

Benchmarking Parameter Control Methods in DE for Mixed-Integer Black-Box Optimization

GECCO 2024 at Melbourne

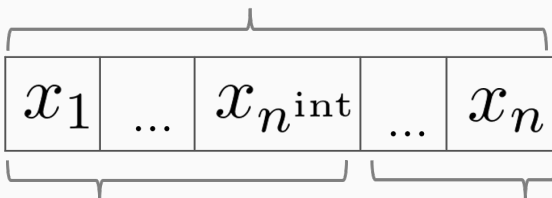
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Mixed-integer black-box optimization in this work

n variables in total



The first n^{int}

variables are integer

The other $n - n^{\text{int}}$

variables are continuous

Note: Categorical variables are not considered

Mixed-integer BBOB (bbob-mixint) [Tušar 19]

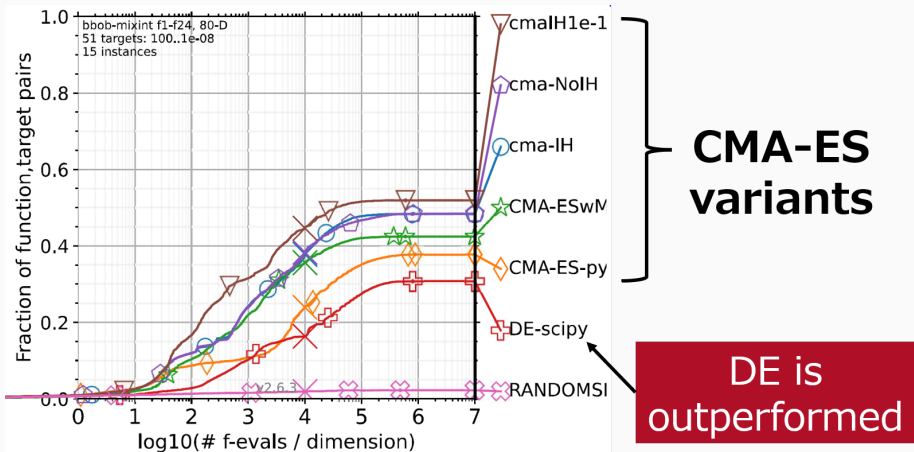
- bbob-mixint consists of 24 mixed-integer functions
 - They are mixed-integer versions of the 24 bbob functions
 - The first $4n/5$ variables are integer

The first 8 variables are integer

The other 2 variables are continuous

x	Range
x_1, x_2	$\{0, 1\}$
x_3, x_4	$\{0, 1, 2, 3\}$
x_5, x_6	$\{0, 1, \dots, 7\}$
x_7, x_8	$\{0, 1, \dots, 15\}$
x_9, x_{10}	$[-5, 5]$

CMA-ES variants outperform the Scipy imple. of differential evolution (DE) on bbob-mixint [Marty 23]



Tristan Marty, Yann Semet, Anne Auger, Sébastien Héron, Nikolaus Hansen : Benchmarking CMA-ES with Basic Integer Handling on a Mixed-Integer Test Problem Suite. GECCO Companion 2023: 1628-1635

Contribution: This work benchmarks parameter control methods (PCMs) in DE on bbob-mixint

- Two critical parameters in DE
 - Scale factor $s > 0$ for differential mutation
 - Crossover rate $c \in [0, 1]$ for crossover
 - Best settings of s and c depend on a problem
- DE requires a PCM that automatically adjusts s and c
 - Their importance is well known for continuous optimization
 - All state-of-the-art DEs (e.g., L-SHADE) use them
- Their effectiveness is unknown for mixed-integer opt.
 - The performance of DE can possibly be improved

Scale factor s and crossover rate c are critical in DE

- For each iteration of DE, the μ individuals $\mathbf{x}_1, \dots, \mathbf{x}_\mu$ generate the μ children $\mathbf{u}_1, \dots, \mathbf{u}_\mu$
 - In environmental selection, \mathbf{u}_i is compared only to \mathbf{x}_i
 - For each $i \in \{1, \dots, \mu\}$, if $f(\mathbf{u}_i) \leq f(\mathbf{x}_i)$, $\mathbf{x}_i \leftarrow \mathbf{u}_i$

- Scale factor s determines the magnitude of mutation

$$\mathbf{v}_i = \mathbf{x}_{r_1} + s (\mathbf{x}_{r_2} - \mathbf{x}_{r_3})$$

- Crossover rate c determines the number of elements inherited from each parent \mathbf{x} to a child \mathbf{u}

$$u_{i,j} = \begin{cases} v_{i,j} & \text{if } \text{rand}[0, 1] \leq c \text{ or } j = j_{\text{rand}} \\ x_{i,j} & \text{otherwise} \end{cases}$$

This work focuses only on PCMs in DE

I am interested in “the PCM in jDE”, not “jDE”



DE instances

DE template

Mutation strategy
Crossover method
PCM
:

jDE

rand/1
binomial
PCM in jDE
:

CoDE

rand/1, rand/2, curr-to-rand/1
binomial
PCM in CoDE
:

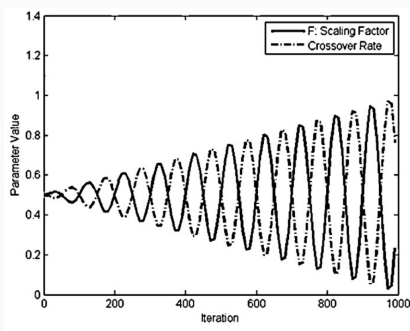
cDE,
SaDE,
EPSDE,
SinDE,
JADE,
...

This work benchmarks 9 PCMs in DE

- 3 deterministic PCMs
 - PCM in CoDE [Wang 11]
 - PCM in SinDE [Draa 15]
 - PCM in CaRS+S [Molina-Pérez 24]
- 6 adaptive PCMs
 - PCM in jDE [Brest 06]
 - PCM in cDE [Tvrđik 06]
 - PCM in JADE [Zhang 09]
 - PCM in EPSDE [Mallipeddi 11]
 - PCM in SHADE [Tanabe 13]
 - PCM in CoBiDE [Wang 14]

Deterministic parameter control methods in DE

- Generating s and c values based on a simple rule
 - without using any information obtained by the search
- Example: The PCM in SinDE [Draa 15]
 - s and c are determined based on the sinusoidal functions



Most adaptive PCMs in DE use a success criterion

- If the child u_i is better than its parent x_i , the pair of s_i and c_i are said to be *successful*
 - Assumption: The better child was generated because the pair of s_i and c_i was suitable for a target problem
- Example: The PCM in jDE [Brest 06]



Rounding operator to repair infeasible solutions

Infeasible

0.4
0.6
1.7
⋮

Rounding



Feasible

0
1
2
⋮

The Lamarckian and Baldwinian repair methods

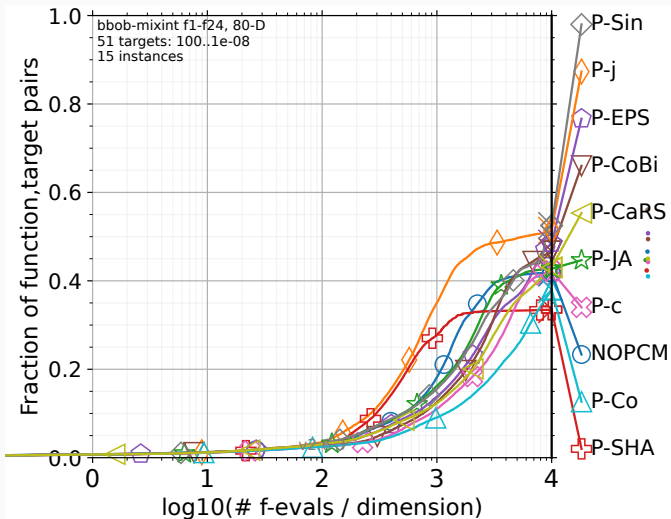
- Let \mathbf{y} be a repaired feasible version of an infeasible \mathbf{x}
- Both of them use $f(\mathbf{y})$ as $f(\mathbf{x})$, but ...
 - The Lamarckian one *replaces* \mathbf{x} with \mathbf{y}
 - The Baldwinian one *doesn't* replace \mathbf{x} with \mathbf{y}
- All individuals in the population are ...
 - **always feasible** for the Lamarckian one
 - **likely to be infeasible** for the Baldwinian one
- There is no clear winner between them [Salcedo-Sanz 09]

Experimental setup

- The COCO platform [Hansen 21]
 - The 24 bbob-mixint functions
 - Num. variables $n \in \{5, 10, 20, 40, 80, 160\}$
- Settings for DE: $(9 + 1) \times 8 \times 2 = 160$ configurations
 - The max. fun. evals. was set to $10^4 \times n$
 - The population size μ : 100
 - Hyper-par. in the 9 PCMs: default
 - For DE with no PCM, $s = 0.5$ and $c = 0.9$
 - Eight representative mutation strategies
 - The Lamarckian and Baldwinian repair methods

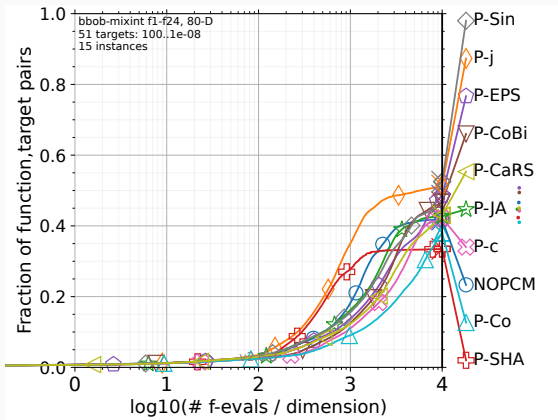
Results for $n = 80$ (rand/1, Baldwinian)

PCMs can improve the performance of DE



Results for $n = 80$ (rand/1, Baldwinian)

- DE with the PCM in SHADE performs poorly
 - P-SHA is SOTA for continuous bbob
 - Rankings for bbob and bbob-mixint are different



Best PCM depends on the type of mutation strategy

Results of the Baldwinian repair method

Strategy	$n = 5$	$n = 10$	$n = 20$	$n = 40$	$n = 80$	$n = 160$
rand/1	NOPCM	P-CoBi	P-c	P-Sin	P-Sin	P-Sin
rand/2	P-Sin	P-Sin	P-Sin	P-Sin	P-j	P-j
best/1	P-Co	P-Co	P-Co	P-Co	P-Co	P-JA
best/2	P-CoBi	P-CoBi	P-EPS	P-CoBi	P-c	P-c
ctb/1	P-CoBi	P-Co	P-Co	P-CoBi	P-Co	P-JA
ctr/1	P-CoBi	P-CoBi	P-CoBi	P-CoBi	P-CoBi	P-CoBi
ctp/1	P-Co	P-CoBi	P-Co	P-Co	P-Co	P-Co
rtp/1	P-Co	P-Co	P-Co	P-Co	P-Co	P-Co

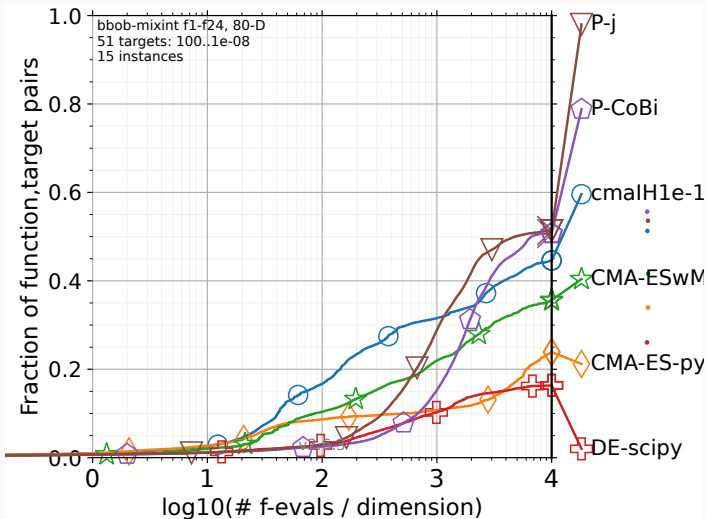
Best PCM depends on the type of repair method

Results of the Lamarckian repair method

Strategy	$n = 5$	$n = 10$	$n = 20$	$n = 40$	$n = 80$	$n = 160$
rand/1	P-CoBi	P-CoBi	P-CoBi	P-CaRS	P-CoBi	P-JA
rand/2	P-Sin	P-Sin	P-CoBi	P-Sin	P-CoBi	P-CoBi
best/1	P-Co	P-Co	P-Co	P-Co	P-Co	P-Co
best/2	P-CoBi	P-CoBi	P-Co	P-CaRS	P-Co	P-Co
ctb/1	P-Co	P-CoBi	P-Co	P-Co	P-Co	P-Co
ctr/1	P-CoBi	P-CoBi	P-Co	P-CoBi	P-CoBi	P-CoBi
ctp/1	P-CoBi	P-CoBi	P-CoBi	P-Co	P-Co	P-Co
rtp/1	P-CoBi	P-SHA	P-Co	P-Co	P-Co	P-Co

Comparison with CMA-ES variants ($n = 80$)

DE with PCMs perform well for larger budgets of fevals



Conclusion: parameter control methods (PCMs)

- 9 PCMs in DE were benchmarked on bbob-mixint
 - Using a PCM can improve the performance of DE
 - The best PCM depends on the type of
 - mutation strategy (e.g., rand/1, rand/2, best/1, etc.)
 - repair method (Lamarckian and Baldwinian)
 - The best PCMs on bbob and bbob-mixint are different
 - PCM in SHADE does not work for bbob-mixint
 - PCM in CoDE, CoBiDE, jDE, and SinDE work well
 - DE with a suitable PCM performs better than CMA-ES on bbob-mixint for larger budgets of fevals
- Many future works: analysis of “DE” and “PCMs”
 - Nobody can explain how the results happened
 - Even analysis of DE for mix-int. opt. has not been done