### Genetic and Evolutionary Computation Conference 2020

**Conference Program** 

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Cancún, México July 8-12, 2020



Association for Computing Machinery



Advancing Computing as a Science & Profession

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GECCO is sponsored by the Association for Computing Machinery Special Interest Group for Genetic and Evolutionary Computation (SIGEVO). SIG Services: 2 Penn Plaza, Suite 701, New York, NY, 10121, USA, 1-800-342-6626 (USA and Canada) or +212-626-0500 (global).

#### 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 Mexico City (México)

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Competitions Chair:	Markus Wagner, University of Adelaide
Hot off The Press:	Heike Trautmann, University of Münster
Late-Breaking Abstracts:	Ke Tang, Southern University of Science and Technology
Annual "Hummies Awards for Human Competitive Results":	John Koza, <i>Standford University</i> Erik Goodman, <i>University of Michigan</i> William B. Langdon, <i>University College London</i>
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- ACO-SI Ant Colony Optimization and Swarm Intelligence

Marde Helbig, Griffith University Christine Solnon, LIRIS, INSA de Lyon

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

Stéphane Doncieux, Sorbonne University, CNRS Dennis G. Wilson, ISAE-SUPAERO

DETA — Digital Entertainment Technologies and Arts Francisco Fernández de Vega, Universidad de Extremadura 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

Fuyuki Ishikawa, National Institute of Informatics Justyna Petke, University College London

#### Theory

Johannes Lengler, ETH Zürich Frank Neumann, The University of Adelaide

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Ender Özcan, University of Nottingham Mathias Pacher, Goethe Universität Frankfurt am Main Federico Pagnozzi, Université Libre de Bruxelles Pramudita Satria Palar, Institut Teknologi Bandung Annibale Panichella, Delft University of Technology Gregor Papa, Jozef Stefan Institute, Jožef Stefan International Postgraduate School Luis Paquete, University of Coimbra Andrew J. Parkes, University of Nottingham Konstantinos Parsopoulos, University of Ioannina Robert M. Patton, Oak Ridge National Laboratory David Pätzel, Universität Augsburg Martín Pedemonte, Instituto de Computación, Facultad de Ingeniería, Universidad de la República Wenbin Pei, Victoria University of Wellington Paola Pellegrini, Université Gustave EIFFEL David Pelta, University of Granada Xingguang Peng, Northwestern Polytechnical University Yiming Peng, Victoria University of Wellington Jordi Pereira, Universidad Adolfo Ibáñez Francisco Baptista Pereira, Instituto Superior de Engenharia de Coimbra, Portugal Leslie Pérez Cáceres, Pontificia Universidad Católica de Valparaíso Diego Perez-Liebana, Queen Mary University of London Justyna Petke, University College London Stjepan Picek, Delft University of Technology Martin Pilat, Charles University, Faculty of Mathematics and Physics Erik Pitzer, University of Applied Sciences Upper Austria Clara Pizzuti, Institute for High Performance Computing and Networking - ICAR National Research Council of Italy -CNR Michal Pluhacek, Tomas Bata University in Zlín Petrica Pop, Technical University of Cluj-Napoca, North University Center at Baia Mare, Romania Daniel Porumbel, CEDRIC, CNAM (Conservatoire National des Arts et Métiers) Petr Pošík, Czech Technical University in Prague Pasqualina Potena, RISE Research Institutes of Sweden AB Matthias Prandtstetter, AIT Austrian Institute of Technology GmbH Steve Prestwich, University College Cork Mike Preuss, Universiteit Leiden Kenneth Price, ESEP Jakob Puchinger, IRT SystemX, CentraleSupélec Robin Purshouse, University of Sheffield Chao Qian, Nanjing University Kai Qin, Swinburne University of Technology Xin Qiu, National University of Singapore B. Y. Qu, Zhongyuan University of Technology Sebastian Raggl, University of Applied Sciences Upper Aus-

tria

Alma Rahat, Swansea University

Günther R. Raidl, TU Wien

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Cristian Ramirez, Otto von Guericke Universität Jeremy Rapin, Facebook Khaled Rasheed, University of Georgia Tapabrata Ray, University of New South Wales Margarita Rebolledo Coy, TH Köln, VU University Amsterdam Patrick M. Reed, Cornell University Frederik Rehbach, TH Koeln Gustavo Reis, School of Technology and Management, Polytechnic Institute of Leiria, Portugal Lewis Rhyd, 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 Rodriguez-Tello, CINVESTAV - Tamaulipas Nicolas Rojas-Morales, Universidad Técnica Federico Santa María Andrea Roli, Alma Mater Studiorum Universita' di Bologna Alejandro Rosales-Pérez, Centro de Investigación en Matemáticas Nick Ross, University of Exeter Franz Rothlauf, University of Mainz Jonathan Rowe, University of Birmingham Proteek Chandan Roy, Michigan State University Guenter 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 Sahraoui, DIRO, 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, University of the Basque Country (UPV/EHU) Alberto Santini, Universitat Pompeu Fabra Haroldo Santos, Federal University of Ouro Preto Valentino Santucci, University for Foreigners of Perugia Yuji Sato, Hosei University Hiroyuki Sato, The University of Electro-Communications Fredéric Saubion, University of Angers, France Saket Saurabh, IMSc, University of Bergen Umberto Scafuri, ICAR-CNR Robert Schaefer, AGH 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-IPN Marco Scirea, University of Southern Denmark

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- Jerry Swan, NNAISENSE SA, University of York

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Taiwan University Guo Yu, Department of Computer Science of University of

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Heiner Zille, Otto-von-Guericke University Magdeburg Nur Zincir-Heywood, Dalhousie University 21

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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 Sydney/Melbourne (Australia), please add 15 hours.
For Shanghai/Beijing (China), please add 13 hours.
For Taipei (Taiwan), 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

Wednesday, July 8	Thursday, July 9	Friday, July 10	Saturday, July 11	Sunday, July 12
		Opening Session 8:30		
Tutorials and Workshops	Tutorials, Workshops, and Competitions	Invited Keynote Leslie Valiant	Invited Keynote Kalyanmoy Deb	Papar Sassions
08:30-10:20	08:30-10:20	08:50-10:10	08:50-10:10	09:00-10:40
Prook	Prook	Break	Break	
Diedk	Dieak			Brook
Tutorials and Workshops	Tutorials and Tutorials, Workshops, and Competitions		Poster Session 2 10:30-12:10	SIGEVO Keynote Darrell Whitley
10:40-12:30	10:40-12:30			11:00-12:20
Lunch break	Women@GECCO 12:30-14:00	Job Market 12:10-13:40	Lunch break	SIGEVO Meeting, Awards and Closing 12:20-13:50
Tutorials and Workshops 14:00-15:50	Tutorials and Workshops 14:00-15:50	Paper Sessions, HOP and HUMIES 13:40-15:20	Paper Sessions 13:40-15:20	
		Break	Break	
Break Tutorials and Workshops	Break Tutorials, Workshops, and LBA	Paper Sessions 15:40-17:20	Paper Sessions 15:40-17:20	
16:10-18:00	16:10-18:00	Break	Break	
		Poster Session 1 17:40-19:20	Paper Sessions 17:40-19:20	

### Workshop and Tutorial Sessions, Wednesday, July 8

08:30-10:20	10:40-12:30	14:00-15:50	16:10-18:00		
EAPwU — Evolutionary Algorithms for Problems with Uncertainty	Visualization in Multiobjective Optimization	GI: Taking real-world source code and improving it using genetic programming	Evolutionary Computation for Digital Art		
p. 42	Filipic, Tušar	Haraldsson, Woodward, Wagner, Alexander	Neumann, Neumann		
NEWK — Neuroevolution at work		SAEOpt — Surrogate- Assisted Evolutionary Optimisation			
p. 42		p. 45			
Hyper-heuristics	Introductory Mathematical Programming for EC	Evolutionary Multi- objective Optimization: Past, Present and Future	Evolutionary Many- Objective Optimization		
Woodward, Tauritz	Shir	Deb	Ishibuchi, Sato		
EvoSoft — Evolutionary Com	putation Software Systems	Fitness landscape analysis to understand and predict alg. performance for single- and multi-objective opt.			
	p. 42	Verel, Derbel			
Evolutionary Computation: A Unified Approach	Model-Based Evolutionary Algorithms	evolutionary Computation and Games	Learning Classifier Systems: From Principles to Modern Systems		
De Jong	Thierens, Bosman	Togelius, Risi, Yannakakis	Stein, Nakata		
ECPERM — Evolutionary Cor Problems	nputation for Permutation p. 43	iGECCO — Interactive Metho	<b>ds @ GECCO</b> p. 45		
Runtime Analysis of Evolutionary Algorithms: Basic Introduction	Introduction to Genetic Programming	Evolution of Neural Networks	Neuroevolution for Deep Reinforcement Learning Problems		
Lehre, Oliveto	O'Reilly, Hemberg	Miikkulainen	На		
	PDEIM — Parallel and Distributed Evolutionary Inspired Methods	VizGEC 2020 — Visualisation Methods in Genetic and Evolutionary Computation			
GreenAI — Evolutionary	Quality-Diversity	Theory and Practice of			
and machine learning solutions in environment, renewable and ecologically- aware scenarios	Optimization	Population Diversity in Evolutionary Computation			
p. 44 IAM — Industrial	Representations for	A Gentle Introduction	BENCHMARK — Good		
Application of Metaheuristics	Evolutionary Algorithms	to Theory (for Non- Theoreticians)	Benchmarking Practices for Evolutionary Computation		
p. 44	Rothlauf	Doerr	p. 46		
Tuto	rials Workshops				

### Workshop and Tutorial Sessions, Thursday, July 9

08:30-10:20	10:40-12:30	14:00-15:50	16:10-18:00		
Decomposition Multi- Objective Optimisation: Current Developments and Future Opportunities	Sequential Experimentation by Evolutionary Algorithms	Statistical Analyses for Meta-heuristic Stochastic Optimization Algorithms	LBA – Late-Breaking Abstracts		
Li, Zhang, Zapotecas	Shir, Bäck	Eftimov, Korošec	p. 51		
RWACMO — Real-world	ECADA — Evolutionary	ADRR — Automated Design	GI@GECCO — The Ninth		
Applications of Continuous	Computation for the	of Robots for the Real-world	Genetic Improvement		
and Mixed-integer	Automated Design of		Workshop (2020)		
p. 47	p. 48	p. 50	p. 52		
Addressing Ethical	Evolutionary Computer	Evolutionary Computation	Evolutionary Algorithms		
Challenges within	Vision	and Machine Learning in	in Biomedical Data Mining:		
Evolutionary Computation		Cryptology	Challenges, Solutions, and		
Applications			Holders		
Torresen		Picek, Jakobovic	Urbanowicz, Sipper		
Dynamic Control Parameter Choices in Evolutionary	with Coevolutionary	Swarm Intelligence in Cybersecurity			
Computation	Algorithms	cybersecurity			
-	°				
Papa, Doerr	Krawiec, Heywood	Zelinka, Senkerik			
Search Based Software	<b>Recent Advances in Particle</b>	Evol. Comp. and Deep	Push		
Engineering: challenges,	Swarm Optimization	Learning for Image Analysis,			
opportunities and recent	Analysis and Understanding	Signal Processing and Pattern Recognition			
	Engelbrecht Cleghorn	Zhang Cagnoni	Spector		
Competitions	Engelbreent, Cleghorn	SecDef 2020 — Constic and			
competitions		Evolutionary Computation in Defense, Security and Risk Management			
	p. 57	p. 51			
SWINGA — Swarm Intelliger Perspectives and Challenges	ice Algorithms: Foundations,		Benchmarking and Analyzing Iterative Optimization Heuristics with IOHprofiler		
	p. 47		Wang, Doerr, Shir, Bäck		
Automated Algorithm	Theory of Estimation-of-		<b>Evolutionary Computation</b>		
Configuration and Design	Distribution Algorithms		for Feature Selection and		
			Feature Construction		
López-Ibáñez, Stutzle	Witt		Xue, Zhang		
Semantic Genetic	Evolutionary Algorithms	Design Principles for Matrix	A Hands-on Guide to		
Programming	and Machine Learning: synergies and challenges	Adaptation Evolution	Distributed Computing Paradigms for Evolutionary		
	synergies und endnenges	otrategies	Computation		
Moraglio, Krawiec	Squillero, Tonda	Beyer	Wang, Zhi		
MedGEC — Workshop on	IWLCS — International Work	shop on Learning Classifier	E-MaOP + DTEO —		
Medical Applications of	Systems		Evolutionary Many-		
Genetic and Evolutionary			objective Optimization +		
Computation			in Evol. Opt.		
p. 48		p. 49	- p. 53		
Tut	orials Workshops	Competitions	LBA session		

Friday         Friday           July 10         July 10           10:30-12:10         13:40-15:20		Friday July 10 15:40-17:20		Saturday July 11 13:40-15:20		Saturday July 11 15:40-17:20		Saturday July 11 17:40-19:20		Sunday July 12 09:00-10:40			
GECH1		GECH2		ENUM1		SBSE1 +ENUM2 +THEORY1	☆	GECH3	☆	CS3		ENUM3	
GA1	p. 69	GA2	p. 71		p. 72	ACO-SI2	p. 74	DETA	p. 76	GA3	p. 80		p. 82
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HOP1		HOP2		RWA1		RWA2		RWA3		RWA4	☆	RWA5	
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		HUMIES		EML1		EML2		EML3		EML4	☆	EML5	
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		GP1		GP2	-0	GP3		GP4	☆	THEORY2		THEORY3	
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CSI		+ACO-SI1	X	ECOMI	X	ECOM2		ECOM3		ECOM4		GECH4	
	p. 68		p. 70		p. 72	l	p. 75		p. 77		p. 81		p. 83
EMO1	☆							EMO2		EMO3		EMO4	
	p. 68								p. 77		p. 80		p. 83
										SBSE2	01		
											p. 81		

### Parallel Sessions, Friday, July 10 through Sunday, July 12



#### **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



#### GECC GECCO KEYNOTE **Biological Evolution as a Form of Learning**

Leslie Valiant, Harvard University, United States

Friday, July 10, 8:50-10:10



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).

#### GECCO KEYNOTE 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-diversed 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 Mamdani 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*.

## sigevo SIGEVO KEYNOTE

**Removing Randomness from Evolutionary Algorithms** 

Sunday, July 12, 11:00-12:20

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.

Saturday, July 11, 8:50-10:10
Tutorials



## Introductory Tutorials

Hyper-heuristics	Wednesday, July 8, 08:30-10:20
John R. Woodward, Queen Mary University of London	
Daniel R. Tauritz, Missouri University of Science and Technology	
Evolutionary Computation: A Unified Approach	Wednesday, July 8, 08:30-10:20
Kenneth De Jong, George Mason University	
Runtime Analysis of Evolutionary Algorithms: Basic Introduction	Wednesday, July 8, 08:30-10:20
Per Kristian Lehre, University of Birmingham	
Pietro S. Oliveto, University of Sheffield	
Introductory Mathematical Programming for EC	Wednesday, July 8, 10:40-12:30
Ofer M. Shir, Tel-Hai College	
Model-Based Evolutionary Algorithms	Wednesday, July 8, 10:40-12:30
Dirk Thierens, Utrecht University	
Peter A. N. Bosman, Centrum Wiskunde & Informatica (CWI)	
Introduction to Genetic Programming	Wednesday, July 8, 10:40-12:30
Una-May O'Reilly, CSAIL, Massachusetts Institute of Technology	
Erik Hemberg, Massachusetts Institute of Technology	
Representations for Evolutionary Algorithms	Wednesday, July 8, 10:40-12:30
Franz Rothlauf, Universität Mainz	
Evolutionary Multi-objective Optimization: Past, Present and Future	Wednesday, July 8, 14:00-15:50
Kalyanmoy Deb, Michigan State University	
Evolutionary Computation and Games	Wednesday, July 8, 14:00-15:50
Julian Togelius, New York University	
Sebastian Risi, IT University of Copenhagen	
Georgios N. Yannakakis, University of Malta	
Evolution of Neural Networks	Wednesday, July 8, 14:00-15:50
Risto Miikkulainen, The University of Texas at Austin	
A Gentle Introduction to Theory (for Non-Theoreticians)	Wednesday, July 8, 14:00-15:50
Benjamin Doerr, Ecole Polytechnique	
Evolutionary Many-Objective Optimization	Wednesday, July 8, 16:10-18:00
Hisao Ishibuchi, Southern University of Science Technology (SUSTech)	
Hiroyuki Sato, The University of Electro-Communications	
Learning Classifier Systems: From Principles to Modern Systems	Wednesday, July 8, 16:10-18:00
Anthony Stein, University of Hohenheim	
Masaya Nakata, Yokohama National University	
Neuroevolution for Deep Reinforcement Learning Problems	Wednesday, July 8, 16:10-18:00
David Ha, <i>Google Brain</i>	

## **Advanced Tutorials**

<b>Visualization in Multiobjective Optimization</b> Bogdan Filipic, <i>Jozef Stefan Institute</i> Tea Tušar, <i>Jozef Stefan Institute</i>	Wednesday, July 8, 10:40-12:30
<b>Quality-Diversity Optimization</b> Antoine Cully, <i>Imperial College</i> Jean-Baptiste Mouret Inria	Wednesday, July 8, 10:40-12:30
Stéphane Doncieux, Sorbonne Université	
<b>Genetic improvement: Taking real-world source code and improving it using genetic programming</b> Saemundur O. Haraldsson, <i>University of Stirling</i> John R. Woodward, <i>Queen Mary University of London</i> Markus Wagner, <i>The University of Adelaide</i> Bradley Alexander, <i>The University of Adelaide</i>	Wednesday, July 8, 14:00-15:50
<b>Fitness landscape analysis to understand and predict algorithm</b> <b>performance for single- and multi-objective optimization</b> Sébastien Verel, Univ. Littoral Côte d'Opale Bilel Derbel, Univ. Lille	Wednesday, July 8, 14:00-15:50
<b>Theory and Practice of Population Diversity in Evolutionary Computation</b> Dirk Sudholt, <i>University of Sheffield</i> Giovanni Squillero, <i>Politecnico di Torino</i>	Wednesday, July 8, 14:00-15:50
<b>Evolutionary Computation for Digital Art</b> Aneta Neumann, <i>The University of Adelaide</i> Frank Neumann, <i>The University of Adelaide</i>	Wednesday, July 8, 16:10-18:00
<b>Decomposition Multi-Objective Optimisation: Current Developments and</b> <b>Future Opportunities</b> 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
<b>Dynamic Control Parameter Choices in Evolutionary Computation</b> Gregor Papa, <i>Jozef Stefan Institute</i> Carola Doerr, <i>CNRS</i>	Thursday, July 9, 08:30-10:20
<b>Semantic Genetic Programming</b> Alberto Moraglio, <i>University of Exeter</i> Krzysztof Krawiec, <i>Poznan University of Technology</i>	Thursday, July 9, 08:30-10:20
<b>Sequential Experimentation by Evolutionary Algorithms</b> Ofer M. Shir, <i>Tel-Hai College</i> Thomas Bäck, <i>Leiden University</i>	Thursday, July 9, 10:40-12:30

<b>Solving Complex Problems with Coevolutionary Algorithms</b> Krzysztof Krawiec, <i>Poznan University of Technology</i> Malcolm Heywood, <i>Dalhousie University</i>	Thursday, July 9, 10:40-12:30
<b>Recent Advances in Particle Swarm Optimization Analysis and Understanding</b> Andries Petrus Engelbrecht, <i>Stellenbosch University</i> Christopher W. Cleghorn, <i>University of Pretoria</i>	Thursday, July 9, 10:40-12:30
<b>Statistical Analyses for Meta-heuristic Stochastic Optimization Algorithms</b> Tome Eftimov, <i>Jozef Stefan Institute</i> Peter Korošec, <i>Jozef Stefan Institute</i>	Thursday, July 9, 14:00-15:50
<b>Design Principles for Matrix Adaptation Evolution Strategies</b> Hans-Georg Beyer, <i>Vorarlberg University of Applied Sciences</i>	Thursday, July 9, 14:00-15:50
<b>A Hands-on Guide to Distributed Computing Paradigms for Evolutionary Computation</b> Rui Wang, <i>Uber AI</i> Jiale Zhi, <i>Uber AI</i>	Thursday, July 9, 16:10-18:00
<b>Benchmarking and Analyzing Iterative Optimization Heuristics with</b> <b>IOHprofiler</b> Hao Wang, <i>Sorbonne Université</i> Carola Doerr, <i>Sorbonne Université</i> Ofer M. Shir, <i>Tel-Hai College</i> Thomas Bäck, <i>Leiden University</i>	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
<b>Automated Algorithm Configuration and Design</b> Manuel López-Ibáñez, <i>University of Manchester</i> Thomas Stutzle, <i>Université Libre de Bruxelles</i>	Thursday, July 9, 08:30-10:20
<b>Evolutionary Computer Vision</b> Gustavo Olague, <i>CICESE</i>	Thursday, July 9, 10:40-12:30
<b>Theory of Estimation-of-Distribution Algorithms</b> Carsten Witt, <i>Technical University Of Denmark</i>	Thursday, July 9, 10:40-12:30

<b>Evolutionary Algorithms and Machine Learning: synergies and challenges</b> Giovanni Squillero, <i>Politecnico di Torino</i>	Thursday, July 9, 10:40-12:30
Alberto Iolida, Ilvitae	
Evolutionary Computation and Machine Learning in Cryptology	Thursday, July 9, 14:00-15:50
Stjepan Picek, Delft University of Technology	<u> </u>
Domagoj Jakobovic, University of Zagreb	
Swarm Intelligence in Cybersecurity	Thursday, July 9, 14:00-15:50
Ivan Zelinka, VSB-TU Ostrava	
Roman Senkerik, UTB in Zlin	
Evolutionary Computation and Evolutionary Deep Learning for Image	Thursday, July 9, 14:00-15:50
Analysis, Signal Processing and Pattern Recognition	
Mengjie Zhang, Victoria University of Wellington	
Stefano Cagnoni, University of Parma	
Evolutionary Algorithms in Biomedical Data Mining: Challenges, Solutions,	Thursday, July 9, 16:10-18:00
and Frontiers	
Ryall J. Oldanowicz, University of Pennsylvania Macha Sinnar, Ban, Currier University	
Mosne Sipper, Ben-Gunon Oniversity	
Push	Thursday, July 9, 16:10-18:00
Lee Spector, Amherst College	
Evolutionary Computation for Feature Selection and Feature Construction	Thursday, July 9, 16:10-18:00
Bing Xue, Victoria University of Wellington	
Mengjie Zhang, Victoria University of Wellington	

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 Đurasević

## 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 Baioletti, 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

Eneko 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

- **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
- **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 Multiobjetive Problems Cosijopii Garcia-Garcia, Alicia Morales-Reyes, Maria-Guadalupe Martínez-Peñaloza

### 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

- 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

### 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

## 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); Ponnuthurai 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** Seved 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-Phuoc 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 IWLCS 2019 to 2020

David Pätzel, 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 Pätzel, 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

### 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

### **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 Jyvaskyla, Finland

### Ask me anything

Una-May O'Reilly, Computer Science & Artificial Intelligence Lab, MIT

### Women@GECCO meets pub quiz

### 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

### 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

Kourigo Moreno, Anures 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

## 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

### 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

### LBA — Late Breaking Abstracts

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

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

- 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, Plínio 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 Hvatov, Alexander Boukhanovsky
- 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

### 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

Judging Panel: Erik D. Goodman, Una-May O'Reilly, Wolfgang Banzhaf, Darrell D. Whitley, Lee Spector

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 twoprizesof* 1,000) Silver (\$3,000) and Gold (\$5,000) awards. The awards will be announced during the GECCO closing ceremony.

## Competitions

### **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 Baeck

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 readjust from one to the other – machines are largely free from such computational bottlenecks. Evolutionary multitasking is an emerging concept in computational intelligence that realises 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 venues of this application domain, or papers describing applications of evolutionary methods 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: https://sig.sigevo.org/index.html/tiki-index.php?page=Job+Ads+Listing



**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)	
<b>A Multiobjective Optimization Approach for Market Timing</b> Ismail Mohamed, Fernando E.B. Otero	Friday, July 10, 13:40-15:20
Complex Systems (CS)	
<b>Diversity Preservation in Minimal Criterion Coevolution through Resource</b> <b>Limitation</b> Jonathan C. Brant, Kenneth O. Stanley	Friday, July 10, 13:40-15:20
Evolutionary Combinatorial Optimization and Metaheuristics (ECOM)	
<b>Golden Parameter Search: Exploiting Structure to Quickly Configure</b> <b>Parameters in Parallel</b> 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)	
<b>Improving Constrained Clustering Via Decomposition-based Multiobjective</b> <b>Optimization with Memetic Elitism</b> Germán González-Almagro, Alejandro Rosáles-Pérez, Julián Luengo, José- Ramón Cano, Salvador García	Saturday, July 11, 17:40-19:20
<b>Neuroevolution of Self-Interpretable Agents</b> Yujin Tang, Duong Nguyen, David Ha	Saturday, July 11, 17:40-19:20
<b>Improving Neuroevolutionary Transfer Learning of Deep Recurrent Neural</b> <b>Networks through Network-Aware Adaptation</b> AbdElRahman ElSaid, Joshua Karns, Zimeng Lyu, Daniel Krutz, Alexander Ororbia II, Travis Desell	Saturday, July 11, 17:40-19:20

Evolutionary Multiobjective Optimization (EMO)	
<b>On the elicitation of indirect preferences in Interactive evolutionary multiple objective optimization</b> Michał K. Tomczyk, Miłosz Kadziński	Friday, July 10, 10:30-12:10
<b>Multi-Objective Hyperparameter Tuning and Feature Selection using Filter Ensembles</b> 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)	
<b>Leveraging Conditional Linkage Models in Gray-box Optimization with the</b> <b>Real-Valued Gene-pool Optimal Mixing Evolutionary Algorithm</b> Anton Bouter, Stefanus C. Maree, Tanja Alderliesten, Peter A. N. Bosman	Friday, July 11, 13:40-15:20
Genetic Algorithms (GA)	
<b>Understanding Transforms of Pseudo-Boolean Functions</b> Darrell Whitley, Hernan Aguirre, Andrew Sutton	Saturday, July 11, 17:40-19:20
<b>Adaptively Preserving Solutions in Both Feasible and Infeasible Regions on Generalized Multiple Constraint Ranking</b> Yohanes Bimo Dwianto, Hiroaki Fukumoto, Akira Oyama	Saturday, July 11, 17:40-19:20
General Evolutionary Computation and Hybrids (GECH)	
<b>Effective Reinforcement Learning through Evolutionary Surrogate-Assisted</b> <b>Prescription</b> Olivier Francon, Santiago Gonzalez, Babak Hodjat, Elliot Meyerson, Risto Miikkulainen, Xin Qiu, Hormoz Shahrzad	Saturday, July 11, 15:40-17:20
ε- <b>shotgun:</b> ε- <b>greedy Batch Bayesian Optimisation</b> George De Ath, Richard M. Everson, Jonathan E. Fieldsend, Alma A. M. Rahat	Saturday, July 11, 15:40-17:20
<b>Analysis of the Performance of Algorithm Configurators for Search</b> <b>Heuristics with Global Mutation Operators</b> George T. Hall, Pietro S. Oliveto, Dirk Sudholt	Saturday, July 11, 15:40-17:20
Genetic Programming (GP)	
<b>Genetic programming approaches to learning fair classifiers</b> William La Cava, Jason H. Moore	Saturday, July 11, 15:40-17:20
<b>A Modular Memory Framework for Time Series Prediction</b> Stephen Paul Kelly, Jacob Newsted, Wolfgang Banzhaf, Cedric Gondro	Saturday, July 11, 15:40-17:20
<b>Semantically-Oriented Mutation Operator in Cartesian Genetic Programming for Evolutionary Circuit Design</b> David Hodan, Vojtech Mrazek, Zdenek Vasicek	Saturday, July 11, 15:40-17:20

## **Real World Applications (RWA)**

<b>Towards Sustainable Forest Management Strategies with MOEAs</b> Philipp Back, Antti Suominen, Pekka Malo, Olli Tahvonen, Julian Blank, Kalyanmoy Deb	Saturday, July 11, 17:40-19:20
<b>Optimisation of Large Wave Farms using a Multi-strategy Evolutionary Framework</b> Mehdi Neshat, Bradley Alexander, Eli O. David, Nataliia Y. Sergiienko, Markus Wagner	Saturday, July 11, 17:40-19:20
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	Friday, July 11, 13:40-15:20
Theory (THEORY)	
<b>Fast Mutation in Crossover-based Algorithms</b> Denis Antipov, Maxim Buzdalov, Benjamin Doerr	Friday, July 11, 13:40-15:20



Papers and Posters

EMO1 Friday, July 10	, 10:30-12:10
Chair: Sanaz Mostaghim (University of Magdeburg) (Best Paper nominees are marked with a star)	
Another Difficulty of Inverted Triangular Pareto Fronts for Decomposition-Based Multi-Objective Algorithms 🔶 Linjun He, Auraham Camacho, Hisao Ishibuchi	10:30-10:55
On the elicitation of indirect preferences in interactive evolutionary multiple objective optimization 🔶 Michał K. Tomczyk, Miłosz Kadziński	10:55-11:20
Multi-Objective Hyperparameter Tuning and Feature Selection using Filter Ensembles 🔶 Martin Binder, Julia Moosbauer, Janek Thomas, Bernd Bischl	11:20-11:45
Constraint Handling within MOEA/D Through an Additional Scalarizing Function Saúl Zapotecas-Martínez, Antonin Sebastien Ponsich,	11:45-12:10
HOP1 Friday, July 10	, 10:30-12:10
Chair: Jakob Bossek (The University of Adelaide) (Best Paper nominees are marked with a star)	
<b>Using Exploratory Landscape Analysis to Visualize Single-objective Problems</b> Urban Škvorc, Tome Eftimov, Peter Korošec	10:30-10:55
<b>Model selection for metabolomics: predicting diagnosis of coronary artery disease using automated machine learning</b> Alena Orlenko, Jason H. Moore	10:55-11:20
<b>Sharp Bounds for Genetic Drift in Estimation of Distribution Algorithms</b> Benjamin Doerr, Weijie Zheng	11:20-11:45
<b>Empirical Linkage Learning</b> Michal Witold Przewozniczek, Marcin Michal Komarnicki	11:45-12:10
CS1 Friday, July 10 Chair: Stéphane Doncieux (Sorbonne Universite; CNRS, ISIR)	, 10:30-12:10
<b>Evolution of Distributed Neural Controllers for Voxel-based Soft Robots</b> Eric Medvet, Alberto Bartoli, Andrea De Lorenzo, Giulio Fidel	10:30-10:55
<b>Quality Diversity for Multi-task Optimization</b> Jean-Baptiste Mouret, Glenn Maguire	10:55-11:20
Scaling MAP-Elites to Deep Neuroevolution Cédric Colas, Joost Huizinga, Vashisht Madhavan, Jeff Clune	11:20-11:45
<b>Learning behaviour-performance maps with meta-evolution</b> David M. Bossens, Jean-Baptiste Mouret, Danesh Tarapore	11:45-12:10

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
<b>CMA-ES for One-Class Constraint Synthesis</b> 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

CS2+ACO-SI1Friday, July 10,	13:40-15:20
Chair: Dennis Wilson (ISAE-SUPAERO) (Best Paper nominees are marked with a star)	
<b>A Multiobjective Optimization Approach for Market Timing</b> 📩 Ismail Mohamed, Fernando E.B. Otero	13:40-14:05
<b>Diversity Preservation in Minimal Criterion Coevolution through Resource Limitation</b> 📩 Jonathan C. Brant, Kenneth O. Stanley	14:05-14:30
<b>Evolving ab initio trading strategies in heterogeneous environments</b> David Rushing Dewhurst, Yi Li, Alex Bogdan, Jasmine Geng	14:30-14:55
HOP2Friday, July 10,Chair: Pascal Kerschke (University of Münster) (Best Paper nominees are marked with a star)	13:40-15:20
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
<b>The Univariate Marginal Distribution Algorithm Copes Well With Deception and Epistasis</b> 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
<b>Large Scale Biomedical Data Analysis with Tree-based Automated Machine Learning</b> Trang T. Le, Weixuan Fu, Jason H. Moore	14:55-15:20
<b>GP1</b> Friday, July 10, Chair: Domagoj Jakobovic (Faculty of Electrical Engineering and Computing, Zagreb)	13:40-15:20
<b>Novelty Search for Automatic Bug Repair</b> Omar Villanueva, Leonardo Trujillo, Daniel Hernández	13:40-14:05
<b>A Study on Graph Representations for Genetic Programming</b> 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
<b>Code Building Genetic Programming</b> Edward Pantridge, Lee Spector	14:55-15:20
GA2 Friday, July 10, Chair: Francisco Chicano (University of Malaga)	13:40-15:20
AutoLR: An Evolutionary Approach to Learning Rate Policies Pedro Carvalho, Nuno Lourenço, Filipe Assunção, Penousal Machado	13:40-14:05
<b>New Search Operators for Node-Depth Based Encoding</b> Gustavo Post Sabin, Telma Woerle de Lima, Anderson da Silva Soares	14:05-14:30
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Comparative Mixing for DSMGA-II Marcin Michal Komarnicki, Michal Witold Przewozniczek, Tomasz Maciej Durda	14:30-14:55
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	14:55-15:20
GECH2 Friday, July 10	0, 13:40-15:20
Chair: Mike Preuss (Universiteit Leiden)	
Initial Design Strategies and their Effects on Sequential Model-Based Optimization: An Exploratory Case Study Based on BBOB Jakob Bossek, Carola Doerr, Pascal Kerschke	13:40-14:05
<b>Using Implicit Multi-Objectives Properties to Mitigate Against Forgetfulness in Coevolutionary</b> <b>Algorithms</b> Adefunke Akinola, Mark Wineberg	14:05-14:30
<b>Model-based Optimization with Concept Drifts</b> Jakob Richter, Junjie Shi, Jian-Jia Chen, Jörg Rahnenführer, Michel Lang	14:30-14:55
On the Choice of the Parameter Control Mechanism in the (1 + $(\lambda, \lambda)$ ) Genetic Algorithm	14:55-15:20

Mario Alejandro Hevia Fajardo, Dirk Sudholt

ENUM1 Friday, July 10	, 15:40-17:20
Chair: Carola Doerr (CNRS, Sorbonne University)	
Towards Dynamic Algorithm Selection for Numerical Black-Box Optimization: Investigating BBOB as a Use Case Diederick L. Vermetten, Hao Wang, Thomas Bäck, Carola Doerr	15:40-16:05
<b>Versatile Black-Box Optimization</b> Jialin Liu, Antoine Moreau, Mike Preuss, Jérémy Rapin, Baptiste Roziere, Fabien Teytaud, Olivier Teytaud	16:05-16:30
<b>Distributed Random Walks for Fitness Landscape Analysis</b> Ryan Dieter Lang, Andries Petrus Engelbrecht	16:30-16:55
A Surrogate-Assisted Metaheuristic for Bilevel Optimization Jesus-Adolfo Mejia-de-Dios, Efren Mezura-Montes	16:55-17:20
ECOM1Friday, July 10Chair: Luis Paquete (University of Coimbra) (Best Paper nominees are marked with a star)	, 15:40-17:20
Golden Parameter Search: Exploiting Structure to Quickly Configure Parameters in Parallel 🖈 Yasha Pushak, Holger H. Hoos	15:40-16:05
Automatic Decomposition of Mixed Integer Programs for Lagrangian Relaxation Using a Multiobjective Approach 🔶 Jake Matt Weiner, Andreas Ernst, Xiaodong Li, Yuan Sun	16:05-16:30
<b>Just-in-Time Batch Scheduling Subject to Batch Size</b> Sergey Polyakovskiy, Dhananjay Thiruvady, Rym M'Hallah	16:30-16:55
Advanced Statistical Analysis of Empirical Performance Scaling Yasha Pushak, Holger H. Hoos	16:55-17:20
RWA1     Friday, July 10       Chair: Bogdan Filipic (Jozef Stefan Institute)	, 15:40-17:20
Automated Design of Multi-Level Network Partitioning Heuristics Employing Self-Adaptive Primitive Granularity Control Aaron Scott Pope, Daniel R. Tauritz	15:40-16:05
A Memetic Level-based Learning Swarm Optimizer for Large-scale Water Distribution Network Optimization Ya-Hui Jia, Yi Mei, Mengjie Zhang	16:05-16:30
Adaptive Augmented Evolutionary Intelligence for the Design of Water Distribution Networks Matthew Barrie Johns, Herman Mahmoud, Edward Keedwell, Dragan Savic	16:30-16:55
<b>Evolving Energy Demand Estimation Models over Macroeconomic Indicators</b> Nuno Lourenço, José Manuel Colmenar, J. Ignacio Hidalgo, Sancho Salcedo Sanz	16:55-17:20

EML1 Friday, July 10	, 15:40-17:20
Chair: Una-May O'Reilly (CSAIL, Massachusetts Institute of Technology; ALFA Group)	
Segmented Initialization and Offspring Modification in Evolutionary Algorithms for Bi-objective Feature Selection Hang Xu, Bing Xue, Mengjie Zhang	15:40-16:05
Relatedness Measures to Aid the Transfer of Building Blocks among Multiple Tasks Trung Nguyen, Will Neil Browne, Mengjie Zhang	16:05-16:30
Feature Standardisation and Coefficient Optimisation for Effective Symbolic Regression Grant Dick, Caitlin A. Owen, Peter A. Whigham	16:30-16:55
<b>Re-purposing Heterogeneous Generative Ensembles with Evolutionary Computation</b> Jamal Toutouh, Erik Hemberg, Una-May O'Reilly	16:55-17:20
GP2 Friday, July 10	, 15:40-17:20
Chair: Mengjie Zhang (Victoria University of Wellington)	
Adaptive Weighted Splines - A New Representation to Genetic Programming for Symbolic Regression Christian Raymand, Qi Chen, Bing Xue, Mengjie Zhang	15:40-16:05
<b>Constructing Efficient Multigrid Solvers with Genetic Programming</b> Jonas Schmitt, Sebastian Kuckuk, Harald Köstler	16:05-16:30
<b>Symbolic Regression Driven by Training Data and Prior Knowledge</b> Jiri Kubalik, Erik Derner, Robert Babuska	16:30-16:55
Improving Symbolic Regression based on Correlation Between Residuals and Variables Qi Chen, Bing Xue, Mengjie Zhang	16:55-17:20

SBSE1+ENUM2+THEORY1 Saturday, July 11, 13	
Chair: Justyna Petke (University College London) (Best Paper nominees are marked with a star)	
Leveraging Conditional Linkage Models in Gray-box Optimization with the Real-Va pool Optimal Mixing Evolutionary Algorithm 📩 Anton Bouter, Stefanus C. Maree, Tanja Alderliesten, Peter A. N. Bosman	lued Gene- 13:40-14:05
Seeding Strategies for Multi-Objective Test Case Selection: An Application on Simultesting $\bigstar$	lation-based 14:05-14:30
Aitor Arrieta, Joseba Andoni Agirre, Goiuria Sagardui	
Fast Mutation in Crossover-based Algorithms 📩 Denis Antipov, Maxim Buzdalov, Benjamin Doerr	14:30-14:55
<b>Causes and Effects of Fitness Landscapes in Unit Test Generation</b> Nasser Albunian, Gordon Fraser, Dirk Sudholt	14:55-15:20
ACO-SI2 S	aturday, July 11, 13:40-15:20
Chair: Jürgen Branke (University of Warwick)	
ACO with automatic parameter selection for a scheduling problem with a job group constraint Lucas Groleaz, Samba-Ndoib Ndiave, Christine Solnon	cumulative 13:40-14:05
An Approach to Assess Swarm Intelligence Algorithms Based on Complex Networks Clodomir Joaquim Santana, Edward Keedwell, Ronaldo Menezes	14:05-14:30
<b>Exploratory Path Planning for Mobile Robots in Dynamic Environments with Ant C</b> <b>Optimization</b> Valéria de Carvalho Santos, Fernando E. B. Otero, Colin Johnson, Fernando Santos Toledo	Colony 14:30-14:55 Osório, Cláudio
A Weighted Population Update Rule for PACO Applied to the Single Machine Total V Tardiness Problem Daniel Abitz, Tom Hartmann, Martin Middendorf	Veighted 14:55-15:20
RWA2 S Chair: Martin Zaefferer (TH Köln)	aturday, July 11, 13:40-15:20
<b>One property to rule them all? On the limits of trade-offs for S-boxes</b> Marko Djurasevic, Domagoj Jakobovic, Stjepan Picek	13:40-14:05
<b>COUGAR: Clustering Of Unknown malware using Genetic Algorithm Routines</b> Zachary Wilkins, Nur Zincir-Heywood	14:05-14:30
Variable Reduction for Surrogate-Based Optimization Frederik Rehbach, Lorenzo Gentile, Thomas Bartz-Beielstein	14:30-14:55
Modeling Wildfires Using Evolutionary Cellular Automata Maxfield E. Green, Todd F. DeLuca, Karl Kaiser	14:55-15:20

<b>EML2</b> Saturday, July 11, 13: Chair: Krzysztof Krawiec (Poznan University of Technology, Center for Artificial Intelligence and Mach Learning)	
Safe Crossover of Neural Networks Through Neuron Alignment Thomas Uriot, Dario Izzo	13:40-14:05
<b>Exploring the Evolution of GANs through Quality Diversity</b> Victor Costa, Nuno Lourenço, João Correia, Penousal Machado	14:05-14:30
XCS Classifier System with Experience Replay Anthony Stein, Roland Maier, Lukas Rosenbauer, Jörg Hähner	14:30-14:55
<b>Evolving Inborn Knowledge For Fast Adaptation in Dynamic POMDP Problems</b> Eseoghene Ben-Iwhiwhu, Pawel Ladosz, Jeffery Dick, Wen-Hua Chen, Praveen Pilly, Andrea Soltoggio	14:55-15:20
GP3 Saturday, July	11, 13:40-15:20
Chair: Lee Spector (Amherst College, Hampshire College)	
<b>Unlabeled Multi-Target Regression with Genetic Programming</b> Uriel López, Leonardo Trujillo, Sara Silva, Pierrick Legrand, Leonardo Vanneschi	13:40-14:05
Synthesis through Unification Genetic Programming Thomas Welsch, Vitaliy Kurlin	14:05-14:30
<b>Towards an evolutionary-based approach for natural language processing</b> Luca Manzoni, Domagoj Jakobovic, Luca Mariot, Stjepan Picek, Mauro Castelli	14:30-14:55
ECOM2 Saturday, July 2	11, 13:40-15:20
Chair: Manuel López-Ibáñez (University of Manchester)	
<b>Do Sophisticated Evolutionary Algorithms Perform Better than Simple Ones?</b> Michael Foster, Matthew Hughes, George O'Brien, Pietro S. Oliveto, James Pyle, Dirk Sudholt, James Williams	13:40-14:05
<b>A Robust Experimental Evaluation of Automated Multi-Label Classification Methods</b> Alex G. C. de Sá, Cristiano G. Pimenta, Gisele L. Pappa, Alex A. Freitas	14:05-14:30
Solving Constrained Combinatorial Reverse Auctions Using MOEAs: A Comparative Study. Elaine Guerrero-Peña, Fernanda Nakano Kazama, Paulo de Barros Correia, Aluizio F.R. Araújo	14:30-14:55
<b>Journey to the Center of the Linear Ordering Problem</b> Leticia Hernando, Alexander Mendiburu, Jose A. Lozano	14:55-15:20

<b>GECH3</b> Saturday, July 11,Chair: Gustavo Olague (CICESE) (Best Paper nominees are marked with a star)	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	15:40-16:05
ε- <b>shotgun:</b> ε- <b>greedy Batch Bayesian Optimisation </b> George De Ath, Richard M. Everson, Jonathan E. Fieldsend, Alma A. M. Rahat	16:05-16:30
<b>Effective Reinforcement Learning through Evolutionary Surrogate-Assisted Prescription</b> $\bigstar$ Olivier Francon, Santiago Gonzalez, Babak Hodjat, Elliot Meyerson, Risto Miikkulainen, Xin Qiu, Hormoz Shahrzad	16:30-16:55
<b>An Evolutionary Optimization Algorithm for Gradually Saturating Objective Functions</b> Dolly Sapra, Andy D. Pimentel	16:55-17:20
GP4Saturday, July 11,Chair: Miguel Nicolau (University College Dublin) (Best Paper nominees are marked with a star)	15:40-17:20
<b>Genetic programming approaches to learning fair classifiers</b> 🗡 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
<b>RWA3</b> Saturday, July 11, Chair: Shahryar Rahnamayan (University of Ontario Institute of Technology (UOIT))	15:40-17:20
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	16:30-16:55
Surrogate-Assisted Asynchronous Multiobjective Algorithm for Nuclear Power Plant Operations Valentin Drouet, Sébastien Verel, Jean-Michel Do	16:55-17:20

EML3 Saturday, July 11	, 15:40-17:20
Chair: Ryan Urbanowicz (University of Pennsylvania)	
<b>Absumption and Subsumption based Learning Classifier Systems</b> Yi Liu, Will Neil Browne, Bing Xue	15:40-16:05
Neural Architecture Search for Sparse DenseNets with Dynamic Compression Damien O'Neill, Bing Xue, Mengjie Zhang	16:05-16:30
Lateralized Learning for Robustness Against Adversarial Attacks in a Visual Classification System Abubakar Siddique, Will N. Browne, Gina M. Grimshaw	16:30-16:55
Improving generalisation of AutoML systems with dynamic fitness evaluations Benjamin Patrick Evans, Bing Xue, Mengjie Zhang	16:55-17:20
DETA Saturday, July 11	, 15:40-17:20
Chair: Francisco Fernández de Vega (Universidad de Extremadura)	
Interactive Evolution and Exploration Within Latent Level-Design Space of Generative Adversarial Networks	15:40-16:05
Jacob Schrum, Jake Gutierrez, Vanessa Volz, Jialin Liu, Simon Lucas, Sebastian Risi	
CPPN2GAN: Combining Compositional Pattern Producing Networks and GANs for Large-scale Pattern Generation Jacob Schrum, Vanessa Volz, Sebastian Risi	16:05-16:30
EMO2 Saturday, July 11	, 15:40-17:20
Chair: Bogdan Filipic (Jozef Stefan Institute)	
<b>Transfer Learning for Gaussian Process Assisted Evolutionary Bi-objective Optimization for Objectives with Different Evaluation Times</b> Xilu Wang, Yaochu Jin, Sebastian Schmitt, Markus Olhofer	15:40-16:05
Designing parallelism in Surrogate-assisted multiobjective optimization based on decomposition Nicolas Berveglieri, Bilel Derbel, Arnaud Liefooghe, Hernan Aguirre, Qingfu Zhang, Kiyoshi Tanaka	16:05-16:30
<b>Computation of the improvement directions of the Pareto front and its application to MOEAs</b> Salvador Botello-Aceves, Arturo Hernandez-Aguirre, S. Ivvan Valdez	16:30-16:55
Data Structures for Non-Dominated Sets: Implementations and Empirical Assessment of Two Decades of Advances Jonathan E. Fieldsend	16:55-17:20
ECOM3 Saturday, July 11	, 15:40-17:20
Chair: Christine Zarges (Aberystwyth University)	
Multi-layer local ontima networks for the analysis of advanced local search-based algorithms	15:40-16:05

 Iulti-layer local optima networks for the analysis of advanced local search-based algorithms
 15:4

 Marcella Scoczynski Ribeiro Martins, Mohamed El Yafrani, Myriam Regattieri De Biase Da Silva
 Delgado, Ricardo Lüders

Solving the Single Row Facility Layout Problem by Differential Evolution Pavel Krömer, Jan Platos, Vaclav Snasel	16:05-16:30
A Deep Learning Approach to Predicting Solutions in Streaming Optimisation Domains Mohamad Alissa, Kevin Sim, Emma Hart	16:30-16:55
<b>Why Many Traveling Salesman Problem Instances Are Easier Than You Think</b> Swetha Varadarajan, Darrell Whitley, Gabriela Ochoa	16:55-17:20

EML4Saturday, July 11,Chair: Bing Xue (Victoria University of Wellington) (Best Paper nominees are marked with a star)5	, 17:40-19:20
<b>Neuroevolution of Self-Interpretable Agents </b> 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 ☆ Germán González-Almagro, Alejandro Rosáles-Pérez, Julián Luengo, José-Ramón Cano, Salvador García	18:30-18:55
GeneCAI: Genetic Evolution for Acquiring Compact AI Mojan Javaheripi, Mohammad Samragh, Tara Javidi, Farinaz Koushanfar	18:55-19:20
RWA4Saturday, July 11Chair: Sushil Louis (UNR; University of Nevada, Reno) (Best Paper nominees are marked with a star)	, 17:40-19:20
<b>Towards Sustainable Forest Management Strategies with MOEAs </b> Philipp Back, Antti Suominen, Pekka Malo, Olli Tahvonen, Julian Blank, Kalyanmoy Deb	17:40-18:05
<b>Optimisation of Large Wave Farms using a Multi-strategy Evolutionary Framework </b> Mehdi Neshat, Bradley Alexander, Nataliia Y. Sergiienko, Markus Wagner	18:05-18:30
<b>Evolutionary Bin Packing for Memory-Efficient Dataflow Inference Acceleration on FPGA</b> Mairin Kroes, Lucian Petrica, Sorin Cotofana, Michaela Blott	18:30-18:55
Simultaneously Searching and Solving Multiple Avoidable Collisions for Testing Autonomous Driving Systems Alessandro Calò, Paolo Arcaini, Shaukat Ali, Florian Hauer, Fuyuki Ishikawa	18:55-19:20
GA3       Saturday, July 11,         Chair: Carlos Segura (Centro de Investigación en Matemáticas (CIMAT))       (Best Paper nominees are marked with a star)	, 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	17:40-18:05
<b>Understanding Transforms of Pseudo-Boolean Functions</b> $\bigstar$ Darrell Whitley, Hernan Aguirre, Andrew Sutton	18:05-18:30
<b>Evolving Diverse Sets of Tours for the Travelling Salesperson Problem</b> Anh Viet Do, Jakob Bossek, Aneta Neumann, Frank Neumann	18:30-18:55
<b>Modelling Parameter Configuration Spaces with Local Optima Networks</b> Germán Treimun-Costa, Elizabeth Montero, Gabriela Ochoa, Nicolás Rojas-Morales	18:55-19:20

CS3 Saturday, July 1	1, 17:40-19:20
Chair: Risto Milkkulainen (The University of Texas at Austin)	
<b>Discovering Representations for Black-box Optimization</b> Adam Gaier, Alexander Asteroth, Jean-Baptiste Mouret	17:40-18:05
<b>Covariance Matrix Adaptation for the Rapid Illumination of Behavior Space</b> Matthew C. Fontaine, Julian Togelius, Stefanos Nikolaidis, Amy K. Hoover	18:05-18:30
<b>Towards Crossing the Reality Gap with Evolved Plastic Neurocontrollers</b> 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	18:55-19:20
THEORY2 Saturday, July 1	1, 17:40-19:20
Chair: Denis Antipov (ITMO University, Ecole Polytechnique)	
The Node Weight Dependent Traveling Salesperson Problem: Approximation Algorithms and Randomized Search Heuristics Jakob Bossek, Katrin Casel, Pascal Kerschke, Frank Neumann	17:40-18:05
The $(1 + (\lambda, \lambda))$ 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
<b>Fixed-Target Runtime Analysis</b> Maxim Buzdalov, Benjamin Doerr, Carola Doerr, Dmitry Vinokurov	18:55-19:20
EMO3 Saturday, July 1	1, 17:40-19:20
Chair: Hiroyuki Sato (The University of Electro-Communications)	
Multiobjective Tree-structured Parzen Estimator for Computationally Expensive Optimization Problems Yoshihiko Ozaki, Yuki Tanigaki, Shuhei Watanabe, Masaki Onishi	17:40-18:05
Surrogate Assisted Evolutionary Algorithm for Medium Scale Multi-Objective Optimisation Problems Vigoran Ruan, Ke Li, Bilel Derbel, Arnaud Liefooghe	18:05-18:30
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	18:55-19:20

Vahid Roostapour, Jakob Bossek, Frank Neumann

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ECOM4	Saturday, July 11, 17:40-19:20
Chair: Francisco Chicano (University of Malaga)	· ·
<b>Dynamic Bi-Objective Routing of Multiple Vehicles</b> Jakob Bossek, Christian Grimme, Heike Trautmann	17:40-18:05
Specific Single- and Multi-Objective Evolutionary Algorithms for the Chance-Cor Knapsack Problem Yue Xie, Aneta Neumann, Frank Neumann	nstrained 18:05-18:30
SBSE2	Saturday, July 11, 17:40-19:20
Chair: Fuyuki Ishikawa (National Institute of Informatics)	
Scalability analysis of Grammatical Evolution Based Test Data Generation Muhammad Sheraz Anjum, Conor Ryan	17:40-18:05
<b>Enhancing Search-Based Product Line Design with Crossover Operators</b> Diego Fernandes da Silva, Luiz Fernando Okada, Thelma Elita Colanzi, Wesley K.	18:05-18:30 G. Assunção
<b>Genetic Algorithms for Redundancy in Interaction Testing</b> Ryan Edward Dougherty	18:30-18:55
<b>Towards Rigorous Validation of Energy Optimisation Experiments</b> Mahmoud Bokhari, Markus Wagner, Bradley Alexander	18:55-19:20

ENUM3 Su Chair: Rvoji Tanabe (Yokohama National University)	nday, July 12, (	09:00-10:40
Analyzing Adaptive Parameter Landscapes in Parameter Adaptation Methods for Diffe Evolution Ryoji Tanabe	erential	09:00-09:25
<b>Deep Generative Model for Non-convex Constraint Handling</b> Naoki Sakamoto, Eiji Semmatsu, Kazuto Fukuchi, Jun Sakuma, Youhei Akimoto		09:25-09:50
RWA5     Su       Chair: Jürgen Branke (University of Warwick)	nday, July 12, (	09:00-10:40
Multi-Objective Optimal Distribution of Materials in Hybrid Components Thomas Gossuin, Didier Garray, Vincent Kelner		09:00-09:25
Hybrid Genetic Algorithm for Ridesharing with Timing Constraints: Efficiency analys World Data Nirav Patel, N. S. Narayanaswamy, Alok Joshi	is with Real-	09:25-09:50
EML5 Su Chair: Jaume Bacardit (Newcastle University)	nday, July 12, (	09:00-10:40
<b>Evolutionary Optimization of Deep Learning Activation Functions</b> Garrett Bingham, William Macke, Risto Miikkulainen		09:00-09:25
Multi-Fitness Learning for Behavior-Driven Cooperation Connor Yates, Reid Christopher, Kagan Tumer,		09:25-09:50
<b>Program Synthesis as Latent Continuous Optimization: Evolutionary Search in Neura</b> <b>Embeddings</b> Pawel Liskowski, Krzysztof Krawiec, Nihat Engin Toklu, Jerry Swan	վ	09:50-10:15
<b>Self-adaptation of XCS learning parameters based on Learning theory</b> Motoki Horiuchi, Masaya Nakata		10:15-10:40
THEORY3 Su	unday, July 12, (	09:00-10:40
Chair: Jakob Bossek (The University of Adelaide)		
A Tight Lower Bound on the Expected Runtime of Standard Steady State Genetic Algor Carsten Witt, Pietro S. Oliveto, Dirk Sudholt	rithms	09:00-09:25
<b>Does Comma Selection Help To Cope With Local Optima?</b> Benjamin Doerr		09:25-09:50
Self-Adjusting Evolutionary Algorithms for Multimodal Optimization Amirhossein Rajabi, Carsten Witt		09:50-10:15

EMO4	Sunday, July 12, 09:00-10:40	
Chair: Jonathan E. Fieldsend (University of Exeter)		
Gap Finding and Validation in Evolutionary Multi- and Many-Objective Optimizati Pablo Valledor Pellicer, Miguel Iglesias Escudero, Silvino Fernández Alzueta, Kalyar	i <b>on</b> 1moy Deb	09:00-09:25
If Unsure, Shuffle: Deductive Sort is $\Theta(MN^3)$ , but $O(MN^2)$ in Expectation over Inpermutations Sumit Mishra, Maxim Buzdalov	ut	09:25-09:50
<b>What is a Good Direction Vector Set for The R2-based Hypervolume Contribution</b> <i>A</i> Yang Nan, Ke Shang, Hisao Ishibuchi	Approximation	09:50-10:15
Surrogate-assisted Multi-objective Combinatorial Optimization based on Decomp Walsh Basis Geoffrey Pruvost, Bilel Derbel, Arnaud Liefooghe, Sébastien Verel, Qingfu Zhang	osition and	10:15-10:40
GECH4	Sunday, July 12,	09:00-10:40
Chair: Fabien Teytaud (University of Lille Nord de France)		
Sensitivity Analysis in Constrained Evolutionary Optimization Julian Schulte, Volker Nissen		09:00-09:25
<b>Expected Improvement versus Predicted Value in Surrogate-Based Optimization</b> Frederik Rehbach, Martin Zaefferer, Boris Naujoks, Thomas Bartz-Beielstein		09:25-09:50
Algorithm Selection of Anytime Algorithms Alexandre D. Jesus, Arnaud Liefooghe, Bilel Derbel, Luís Paquete		09:50-10:15
From Understanding Genetic Drift to a Smart-Restart Parameter-less Compact Generation Doerr, Weijie Zheng	netic Algorithm	10:15-10:40

### Poster Session 1

Friday, July 10, 17:40-19:20

## Poster Group 1

### 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

## Poster Group 2

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,

### Poster Group 3

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 Pätzel, 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

### Poster Group 5

# 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

### Poster Group 8

- **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

### Poster Group 9

- 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 Garcia, 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 Moataz Chouchen, Ali Ouni, Mohamed Wiem Mkaouer, Raula Gaikovina Kula, Katsuro Inoue

### Poster Group 10

### **Crash Reproduction Using Helper Objectives**

Pouria Derakhshanfar, Xavier Devroey, Andy Zaidman, Arie van Deursen, Annibale Panichella

## **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** Vida Vukasinovic, David Kuboň, Riste Škrekovski, Petr Gregor

### Poster Session 2

Saturday, July 11, 10:30-12:10

## Poster Group 1

- **Evolutionary Curriculum Learning Approach for Transferable Cellular Automata Rule Optimization** 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

**Ådea — 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 Paretooptimal 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, Márcia 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 Duflo, 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

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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
  Seved Jalaleddin Mousavirad, Gerald Schaefer, Jakov Korovin
Evolved Ensemble of Detectors for Gross Error Detection
  Thanh Tien Nguyen, John McCall, Allan Wilson, Laud 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
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A New Approach to Distribute MOEA Pareto Front Computation
Federica Sarro, Alessio Petrozziello, Dan-Qi He, Shin Yoo
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**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

## 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

Lucas Groleaz, *LIRIS*, Samba-Ndojh Ndiaye, *LIRIS*, Christine Solnon, *CITI* 

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 exceeds 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 parameter 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

Clodomir Joaquim Santana, *University of Exeter*, Edward Keedwell, *University of Exeter*, Ronaldo Menezes, *University of Exeter* 

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

Valéria de Carvalho Santos, Universidade Federal de Ouro Preto, Fernando Esteban Barril Otero, University of Kent, Colin Johnson, University of Kent, Fernando Santos Osório, University of São Paulo, Cláudio Toledo, University of São Paulo

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. The proposed algorithm was evaluated in static and dynamic environments, showing promising results in both of them. Experiments in dynamic environments shows the adaptability of our proposal.

#### A Weighted Population Update Rule for PACO Applied to the Single Machine Total Weighted Tardiness Problem

Daniel Abitz, *Leipzig University*, Tom Hartmann, *Leipzig University*, Martin Middendorf, *Leipzig University* 

In this paper a new population update rule for population based ant colony optimization (PACO) is proposed. PACO is a well known alternative to the standard ant colony optimization 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)

#### CS1

Friday, July 10, 10:30-12:10

#### 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 sparkled 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 gaits. 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 stateof-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 behaviour 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.

#### CS2+ACO-SI1

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

## Diversity Preservation in Minimal Criterion Coevolution through Resource Limitation $\bigstar$

## 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, Mass-Mutual 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.

#### CS3

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

#### **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 highperforming 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,

### Digital Entertainment Technologies and Arts

DETA

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

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 Laflaquière, Softbank Robotics Europe, Alexandre Coninx, Sorbonne Universite; 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 straight-forward 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.

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

Jacob Schrum, Southwestern University, Jake Gutierrez, South-

western University, Vanessa Volz, modl.ai, Jialin Liu, Southern University of Science and Technology, Simon Lucas, Queen Mary University of London, 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.

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

Jacob Schrum, Southwestern University, Vanessa Volz, modl.ai, Sebastian Risi, IT University of Copenhagen

Generative Adversarial Networks (GANs) are proving to be a powerful indirect genotype-to-phenotype mapping for evolutionary search, but they have limitations. In particular, GAN output does not scale to arbitrary dimensions, and there is no obvious way of combining multiple GAN outputs into a cohesive whole, which would be useful in many areas, such as the generation of video game levels. Game levels often consist of several segments, sometimes repeated directly or with variation, organized into an engaging pattern. Such patterns can be produced with Compositional Pattern Producing Networks (CPPNs). Specifically, a CPPN can define latent vector GAN inputs as a function of geometry, which provides a way to organize level segments output by a GAN into a complete level. This new CPPN2GAN approach is validated in both Super Mario Bros. and The Legend of Zelda. Specifically, divergent search via MAP-Elites demonstrates that CPPN2GAN can better cover the space of possible levels. The layouts of the resulting levels are also more cohesive and aesthetically consistent.

### Evolutionary Combinatorial Optimization and Metaheuristics

#### ECOM1

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

## Golden Parameter Search: Exploiting Structure to Quickly Configure Parameters in Parallel $\bigstar$

#### 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.

### 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 mixedinteger 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).

#### ECOM2

Saturday, July 11, 13:40-15:20

## 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, Universidade Federal de Pernambuco, Fernanda Nakano Kazama, University of Campinas, Paulo de Barros Correia, University of Campinas, Aluizio F.R. Araújo, 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 multiobjective 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.

#### ECOM3

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

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

Marcella Scoczynski Ribeiro Martins, *The Federal University* of *Technology - Paraná*, Mohamed El Yafrani, *Aalborg Univer*sity, 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 result to suggest new strategies for generating high quality solutions for Travelling Salesman Problems.

#### ECOM4

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

#### **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 singlevehicle 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 a. 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 problemspecific 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 knowledgetransfer 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 heuristicbased methods. When comparing both evolutionary, the one with only an upper size bound on the ensemble size is the best.

#### EML2

Saturday, July 11, 13:40-15:20

## 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 highquality 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-Iwhiwhu, 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 metareinforcement 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.

## 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-discernible 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 overspecific rules. It is hypothesized that utilizing just assumption & 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. ACSC'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

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 skipconnection 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.

## 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 miss-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.

Improving generalisation of AutoML systems with dynamic
#### 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.

#### EML4

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

#### 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 selfattention 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/

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

AbdElRahman ElSaid, Rochester Institute of Technology, Joshua Karns, Rochester Institute of Technology, Zimeng Lyu, Rochester Institute of Technology, Travis Desell, Rochester Institute of Technology, Daniel Krutz, 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 e"orts 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.

### Improving Constrained Clustering Via Decomposition-based Multiobjective Optimization with Memetic Elitism 📩

Germán González-Almagro, University of Granada, Alejandro Rosáles-Pérez, Tecnológico de Monterrey, Julián Luengo, University of Granada, José-Ramón Cano, University of Jaén; Dept. of Computer Science, Salvador García, University of Granada

Clustering has always been a topic of interest in knowledge discovery, it is able to provide us with valuable information within the unsupervised machine learning framework. It received renewed attention when it was shown to produce better results in environments where partial information about how to solve the problem is available, thus leading to a new machine learning paradigm: semi-supervised machine learning. This new type of information can be given in the form of constraints, which guide the clustering process towards quality solutions. In particular, this study considers the pairwise instance-level must-link and cannot-link constraints. Given the ill-posed nature of the constrained clustering problem, we approach it from the multiobjective optimization point of view. Our proposal consists in a memetic elitist evolutionary strategy that favors exploitation by applying a local search procedure to the elite of the population and transferring its results only to the external population, which will also be used to generate new individuals. We show the capability of this method to produce quality results for the constrained clustering problem when

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 computeintensive 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.

#### EML5

Sunday, July 12, 09:00-10:40

#### 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, Kagan 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.

## **Evolutionary Multiobjective Optimization**

## EMO1

Friday, July 10, 10:30-12:10

# 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 Braininspired Intelligent Computation

A set of uniformly sampled weight vectors from a unit simplex has been frequently used in decomposition-based multiobjective 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, welldistributed 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, bestof-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 multiobjective 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

given data.

# Constraint Handling within MOEA/D Through an Additional Scalarizing Function

Saúl Zapotecas-Martínez, Universidad Autónoma Metropolitana Unidad Cuajimalpa, Antonin 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 multiobjective 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-ofthe-art MOEAs adopted in our comparative study.

#### EMO2

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

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

Xilu Wang, University of Surrey, Yaochu Jin, University of Surrey, Sebastian Schmitt, 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-SAEA) 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 Berveglieri, University of Lille, Bilel Derbel, University of Lille, Arnaud Liefooghe, 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 Geospacial* 

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.

#### EMO3

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

### Multiobjective Tree-structured Parzen Estimator for Computationally Expensive Optimization Problems

Yoshihiko Ozaki, AI Research Center, Yuki Tanigaki, AI Research Center, Shuhei Watanabe, AI Research Center, Masaki Onishi, AI Research Center

Practitioners often encounter computationally expensive multiobjective optimization problems to be solved in a variety of realworld applications. On the purpose of challenging these problems, we propose a new surrogate-based multiobjective optimization algorithm that does not require a large evaluation budget. It is called Multiobjective Tree-structured Parzen Estimator (MOTPE) and is an extension of the tree-structured Parzen estimator widely used to solve expensive single-objective optimization problems. Our empirical evidences reveal that MOTPE can approximate Pareto fronts of many benchmark problems better than existing methods with a limited budget. In this paper, we discuss furthermore the influence of MOTPE configurations to understand its behavior.

# 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 mediumscale 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.

#### EMO4

Sunday, July 12, 09:00-10:40

# Gap Finding and Validation in Evolutionary Multi- and Many-Objective Optimization

Pablo Valledor Pellicer, *ArcelorMittal*, Miguel Iglesias Escudero, *ArcelorMittal*, Silvino Fernández Alzueta, *ArcelorMittal*, Kalyanmoy Deb, *Michigan State University* 

Over 30 years, evolutionary multi- and many-objective optimization (EMO/EMaO) algorithms have been extensively applied to find well-distributed Pareto-optimal (PO) solutions in a single run. However, in real-world problems, the PO front may not always be one continuous hyper-surface, rather several irregularities may exist involving disjointed surfaces, holes within the surface or mixed-dimensional surfaces. When a set of trade-off solutions are obtained by EMO/EMaO algorithms, there may exist less dense or no solutions (we refer as 'gaps') in different parts of the front. This can happen for at least two reasons: (i) gaps do exist in the PO front, or (ii) no gaps exist, but EMO/EMaO is not able to find any solution there. To make a confident judgement, we propose a three-step procedure. First, we suggest a computational procedure to identify gaps in a PO front. Second, we propose a method to identify well-distributed gap-points. Third, we apply a focused EMO/EMaO algorithm to find representative trade-off points in the gaps. The location of the final points reveals the true reason for the existence of gaps on the original fronts. Procedures are supported by results on two to eight-objective test problems and on a five-objective scheduling problem from a steelmaking industry.

# 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 wellunderstood, 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 Braininspired 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 vectors 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, Bilel 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 algo-

## **Evolutionary Numerical Optimization**

## ENUM1

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

#### 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 rithms 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.

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.

#### SBSE1+ENUM2+THEORY1

#### Saturday, July 11, 13:40-15:20

### Leveraging Conditional Linkage Models in Gray-box Optimization with the Real-Valued Gene-pool Optimal Mixing Evolutionary Algorithm 📩

Anton Bouter, Centrum Wiskunde & Informatica (CWI), Stefanus C. Maree, Amsterdam UMC, Tanja Alderliesten, Leiden University Medical Center, Peter A. N. Bosman, Centrum Wiskunde & Informatica (CWI)

Often, real-world problems are of the gray-box type. It has been shown that the Real-Valued Gene-pool Optimal Mixing Evolutionary Algorithm (RV-GOMEA) is in principle capable of exploiting such a Gray-Box Optimization (GBO) setting using linkage models that capture dependencies between problem variables, resulting in excellent performance and scalability on both benchmark and real-world problems that allow for partial evaluations. However, linkage models proposed for RV-GOMEA so far cannot explicitly capture overlapping dependencies. Consequently, performance degrades if such dependencies exist. 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 VkD-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.

#### ENUM3

Sunday, July 12, 09:00-10:40

#### 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 adaptive parameter landscape. Second, we propose a method of analyzing adaptive parameter landscapes using a 1-steplookahead 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

## GECH1

Friday, July 10, 10:30-12:10

### 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 machinetrained 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-target 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-ofthe-art method.

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

## Benjamin Doerr, École Polytechniqu, 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 algorithms (EDAs) are an alternative, maintaining a probabilistic model of the solution space instead of an explicit population. Such a model is able to implicitly represent a solution set that is far larger than any realistic population size. To support the study of how optimization algorithms handle large sets of optima, we propose the test function EqualBlocksOneMax (EBOM). It has an easy to optimize fitness landscape, however, with an exponential number of optima. We show that the bivariate EDA mutual-information-maximizing input clustering (MIMIC), without any problem-specific modification, quickly generates a model that behaves very similarly to a theoretically ideal model for that function, which samples each of the exponentially many optima with the same maximal probability.

### GECH2

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

### Initial Design Strategies and their Effects on Sequential Model-Based Optimization: An Exploratory Case Study Based on BBOB

Jakob Bossek, The University of Adelaide, Carola Doerr, Sorbonne Université, Pascal Kerschke, University of Münster

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 had 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-Queen 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-Queen 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 $(1+(\lambda,\lambda))$ Genetic Algorithm

Mario Alejandro Hevia Fajardo, *The University of Sheffield*, Dirk Sudholt, *The University of Sheffield* 

The self-adjusting  $(1+(\lambda,\lambda))$ 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  $\lambda$  that governs the mutation strength and the number of offspring. However, on multimodal problems, the parameter control mechanism tends to increase  $\lambda$  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  $\lambda = n$  is reached. This is ineffective for problems where large jumps are required. Capping  $\lambda$  at smaller

values is beneficial for such problems. Finally, resetting  $\lambda$  to 1 allows the parameter to cycle through the parameter space. We show that this strategy is effective for all Jump problems: the  $(1+(\lambda,\lambda))$ 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  $(1+(\lambda,\lambda))$ GA that allows to translate existing runtime bounds from the (1+1)EA to the self-adjusting  $(1+(\lambda,\lambda))$ GA. Our methods are easy to use and give upper bounds for novel classes of functions.

#### GECH3

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

# Analysis of the Performance of Algorithm Configurators for Search Heuristics with Global Mutation 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  $\chi/n$  and the best value for  $\chi$  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.

#### ε-shotgun: ε-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  $\varepsilon$ -greedy procedure for Bayesian optimisation in batch settings in which the black-box function can be evaluated multiple times in parallel. Our  $\varepsilon$ -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  $\epsilon$  – 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  $\epsilon$ -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 Francon, 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 population diversity, 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.

#### GECH4

Sunday, July 12, 09:00-10:40

#### 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 Polytechniqu, 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

## GA1

Friday, July 10, 10:30-12:10

# 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 wellknown 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.

### On measuring and improving the quality of linkage learning in modern evolutionary algorithms applied to solve partially additively separable problems

Michal Witold Przewozniczek, Wroclaw University of Science and Technology, Bartosz Frej, Wroclaw University of Science and Technology, Marcin Michal Komarnicki, Wroclaw University of Science and Technology

Linkage learning is frequently employed in modern evolutionary algorithms. High linkage quality may be the key to an evolutionary method's effectiveness. Similarly, the faulty linkage may be the reason for its poor performance. Many state-of-theart evolutionary methods use a Dependency Structure Matrix (DSM) to obtain linkage. In this paper, we propose a quality measure for DSM. Based on this measure, we analyze the behavior of modern evolutionary methods. We show the dependency between the linkage and the effectiveness of the considered methods. Finally, we propose a framework that improves the quality of the linkage.

# 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 homogenous 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.

## Genetic Algorithm for the Weight Maximization Problem on Weighted Automata

Elena Gutiérrez, IMDEA Software Institute, Takamasa Okudono, National Institute of Informatics, Masaki Waga, National Institute of Informatics, Ichiro Hasuo, National Institute of Informatics

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.

#### GA2

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

## AutoLR: An Evolutionary Approach to Learning Rate Policies

Pedro Carvalho, University of Coimbra, Nuno Lourenço, University of Coimbra, Filipe Assunção, University of Coimbra, Penousal Machado, CISUC

The choice of a proper learning rate is paramount for good Artificial Neural Network training and performance. In the past, one had to rely on experience and trial-and-error to find an adequate learning rate. Presently, a plethora of state of the art automatic methods exist that make the search for a good learning rate easier. While these techniques are effective and have yielded good results over the years, they are general solutions. This means the optimization of learning rate for specific network topologies remains largely unexplored. This work presents AutoLR, a framework that evolves Learning Rate Schedulers for a specific Neural Network Architecture using Structured Grammatical Evolution. The system was used to evolve learning rate policies that were compared with a commonly used baseline value for learning rate. Results show that training performed using certain evolved policies is more efficient than the established baseline and suggest that this approach is a viable means of improving a neural network's performance.

### 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.

#### **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.

### A Biased Random Key Genetic Algorithm applied to the VRPTW with Skill Requirements and Synchronization Constraints

Alberto Francisco Kummer, Federal University of Rio Grande do

Sul, Luciana Salete Buriol, Federal University of Rio Grande do Sul, Olinto César Bassi de Araújo, Federal University of Santa Maria

We applied a Biased Random Key Genetic Algorithm (BRKGA) to solve the Vehicle Routing Problem with Time Windows and Synchronization Constraints. Additionally, both vehicles and clients are skilled, and each client can require up to two distinct skills to be serviced. On double-skilled clients, the operations of each skill should be performed by different vehicles, either simultaneously or respecting a precedence order. Those requirements introduce nonlinearities on the problem, in the sense that a small change on a single route potentially impacts all the other routes of the solution, making it hard to define an effective local search procedure. To circumvent this difficulty, we approached the problem using a genetic algorithm that evolves the sequence in which the services are inserted into the routes. We assessed the performance of our solution method using instances from the literature of the home health care problem. The genetic algorithm outperformed the previous best-known solutions found by a fix-and-optimize matheuristic by up to 25%, using less than half of computational times reported previously. The BRKGA demonstrated to be able to perform well both in exploration and exploitation in the solution space of the problem.

#### GA3

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

### 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 singleobjective 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

Anh Viet Do, *The University of Adelaide*, Jakob Bossek, *The University of Adelaide*, Aneta Neumann, *The University of Adelaide*, Frank Neumann, *The University of Adelaide* 

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

Germán Treimun-Costa, Universidad Técnica Federico Santa María, Elizabeth Montero, Universidad Andres Bello, Gabriela Ochoa, Stirling University, Nicolás Rojas-Morales, Universidad Técnica Federico Santa María

Most algorithms proposed for solving complex problems require the definition of some parameter values. The process of finding suitable parameter values is a 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 optima and edges represent transitions between them. In this work, we adapt the Local Optima Network model to analyze and visualize the global structure of parameter configuration 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**

## GP1

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

#### Novelty Search for Automatic Bug Repair

Omar Villanueva, Tecnológico Nacional de México/Instituto Tecnologico de Tijuana, Leonardo Trujillo, Instituto Tecnológico de Tijuana, Daniel Hernández, Tecnológico Nacional de México/Instituto Tecnologico de Tijuana

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

Léo Françoso Sotto, Federal University of São Paulo, Paul Kaufmann, Mainz University, Timothy Atkinson, University of Manchester, Roman Kalkreuth, TU Dortmund, Márcio Porto Basgalupp, Federal University of São Paulo

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  $(1 + \lambda)$ . 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  $(1 + \lambda)$  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 Programming, can produce programs that solve problems typically assigned in introductory academic settings. These problems focus on a narrow, predetermined set of simple data structures, basic control flow patterns, and primitive, non-overlapping data types (without, for example, inheritance or composite types). Few, if any, genetic programming methods for program synthesis have convincingly demonstrated the capability of synthesizing programs that use arbitrary data types, data structures, and specifications that are drawn from existing codebases. In this paper, we introduce Code Building Genetic Programming (CBGP) as a framework within which this can be done, by leveraging programming language features such as reflection and first-class specifications. CBGP produces a computational graph that can be executed or translated into source code of a host language. To demonstrate the novel capabilities of CBGP, we present results on new benchmarks that use nonprimitive, polymorphic data types as well as some standard program synthesis benchmarks.

#### GP2

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

### Adaptive Weighted Splines - A New Representation to Genetic Programming for Symbolic Regression

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 been devoted to 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-theart ensemble regression methods Random Forests and Gradient Boosting.

### Constructing Efficient Multigrid Solvers with Genetic Programming

Jonas Schmitt, University of Erlangen-Nuremberg, Sebastian Kuckuk, University of Erlangen-Nuremberg, Harald Köstler,

#### University of Erlangen-Nuremberg

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 grammarguided 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 Vand 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 steadystate 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 multiobjective 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, Mengjie Zhang, Victoria University of Wellington

In traditional regression analysis, a detailed examination of the residuals can provide an important way of validating the model quality. However, it has not been utilised in genetic programming based symbolic regression. This work aims to fill this gap and propose a new evaluation criterion of minimising the correlation between the residuals of regression models and the independent variables. Based on a recent association detection measure, maximal information coefficient which provides an accurate estimation of the correlation, the new evaluation measure is expected to enhance the generalisation of genetic programming by driving the evolutionary process towards models that are without unnecessary complexity and less likely learning from noise in data. The experiment results show that, compared with standard genetic programming which selects model based on the training error only and two state-of-the-art multiobjective genetic programming methods with mechanisms to prefer models with adequate structures, our new multiobjective genetic programming method minimising both the correlation between residuals and variable, and the training error has a consistently better generalisation performance and evolves simpler models.

## GP3

Saturday, July 11, 13:40-15:20

#### 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.

#### GP4

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

# Genetic programming approaches to learning fair classifiers $\bigstar$

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 pa- per, 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 lexi- case 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 con- straints 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* 

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 continuousaction 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.

## Hot off The Press

#### HOP1

Friday, July 10, 10:30-12:10

### Using Exploratory Landscape Analysis to Visualize Single-objective Problems

### Urban Škvorc, Jozef Stefan Institute, Tome Eftimov, 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 nonobstructive 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, populationbased 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**

Michal 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].

#### HOP2

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

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

Stefano Ruberto, University of Pennsylvania, Valerio Terragni, Universitá della Svizzera italiana, Jason H. Moore, University of Pennsylvania

Semantic Genetic Programming (SGP) approaches demonstrated remarkable results in different omains. 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

# Benjamin Doerr, École Polytechnique, Martin S. Krejca, Hasso Plattner Institute

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 DeceptiveLeadingBlocks (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 Eftimov, 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 realworld 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

Trang T. Le, University of Pennsylvania, Weixuan Fu, University of Pennsylvania, Jason H. Moore, University of Pennsylvania

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.

## **Real World Applications**

## RWA1

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

### 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 target 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. Aimed 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 EA's mutation operator. This work focuses on the development of an adaptive method for applying multiple HDHs throughout an EA's 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, J. 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 uttermost 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.

#### RWA2

Saturday, July 11, 13:40-15:20

## One property to rule them all? On the limits of trade-offs for S-boxes

Marko Djurasevic, University of Zagreb, Domagoj Jakobovic, University of Zagreb, Stjepan Picek, Delft University of Technology

Substitution boxes (S-boxes) are nonlinear mappings that represent one of the core parts of many cryptographic algorithms (ciphers). If S-box does not possess good properties, a cipher would be susceptible to attacks. To design suitable S-boxes, 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 Krigingbased 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.

### RWA3

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

#### 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.

### Non-deterministic Journey Planning in Multi-modal Transportation Networks: A Meta-heuristic Approach

Mohammad Haqqani, *RMIT University*, Xiaodong Li, *RMIT University*, Xinghuo Yu, *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 multimodal 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, CO<sub>2</sub>, 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 transportation 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.

#### A Genetic Programming Approach to Feature Construction for Ensemble Learning in Skin Cancer Detection

Qurrat Ul Ain, Victoria University of Wellington, Harith Alsahaf, 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.

# 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.

### RWA4

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

# Towards Sustainable Forest Management Strategies with MOEAs $\Rightarrow$

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 $\bigstar$

Mehdi Neshat, *The University of Adelaide*, Bradley Alexander, *The University of Adelaide*, Nataliia Y. Sergiienko, *The University of Adelaide*, Markus Wagner, *The University of Adelaide* 

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 searchbased 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.

#### RWA5

Sunday, July 12, 09:00-10:40

#### 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. multimaterial, components under technological constraints is performed to guide engineers towards the optimal design of manufactured parts in competition-driven industries, like in the automotive or the aerospace sector. Specifically, three of the main challenges Original Equipment Manufacturers (OEMs) face nowadays are met : simultaneously minimizing compliance, weight and cost. This is achieved by replacing pure metallic components with lightweight materials such as thermoplastics and composites. However, a mere substitution would not be appropriate because it would usually result in insufficient performances or expensive designs. The task of the genetic algorithm is hence to find the optimal material distribution using Voronoï tessellations on a fixed Finite Element (FE) mesh while complying with the manufacturing methods of thermoplastics and composites. The Voronoï encoding has a great advantage over traditional Bit-Array genotypes : its size is independent of the FE mesh granularity, therefore refining the mesh has no impact on the computational cost of the genetic algorithm's operators. Experimental results on the cantilever beam test case show Pareto optimal material distributions.

#### Hybrid Genetic Algorithm for Ridesharing with Timing Constraints: Efficiency analysis with Real-World Data

Nirav Patel, Indian Institute of Technology, Narayanaswamy N.S., Indian Institute of Technology, Alok Joshi, Indian Institute of Technology Ridesharing is an effective, affordable and environmentfriendly way of transportation. This paper proposes a Hybrid Genetic Algorithm(HGA) using elements from Simulated Annealing(SA) and Local search(LS) which is very suitable for Ridesharing related applications. The designed algorithm efficiently handles advanced constraints like timing windows for pick-up, detour time(or distance), and waiting-time minimization. To quantify the effectiveness of HGA for the Ridesharing problem, we study Ridesharing as a close variant of the Orienteering Problem(OP) and Orienteering Problem with Time Windows(OPTW). Both these problems have been extensively studied and have many benchmark data sets. The experiments performed on popular benchmark instances of orienteering problems show that HGA for orienteering problems outperforms all the available algorithms in the literature. Further, for many of the instances, our HGA has discovered solutions that are better than the current Best Known solutions. Using our HGA for orienteering problems, we have developed a prototype system for Ridesharing. The experiments using real-world New York taxi trip data shows that the system ensures that about 40%-50% of the rides could have been shared, and 60%-70% of customers could have shared rides. The detailed data analysis shows how to tune the parameters for the system to achieve the best possible performance.

## Search-Based Software Engineering

#### SBSE1+ENUM2+THEORY1

Saturday, July 11, 13:40-15:20

### 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 costeffectiveness 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.

#### SBSE2

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

# 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 wellstudied 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.

## Theory

### SBSE1+ENUM2+THEORY1

Saturday, July 11, 13:40-15:20

### Fast Mutation in Crossover-based Algorithms 🖈

Denis Antipov, ITMO University, Maxim Buzdalov, ITMO University, Benjamin Doerr, École Polytechniqu

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 crossoverbased 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  $(1 + (\lambda, \lambda))$  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.

#### THEORY2

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

#### 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 3.59approximation 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 $(1 + (\lambda, \lambda))$ GA Is Even Faster on Multimodal Problems

Denis Antipov, ITMO University, Benjamin Doerr, École Polytechniqu, Vitalii Karavaev, ITMO University

For the  $(1 + (\lambda, \lambda))$  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  $(1 + (\lambda, \lambda))$  GA optimizes any jump function with jump size  $2 \le k \le n/16$  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  $(1 + (\lambda, \lambda))$  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  $\Theta(m)$  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 Polytechniqu*, Carola Doerr, *Sorbonne Université*, Dmitry Vinokurov, *ITMO University* 

Runtime analysis aims at contributing to our understanding of evolutionary algorithms through mathematical analyses of

their runtimes. In the context of discrete optimization problems, runtime analysis classically studies the time needed to find an optimal solution. However, both from a practical and a theoretical viewpoint, more fine-grained performance measures are needed. Two complementary approaches have been suggested: fixed-budget analysis and fixed-target analysis. In this work, we conduct an in-depth study on the advantages and limitations of fixed-target analyses. We show that, different from fixed-budget analyses, many classical methods from the runtime analysis of discrete evolutionary algorithms yield fixedtarget results without greater effort. We use this to conduct a number of new fixed-target analyses. However, we also point out examples where an extension of the existing runtime result to a fixed-target result is highly non-trivial.

#### THEORY3

Sunday, July 12, 09:00-10:40

### A Tight Lower Bound on the Expected Runtime of Standard Steady State Genetic Algorithms

Carsten Witt, Technical University of Denmark, Pietro S. Oliveto, University of Sheffield, Dirk Sudholt, The University of Sheffield

Recent progress in the runtime analysis of evolutionary algorithms (EAs) has allowed the derivation of upper bounds on the expected runtime of standard steady-state GAs. These upper bounds have shown speed-ups of the GAs using crossover and mutation over the same algorithms that only use mutation operators (i.e., steady-state EAs) both for standard unimodal (i.e., OneMax) and multimodal (i.e., Jump) benchmark functions. These upper bounds suggest that populations are beneficial to the GA as well as higher mutation rates than the default 1/n rate. However, making rigorous claims was not possible because matching lower bounds were not available. Proving lower bounds on crossover-based EAs is a notoriously difficult task as it is hard to capture the progress that a diverse population can make. We use a potential function approach to prove a tight lower bound on the expected runtime of the  $(2 + 1) \sim GA$ for OneMax for all mutation rates c/n with c < 1.422. This provides the last piece of the puzzle that completes the proof that larger population sizes improve the performance of the standard steady-state GA for OneMax for various mutation rates, and it proves that the optimal mutation rate for the  $(2 + 1) \sim GA$ on OneMax is  $(\sqrt{97} - 5)/(4n) \approx 1.2122/n$ .

#### Does Comma Selection Help To Cope With Local Optima?

#### Benjamin Doerr, École Polytechniqu

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

Amirhossein Rajabi, Technical University of Denmark, Carsten Witt, Technical University of Denmark

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 selfadjusting 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 wellknown 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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