## CEC 2016 Special Session on Learning-based Real-Parameter Single Objective Optimization

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#### Introduction

- CEC 2015 Competition on Learning-based Real-Parameter Single Objective Optimization including 15 benchmark functions <a href="http://www.ntu.edu.sg/home/EPNSugan/index\_files/CEC2015">http://www.ntu.edu.sg/home/EPNSugan/index\_files/CEC2015</a>
- J. J. Liang, B. Y. Qu, P. N. Suganthan, Q. Chen, "Problem Definitions and Evaluation Criteria for the CEC 2015 Competition on Learning-based Real-Parameter Single Objective Optimization", Technical Report, computational Intelligence Laboratory, Zhengzhou University, Zhengzhou China and Technical Report, Nanyang Technological University, Singapore, Nov 2014.

#### Introduction

#### Unimodal Functions

- Rotated High Conditioned Elliptic Function
- Rotated Cigar Function

### Simple Multimodal Functions

- Shifted and Rotated Ackley's Function
- Shifted and Rotated Rastrigin's Function
- Shifted and Rotated Schwefel's Function

#### Introduction

### Hybrid Functions

- Hybrid Function 1 (N=3)
- Hybrid Function 2 (N=4)
- Hybrid Function 3 (N=5)

#### Composition Functions

- Composition Function 1 (N=3)
- Composition Function 2 (N=3)
- Composition Function 3 (N=5)
- Composition Function 4 (N=5)
- Composition Function 5 (N=5)
- Composition Function 6 (N=7)
- $\circ$  Composition Function 7 (N=10)

## **Algorithms**

- 20 papers were submitted
- 5 algorithms are involved in the CEC 2015 benchmark problems
- 9 algorithms are involved in the CEC 2014 benchmark problems
- 6 algorithms are involved in the expensive cost competition

# Algorithms

ID	Algorithm	Paper Title
	MVMO	Solving the CEC2016 Real-Parameter Single Objective Optimization Problems through MVMO-PHM (Technical Report)
E-16539	CCLSHADE	Cooperative Co-evolution using LSHADE with Restarts For The CEC15 Benchmarks
E-16554	LSHADE44	LSHADE with Competing Strategies Applied to CEC2015 Learning based Test Suite
E-16541	AsAMP-dD	An Asynchronous Adaptive Multi population Model for Distributed Differential Evolution
E-16621	SOMA	Competition On Learning-based Real-Parameter Single Objective Optimization by SOMA Swarm Based Algorithm with SOMARemove Strategy

## Friedman Ranking

Algorithm	Ranking	
SOMA	3.7500	
AsAMPdD	3.9833	
CCLSHADE	2.4417	
LSHADE44	2.6583	
MVMO	2.1667	

Best
Second
Third

## Ranking Method-1

Rank all the algorithms based on overall fitness summation for all the dimensions

	SOMA	AsAMP-dD	CCLSHADE	LSHADE44	MVMO
D=10	8.2752E+02	3.0772E+03	5.0833E+03	7.0448E+03	6.4631E+02
D=30	4.9992E+04	3.9158E+04	3.4581E+04	4.3470E+04	2.3868E+03
D=50	2.2369E+05	8.5913E+04	6.8460E+04	5.8696E+04	5.7325E+03
D=100	3.0760E+06	1.7741E+05	1.2021E+05	2.5500E+05	2.0087E+04
Overall	3.3505E+06	3.0556E+05	2.2833E+05	3.6421E+05	2.8852E+04
Rank	5	3	2	4	1

Best
Second
Third

## Ranking Method- 2 (Wilcoxon test)

 Use MVMO as a control method and compare CCLSHADE algorithm

CCLSHADE	D=10	D=30	D=50	D=100
+ (win)	9	2	2	1
= (tie)	1	1	1	3
- (lose)	5	12	12	11
Overall	4	-10	-10	-10