the performance degrades for unseen data. Automated machine learning aims to free (or at least ease) the developer from the burden of pipeline creation, but this overfitting problem can persist. In fact, this can become more of a problem as we look to iteratively optimise the performance of an internal cross-validation (most often k-fold). While this internal cross-validation hopes to reduce this overfitting, we show we can still risk overfitting to the particular folds used. In this work, we aim to remedy this problem by introducing dynamic fitness evaluations which approximate repeated k-fold cross-validation, at little extra cost over single k-fold, and far lower cost than typical repeated k-fold. The results show that when time equated, the proposed fitness function results in significant improvement over the current state-of-the-art baseline method which uses an internal single k-fold. Furthermore, the proposed extension is very simple to implement on top of existing evolutionary computation methods, and can provide essentially a free boost in generalisation/testing performance.

Improving Neuroevolutionary Transfer Learning of Deep Recurrent Neural Networks through Network-Aware Adaptation

AbdelRahman EL-Said, Rochester Institute of Technology, Joshua Karns, Rochester Institute of Technology, Zimeng Lyu, Rochester Institute of Technology, Travis Desell, Rochester Institute of Technology, Daniel Kruitz, Rochester Institute of Technology, Alexander Ororbia II, Rochester Institute of Technology

Transfer learning entails taking an artificial neural network (ANN) that is trained on a source dataset and adapting it to a new target dataset. While this has been shown to be quite powerful, its use has generally been restricted by architectural constraints. Previously, in order to reuse and adapt an ANN’s internal weights and structure, the underlying topology of the ANN being transferred across tasks must remain mostly the same while a new output layer is attached, discarding the old output layer’s weights. This work introduces network-aware adaptive structure transfer learning (N-ASTL), an advancement over prior efforts to remove this restriction. N-ASTL utilizes statistical information related to the source network’s topology and weight distribution in order to inform how new input and output neurons are to be integrated into the existing structure. Results show improvements over prior state-of-the-art, including the ability to transfer in challenging real-world datasets not previously possible and improved generalization over RNNs without transfer.

Neuroevolution of Self-Interpretable Agents

Yujin Tang, Google, Duong Nguyen, Google, David Ha, Google Brain

Inattentional blindness is the psychological phenomenon that causes one to miss things in plain sight. It is a consequence of the selective attention in perception that lets us remain focused on important parts of our world without distraction from irrelevant details. Motivated by selective attention, we study the properties of artificial agents that perceive the world through the lens of a self-attention bottleneck. By constraining access to only a small fraction of the visual input, we show that their policies are directly interpretable in pixel space. We find neuroevolution ideal for training self-attention architectures for vision-based reinforcement learning tasks, allowing us to incorporate modules that can include discrete, non-differentiable operations which are useful for our agent. We argue that self-attention has similar properties as indirect encoding, in the sense that large implicit weight matrices are generated from a small number of key-query parameters, thus enabling our agent to solve challenging vision based tasks with at least 1000x fewer parameters than existing methods. Since our agent attends to only task critical visual hints, they are able to generalize to environments where task irrelevant elements are modified while conventional methods fail. Videos of our results and source code available at https://attentionagent.github.io/
considering incremental levels of constraint-based information. For the comparison with state-of-the-art methods, we include previous multiobjective approaches, single-objective genetic algorithms and classic constrained clustering methods.

**GeneCAI: Genetic Evolution for Acquiring Compact AI**

Mojan Javaheripi, UC San Diego, Mohammad Samragh, UC San Diego, Tara Javidi, UC San Diego, Farinaz Koushanfar, UC San Diego

In the contemporary big data realm, Deep Neural Networks (DNNs) are evolving towards more complex architectures to achieve higher inference accuracy. Model compression techniques can be leveraged to efficiently deploy these compute-intensive architectures on resource-limited mobile devices. Such methods comprise various hyperparameters that require per-layer customization to ensure high accuracy. Choosing the hyperparameters is cumbersome as the pertinent search space grows exponentially with model layers. This paper introduces GeneCAI, a novel optimization method that automatically learns how to tune per-layer compression hyperparameters. We devise a bijective translation scheme that encodes compressed DNNs to the genotype space. Each genotype's optimality is measured using a multi-objective score based on the accuracy and number of floating-point operations. We develop customized genetic operations to iteratively evolve the non-dominated solutions towards the optimal Pareto front, thus, capturing the optimal trade-off between model accuracy and complexity. GeneCAI optimization method is highly scalable and can achieve a near-linear performance boost on distributed multi-GPU platforms. Our extensive evaluations demonstrate that GeneCAI outperforms existing rule-based and reinforcement learning methods in DNN compression by finding models that lie on a better accuracy/complexity Pareto curve.

**Evolutionary Optimization of Deep Learning Activation Functions**

Garrett Bingham, The University of Texas at Austin, William Macke, The University of Texas at Austin, Risto Miikkulainen, Cognizant Technology Solutions

The choice of activation function can have a large effect on the performance of a neural network. While there have been some attempts to hand-engineer novel activation functions, the Rectified Linear Unit (ReLU) remains the most commonly-used in practice. This paper shows that evolutionary algorithms can discover novel activation functions that outperform ReLU. A tree-based search space of candidate activation functions is defined and explored with mutation, crossover, and exhaustive search. Experiments on training wide residual networks on the CIFAR-10 and CIFAR-100 image datasets show that this approach is effective. Replacing ReLU with evolved activation functions results in statistically significant increases in network accuracy. Optimal performance is achieved when evolution is allowed to customize activation functions to a particular task; however, these novel activation functions are shown to generalize, achieving high performance across tasks. Evolutionary optimization of activation functions is therefore a promising new dimension of metalearning in neural networks.

**Multi-Fitness Learning for Behavior-Driven Cooperation**

Connor Yates, Oregon State University, Reid Christopher, Oregon State University, Kağan Tumer, Oregon State University

Evolutionary learning algorithms have been successfully applied to multiagent problems where the desired system behavior can be captured by a single fitness signal. However, the complexity of many real world applications cannot be reduced to a single number, particularly when the fitness (i) arrives after a lengthy sequence of actions, and (ii) depends on the joint-action of multiple teammates. In this paper, we introduce the multi-fitness learning paradigm to enable multiagent teams to identify which fitness matters when in domains that require long-term, complex coordination. We demonstrate that multi-fitness learning efficiently solves a cooperative exploration task where teams of rovers must coordinate to observe various points of interest in a specific but unknown order.

**Program Synthesis as Latent Continuous Optimization: Evolutionary Search in Neural Embeddings**

Pawel Liskowski, NNAISENSE, Krzysztof Krawiec, Poznan University of Technology, Nihat Engin Toklu, NNAISENSE, Jerry Swan, NNAISENSE

In optimization and machine learning, the divide between discrete and continuous problems and methods is deep and persistent. We attempt to remove this distinction by training neural network autoencoders that embed discrete candidate solutions in continuous latent spaces. This allows us to take advantage of state-of-the-art continuous optimization methods for solving discrete optimization problems, and mitigates certain challenges in discrete optimization, such as design of bias-free search operators. In the experimental part, we consider program synthesis as the special case of combinatorial optimization. We train an autoencoder network on a large sample of programs in a problem-agnostic, unsupervised manner, and then use it with an evolutionary continuous optimization algorithm (CMA-ES) to map the points from the latent space to programs. We propose also a variant in which semantically similar programs are more likely to have similar embeddings. Assessment on a range of benchmarks in two domains indicates the viability of this approach and the usefulness of involving program semantics.

**Self-adaptation of XCS learning parameters based on Learning theory**

Motoki Horiuchi, Yokohama National University, Masaya Nakata, Yokohama National University
This paper proposes a self-adaptation technique of parameter settings used in the XCS learning scheme. Since we adaptively set those settings to their optimum values derived by the recent XCS learning theory, our proposal does not require any trial and error process to find their proper values. Thus, our proposal can always satisfy the optimality of XCS learning scheme, i.e. to distinguish accurate rules from inaccurate rules with the minimum update number of rules. Experimental results on artificial classification problems including overlapping problems show that XCS with our self-adaptation technique significantly outperforms the standard XCS.

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Abstracts

Evolutionary Multiobjective Optimization

Another Difficulty of Inverted Triangular Pareto Fronts for Decomposition-Based Multi-Objective Algorithms

Linjun He, Southern University of Science and Technology, Auraham Camacho, CINVESTAV Unidad Tamaulipas, Hisao Ishibuchi, Guangdong Provincial Key Laboratory of Brain-inspired Intelligent Computation

A set of uniformly sampled weight vectors from a unit simplex has been frequently used in decomposition-based multi-objective algorithms. The number of the generated weight vectors is controlled by a user-defined parameter H. In the literature, good results are often reported on test problems with triangular Pareto fronts since the shape of the Pareto fronts is consistent with the distribution of the weight vectors. However, when a problem has an inverted triangular Pareto front, well-distributed solutions over the entire Pareto front are not obtained due to the inconsistency between the Pareto front shape and the weight vector distribution. In this paper, we demonstrate that the specification of H has an unexpected large effect on the performance of decomposition-based multi-objective algorithms when the test problems have inverted triangular Pareto fronts. We clearly explain why their performance is sensitive to the specification of H in an unexpected manner (e.g., H = 3 is bad but H = 4 is good for three-objective problems whereas H = 3 is good but H = 4 is bad for four-objective problems). After these discussions, we suggest a simple weight vector specification method for inverted triangular Pareto fronts.

On the elicitation of indirect preferences in interactive evolutionary multiple objective optimization

Michał K. Tomczyk, Poznan University of Technology, Miłosz Kadziński, Poznan University of Technology

We consider essential challenges related to the elicitation of indirect preference information in interactive evolutionary algorithms for multiple objective optimization. The methods in this stream use holistic judgments provided by the Decision Maker (DM) to progressively bias the evolutionary search toward his/her most preferred region in the Pareto front. We enhance such an interactive process using three targeted developments and illustrate their efficiency in the context of a decomposition-based evolutionary framework. Firstly, we present some active learning strategies for selecting solutions from the current population that should be critically compared by the DM. These strategies implement the paradigm of maximizing the potential information gain derived from the DM’s answer. Secondly, we discuss the procedures for deciding when the DM should be questioned for preference information. In this way, we refer to a more general problem of distributing the DM’s interactions with the method in a way that ensures sufficient evolutionary pressure. Thirdly, we couple the evolutionary schemes with different types of indirect preferences, including pairwise comparisons, preference intensities, best-of-k judgments, and complete orders of a small subset of solutions. A thorough experimental analysis indicates that the three introduced advancements have a positive impact on the DM-perceived quality of constructed solutions.

Multi-Objective Hyperparameter Tuning and Feature Selection using Filter Ensembles

Martin Binder, Ludwig–Maximilians-Universität München, Julia Moosbauer, Ludwig–Maximilians-Universität München, Janek Thomas, Fraunhofer Institute for Integrated Circuits IIS, Bernd Bischl, Ludwig–Maximilians-Universität München

Both feature selection and hyperparameter tuning are key tasks in machine learning. Hyperparameter tuning is often useful to increase model performance, while feature selection is undertaken to attain sparse models. Sparsity may yield better model interpretability and lower cost of data acquisition, data handling and model inference. While sparsity may have a beneficial or detrimental effect on predictive performance, a small drop in performance may be acceptable in return for a substantial gain in sparseness. We therefore treat feature selection as a multi-objective optimization task. We perform hyperparameter tuning and feature selection simultaneously because the choice of features of a model may influence what hyperparameters perform well. We present, benchmark, and compare two different approaches for multi-objective joint hyperparameter optimization and feature selection: The first uses multi-objective model-based optimization. The second is an evolutionary NSGA-II-based wrapper approach to feature selection which incorporates specialized sampling, mutation and recombination operators. Both methods make use of parameterized filter ensembles. While model-based optimization needs fewer objective evaluations to achieve good performance, it incurs computational overhead compared to the NSGA-II, so the preferred choice depends on the cost of evaluating a model on
Constraint Handling within MOEA/D Through an Additional Scalarizing Function

Saulí Zapotecas-Martínez, Universidad Autónoma Metropolitana Unidad Cuajimalpa, Antonín Sebastien Ponsich, Universidad Autónoma Metropolitana Unidad Azcapotzalco

The Multi-Objective Evolutionary Algorithm based on Decomposition (MOEA/D) has shown high-performance levels when solving complicated multi-objective optimization problems. However, its adaptation for dealing with constrained multi-objective optimization problems (cMOPs) keeps being under the scope of recent investigations. This paper introduces a novel selection mechanism inspired by the epsilon-constraint method, which builds a bi-objective problem considering the scalarizing function (used into the decomposition approach of MOEA/D) and the constraint violation degree as an objective function. During the selection step of MOEA/D, the scalarizing function is considered to choose the best solutions to the cMOP. Preliminary results obtained over a set of complicated test problems drawn from the CF test suite indicate that the proposed algorithm is highly competitive regarding state-of-the-art MOEAs adopted in our comparative study.

Transfer Learning for Gaussian Process Assisted Evolutionary Bi-objective Optimization for Objectives with Different Evaluation Times

Xílú Wang, University of Surrey, Yaochu Jin, University of Surrey, Sebastian Schmidt, Honda Research Institute Europe GmbH, Markus Olhofer, Honda Research Institute Europe GmbH

Despite the success of evolutionary algorithms (EAs) for solving multi-objective problems, most of them are based on the assumption that all objectives can be evaluated within the same period of time. However, in many real-world applications, such an assumption is unrealistic since different objectives must be evaluated using different computer simulations or physical experiments with various time complexities. To address this issue, a surrogate assisted evolutionary algorithm along with a parameter-based transfer learning (T-SAE) is proposed in this work. While the surrogate for the cheap objective can be updated on sufficient training data, the surrogate for the expensive one is updated by either the training data set or a transfer learning approach. To find out the transferable knowledge, a filter-based feature selection algorithm is used to capture the pivotal features of each objective, and then use the common important features as a carrier for knowledge transfer between the cheap and expensive objectives. Then, the corresponding parameters in the surrogate models are adaptively shared to enhance the quality of the surrogate models. The experimental results demonstrate that the proposed algorithm outperforms the compared algorithms on the bi-objective optimization problems whose objectives have a large difference in computational complexities.

Designing parallelism in Surrogate-assisted multiobjective optimization based on decomposition

Nicolas Bervecglieri, University of Lille, Bilel Derbel, University of Lille, Arnaud Leflooghe, JFLI - CNRS - University of Tokyo, Hernan Aguirre, Shinshu University, Qingfu Zhang, City University of Hong Kong, Kiyoshi Tanaka, Shinshu University

On the one hand, surrogate-assisted evolutionary algorithms are established as a method of choice for expensive black-box optimization problems. On the other hand, the growth in computing facilities has seen a massive increase in potential computational power, granted the users accommodate their approaches with the offered parallelism. While a number of studies acknowledge the impact of parallelism for single-objective expensive optimization-assisted by surrogates, extending such techniques to the multi-objective setting has not yet been properly investigated, especially within the state-of the-art decomposition framework. We first highlight the different degrees of parallelism in existing surrogate-assisted multi-objective evolutionary algorithms based on decomposition (S-MOEA/D). We then provide a comprehensive analysis of the key steps towards a successful parallel S-MOEA/D approach. Through an extensive benchmarking effort relying on the well-established bbob-biobj test functions, we analyze the performance of the different algorithm designs with respect to the problem dimensionality and difficulty, the amount of parallel cores available, and the supervised learning models considered. In particular, we show the difference in algorithm scalability based on the selected surrogate-assisted approaches, the performance impact of distributing the model training task and the efficacy of the designed parallel-surrogate methods.

Computation of the improvement directions of the Pareto front and its application to MOEAs

Salvador Botello-Aceves, Center for Research in Mathematics, Arturo Hernandez-Aguirre, Center for Research in Mathematics, S. Ivvan Valdez, Centro de Investigación en Ciencias de Información Geospatial

This paper introduces the mathematical development and algorithm of the Improvement-Directions Mapping (IDM) method, which computes improvement directions to "push" the current solutions toward the true Pareto front. The main idea is to compute normal vectors to the front, as improvement directions in the objective space, to be then transformed into search directions in the variable space through a transformation tensor. The main contributions of the IDM as a local search operator versus previous approaches are the following: 1) It does not require of a priori information about improvement directions or location of the true Pareto front, 2) It uses a local quadratic approximation of the Pareto front to compute
the transformation tensor, thus, reducing numerical problems and avoiding abrupt changes in the search direction which could lead to erratic searches. These features allow the IDM to be implemented as a local search operator within any Multiobjective Evolutionary Algorithm (MOEA). The potential of the IDM is shown by hybridizing two well-known multi-objective algorithms: a) MOEA/D + IDM; b) NSGA-II + IDM. In the first approach, IDM "pushes" the offspring population in each iteration. A similar experiment is performed with the second approach. Furthermore, one more experiment evaluates the IDM as a refinement step that is applied to the last Pareto front delivered by NSGA-II.

Data Structures for Non-Dominated Sets: Implementations and Empirical Assessment of Two Decades of Advances
Jonathan Edward Fieldsend, University of Exeter

Many data structures have been developed over the last two decades for the storage and efficient update of unconstrained sets of mutually non-dominating solutions. Typically, analysis has been provided in the original works for these data structures in terms of worst/average case complexity performance. Often, however, other aspects such as rebalancing costs of underlying data structures, cache sizes, etc., can also significantly affect behaviour. Empirical performance comparison has often (but not always) been limited to run-time comparison with a basic linear list. No comprehensive comparison between the different specialised data structures proposed in the last two decades has thus far been undertaken. We take significant strides in addressing this here. Eight data structures from the literature are implemented within the same overarching open source Java framework. We additionally highlight and rectify some errors in published work — and offer additional efficiency gains. Run-time performances are compared and contrasted, using data sequences embodying a number of different characteristics. We show that in different scenarios different data structures are preferable, and that those with the lowest big O complexity are not always the best performing. We also find that performance profiles can vary drastically with computational architecture, in a non-linear fashion.

Surrogate Assisted Evolutionary Algorithm for Medium Scale Multi-Objective Optimisation Problems
Xiaoran Ruan, University of Electronic Science and Technology of China, Ke Li, University of Exeter, Bilel Derbel, University of Lille, Arnaud Liefooghe, JFLI - CNRS - University of Tokyo

Building a surrogate model of an objective function has shown to be effective to assist evolutionary algorithms (EAs) to solve real-world complex optimisation problems which involve either computationally expensive numerical simulations or costly physical experiments. However, their effectiveness mostly focuses on small-scale problems with less than 10 decision variables. The scalability of surrogate assisted EAs (SAEAs) have not been well studied yet. In this paper, we propose a Gaussian process surrogate model assisted EA for medium-scale expensive multi-objective optimisation problems with up to 50 decision variables. There are three distinctive features of our proposed SAEA. First, instead of using all decision variables in surrogate model building, we only use those correlated ones to build the surrogate model for each objective function. Second, rather than directly optimising the surrogate objective functions, the original multi-objective optimisation problem is transformed to a new one based on the surrogate models. Last but not the least, a subset selection method is developed to choose a couple of promising candidate solutions for actual objective function evaluations thus to update the training dataset. The effectiveness of our proposed algorithm is validated on benchmark problems with 10, 20, 50 variables, comparing with three state-of-the-art SAEAs.

Effects of Dominance Resistant Solutions on the Performance of Evolutionary Multi-Objective and Many-Objective Algorithms
Hisao Ishibuchi, Guangdong Provincial Key Laboratory of Brain-inspired Intelligent Computation, Takashi Matsumoto, Osaka Prefecture University, Naoki Masuyama, Osaka Prefecture University, Yusuke Nojima, Osaka Prefecture University

Dominance resistant solutions (DRSs) in multi-objective problems have very good values for some objectives and very bad values for other objectives. Whereas DRSs are far away from the Pareto front, they are hardly dominated by other solutions due to some very good objective values. It is well known that the existence of DRSs severely degrades the search ability of Pareto dominance-based algorithms such as NSGA-II and SPEA2. In this paper, we examine the effect of DRSs on the search ability of NSGA-II on the DTLZ test problems with many objectives. We slightly change their problem formulation to increase
the size of the DRS region. Through computational experiments, we show that DRSs have a strong negative effect on the search ability of NSGA-II whereas they have almost no effect on MOEA/D with the PBI function. We also show that a slightly modified NSGA-II for decreasing the negative effect of DRSs works well on many-objective DTLZ test problems (its performance is similar to NSGA-III and MOEA/D). These results suggest that DTLZ is not an appropriate test suite for evaluating many-objective evolutionary algorithms. This issue is further addressed through computational experiments on newly formulated test problems with no distance function.

Runtime Analysis of Evolutionary Algorithms with Biased Mutation for the Multi-Objective Minimum Spanning Tree Problem

Vahid Roostapour, The University of Adelaide, Jakob Bossek, The University of Adelaide, Frank Neumann, The University of Adelaide

Evolutionary algorithms (EAs) are general-purpose problem solvers that usually perform an unbiased search. This is reasonable and desirable in a black-box scenario. For combinatorial optimization problems, often more knowledge about the structure of optimal solutions is given, which can be leveraged by means of biased search operators. We consider the Minimum Spanning Tree (MST) problem in a single- and multi-objective version, and introduce a biased mutation, which puts more emphasis on the selection of edges of low rank in terms of low domination number. We present example graphs where the biased mutation can significantly speed up the expected runtime until (Pareto-)optimal solutions are found. On the other hand, we demonstrate that bias can lead to exponential runtime if “heavy” edges are necessarily part of an optimal solution. However, on general graphs in the single-objective setting, we show that a combined mutation operator which decides for unbiased or biased edge selection in each step with equal probability exhibits a polynomial upper bound – as unbiased mutation – in the worst case and benefits from bias if the circumstances are favorable.

If Unsure, Shuffle: Deductive Sort is $\Theta(MN^3)$, but $O(MN^2)$ in Expectation over Input Permutations

Sumit Mishra, Indian Institute of Information Technology Guwahati, Maxim Buzdalov, ITMO University

Despite significant advantages in theory of evolutionary computation, many papers related to evolutionary algorithms still lack proper analysis and limit themselves by rather vague reflections on why making a certain design choice improves the performance. While this seems to be unavoidable when talking about the behavior of an evolutionary algorithm on a practical optimization problem, doing the same for computational complexities of parts of evolutionary algorithms is harmful and should be avoided. Non-dominated sorting is one of such parts, commonly used in various evolutionary multiobjective algorithms. The complexity of the problem as such is not well-understood, and many algorithms were proposed for solving it in recent years. Unfortunately, the analysis of some of them is imperfect. In this paper, we prove that, contrary to what is claimed by its authors, the algorithm known as Deductive Sort has the worst-case time complexity of $\Theta(MN^3)$, where $M$ is the number of objectives and $N$ is the population size. However, if one shuffles the input, the worst-case expected running time drops down to $O(MN^2)$.

What is a Good Direction Vector Set for The R2-based Hypervolume Contribution Approximation

Yang Nan, Guangdong Provincial Key Laboratory of Brain-inspired Intelligent Computation, Ke Shang, Guangdong Provincial Key Laboratory of Brain-inspired Intelligent Computation, Hisao Ishibuchi, Guangdong Provincial Key Laboratory of Brain-inspired Intelligent Computation

The hypervolume contribution is an important concept in hypervolume-based evolutionary multi-objective optimization algorithms. It describes the loss of the hypervolume when a solution is removed from the current population. Since its calculation is $\#P$-hard in the number of objectives, its approximation is necessary for many-objective optimization problems. Recently, an R2-based hypervolume contribution approximation method was proposed. This method relies on a set of direction
We consider the design and analysis of surrogate-assisted algorithms for the approximation. However, the influence of different direction vector generation methods on the approximation quality has not been studied yet. This paper aims to investigate this issue. Five direction vector generation methods are investigated, including Das and Dennis’s method (DAS), unit normal vector method (UNV), JAS method, maximally sparse selection method with DAS (MSS-D), and maximally sparse selection method with UNV (MSS-U). Experimental results suggest that the approximation quality strongly depends on the direction vector generation method. The JAS and UNV methods show the best performance whereas the DAS method shows the worst performance. The reasons behind the results are also analyzed.

**Surrogate-assisted Multi-objective Combinatorial Optimization based on Decomposition and Walsh Basis**

Geoffrey Pruvost, University of Lille - CNRS - CRISTAL, Bilal Derbel, University of Lille - CNRS - CRISTAL, Arnaud Liefooghe, JFLI - CNRS - University of Tokyo, Sébastien Verel, Université du Littoral Côte d’Opale, Qingfu Zhang, City University of Hong Kong

We consider the design and analysis of surrogate-assisted algorithms for expensive multi-objective combinatorial optimization. Focusing on pseudo-boolean functions, we leverage existing techniques based on Walsh basis to operate under the decomposition framework of MOEA/D. We investigate two design components for the cheap generation of a promising pool of offspring and the actual selection of one solution for expensive evaluation. We propose different variants, ranging from a filtering approach that selects the most promising solution at each iteration by using the constructed Walsh surrogates to discriminate between a pool of offspring generated by variation, to a substitution approach that selects a solution to evaluate by optimizing the Walsh surrogates in a multi-objective manner. Considering bi-objective NK landscapes as benchmark problems offering different degree of non-linearity, we conduct a comprehensive empirical analysis including the properties of the achievable approximation sets, the anytime performance, and the impact of the order used to train the Walsh surrogates. Our empirical findings show that, although our surrogate-assisted design is effective, the optimal integration of Walsh models within a multi-objective evolutionary search process gives rise to particular questions for which different trade-off answers can be obtained.

**Evolutionary Numerical Optimization**

**Towards Dynamic Algorithm Selection for Numerical Black-Box Optimization: Investigating BBOB as a Use Case**

Diederick L. Vermetten, Leiden University, Hao Wang, Sorbonne Université, Thomas Bäck, Leiden University, Carola Doerr, Sorbonne Université

One of the most challenging problems in evolutionary computation is to select from its family of diverse solvers one that performs well on a given problem. This algorithm selection problem is complicated by the fact that different phases of the optimization process require different search behavior. While this can partly be controlled by the algorithm itself, there exist large differences between algorithm performance. It can therefore be beneficial to swap the configuration or even the entire algorithm during the run. Long deemed impractical, recent advances in Machine Learning and in exploratory landscape analysis give hope that this dynamic algorithm configuration (dynAC) can eventually be solved by automatically trained configuration schedules. With this work we aim at promoting research on dynAC, by introducing a simpler variant that focuses only on switching between different algorithms, not configurations. Using the rich data from the Black Box Optimization Benchmark (BBOB) platform, we show that even single-switch dynamic Algorithm selection (dynAS) can potentially result in significant performance gains. We also discuss key challenges in dynAS, and argue that the BBOB-framework can become a useful tool in overcoming these.

**Versatile Black-Box Optimization**

Jialin Liu, Southern University of Science and Technology, Antoine Moreau, Université Clermont Auvergne, Mike Preuss, Leiden Institute of Advanced Computer Science, Jérémy Rapin, Facebook AI Research, Baptiste Roziere, Facebook AI Research, Fabien Teytaud, Université du Littoral Côte d’Opale, Olivier Teytaud, Facebook AI Research

Choosing automatically the right algorithm using problem descriptors is a classical component of combinatorial optimization. It is also a good tool for making evolutionary algorithms fast, robust and versatile. We present Shiva, an algorithm good at both discrete and continuous, noisy and noise-free, sequential and parallel, black-box optimization. Our algorithm is experimentally compared to competitors on YABBOB, a BBOB comparable testbed, and on some variants of it, and then validated on several real world testbeds.

**Distributed Random Walks for Fitness Landscape Analysis**

Ryan Dieter Lang, Stellenbosch University, Andries Petrus Engelbrecht, Stellenbosch University

Fitness landscape analysis is used to mathematically characterize optimization problems. In order to perform fitness landscape analysis on continuous-valued optimization problems, a sample of the fitness landscape needs to be taken. A common way to perform this sampling is to use random walk algorithms. This paper proposes a new random walk algorithm
for continuous-valued optimization problems, called the distributed random walk algorithm. The algorithm is based on the premise that multiple short random walks of the same type will provide better coverage of the decision space and more robust fitness landscape measures than a single long random walk. The distributed random walk algorithm is simple to implement, and the computational overhead is insignificant compared to random walk algorithms in the literature. The results of the study indicate that the distributed random walk algorithm achieves both of these objectives. Furthermore, the benefits of the distributed random walk algorithm are shown to be much more significant when small step sizes are used in the random walks.

A Surrogate-Assisted Metaheuristic for Bilevel Optimization
Jesus-Adolfo Mejia-de-Dios, University of Veracruz, Efren Mezura-Montes, University of Veracruz

A Bilevel Optimization Problem (BOP) is related to two optimization problems in a hierarchical structure. A BOP is solved when an optimum of the upper level problem is found, subject to the optimal response of the respective lower level problem. This paper presents a metaheuristic method assisted by a kernel interpolation numerical technique to approximate optimal solutions of a BOP. Two surrogate methods approximate upper and lower level objective functions on solutions obtained by a population-based algorithm adapted to save upper level objective function evaluations. Some theoretical properties about kernel interpolation are used to study global convergence in some particular BOPs. The empirical results of this approach are analyzed when representative test functions for bilevel optimization are solved. The overall performance provided by the proposal is competitive.

In this paper, we therefore introduce various ways of using conditional linkage models in RV-GOMEA. Their use is compared to that of non-conditional models, and to VxD-CMA, whose performance is among the state of the art, on various benchmark problems with overlapping dependencies. We find that RV-GOMEA with conditional linkage models achieves the best scalability on most problems, with conditional models leading to similar or better performance than non-conditional models. We conclude that the introduction of conditional linkage models to RV-GOMEA is an important contribution, as it expands the set of problems for which optimization in a GBO setting results in substantially improved performance and scalability. In future work, conditional linkage models may prove to benefit the optimization of real-world problems.

Analyzing Adaptive Parameter Landscapes in Parameter Adaptation Methods for Differential Evolution
Ryoji Tanabe, Yokohama National University

Since the scale factor and the crossover rate significantly influence the performance of differential evolution (DE), parameter adaptation methods (PAMs) for the two parameters have been well studied in the DE community. Although PAMs can sufficiently improve the effectiveness of DE, PAMs are poorly understood (e.g., the working principle of PAMs). One of the difficulties in understanding PAMs comes from the unclarity of the parameter space that consists of the scale factor and the crossover rate. This paper addresses this issue by analyzing adaptive parameter landscapes in PAMs for DE. First, we propose a concept of an adaptive parameter landscape, which captures a moment in a parameter adaptation process. For each iteration, each individual in the population has its own adaptive parameter landscape. Second, we propose a method of analyzing adaptive parameter landscapes using a 1-step lookahead greedy improvement metric. Third, we examine adaptive parameter landscapes in three PAMs by using the proposed method. Results provide insightful information about PAMs in DE.

Deep Generative Model for Non-convex Constraint Handling
Naoki Sakamoto, University of Tsukuba, Eiji Semmatsu, Shinshu University, Kazuto Fukuchi, University of Tsukuba, Jun Sakuma, University of Tsukuba, Youhei Akimoto, University of Tsukuba

In this study, we consider black-box minimization problems with non-convex constraints, where the constraints are significantly cheaper to evaluate than the objective. Non-convex constraints generally make it difficult to solve problems using evolutionary approaches. In this paper, we revisit a conventional technique called decoder constraint handling, which transforms a feasible non-convex domain into an easy-to-control
convex set. This approach is promising because it transforms a constrained problem into an almost unconstrained one. However, its application has been considerably limited, because designing or training such a nonlinear decoder requires domain knowledge or manually prepared training data. To fully automate the decoder design, we use deep generative models. We propose a novel scheme to train a deep generative model without using manually prepared training data. For this purpose, we first train feasible solution samplers, which are deep neural networks, using the constraint functions. Subsequently, we train another deep generative model using the data generated from the trained samplers as the training data. The proposed framework is applied to tasks inspired by topology optimization problems. The empirical study demonstrates that the proposed approach can locate better solutions with fewer objective function evaluations than the existing approach.

**General Evolutionary Computation and Hybrids**

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### Landscape-Aware Fixed-Budget Performance Regression and Algorithm Selection for Modular CMA-ES Variants

Anja Jankovic, Sorbonne Université, Carola Doerr, Sorbonne Université

Automated algorithm selection promises to support the user in the decisive task of selecting a most suitable algorithm for a given problem. A common component of these machine-trained techniques are regression models which predict the performance of a given algorithm on a previously unseen problem instance. In the context of numerical black-box optimization, such regression models typically build on exploratory landscape analysis (ELA), which quantifies several characteristics of the problem. These measures can be used to train a supervised performance regression model. First steps towards ELA-based performance regression have been made in the context of a fixed-budget setting. In many applications, however, the user needs to select an algorithm that performs best within a given budget of function evaluations. Adopting this fixed-budget setting, we demonstrate that it is possible to achieve high-quality performance predictions with off-the-shelf supervised learning approaches, by suitably combining two differently trained regression models. We test this approach on a very challenging problem: algorithm selection on a portfolio of very similar algorithms, which we choose from the family of modular CMA-ES algorithms.

### Integrated vs. Sequential Approaches for Selecting and Tuning CMA-ES Variants

Diederick L. Vermetten, Leiden University, Hao Wang, Sorbonne Université, Carola Doerr, Sorbonne Université, Thomas Bäck, Leiden University

When faced with a specific optimization problem, deciding which algorithm to apply is always a difficult task. Not only is there a vast variety of algorithms to select from, but these algorithms are often controlled by many hyperparameters, which need to be suitably tuned in order to achieve peak performance. Usually, the problem of selecting and configuring the optimization algorithm is addressed sequentially, by first selecting a suitable algorithm and then tuning it for the application at hand. Integrated approaches, commonly known as Combined Algorithm Selection and Hyperparameter (CASH) solvers, have shown promise in several applications. In this work we compare sequential and integrated approaches for selecting and tuning the best out of the 4,608 variants of the modular Covariance Matrix Adaptation Evolution Strategy (CMA-ES). We show that the ranking of these variants depends to a large extent on the quality of the hyperparameters. Sequential approaches are therefore likely to recommend sub-optimal choices. Integrated approaches, in contrast, manage to provide competitive results at much smaller computational cost. We also highlight important differences in the search behavior of two CASH approaches, which build on racing (irace) and on model-based optimization (MIP-EGO), respectively.

### CMA-ES for One-Class Constraint Synthesis

Marcin Karmelita, Institute of Computing Science, Tomasz P. Pawlak, Institute of Computing Science

We propose CMA-ES for One-Class Constraint Synthesis (CMAESOCCS), a method that synthesizes Mixed-Integer Linear Programming (MILP) model from exemplary feasible solutions to this model using Covariance Matrix Adaptation – Evolutionary Strategy (CMA-ES). Given a one-class training set, CMAESOCCS adaptively detects partitions in this set, synthesizes independent Linear Programming models for all partitions and merges these models into a single MILP model. CMAESOCCS is evaluated experimentally using synthetic problems. A practical use case of CMAESOCCS is demonstrated based on a problem of synthesis of a model for a rice farm. The obtained results are competitive when compared to a state-of-the-art method.

### Bivariate Estimation-of-Distribution Algorithms Can Find an Exponential Number of Optima

Benjamin Doerr, Ecole Polytechnique, Martin S. Krejca, Hasso Plattner Institute

Finding a large set of optima in a multimodal optimization landscape is a challenging task. Classical population-based evolutionary algorithms (EAs) typically converge only to a single solution. While this can be counteracted by applying niching strategies, the number of optima is nonetheless trivially bounded by the population size. Estimation-of-distribution
Sequential model-based optimization (SMBO) approaches are algorithms for solving problems that require computationally or otherwise expensive function evaluations. The key design principle of SMBO is a substitution of the true objective function by a surrogate, which is used to propose the point(s) to be evaluated next. SMBO algorithms are intrinsically modular, leaving the user with many important design choices. Significant research efforts go into understanding which settings perform best for which type of problems. Most works, however, focus on the choice of the model, the acquisition function, and the strategy used to optimize the latter. The choice of the initial sampling strategy, however, receives much less attention. Not surprisingly, quite diverging recommendations can be found in the literature. We analyze in this work how the size and the distribution of the initial sample influences the overall quality of the efficient global optimization (EGO) algorithm, a well-known SMBO approach. While, overall, small initial budgets using Halton sampling seem preferable, we also observe that the performance landscape is rather unstructured. We furthermore identify several situations in which EGO performs unfavorably against random sampling. Both observations indicate that an adaptive SMBO design could be beneficial, making SMBO an interesting test-bed for automated algorithm design.

Using Implicit Multi-Objectives Properties to Mitigate Against Forgetfulness in Coevolutionary Algorithms
Adefunke Akinola, University of Guelph, Mark Wineberg, University of Guelph

It has been noticed that, while coevolutionary computational systems have only a single objective when evaluating, there is a subtle multi-objective aspect to evaluation since different pairings can be thought of as different objectives (all in support of the single original objective). Previously researchers used this to identify pairings of individuals during evaluation within a single generation. However, because of the problems of forgetfulness and the Red-Queens effect, this does not allow for the proper control that the technique promises. In this research, this implicit multi-objective approach is extended to function between generations as well as within. This makes it possible to implement a more powerful form of elitism as well as mitigate against some of the pathologies of Coevolutionary systems that forgetfulness and the Red-Queens effect engender, thus providing more robust solutions.

Model-based Optimization with Concept Drifts
Jakob Richter, TU Dortmund, Junjie Shi, TU Dortmund, Jian-Jia Chen, TU Dortmund, Jörg Rahnenführer, TU Dortmund, Michel Lang, TU Dortmund

Model-based Optimization (MBO) is a method to optimize expensive black-box functions that uses a surrogate to guide the search. We propose two practical approaches that allow MBO to optimize black-box functions where the relation between input and output changes over time, which are known as dynamic optimization problems (DOPs). The window approach trains the surrogate only on the most recent observations, and the time-as-covariate approach includes the time as an additional input variable in the surrogate, giving it the ability to learn the effect of the time on the outcomes. We focus on problems where the change happens systematically and label this systematic change concept drift. To benchmark our methods we define a set of benchmark functions built from established synthetic static functions that are extended with controlled drifts. We evaluate how the proposed approaches handle scenarios of no drift, sudden drift and incremental drift. The results show that both new methods improve the performance if a drift is present. For higher-dimensional multimodal problems the window approach works best and on lower-dimensional problems, where it is easier for the surrogate to capture the influence of the time, the time-as-covariate approach works better.

On the Choice of the Parameter Control Mechanism in the (1 + (λ, λ)) Genetic Algorithm
Mario Alejandro Hevia Fajardo, The University of Sheffield, Dirk Sudholt, The University of Sheffield

The self-adjusting (1+(λ,λ))GA is the best known genetic algorithm for problems with a good fitness-distance correlation as in OneMax. It uses a parameter control mechanism for the parameter λ that governs the mutation strength and the number of offspring. However, on multimodal problems, the parameter control mechanism tends to increase λ uncontrollably. We study this problem and possible solutions to it using rigorous runtime analysis for the standard jump benchmark problem class. The original algorithm behaves like a (1+n)EA whenever the maximum value λ = n is reached. This is ineffective for problems where large jumps are required. Capping λ at smaller
values is beneficial for such problems. Finally, resetting λ to 1 allows the parameter to cycle through the parameter space. We show that this strategy is effective for all jump problems: the (1+λ,λ)GA performs as well as the (1+1)EA with the optimal mutation rate and fast evolutionary algorithms, apart from a small polynomial overhead. Along the way, we present new general methods for bounding the runtime of the (1+λ,λ)GA that allows to translate existing runtime bounds from the (1+1)EA to the self-adjusting (1+λ,λ)GA. Our methods are easy to use and give upper bounds for novel classes of functions.

Analysis of the Performance of Algorithm Configurators for Search Heuristics with Global Optimization Operators

George T. Hall, The University of Sheffield, Pietro S. Oliveto, University of Sheffield, Dirk Sudholt, The University of Sheffield

Recently it has been proved that a simple algorithm configurator called ParamRLS can efficiently identify the optimal neighbourhood size to be used by stochastic local search to optimise two standard benchmark problem classes. In this paper we analyse the performance of algorithm configurators for tuning the more sophisticated global mutation operator used in standard evolutionary algorithms, which flips each of the n bits independently with probability χ/n and the best value for χ has to be identified. We compare the performance of configurators when the best-found fitness values within the cutoff time x are used to compare configurations against the actual optimisation time for two standard benchmark problem classes, Ridge and LeadingOnes. We rigorously prove that all algorithm configurators that use optimisation time as performance metric require cutoff times that are at least as large as the expected optimisation time to identify the optimal configuration. Matters are considerably different if the fitness metric is used. To show this we prove that the simple ParamRLS-F configurator can identify the optimal mutation rates even when using cutoff times that are considerably smaller than the expected optimisation time of the best parameter value for both problem classes.

ɛ-shotgun: ɛ-greedy Batch Bayesian Optimisation

George De Ath, University of Exeter, Richard M. Everson, University of Exeter, Jonathan Edward Fieldsend, University of Exeter, Alma A. M. Rahat, Swansea University

Bayesian optimisation is a popular, surrogate model-based approach for optimising expensive black-box functions. Given a surrogate model, the next location to expensively evaluate is chosen via maximisation of a cheap-to-query acquisition function. We present an ɛ-greedy procedure for Bayesian optimisation in batch settings in which the black-box function can be evaluated multiple times in parallel. Our ɛ-shotgun algorithm leverages the model’s prediction, uncertainty, and the approximated rate of change of the landscape to determine the spread of batch solutions to be distributed around a putative location. The initial target location is selected either in an exploitative fashion on the mean prediction, or – with probability ɛ – from elsewhere in the design space. This results in locations that are more densely sampled in regions where the function is changing rapidly and in locations predicted to be good (i.e close to predicted optima), with more scattered samples in regions where the function is flatter and/or of poorer quality. We empirically evaluate the ɛ-shotgun methods on a range of synthetic functions and two real-world problems, finding that they perform at least as well as state-of-the-art batch methods and in many cases exceed their performance.

Effective Reinforcement Learning through Evolutionary Surrogate-Assisted Prescription

Olivier Franccon, Cognizant Technology Solutions, Santiago Gonzalez, Cognizant Technology Solutions, Babak Hodjat, Cognizant Technology Solutions, Elliot Meyerson, Cognizant Technology Solutions, Risto Miikkulainen, Cognizant Technology Solutions, Xin Qiu, Cognizant Technology Solutions, Hormoz Shahrzad, Cognizant Technology Solutions

There is now significant historical data available on decision making in organizations, consisting of the decision problem, what decisions were made, and how desirable the outcomes were. Using this data, it is possible to learn a surrogate model, and with that model, evolve a decision strategy that optimizes the outcomes. This paper introduces a general such approach, called Evolutionary Surrogate-Assisted Prescription, or ESP. The surrogate is, for example, a random forest or a neural network trained with gradient descent, and the strategy is a neural network that is evolved to maximize the predictions of the surrogate model. ESP is further extended in this paper to sequential decision-making tasks, which makes it possible to evaluate the framework in reinforcement learning (RL) benchmarks. Because the majority of evaluations are done on the surrogate, ESP is more sample efficient, has lower variance, and lower regret than standard RL approaches. Surprisingly, its solutions are also better because both the surrogate and the strategy network regularize the decision making behavior. ESP thus forms a promising foundation to decision optimization in real-world problems.

An Evolutionary Optimization Algorithm for Gradually Saturating Objective Functions

Dolly Sapra, University of Amsterdam, Andy D. Pimentel, University of Amsterdam

Evolutionary algorithms have been actively studied for dynamic optimization problems in the last two decades, however the research is mainly focused on problems with large, periodical or abrupt changes during the optimization. In contrast, this paper concentrates on gradually changing environments with an additional imposition of a saturating objective function. This work is motivated by an evolutionary neural
architecture search methodology where a population of Convolutional Neural Networks (CNNs) is evaluated and iteratively modified using genetic operators during the training process. The objective of the search, namely the prediction accuracy of a CNN, is a continuous and slow moving target, increasing with each training epoch and eventually saturating when the training is nearly complete. Population diversity is an important consideration in dynamic environments wherein a large diversity restricts the algorithm from converging to a small area of the search space while the environment is still transforming. Our proposed algorithm adaptively influences the objective function, depending on the rate of change of the objective function, using disruptive crossovers and non-elitist population replacements. We compare the results of our algorithm with a traditional evolutionary algorithm and demonstrate that the proposed modifications improve the algorithm performance in gradually saturating dynamic environments.

**Sensitivity Analysis in Constrained Evolutionary Optimization**

Julian Schulte, Technische Universität Ilmenau, Volker Nissen, Technische Universität Ilmenau

Sensitivity analysis deals with the question of how changes in input parameters of a model affect its outputs. For constrained optimization problems, one question may be how variations in budget or capacity constraints influence the optimal solution value. Although well established in the domain of linear programming, it is hardly addressed in evolutionary computation. In this paper, a general approach is proposed which allows to identify how the outcome of an evolutionary algorithm is affected when model parameters, such as constraints, are changed. Using evolutionary bilevel optimization in combination with data mining and visualization techniques, the recently suggested concept of bilevel innovization allows to find trade-offs among constraints and objective value. Additionally, it enables decision-makers to gain insights into the overall model behavior under changing framework conditions. The concept of bilevel innovization as a tool for sensitivity analysis is illustrated, without loss of generality, by the example of the multidimensional knapsack problem. The experimental results show that by applying bilevel innovization it is possible to determine how the solution values are influenced by changes of different constraints. Furthermore, rules were obtained that provide information on how parameters can be modified to achieve efficient trade-offs between constraints and objective value.

**Expected Improvement versus Predicted Value in Surrogate-Based Optimization**

Frederik Rehbach, TH Köln, Martin Zaefferer, TH Köln, Boris Naujoks, TH Köln, Thomas Bartz-Beielstein, TH Köln

Surrogate-based optimization relies on so-called infill criteria (acquisition functions) to decide which point to evaluate next. When Kriging is used as the surrogate model of choice (also called Bayesian optimization), one of the most frequently chosen criteria is expected improvement. We argue that the popularity of expected improvement largely relies on its theoretical properties rather than empirically validated performance. Few results from the literature show evidence, that under certain conditions, expected improvement may perform worse than something as simple as the predicted value of the surrogate model. We benchmark both infill criteria in an extensive empirical study on the ‘BBOB’ function set. This investigation includes a detailed study of the impact of problem dimensionality on algorithm performance. The results support the hypothesis that exploration loses importance with increasing problem dimensionality. A statistical analysis reveals that the purely exploitative search with the predicted value criterion performs better on most problems of five or higher dimensions. Possible reasons for these results are discussed. In addition, we give an in-depth guide for choosing the infill criteria based on prior knowledge about the problem at hand, its dimensionality, and the available budget.

**Algorithm Selection of Anytime Algorithms**

Alexandre D. Jesus, University of Lille, Arnaud Liefooghe, JFLI - CNRS - University of Tokyo, Bilel Derbel, University of Lille, Luís Paquete, University of Coimbra

Anytime algorithms for optimization problems are of particular interest since they allow to trade off execution time with result quality. However, the selection of the best anytime algorithm for a given problem instance has been focused on a particular budget for execution time or particular target result quality. Moreover, it is often assumed that these anytime preferences are known when developing or training the algorithm selection methodology. In this work, we study the algorithm selection problem in a context where the decision maker’s anytime preferences are defined by a general utility function, and only known at the time of selection. To this end, we first examine how to measure the performance of an anytime algorithm with respect to this utility function. Then, we discuss approaches for the development of selection methodologies that receive a utility function as an argument at the time of selection. Then, to illustrate one of the discussed approaches, we present a preliminary study on the selection between an exact and a heuristic algorithm for a bi-objective knapsack problem. The results show that the proposed methodology has an accuracy greater than 96% in the selected scenarios, but we identify room for improvement.

**From Understanding Genetic Drift to a Smart-Restart Parameter-less Compact Genetic Algorithm**

Benjamin Doerr, École Polytechnique, Weijie Zheng, Southern University of Science and Technology
One of the key difficulties in using estimation-of-distribution algorithms is choosing the population sizes appropriately: Too small values lead to genetic drift, which can cause enormous difficulties. In the regime with no genetic drift, however, often the runtime is roughly proportional to the population size, which renders large population sizes inefficient. Based on a recent quantitative analysis which population sizes lead to genetic drift, we propose a parameter-less version of the compact genetic algorithm that automatically finds a suitable population size without spending too much time in situations unfavorable due to genetic drift. We prove an easy mathematical runtime guarantee for this algorithm and conduct an extensive experimental analysis on four classic benchmark problems. The former shows that under a natural assumption, our algorithm has a performance similar to the one obtainable from the best population size. The latter confirms that missing the right population size can be highly detrimental and shows that our algorithm as well as a previously proposed parameter-less one based on parallel runs avoids such pitfalls. Comparing the two approaches, ours profits from its ability to abort runs which are likely to be stuck in a genetic drift situation.

Genetic Algorithms

**An Improved GPU-Accelerated Heuristic Technique Applied to the Capacitated Vehicle Routing Problem**

Marwan Fouad Abdelatti, University of Rhode Island, Manbir Singh Sodhi, University of Rhode Island

The capacitated vehicle routing problem (CVRP) is a well-known NP-hard combinatorial problem. Genetic algorithms (GAs) are often used in solving CVRPs. However, the computational effort required to find a feasible solution becomes problematic for very large instances. Parallel-computation technology can significantly improve the performance of CVRP solution algorithms to deal with large problem sets, especially when using GAs. In this paper, an improved genetic algorithm is designed to be entirely executed on an NVIDIA GPU, taking advantage of the special CUDA GPU architecture to solving the CVRP. By distributing array elements over the GPU grid and using GPU kernel functions, the proposed algorithm successfully provides high-quality solutions within reasonable computational times, and near-optimal solutions for smaller benchmark problems. Within this framework, we address the execution speed problem in CVRPs by developing an algorithm that is entirely running on an NVIDIA GPU, investigate how to incorporate local search algorithms with the GA, and develop comparisons between the algorithm performance on both CPU and GPU. The utility of improved local search algorithms in the overall performance of the GA is also investigated.

**Multi-layer Heterogeneous Ensemble with Classifier and Feature Selection**

Thanh Tien Nguyen, Robert Gordon University, Van Nang Pham, Hanoi University of Science and Technology, Truong Manh Dang, Robert Gordon University, Vu Anh Luong, Griffith University, John McCall, Robert Gordon University, Alan Wee-Chung Liew, Griffith University

Deep Neural Networks have achieved many successes when applying to visual, text, and speech information in various domains. The crucial reasons behind these successes are the multi-layer architecture and the in-model feature transformation of deep learning models. These design principles have inspired other sub-fields of machine learning including ensemble learning. In recent years, there are some deep homogeneous ensemble models introduced with a large number of classifiers in each layer. These models, thus, require a costly computational classification. Moreover, the existing deep ensemble models use all classifiers including unnecessary ones which can reduce the predictive accuracy of the ensemble. In this study, we propose a multi-layer ensemble learning framework called Multi-Layer heterogeneous Ensemble System (MULES) to solve the classification problem. The proposed system works with a small number of heterogeneous classifiers to obtain ensemble diversity, therefore being efficiency in resource usage. We also propose an Evolutionary Algorithm-based selection method to select the subset of suitable classifiers and features at each layer to enhance the predictive performance of MULES. The selection method uses NSGA-II algorithm to optimize two objectives concerning classification accuracy and ensemble diversity. Experiments on 33 datasets confirm that MULES is better than a number of well-known benchmark algorithms.
The weight maximization problem (WMP) is the problem of finding the word of highest weight on a weighted finite state automaton (WFA). It is an essential question that emerges in many optimization problems in automata theory. Unfortunately, the general problem can be showed to be undecidable, whereas its bounded decisional version is NP-complete. Designing efficient algorithms that produce approximate solutions to the WMP in reasonable time is an appealing research direction that can lead to several new applications including formal verification of systems abstracted as WFAs. In particular, in combination with a recent procedure that translates a recurrent neural network into a weighted automaton, an algorithm for the WMP can be used to analyze and verify the network by exploiting the simpler and more compact automata model. In this work, we propose, implement and evaluate a meta-heuristic based on genetic algorithms to approximate solutions to the WMP. We experimentally evaluate its performance on examples from the literature and show its potential on different applications.

**Comparative Mixing for DSMGA-II**

Marcin Michal Komarnicki, Wroclaw University of Science and Technology, Michal Witold Przewozniczek, Wroclaw University of Science and Technology, Tomasz Maciej Durda, Wroclaw University of Science and Technology

Dependency Structure Matrix Genetic Algorithm-II (DSMGA-II) is a recently proposed optimization method that builds the linkage model on the base of the Dependency Structure Matrix (DSM). This model is used during the Optimal Mixing (OM) operators, such as the Restricted Mixing (RM) and the Back Mixing (BM). DSMGA-II was shown to solve theoretical and real-world optimization problems effectively. In this paper, we show that the effectiveness of DSMGA-II and its improved version, namely Two-edge Dependency Structure Matrix Genetic Algorithm-II (DSMGA-IIe), is relatively low for NK-landscape problems. Thus, we propose the Comparative Mixing (CM) operator that extends the RM operator. The CM operator modifies the linkage information obtained from the DSM-based linkage model by comparing the receiver individual with a randomly selected member of the population. Such modification enables DSMGA-II to solve NK-landscape problems effectively and does not limit DSMGA-II performance on most problems for which it was already shown effective.

**New Search Operators for Node-Depth Based Encoding**

Gustavo Post Sabin, Universidade Federal de Mato Grosso, Telma Woerle de Lima, Universidade Federal de Goiás, Anderson da Silva Soares, Universidade Federal de Goiás

Node-Depth Based Encoding is a representation for Evolutionary Algorithms applied to problems modelled by trees, storing nodes and their respective depths in an appropriately ordered list. This representation was highlighted by the results obtained, whose mutation and recombination operators have low time complexity in relation to other representations for the same problems. This work proposes new search operators. A high locality mutation operator, having as its main differential the possibility of including any edge present in a graph and a recombination operator with the ability to generate solutions with as much heritability as possible considering the edges set in the solutions, making it possible to recombine any solutions in the population with the aim of improving search convergence. All proposed operators also allow for dealing with problems modelled by forests with many trees and complete or noncomplete graphs, always generating feasible solutions. This study performed bias, locality and heritability investigations for proposed operators showing that they have adequate characteristics for dealing with high dimensionality problems. In order to evaluate the performance of proposed operators, the results of preliminary tests were obtained applying to a benchmark problem in the literature. The use of proposed operators resulted in faster convergence.
Adaptively Preserving Solutions in Both Feasible and Infeasible Regions on Generalized Multiple Constraint Ranking

Yohanes Bimo Dwianto, The University of Tokyo, Hiroaki Fukumoto, Japan Aerospace Exploration Agency, Akira Oyama, Japan Aerospace Exploration Agency

In the present work, we propose some new modifications of an existing constraint handling technique (CHT) for single-objective optimization problems. The base CHT is generalized multiple constraint ranking (G-MCR), which is already a modified version of the original CHT, MCR. Despite that G-MCR significantly outperformed the original MCR in the previous study, it is found that G-MCR tends to generate very few feasible individuals on a certain real-world like engineering design problem. In the present work, G-MCR is further modified to strike a better balance between feasible and infeasible individuals on each generation in an adaptive way so that the interaction between feasible and infeasible regions can be maintained, thus providing more efficient search towards constrained global optimum. Based on the investigation on 78 benchmark problems, we obtain that some of the proposed modifications produce more robust convergence performance by obtaining significant superiority on many types of problems. On real-world like engineering design problems, we also observe that the feasibility ratio generated on each generation might have an important role in improving the convergence performance.

Understanding Transforms of Pseudo-Boolean Functions

Darrell Whitley, Colorado State University, Hernan Aguirre, Shinshu University, Andrew Sutton, University of Minnesota

There exists general transforms that converts pseudo-Boolean functions into \( k \)-bounded pseudo-Boolean functions, for all \( k > 1 \). In addition to these general transforms, there can also exist specialized transforms that can be applied in special cases. New results are presented examining what happens to the "bit flip" neighborhood when transforms are applied. Transforms condense variables in a particular order. We show that different variable orderings produce different results in terms of problem difficulty. We also prove new results about the embedding of the original function in the new \( k \)-bounded function. Finally, this paper also looks at how parameter optimization problems can be expressed as high precision \( k \)-bounded pseudo-Boolean functions. This paper lays a foundation for the much wider application of evolutionary algorithms to \( k \)-bounded pseudo-Boolean functions.

Evolving Diverse Sets of Tours for the Travelling Salesperson Problem


Evolving diverse sets of high quality solutions has gained increasing interest in the evolutionary computation literature in recent years. With this paper, we contribute to this area of research by examining evolutionary diversity optimisation approaches for the classical Traveling Salesperson Problem (TSP). We study the impact of using different diversity measures for a given set of tours and the ability of evolutionary algorithms to obtain a diverse set of high quality solutions when adopting these measures. Our studies show that a large variety of diverse high quality tours can be achieved by using our approaches. Furthermore, we compare our approaches in terms of theoretical properties and the final set of tours obtained by the evolutionary diversity optimisation algorithm.

Modelling Parameter Configuration Spaces with Local Optima Networks

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Most algorithms proposed for solving complex problems require the definition of some parameter values. The process of finding suitable parameter values is an optimization problem by itself. Understanding the global structure of search spaces of complex optimization problems remains a challenge. Moreover, understanding the relationship between parameter values and the performance of metaheuristics is a key issue on their development. Local Optima Networks propose a scheme to model search spaces as networks whose nodes represent local
optimization spaces. Our main objectives are to understand the structure of these networks and explore the difficulty of different tuning scenarios using common indicators previously proposed in local optima networks studies (e.g., number of local optima, number of global optima and presence of local and global funnels). For this, we use the well-known tuning method ParamILS to analyze configuration search spaces of a standard genetic algorithm that solves continuous optimization problems.

Genetic Programming

Novelty Search for Automatic Bug Repair

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Genetic Improvement (GI) focuses on the development of evolutionary methods to automate software engineering tasks, such as performance improvement or software bugs removal. Concerning the latter, one of the earliest and most well-known methods in this area is the Genetic Program Repair (GenProg), a variant of Genetic Programming (GP). However, most GI systems encounter problems that are derived from the fact that they operate directly at source code level. These problems include highly neutral fitness landscapes and loss of diversity during the search, which are always undesirable in search and optimization tasks. This paper explores the use of Novelty Search (NS) with GenProg, since it can allow a search process to overcome these type of issues. While NS has been combined with GP before, and recently used with other GI systems, in the area of automatic bug repair NS has not been used until this work. Results show that GenProg with NS outperforms the original algorithm in some cases, based on an extensive experimental evaluation.

A Study on Graph Representations for Genetic Programming

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Graph representations promise several desirable properties for Genetic Programming (GP): multiple-output programs, natural representations of code reuse and, in many cases, an innate mechanism for neutral drift. Each graph GP technique provides a program representation, genetic operators and overarching evolutionary algorithm. This makes it difficult to identify the individual causes of empirical differences, both between these methods and in comparison to traditional GP. In this work, we empirically study the behavior of Cartesian Genetic Programming (CGP), Linear Genetic Programming (LGP), Evolving Graphs by Graph Programming (EGGP) and traditional GP. By fixing some aspects of the configurations, we study the performance of each graph GP method and GP in combination with three different EAs: generational, steady-state and (1 + λ). In general, we find that the best choice of representation, genetic operator and evolutionary algorithm depends on the problem domain. Further, we find that graph GP methods, particularly in combination with the (1 + λ) EA are significantly better on digital circuit synthesis tasks.

DAE-GP: Denoising Autoencoder LSTM Networks as Probabilistic Models in Estimation of Distribution Genetic Programming

David Wittenberg, Johannes Gutenberg-Universität, Franz Rothlauf, Johannes Gutenberg-Universität, Dirk Schweim, Johannes Gutenberg-Universität

Estimation of distribution genetic programming (EDA-GP) algorithms are metaheuristics where sampling new solutions from a learned probabilistic model replaces the standard mutation and recombination operators of genetic programming (GP). This paper presents DAE-GP, a new EDA-GP which uses denoising autoencoder long short-term memory networks (DAE-LSTMs) as probabilistic model. DAE-LSTMs are artificial neural networks that first learn the properties of a parent population by mapping promising candidate solutions to a latent space and reconstructing the candidate solutions from the latent space. The trained model is then used to sample new offspring solutions. We show on a generalization of the royal tree problem that DAE-GP outperforms standard GP and that performance differences increase with higher problem complexity. Furthermore, DAE-GP is able to create offspring with higher fitness from a learned model in comparison to standard GP. We believe that the key reason for the high performance of DAE-GP is that we do not impose any assumptions about the relationships between learned variables which is different to previous EDA-GP models. Instead, DAE-GP flexibly identifies and models relevant dependencies of promising candidate solutions.

Code Building Genetic Programming

Edward Pantridge, Swoop, Lee Spector, Amherst College

In recent years the field of genetic programming has made significant advances towards automatic programming. Research and development of contemporary program synthesis methods, such as PushGP and Grammar Guided Genetic Program-
Adaptive Weighted Splines - A New Representation to

Christian Raymand, Victoria University of Wellington, Qi Chen, Victoria University of Wellington, Bing Xue, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington.

Genetic Programming for Symbolic Regression is often prone to overfit the training data, resulting in poor generalization on unseen data. To address this issue, many pieces of research have focused on regularization via controlling the model complexity. However, due to the unstructured tree-based representation of individuals the model complexity cannot be directly computed, rather approximation of the complexity must be taken. This paper proposes a new novel representation called Adaptive Weighted Splines which enables explicit control over the complexity of individuals using splines. The experimental results confirm that this new representation is significantly better than the tree-based representation at avoiding overfitting and generalizing on unseen data, demonstrating notably better and far more consistent generalization performances on all the benchmark problems. Further analysis also shows that in most cases, the new Genetic Programming method outperforms classical regression techniques such as Linear Regression, Support Vector Regression, K-Nearest Neighbour and Decision Tree Regression and performs competitively with state-of-the-art ensemble regression methods Random Forests and Gradient Boosting.

Constructing Efficient Multigrid Solvers with Genetic Programming

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For many linear and nonlinear systems that arise from the discretization of partial differential equations the construction of an efficient multigrid solver is a challenging task. Here we present a novel approach for the optimization of geometric multigrid methods that is based on evolutionary computation, a generic program optimization technique inspired by the principle of natural evolution. A multigrid solver is represented as a tree of mathematical expressions which we generate based on a tailored grammar. The quality of each solver is evaluated in terms of convergence and compute performance using automated local Fourier analysis (LFA) and roofline performance modeling, respectively. Based on these objectives a multi-objective optimization is performed using grammar-guided genetic programming with a non-dominated sorting based selection. To evaluate the model-based prediction and to target concrete applications, scalable implementations of an evolved solver can be automatically generated with the ExaStencils framework. We demonstrate the effectiveness of our approach by constructing multigrid solvers for a linear elastic boundary value problem that are competitive with common V- and W-cycles.

Symbolic Regression Driven by Training Data and Prior Knowledge

Jiri Kubalik, Czech Technical University, Erik Derner, Czech Technical University, Robert Babuska, Czech Technical University

In symbolic regression, the search for analytic models is typically driven purely by the prediction error observed on the training data samples. However, when the data samples do not sufficiently cover the input space, the prediction error does not provide sufficient guidance toward desired models. Standard symbolic regression techniques then yield models that are partially incorrect, for instance, in terms of their steady-state characteristics or local behavior. If these properties were considered already during the search process, more accurate and relevant models could be produced. We propose a multi-objective symbolic regression approach that is driven by both the training data and the prior knowledge of the properties the desired model should manifest. The properties given in the form of formal constraints are internally represented by a set of discrete data samples on which candidate models are exactly checked. The proposed approach was experimentally evaluated on three test problems with results clearly demonstrating its capability to evolve realistic models that fit the training data well while complying with the prior knowledge of the desired model characteristics at the same time. It outperforms standard symbolic regression by several orders of magnitude in terms of the mean squared deviation from a reference model.

Improving Symbolic Regression based on Correlation Between Residuals and Variables

Qi Chen, Victoria University of Wellington, Bing Xue, Victoria University of Wellington
Unlabeled Multi-Target Regression with Genetic Programming

Uriel López, Tecnológico Nacional de México/IT de Tijuana, Leonardo Trujillo, Instituto Tecnológico de Tijuana, Sara Silva, University of Lisbon, Pierrick Legrand, Université de Bordeaux, Leonardo Vanneschi, Universidade Nova de Lisboa

Machine Learning (ML) has now become an important and ubiquitous tool in science and engineering, with successful applications in many real-world domains. However, there are still areas in need of improvement, and problems that are still considered difficult with off-the-shelf methods. One such problem is Multi Target Regression (MTR), where the target variable is a multidimensional tuple instead of a scalar value. In this work, we propose a more difficult variant of this problem which we call Unlabeled MTR (uMTR), where the structure of the target space is not given as part of the training data. This version of the problem lies at the intersection of MTR and clustering, an unexplored problem type. Moreover, this work proposes a solution method for uMTR, a hybrid algorithm based on Genetic Programming and RANdom Sample Consensus (RANSAC). Using a set of benchmark problems, we are able to show that this approach can effectively solve the uMTR problem.

Synthesis through Unification Genetic Programming

Thomas Welsch, University of Liverpool, Vitaliy Kurlin, University of Liverpool

We present a new method, Synthesis through Unification Genetic Programming (STUN GP), which synthesizes provably correct programs using a Divide and Conquer approach. This method first splits the input space by undergoing a discovery phase that uses Counterexample-Driven Genetic Programming (CDGP) to identify a set of programs that are provably correct under unknown unification constraints. The STUN GP method then computes these restraints by synthesizing predicates with CDGP that strictly map inputs to programs where the output will be correct. This method builds on previous work towards applying Genetic Programming (GP) to Syntax Guided Synthesis (SyGus) problems that seek to synthesize programs adhering to a formal specification rather than a fixed set of input-output examples. We show that our method is more scalable than previous CDGP variants, solving several benchmarks from the SyGus Competition that cannot be solved by CDGP. STUN GP significantly cuts into the gap between GP and state-of-the-art SyGus solvers and further demonstrates Genetic Programming's potential for Program Synthesis.

Towards an evolutionary-based approach for natural language processing

Luca Manzoni, University of Trieste, Domagoj Jakobovic, University of Zagreb, Luca Mariot, Delft University of Technology, Stjepan Picek, Delft University of Technology, Mauro Castelli, Universidade Nova de Lisboa

Tasks related to Natural Language Processing (NLP) have recently been the focus of a large research endeavor by the machine learning community. The increased interest in this area is mainly due to the success of deep learning methods. Genetic Programming (GP), however, was not under the spotlight with respect to NLP tasks. Here, we propose a first proof-of-concept that combines GP with the well established NLP tool word2vec for the next word prediction task. The main idea is that, once words have been moved into a vector space, traditional GP operators can successfully work on vectors, thus producing meaningful words as the output. To assess the suitability of this approach, we perform an experimental evaluation on a set of existing newspaper headlines. Individuals resulting from this (pre-)training phase can be employed as the initial population in other NLP tasks, like sentence generation, which will be the focus of future investigations, possibly employing adversarial co-evolutionary approaches.

Genetic programming approaches to learning fair classifiers

William La Cava, University of Pennsylvania, Jason H. Moore, University of Pennsylvania

Society has come to rely on algorithms like classifiers for important decision making, giving rise to the need for ethical
guarantees such as fairness. Fairness is typically defined by asking that some statistic of a classifier be approximately equal over protected groups within a population. In this paper, current approaches to fairness are discussed and used to motivate algorithmic proposals that incorporate fairness into genetic programming for classification. We propose two ideas. The first is to incorporate a fairness objective into multi-objective optimization. The second is to adapt lexicase selection to define cases dynamically over intersections of protected groups. We describe why lexicase selection is well suited to pressure models to perform well across the potentially infinitely many subgroups over which fairness is desired. We use a recent genetic programming approach to construct models on four datasets for which fairness constraints are necessary, and empirically compare performance to prior methods utilizing game-theoretic solutions. Methods are assessed based on their ability to generate trade-offs of subgroup fairness and accuracy that are Pareto optimal. The result show that genetic programming methods in general, and random search in particular, are well suited to this task.

**Semantically-Oriented Mutation Operator in Cartesian Genetic Programming for Evolutionary Circuit Design**

David Hodan, Brno University of Technology, Vojtech Mrazek, Brno University of Technology, Zdenek Vasicek, Brno University of Technology

Despite many successful applications, Cartesian Genetic Programming (CGP) suffers from limited scalability, especially when used for evolutionary circuit design. Considering the multiplier design problem, for example, the 5x5-bit multiplier represents the most complex circuit evolved from a randomly generated initial population. The efficiency of CGP highly depends on the performance of the point mutation operator, however, this operator is purely stochastic. This contrasts with the recent developments in Genetic Programming (GP), where advanced informed approaches such as semantic-aware operators are incorporated to improve the search space exploration capability of GP. In this paper, we propose a semantically-oriented mutation operator (SOMO) suitable for the evolutionary design of combinational circuits. SOMO uses semantics to determine the best value for each mutated gene. Compared to the common CGP and its variants as well as the recent versions of Semantic GP, the proposed method converges on common Boolean benchmarks substantially faster while keeping the phenotype size relatively small. The successfully evolved instances presented in this paper include 10-bit parity, 10+10-bit adder and 5x5-bit multiplier. The most complex circuits were evolved in less than one hour with a single-thread implementation running on a common CPU.

**A Modular Memory Framework for Time Series Prediction**

Stephen Paul Kelly, Michigan State University, Jacob Newsted, Michigan State University, Wolfgang Banzhaf, Michigan State University

Tangled Program Graphs (TPG) is a framework for genetic programming which has shown promise in challenging reinforcement learning problems with discrete action spaces. The approach has recently been extended to incorporate temporal memory mechanisms that enable operation in environments with partial-observability at multiple timescales. Here we propose a highly-modular memory structure that manages temporal properties of a task and enables operation in problems with continuous action spaces. This significantly broadens the scope of real-world applications for TPGs, from continuous-action reinforcement learning to time series forecasting. We begin by testing the new algorithm on a suite of symbolic regression benchmarks. Next, we evaluate the method in 3 challenging time series forecasting problems. Results generally match the quality of state-of-the-art solutions in both domains. In the case of time series prediction, we show that temporal memory eliminates the need to pre-specify a fixed-size sliding window of previous values, or autoregressive state, which is used by all compared methods. This is significant because it implies that no prior model for a time series is necessary, and the forecaster may adapt more easily if the properties of a series change significantly over time.

**Multi-Tree Genetic Programming for Feature Construction-Based Domain Adaptation in Symbolic Regression with Incomplete Data**

Baligh Al-Helali, Victoria University of Wellington, Qi Chen, Victoria University of Wellington, Bing Xue, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

Nowadays, transfer learning has gained a rapid popularity in tasks with limited data available. While traditional learning limits the learning process to knowledge available in a specific (target) domain, transfer learning can use parts of knowledge extracted from learning in a different (source) domain to help learning in the target domain. This concept is of special importance when there is a lack of knowledge in the target domain. Consequently, since data incompleteness is a serious cause of knowledge shortage in real-world learning tasks, it can be typically addressed using transfer learning. One way to achieve that is feature construction-based domain adaptation. However, although it is considered as a powerful feature construction algorithm, Genetic Programming has not been fully utilized for domain adaptation. In this work, a multi-tree genetic programming method is proposed for feature construction-based domain adaptation. The main idea is to construct a transformation from the source feature space to the target feature space, which maps the source domain close to the target domain. This method is utilized for symbolic regression with missing values. The experimental work shows encouraging potential of the proposed approach when applied to real-world tasks considering different transfer learning scenarios.
Using Exploratory Landscape Analysis to Visualize Single-objective Problems

Urban Škvorc, Jožef Stefan Institute, Tome Eitimov, Jožef Stefan Institute, Peter Korošec, Jožef Stefan Institute

Selecting the relevant algorithm for a given problem is a crucial first step for achieving good optimization algorithm performance. Exploratory Landscape Analysis can help with this problem by calculating landscape features that numerically describe individual problems. To understand the problem space in single-objective numerical optimization, we recently presented a preliminary study of how Exploratory Landscape Analysis can be used to visualize different optimization benchmark problems, with the ultimate goal of visualizing problems that are similar to one another close together. In this paper, we examine how the selection of landscape features affects such a visualization, and show that proper preprocessing of landscape features is crucial for producing a good visualization. In particular, we show that only a subset of landscape features is invariant to simple transforms such as translation and scaling. Further, we examine how such an approach can be used to visually compare problems from different optimization benchmark sets.

Model selection for metabolomics: predicting diagnosis of coronary artery disease using automated machine learning

Alena Orlenko, University of Pennsylvania, Jason H. Moore, University of Pennsylvania

Automated machine learning (AutoML) is emerging as a powerful tool for enabling the automatic selection of machine learning (ML) methods and parameter settings for the prediction of biomedical endpoints. AutoML approaches are important because it isn’t always obvious what ML methods are optimal for any given dataset. Here, we apply the tree-based pipeline optimization tool (TPOT) to the prediction of angiographic diagnosis of coronary artery disease (CAD). With TPOT, ML models are represented in the computer as expression trees and optimal pipelines discovered using a stochastic search method called genetic programming. We provide some guidelines for TPOT-based ML pipeline selection and optimization based on various clinical phenotypes and high-throughput metabolic profile in the Angiography and Genes Study (ANGES). We analyzed the metabolic profile of the ANGES cohort represented by 73 serum metabolites with a goal to identify the role of non-obstructive CAD patients in CAD diagnostics. We performed a comparative analysis of TPOT-generated ML pipelines with selected ML classifiers, optimized with grid search approach applied to two phenotypic profiles outlining the different role for non-obstructive CAD patients. As a result, TPOT generated ML pipelines that outperformed grid search optimized models across multiple performance metrics including balanced accuracy and precision-recall curve.

Sharp Bounds for Genetic Drift in Estimation of Distribution Algorithms (Hot-off-the-Press Track at GECCO 2020)

Benjamin Doerr, École Polytechnique, Weijie Zheng, Southern University of Science and Technology

Estimation of distribution algorithms (EDAs) are a successful branch of evolutionary algorithms (EAs) that evolve a probabilistic model instead of a population. Analogous to genetic drift in EAs, EDAs also encounter the phenomenon that the random sampling in the model update can move the sampling frequencies to boundary values not justified by the fitness. This can result in a considerable performance loss. This work gives the first tight quantification of this effect for three EDAs and one ant colony optimizer, namely for the univariate marginal distribution algorithm, the compact genetic algorithm, population-based incremental learning, and the max-min ant system with iteration-best update. Our results allow to choose the parameters of these algorithms in such a way that within a desired runtime, no sampling frequency approaches the boundary values without a clear indication from the objective function. This paper for the Hot-off-the-Press track at GECCO 2020 summarizes the work “Sharp Bounds for Genetic Drift in Estimation of Distribution Algorithms” by B. Doerr and W. Zheng, which has been accepted for publication in the IEEE Transactions on Evolutionary Computation.

Empirical Linkage Learning

Michał Witold Przewozniczek, Wrocław University of Science and Technology, Marcin Michal Komarnicki, Wrocław University of Science and Technology

Linkage learning is employed by many state-of-the-art evolutionary methods designed for solving problems in discrete domains. The effectiveness of these methods is dependent on linkage quality. The linkage may suffer to two different inaccuracy types. If some of the gene dependencies are not discovered, then the missing linkage takes place. Oppositely, if linkage identifies some gene dependencies that do not exist, then the false linkage takes place. To the best of our knowledge, all linkage learning techniques proposed that far predict that some genes are dependent and can commit both of the mistake types. Therefore, we propose a more direct approach. We disturb the genotype and check how these disturbances have influenced the effects of local search. We prove that the proposed Linkage Learning based on Local Optimization (3LO) will never report any false linkage, although it may still miss some true gene dependencies. 3LO is fundamentally different from other linkage learning techniques. Its main disadvantage
is a high computational cost and it is not suitable for already known evolutionary methods that frequently compute linkage. Therefore, we propose an evolutionary method that employs 3LO. More details considering 3LO, linkage quality and diversity may be found in the original paper [5].

**SGP-DT: Towards Effective Symbolic Regression with a Semantic GP Approach Based on Dynamic Targets**


Semantic Genetic Programming (SGP) approaches demonstrated remarkable results in different domains. SGP-DT is one of the latest of such approaches. Notably, SGP-DT proposes a dynamic-target approach that combines multiple GP runs without relying on any form of crossover. On eight well-known datasets SGP-DT achieves small RMSE, on average 25% smaller than state-of-the-art approaches.

**The Univariate Marginal Distribution Algorithm Copes Well With Deception and Epistasis**


In their recent work, Lehre and Nguyen (FOGA 2019) show that the univariate marginal distribution algorithm (UMDA) needs time exponential in the parent populations size to optimize the Deceptive Leading Blocks (DLB) problem. They conclude from this result that univariate EDAs have difficulties with deception and epistasis. In this work, we show that this negative finding is caused by an unfortunate choice of the parameters of the UMDA. When the population sizes are chosen large enough to prevent genetic drift, then the UMDA optimizes the DLB problem with high probability with at most $\lambda (\frac{n}{2} + 2e\ln n)$ fitness evaluations. Since an offspring population size $\lambda$ of order $n\log n$ can prevent genetic drift, the UMDA can solve the DLB problem with $O(n^2 \log n)$ fitness evaluations. Together with the result of Lehre and Nguyen, our result for the first time rigorously proves that running EDAs in the regime with genetic drift can lead to drastic performance losses. This extended abstract summarizes our work “The Univariate Marginal Distribution Algorithm Copes Well with Deception and Epistasis”, which appeared in the Proceedings of Evolutionary Computation in Combinatorial Optimization (EvoCOP), 2020, pp. 51–66, and won the conference’s best-paper award.

**Is the statistical significance between stochastic optimization algorithms’ performances also significant in practice?**

Tome Efthimov, *Jožef Stefan Institute*, Peter Korošec, *Jožef Stefan Institute*

To transfer the learned knowledge that is coming from benchmarking studies, which involve stochastic optimization algorithms, we should find a way to decide if the statistical significance between their performance is also important for real-world applications. For this reason, we have recently proposed a practical Deep Statistical Comparison (pDSC). It takes into account practical significance when making a statistical comparison of meta-heuristic stochastic optimization algorithms for single-objective optimization problems. Experimental results showed that our recently proposed approach provided very promising results.

**Large Scale Biomedical Data Analysis with Tree-based Automated Machine Learning**


Tree-based Pipeline Optimization Tool (TPOT) is an automated machine learning (AutoML) system that recommends optimal pipeline for supervised learning problems by scanning data for novel features, selecting appropriate models and optimizing their parameters. However, like other AutoML systems, TPOT may reach computational resource limits when working on big data such as whole-genome expression data. We develop two novel features for TPOT, Feature Set Selector and Template, which leverage domain knowledge, greatly reduce the computational expense and flexibly extend TPOT’s application to biomedical big data analysis.

**Automated Design of Multi-Level Network Partitioning Heuristics Employing Self-Adaptive Primitive Granularity Control**

Aaron Scott Pope, *Los Alamos National Laboratory*, Daniel R. Tauritz, *Auburn University*

Network segmentation has a variety of applications, including computer network security. A well segmented computer network is less likely to result in information leaks and more resilient to adversarial traversal. Conventionally network segmentation approaches rely on graph partitioning algorithms. However, general-purpose graph partitioning solutions are just that, general purpose. These approaches do not exploit specific topological characteristics present in certain classes of networks. Tailored partition methods can be developed to tar-
get specific domains, but this process can be time consuming and difficult. This work builds on previous research employing generative hyper-heuristics in the form of genetic programming for automating the development of customized graph partitioning heuristics by incorporating a dynamic approach to controlling the granularity of the heuristic search. The potential of this approach is demonstrated using two real-world complex network applications. Results show that the automated design process is capable of fine-tuning graph partitioning heuristics that sacrifice generality for improved performance on targeted networks.

A Memetic Level-based Learning Swarm Optimizer for Large-scale Water Distribution Network Optimization
Ya-Hui Jia, Victoria University of Wellington, Yi Mei, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

Potable water distribution networks are requisites of modern cities. Because of the city expansion, nowadays, the scale of the network grows rapidly, which brings great difficulty to its optimization. Evolutionary computation methods have been widely investigated on small-scale networks, but their performance is far from satisfactory on large-scale networks. Aiming at addressing this difficulty, a new memetic algorithm, called level-based learning swarm optimizer with restart and local search, is proposed in this paper to solve the large-scale water distribution network optimization problem. Instead of using traditional evolutionary computation algorithms, the level-based learning swarm optimizer that is especially proposed for large-scale optimization problems is applied as the population-based optimizer. Two restart strategies are incorporated to make the algorithm more effective. They can help the algorithm jump out from local optima thus to increase its exploration ability. Moreover, a simple yet effective local search algorithm is proposed based on the domain knowledge to further refine the solutions after the algorithm converges. Experimental results on both single-source and multi-source large-scale water distribution networks show that the proposed algorithm is more effective than the state-of-the-art evolutionary computation algorithms.

Adaptive Augmented Evolutionary Intelligence for the Design of Water Distribution Networks
Matthew Barrie Johns, University of Exeter, Herman Mahmoud, University of Exeter, Edward Keedwell, University of Exeter, Dragan Savic, University of Exeter

The application of Evolutionary Algorithms (EAs) to real-world problems comes with inherent challenges, primarily the difficulty in defining the large number of considerations needed when designing complex systems such as Water Distribution Networks (WDN). One solution is to use an Interactive Evolutionary Algorithm (IEA), which integrates a human expert into the optimisation process and helps guide it to solutions more suited to real-world application. The involvement of an expert provides the algorithm with valuable domain knowledge, however it is an intensive task requiring extensive interaction, leading to user fatigue and reduced effectiveness. To address this, the authors have developed methods for capturing human expertise from user interactions utilising machine learning to produce Human-Derived Heuristics (HDH) which are integrated into an EAs mutation operator. This work focuses on the development of an adaptive method for applying multiple HDHs throughout an EAs search. The new adaptive approach is shown to outperform both singular HDH approaches and traditional EAs on a range of large scale WDN design problems. This work paves the way for the development of a new type of IEA that has the capability of learning from human experts whilst minimising user fatigue.

Evolving Energy Demand Estimation Models over Macroeconomic Indicators
Nuno Lourenço, University of Coimbra, José Manuel Colmenar, Universidad Rey Juan Carlos, Ignacio Hidalgo, Universidad Complutense de Madrid, Sancho Salcedo Sanz, Universidad de Alcalá

Energy is essential for all countries, since it provides sustainable social, economic and/or environmental developments. Since the industrial revolution, the demand for energy has increased exponentially. It is expected that the energy consumption in the world increases by 50% by 2030. As such, managing the demand of energy is of the utmost importance. The development of tools to model and accurately predict the demand of energy is very important to policy makers. In this paper we propose the usage of the Structured Gramatical Evolution (SGE) algorithm to evolve models of energy demand, over macro-economic indicators. The proposed SGE is hybridized with a Differential Evolution approach in order to obtain the parameters of the models evolved which better fit with the real energy demand. We have tested the performance of the proposed approach in a problem of total energy demand estimation in Spain, where we show that the SGE is able to generate extremely accurate models for the energy prediction within one year time-horizon.

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we can use heuristics as it allows significant freedom in the selection of required cryptographic properties. Unfortunately, with heuristics, one is seldom sure how good a trade-off between cryptographic properties is reached or if optimizing for one property optimizes implicitly for another property. In this paper, we consider what is to the best of our knowledge, the most detailed analysis of trade-offs among S-box cryptographic properties. More precisely, we ask questions if one property is optimized, what is the worst possible value for some other property, and what happens if all properties are optimized. Our results show that while it is possible to reach a large variety of possible solutions, optimizing for a certain property would commonly result in good values for other properties. In turn, this suggests that a single-objective approach should be a method of choice unless some precise values for multiple properties are needed.

COUGAR: Clustering Of Unknown malware using Genetic Algorithm Routines
Zachary Wilkins, Dalhousie University, Nur Zincir-Heywood, Dalhousie University
Through malware, cyber criminals can leverage our computing resources to disrupt our work, steal our information, and even hold it hostage. Security professionals seek to classify these malicious software so as to prevent their distribution and execution, but the sheer volume of malware complicates these efforts. In response, machine learning algorithms are actively employed to alleviate the workload. One such approach is evolutionary computation, where solutions are bred, rather than built. In this paper, we design, develop and evaluate a system, COUGAR, to reduce high-dimensional malware behavioural data, and optimize clustering behaviour using a multi-objective genetic algorithm. Evaluations demonstrate that each of our chosen clustering algorithms can successfully highlight groups of malware. We also present an example real-world scenario, based on the testing data, to demonstrate practical applications.

Variable Reduction for Surrogate-Based Optimization
Frederik Rehbach, TH Köln, Lorenzo Gentile, TH Köln, Thomas Bartz-Beielstein, TH Köln
Real-world problems such as computational fluid dynamics simulations and finite element analyses are computationally expensive. A standard approach to mitigating the high computational expense is Surrogate-Based Optimization (SBO). Yet, due to the high-dimensionality of many simulation problems, SBO is not directly applicable or not efficient. Reducing the dimensionality of the search space is one method to overcome this limitation. In addition to the applicability of SBO, dimensionality reduction enables easier data handling and improved data and model interpretability. Regularization is considered as one state-of-the-art technique for dimensionality reduction. We propose a hybridization approach called Regularized-Surrogate-Optimization (RSO) aimed at overcoming difficulties related to high-dimensionality. It couples standard Kriging-based SBO with regularization techniques. The employed regularization methods are based on three adaptations of the least absolute shrinkage and selection operator (LASSO). In addition, tree-based methods are analyzed as an alternative variable selection method. An extensive study is performed on a set of artificial test functions and two real-world applications: the electrostatic precipitator problem and a multilayered composite design problem. Experiments reveal that RSO requires significantly less time than standard SBO to obtain comparable results. The pros and cons of the RSO approach are discussed, and recommendations for practitioners are presented.

Modeling Wildfires Using Evolutionary Cellular Automata
Maxfield E. Green, University of Vermont, Todd F. DeLuca, University of Vermont, Karl Kaiser, University of Vermont
With the increased size and frequency of wildfire events worldwide, accurate real-time prediction of evolving wildfire fronts is a crucial component of firefighting efforts and forest management practices. We propose a cellular automaton (CA) that simulates the spread of wildfire. We embed the CA inside of a genetic program (GP) that learns the state transition rules from spatially registered synthetic wildfire data. We demonstrate this model’s predictive abilities by testing it on unseen synthetically generated landscapes. We compare the performance of a genetic program (GP) based on a set of primitive operators and restricted expression length to null and logistic models. We find that the GP is able to closely replicate the spreading behavior driven by a balanced logistic model. Our method is a potential alternative to current benchmark physics-based models.

Impact of NSGA-II Objectives on EEG Feature Selection Related to Motor Imagery
Miguel Leon, Malardalen University, Christoffer Parkkila, Malardalen University, Jonatan Tidare, Malardalen University, Ning Xiong, Malardalen University, Elaine Astrand, Malardalen University
The selection of ElectroEncephaloGram (EEG) features with functional relevance to Motor Imagery (MI) is a crucial task for a successful outcome in Brain-Computer Interface (BCI)-based motor rehabilitation. Individual EEG patterns during MI require subject-dependent feature selection, which is an arduous task due to the complexity and large number of features. One solution is to use metaheuristics, e.g., Genetic Algorithm (GA), to avoid an exhaustive search which is impractical. In this work, one of the most widely used GA, NSGA-II, is used with a hierarchical individual representation to facilitate the exclusion of EEG channels irrelevant for MI. In essence, the performance of different objectives in NSGA-II was evaluated on a previously recorded MI EEG data set. Empirical results show that
k-Nearest Neighbors (k-NN) combined with Pearson's Correlation (PCFS) as objective functions yielded higher classification accuracy as compared to the other objective-combinations (73% vs. 69%). Linear Discriminant Analysis (LDA) combined with Feature Reduction (FR) as objective functions maximized the reduction of features (99.6%) but reduced classification performance (65.6%). All classifier objectives combined with PCFS selected similar features in accordance with expected activity patterns during MI. In conclusion, PCFS and a classifier as objective functions constitutes a good trade-off solution for MI data.

A Genetic Programming Approach to Feature Construction for Ensemble Learning in Skin Cancer Detection
Qurrat Ul Ain, Victoria University of Wellington, Harith Al-sahaf, Victoria University of Wellington, Bing Xue, Victoria University of Wellington, Mengjie Zhang, Victoria University of Wellington

Ensembles of classifiers have proved to be more effective than a single classification algorithm in skin image classification problems. Generally, the ensembles are created using the whole set of original features. However, some original features can be redundant and may not provide useful information in building good ensemble classifiers. To deal with this, existing feature construction methods that usually generate new features for only a single classifier have been developed but they fit the training data too well, resulting in poor test performance. This study develops a new classification method that combines feature construction and ensemble learning using genetic programming (GP) to address the above limitations. The proposed method is evaluated on two benchmark real-world skin image datasets. The experimental results reveal that the proposed algorithm has significantly outperformed two existing GP approaches, two state-of-the-art convolutional neural network methods, and ten commonly used machine learning algorithms. The evolved individual that is considered as a set of constructed features helps identify prominent original features which can assist dermatologists in making a diagnosis.

Non-deterministic Journey Planning in Multi-modal Transportation Networks: A Meta-heuristic Approach
Mohammad Haqqani, RMIT University, Xiaodong Li, RMIT University

Multi-modal journey planning, which allows multiple modes of transport to be used within a single trip, is becoming increasingly popular, due to a strong practical interest and the increasing availability of data. In real-life situations, transport networks often involve uncertainty, and yet, most approaches assume a deterministic environment, making plans more prone to failures such as major delays in the arrival or waiting for a long time at stations. In this paper, we tackle the multi-criteria stochastic journey planning problem in multi-modal transportation networks. We represent the problem as a probabilistic conditional planning problem and we use Markov Decision Processes (MDPs) to model the problem. Journey plans are optimised simultaneously against five criteria, namely: travel time, journey convenience, monetary cost, CO2, and personal energy expenditure (PEE). We develop an NSGA-III-based solver as a baseline search method for producing optimal policies for travelling from a given origin to a given destination. Our empirical evaluation uses Melbourne transport network using probabilistic density functions for estimated departure/arrival time of the trips. Numerical results suggest that the proposed method is effective for practical purposes and provide strong evidence in favour of switching from deterministic to non-deterministic planning.

Surrogate-Assisted Asynchronous Multiobjective Algorithm for Nuclear Power Plant Operations
Valentin Drouet, Commissariat à l’Énergie Atomique, Sébastien Verel, Université du Littoral Côte d’Opale, Jean-Michel Do, Commissariat à l’Énergie Atomique

In the context of the introduction of renewable energies in France, Nuclear Power Plant Operations (NPPO) are a key component for the compensation of the intermittent production of solar and wind power. In this work, we focus on the optimization of the operation cost and stability of power of a real-life power transient, while maintaining safety standards. From an optimization point of view, the NPPO problem is a typical example of a discrete constrained bi-objective problem based on time expensive computation simulation. We propose a massive asynchronous parallel master/workers MOEA/D assisted by a surrogate models. The algorithm design components are discussed and argued in this work. We show that our proposed surrogate assistance is able to improve algorithm performance and reliability, allowing us to extend our approach to a large range of strategic future real-life operations.

Towards Sustainable Forest Management Strategies with MOEAs ★
Philipp Back, Aalto University School of Business, Antti Suominen, Aalto University School of Business, Pekka Malo, Aalto University School of Business, Olli Tahvonen, University of Helsinki, Julian Blank, Michigan State University, Kalyanmoy Deb, Michigan State University

Sustainable forest management is a crucial element in combating climate change, plastic pollution, and other unsolved challenges of the 21st century. Forests not only produce wood - a renewable resource that is increasingly replacing fossil-based materials - but also preserve biodiversity and store massive amounts of carbon. Thus, a truly optimal forest policy has to balance profit-oriented logging with ecological and societal interests, and should thus be solved as a multi-objective optimization problem. Economic forest research, however,
has largely focused on profit maximization. Recent publications still scalarize the problem a priori by assigning weights to objectives. In this paper, we formulate a multi-objective forest management problem where profit, carbon storage, and biodiversity are maximized. We obtain Pareto-efficient forest management strategies by utilizing three state-of-the-art Multi-Objective Evolutionary Algorithms (MOEAs), and by incorporating domain-specific knowledge through customized evolutionary operators. An analysis of Pareto-efficient strategies and their harvesting schedules in the design space clearly shows the benefits of the proposed approach. Unlike many EMO application studies, we demonstrate how a systematic post-optimality trade-off analysis can be applied to choose a single preferred solution. Our pioneering work on sustainable forest management explores an entirely new application area for MOEAs with great societal impact.

Optimisation of Large Wave Farms using a Multi-strategy Evolutionary Framework


Wave energy is a fast-developing and promising renewable energy resource. The primary goal of this research is to maximise the total harnessed power of a large wave farm consisting of fully-submerged three-tether wave energy converters (WECs). Energy maximisation for large farms is a challenging search problem due to the costly calculations of the hydrodynamic interactions between WECs in a large wave farm and the high dimensionality of the search space. To address this problem, we propose a new hybrid multi-strategy evolutionary framework combining smart initialisation, discrete local search and continuous global optimisation. For assessing the performance of the proposed hybrid method, we compare it with a wide variety of state-of-the-art optimisation approaches, including six continuous evolutionary algorithms, four discrete search techniques and three hybrid optimisation methods. The results show that the proposed method performs considerably better in terms of both efficiency, convergence speed and farm output.

Evolutionary Bin Packing for Memory-Efficient Dataflow Inference Acceleration on FPGA

Mairin Kroes, Delft University of Technology, Lucian Petrica, Xilinx, Sorin Cotofana, Delft University of Technology, Michaela Blott, Xilinx

Convolutional neural network (CNN) dataflow inference accelerators implemented in Field Programmable Gate Arrays (FPGAs) have demonstrated increased energy efficiency and lower latency compared to CNN execution on CPUs or GPUs. However, the complex shapes of CNN parameter memories do not map well to FPGA on-chip memories (OCM), which results in poor OCM utilization and ultimately limits the size and types of CNNs which can be effectively accelerated on FPGAs. In this work, we present a design methodology that improves the mapping efficiency of CNN parameters to FPGA OCM. We frame the mapping as a bin packing problem and determine that traditional bin packing algorithms are not well suited to solve the problem within FPGA- and CNN-specific constraints. We hybridize genetic algorithms and simulated annealing with traditional bin packing heuristics to create flexible mappers capable of grouping parameter memories such that each group optimally fits FPGA on-chip memories. We evaluate these algorithms on a variety of FPGA inference accelerators. Our hybrid mappers converge to optimal solutions in a matter of seconds for all CNN use-cases, achieve an increase of up to 65% in OCM utilization efficiency for deep CNNs, and are up to 200x faster than current state-of-the-art simulated annealing approaches.

Simultaneously Searching and Solving Multiple Avoidable Collisions for Testing Autonomous Driving Systems

Alessandro Calò, Technical University of Munich, Paolo Arcaini, National Institute of Informatics, Shaukat Ali, Simula Research Laboratory, Florian Hauer, Technical University of Munich, Fuyuki Ishikawa, National Institute of Informatics

The oracle problem is a key issue in testing Autonomous Driving Systems (ADS): when a collision is found, it is not always clear whether the ADS is responsible for it. Our recent search-based testing approach offers a solution to this problem by defining a collision as avoidable if a differently configured ADS would have avoided it. This approach searches for both collision scenarios and the ADS configurations capable of avoiding them. However, its main problem is that the ADS configurations generated for avoiding some collisions are not suitable for preventing other ones. Therefore, it does not provide any guidance to automotive engineers for improving the safety of the ADS. To this end, we propose a new search-based approach to generate configurations of the ADS that can avoid as many different types of collisions as possible. We present two versions of the approach, which differ in the way of searching for collisions and alternative configurations. The approaches have been experimented on the path planner component of an ADS provided by our industry partner.

Multi-Objective Optimal Distribution of Materials in Hybrid Components

Thomas Gossuin, CRIG, Didier Garray, Sirris, Vincent Kelner, HELMo

Genetic algorithms are 0th-order methods and therefore praised in many non-differentiable optimization problems, which encompass the majority of real world applications. In this work, a multi-objective optimization of hybrid, i.e. multi-material, components under technological constraints is performed to guide engineers towards the optimal design of man-
Seeding Strategies for Multi-Objective Test Case Selection: An Application on Simulation-based testing

Aitor Arrieta, Mondragon University, Joseba Andoni Agirre, Mondragon University, Goiuria Sagardui, Mondragon University

The time it takes software systems to be tested is usually long. This is often caused by the time it takes the entire test suite to be executed. To optimize this, regression test selection approaches have allowed for improvements to the cost-effectiveness of verification and validation activities in the software industry. In this area, multi-objective algorithms have played a key role in selecting the appropriate subset of test cases from the entire test suite. In this paper, we propose a set of seeding strategies for the test case selection problem that generate the initial population of multi-objective algorithms. We integrated these seeding strategies with an NSGA-II algorithm for solving the test case selection problem in the context of simulation-based testing. We evaluated the strategies with six case studies and a total of 21 fitness combinations for each case study (i.e., a total of 126 problems). Our evaluation suggests that these strategies are indeed helpful for solving the multi-objective test case selection problem. In fact, two of the proposed seeding strategies outperformed the NSGA-II algorithm without seeding population with statistical significance for 92.8 and 96% of the problems.

Causes and Effects of Fitness Landscapes in Unit Test Generation

Nasser Albunian, The University of Sheffield, Gordon Fraser, University of Passau, Dirk Sudholt, The University of Sheffield

Search-based unit test generation applies evolutionary search to maximize code coverage. Although the performance of this approach is often good, sometimes it is not, and how the fitness landscape affects this performance is poorly understood. This paper presents a thorough analysis of 331 Java classes by (i) characterizing their fitness landscape using six established fitness landscape measures, (ii) analyzing the impact of these fitness landscape measures on the search, and (iii) investigating the underlying properties of the source code influencing these measures. Our results reveal that classical indicators for rugged fitness landscapes suggest well searchable problems in the case of unit test generation, but the fitness landscape for most problem instances is dominated by detrimental plateaus. A closer look at the underlying source code suggests that these plateaus are frequently caused by code in private methods, methods throwing exceptions, and boolean flags. This suggests that inter-procedural distance metrics and testability transformations could improve search-based test generation.
Scalability analysis of Grammatical Evolution Based Test Data Generation
Muhammad Sheraz Anjum, University of Limerick, Conor Ryan, University of Limerick

Heuristic-based search techniques have been increasingly used to automate different aspects of software testing. Several studies suggest that variable interdependencies may exist in branching conditions of real life programs, and these dependencies result in the need for highly precise data values (such as of the form i=j=k) for code coverage analysis. This requirement makes it very difficult for Genetic Algorithm (GA)-based approach to successfully search for the required test data from vast search spaces of real life programs. Ariadne is the only Grammatical Evolution (GE)-based test data generation system, proposed to date, that uses grammars to exploit variable interdependencies to improve code coverage. Ariadne has been compared favourably to other well-known test data generation techniques in the literature; however, its scalability has not yet been tested for increasingly complex programs. This paper presents the results of a rigorous analysis performed to examine Ariadne's scalability. We also designed and employed a large set of highly scalable 18 benchmark programs for our experiments. Our results suggest that Ariadne is highly scalable as it exhibited 100% coverage across all the programs of increasing complexity with significantly smaller search costs than GA-based approaches, which failed even with huge search budgets.

Enhancing Search-Based Product Line Design with Crossover Operators
Diego Fernandes da Silva, Universidade Estadual de Maringá, Luiz Fernando Okada, Universidade Estadual de Maringá, Thelma Elita Colanzi, Universidade Estadual de Maringá, Wesley K. G. Assunção, Federal University of Technology - Paraná

The Product Line Architecture (PLA) is one of the most important artifacts of a Software Product Line. PLA designing has been formulated as a multi-objective optimization problem and successfully solved by a state-of-the-art search-based approach. However, the majority of empirical studies optimize PLA designs without applying one of the most important genetic operators: the crossover. A crossover operator for PLA design, named Feature-driven Crossover, was proposed in a previous study. In spite of the promising results, the authors of such an operator acknowledged that some improvements in their operator would prevent some incomplete solutions that were generated and lead to better performance. To overcome these limitations, this paper aims to enhance the search-based PLA design optimization by improving the Feature-driven Crossover and introducing a novel crossover operator specific for PLA design. The proposed operators were evaluated in two well-studied PLA designs, considering three experimental configurations using NSGA-II in comparison with a baseline experiment that uses only mutation operators. Empirical results show the usefulness and efficiency of the presented operators reaching consistent solutions. In addition, the results pointed that the two operators complement each other, leading to PLA design solutions with better feature modularization than the baseline experiment.

Genetic Algorithms for Redundancy in Interaction Testing
Ryan Edward Dougherty, Colgate University

It is imperative for testing to determine if the components within large-scale software systems operate functionally. Interaction testing involves designing a suite of tests, which guarantees to detect a fault if one exists among a small number of components interacting together. The cost of this testing is typically modeled by the number of tests, and thus much effort has been taken in reducing this number. Here, we incorporate redundancy into the model, which allows for testing in non-deterministic environments. Existing algorithms for constructing these test suites usually involve one “fast” algorithm for generating most of the tests, and another “slower” algorithm to “complete” the test suite. We employ a genetic algorithm that generalizes these approaches that also incorporates redundancy by increasing the number of algorithms chosen, which we call “stages.” By increasing the number of stages, we show that not only can the number of tests be reduced compared to existing techniques, but the computational time in generating them is also greatly reduced.

Towards Rigorous Validation of Energy Optimisation Experiments
Mahmoud Bokhari, The University of Adelaide, Markus Wagner, The University of Adelaide, Brad Alexander, The University of Adelaide

The optimisation of software energy consumption is of growing importance across all scales of modern computing, i.e., from embedded systems to data-centres. Practitioners in the field of Search-Based Software Engineering and Genetic Improvement of Software acknowledge that optimising software energy consumption is difficult due to noisy and expensive fitness evaluations. However, it is apparent from results to date that more progress needs to be made in rigorously validating optimisation results. This problem is pressing because modern computing platforms have highly complex and variable behaviour with respect to energy consumption. To compare solutions fairly we propose in this paper a new validation approach called R3-validation which exercises software variants in a rotated-round-robin order. Using a case study, we present an in-depth analysis of the impacts of changing system states on software energy usage, and we show how R3-validation mitigates these. We compare it with current validation approaches across multiple devices and operating systems, and we show that it aligns best with actual platform behaviour.
Fast Mutation in Crossover-based Algorithms
Denis Antipov, ITMO University, Maxim Buzdalov, ITMO University, Benjamin Doerr, École Polytechnique

The heavy-tailed mutation operator proposed in Doerr et al. (GECCO 2017), called fast mutation to agree with the previously used language, so far was successfully used only in purely mutation-based algorithms. There, it can relieve the algorithm designer from finding the optimal mutation rate and nevertheless obtain a performance close to the one that the optimal mutation rate gives. In this first runtime analysis of a crossover-based algorithm using a heavy-tailed choice of the mutation rate, we show an even stronger impact. With a heavy-tailed mutation rate, the runtime of the $(1 + \lambda \mu)$ genetic algorithm on the OneMax benchmark function becomes linear in the problem size. This is asymptotically faster than with any static mutation rate and is the same asymptotic runtime that can be obtained with a self-adjusting choice of the mutation rate. This result is complemented by an empirical study which shows the effectiveness of the fast mutation also on random MAX-3SAT instances.

The Node Weight Dependent Traveling Salesperson Problem: Approximation Algorithms and Randomized Search Heuristics
Jakob Bossek, The University of Adelaide, Katrin Casel, Hasso Plattner Institute, Pascal Kerschke, University of Münster, Frank Neumann, The University of Adelaide

Several important optimization problems in the area of vehicle routing can be seen as variants of the classical Traveling Salesperson Problem (TSP). In the area of evolutionary computation, the Traveling Thief Problem (TTP) has gained increasing interest over the last 5 years. In this paper, we investigate the effect of weights on such problems, in the sense that the cost of traveling increases with respect to the weights of nodes already visited during a tour. This provides abstractions of important TSP variants such as the Traveling Thief Problem and time dependent TSP variants, and allows to study precisely the increase in difficulty caused by weight dependence. We provide a 3.59-approximation for this weight dependent version of TSP with metric distances and bounded positive weights. Furthermore, we conduct experimental investigations for simple randomized local search with classical mutation operators and two variants of the state-of-the-art evolutionary algorithm EAX adapted to the weighted TSP. Our results show the impact of the node weights on the position of the nodes in the resulting tour.

The $(1 + (\lambda, \mu))$ GA is Even Faster on Multimodal Problems
Denis Antipov, ITMO University, Benjamin Doerr, École Polytechnique, Vitalii Karavaev, ITMO University

For the $(1 + (\lambda, \mu))$ genetic algorithm rigorous runtime analyses on unimodal fitness functions have shown that it can be faster than classical evolutionary algorithms, though on these simple problems the gains are only moderate. In this work, we conduct the first runtime analysis of this algorithm on a multimodal problem class, the jump functions benchmark. We show that with the right parameters, the $(1 + (\lambda, \mu))$ GA optimizes any jump function with jump size $2 \leq k \leq n/16$ in expected time $O(n^{(k+1)/2}e^{O(k)k^{-k/2}})$, which significantly and already for constant $k$ outperforms standard mutation-based algorithms with their $\Theta(n^k)$ runtime and standard crossover-based algorithms with their $O(n^{k-1})$ runtime. Our work also suggests some general advice on how to set the parameters of the $(1 + (\lambda, \mu))$ GA, which might ease the further use of this algorithm.

More Effective Randomized Search Heuristics for Graph Coloring Through Dynamic Optimization
Jakob Bossek, The University of Adelaide, Frank Neumann, The University of Adelaide, Pan Peng, University of Sheffield, Dirk Sudholt, The University of Sheffield

Dynamic optimization problems have gained significant attention in evolutionary computation as evolutionary algorithms (EAs) can easily adapt to changing environments. We show that EAs can solve the graph coloring problem for bipartite graphs more efficiently by using dynamic optimization. In our approach the graph instance is given incrementally such that the EA can reoptimize its coloring when a new edge introduces a conflict. We show that, when edges are inserted in a way that preserves graph connectivity, Randomized Local Search (RLS) efficiently finds a proper 2-coloring for all bipartite graphs. This includes graphs for which RLS and other EAs need exponential expected time in a static optimization scenario. We investigate different ways of building up the graph by popular graph traversals such as breadth-first-search and depth-first-search and analyse the resulting runtime behavior. We further show that offspring populations (e.g. a $(1+\lambda)$RLS) lead to an exponential speedup in $\lambda$. Finally, an island model using 3 islands succeeds in an optimal time of $\Theta(n)$ on every $m$-edge bipartite graph, outperforming offspring populations. This is the first example where an island model guarantees a speedup that is not bounded in the number of islands.

Fixed-Target Runtime Analysis
Maxim Buzdalov, ITMO University, Benjamin Doerr, École Polytechnique, Carola Doerr, Sorbonne Université, Dmitry Vinokurov, ITMO University

Runtime analysis aims at contributing to our understanding of evolutionary algorithms through mathematical analyses of
Does Comma Selection Help To Cope With Local Optima?
Benjamin Doerr, École Polytechnique

One hope of using non-elitism in evolutionary computation is that it aids leaving local optima. We perform a rigorous runtime analysis of a basic non-elitist evolutionary algorithm (EA), the \((\mu, \lambda)\) EA, on the most basic benchmark function with a local optimum, the jump function. We prove that for all reasonable values of the parameters and the problem, the expected runtime of the \((\mu, \lambda)\) EA is, apart from lower order terms, at least as large as the expected runtime of its elitist counterpart, the \((\mu + \lambda)\) EA (for which we conduct the first runtime analysis to allow this comparison). Consequently, the ability of the \((\mu, \lambda)\) EA to leave local optima to inferior solutions does not lead to a runtime advantage. We complement this lower bound with an upper bound that, for broad ranges of the parameters, is identical to our lower bound apart from lower order terms. This is the first runtime result for a non-elitist algorithm on a multi-modal problem that is tight apart from lower order terms.

Self-Adjusting Evolutionary Algorithms for Multimodal Optimization
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Recent theoretical research has shown that self-adjusting and self-adaptive mechanisms can provably outperform static settings in evolutionary algorithms for binary search spaces. However, the vast majority of these studies focuses on unimodal functions which do not require the algorithm to flip several bits simultaneously to make progress. In fact, existing self-adjusting algorithms are not designed to detect local optima and do not have any obvious benefit to cross large Hamming gaps. We suggest a mechanism called stagnation detection that can be added as a module to existing evolutionary algorithms (both with and without prior self-adjusting schemes). Added to a simple \((1+1)\)EA, we prove an expected runtime on the well-known Jump benchmark that corresponds to an asymptotically optimal parameter setting and outperforms other mechanisms for multimodal optimization like heavy-tailed mutation.

We also investigate the module in the context of a self-adjusting \((1+\lambda)\)EA and show that it combines the previous benefits of this algorithm on unimodal problems with more efficient multimodal optimization. To explore the limitations of the approach, we additionally present an example where both self-adjusting mechanisms, including stagnation detection, do not help to find a beneficial setting of the mutation rate. Finally, we investigate our module for stagnation detection experimentally.
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