

goa
(Global Optimization Animations)
Optimization Methods - Global Optimization Project

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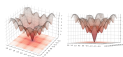
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Introduction

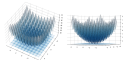
What is **goa** (Global Optimization Animations)?

- **goa** is a Python package that implements:
 - 1 some problems (optimization test functions)

- Ackley function



- Rastrigin function



- 2 some solutions (optimization algorithms)

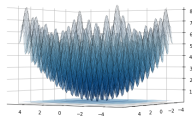
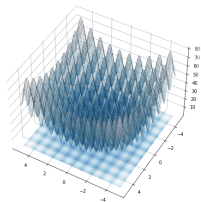
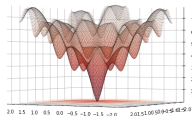
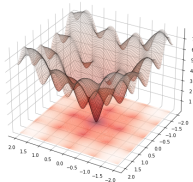
- Differential Evolution [1]
- Memetic Differential Evolution [3]
- Coordinate Method with Simple Descent Direction [5]

- **goa** aims to visualize the execution of these algorithms on a given problem

Ackley and Rastrigin functions

Ackley and Rastrigin functions have common similarities:

- multimodal ($f : \mathbb{R}^n \rightarrow \mathbb{R}$)
- many local minima
- unique global minimum at $f(x^*) = 0$, with $x^* = (0, \dots, 0)$



Differential Evolution algorithm

Given a population of p vectors $x_i \in \mathbb{R}^n, i \in \{0, \dots, p-1\}$,
a high level description of one iteration of the algorithm is the following:

- for each x_i :
 - 1 apply to x_i a linear transformation and call it *Trial*
 - 2 restore randomly chosen dimensions of *Trial*
 - 3 if $f(\textit{Trial}) < f(x_i)$, then set $x_i := \textit{Trial}$

Memetic Differential Evolution vs Differential Evolution

The two algorithms differ only by the 10th line of code.

```
Data:  $F \in (0, 2)$ ,  $CR \in [0, 1]$ 
1 foreach  $i \in 1, \dots, p$  do
2   let  $ii := \mathcal{U}(1, \dots, n)$ ;
3   randomly choose  $k_1, k_2, k_3 \in \{1, \dots, p\} \setminus \{i\}$ ;
4   let  $Trial := x_{k_1} + F(x_{k_2} - x_{k_3})$ ;
5   for  $j \in 1, \dots, n : j \neq ii$  do
6     if  $\mathcal{U}(0, 1) < CR$  then
7       let  $Trial^{(j)} := x_i^{(j)}$ ;
8     end
9   end
10  let  $Trial := \mathcal{L}(f, Trial)$ ;
11  if  $f(Trial) < f(x_i)$  then
12    let  $x_i := Trial$ ;
13  end
14 end
```

Algorithm 1: Memetic Differential Evolution

```
Data:  $F \in (0, 2)$ ,  $CR \in [0, 1]$ 
1 foreach  $i \in 1, \dots, p$  do
2   let  $ii := \mathcal{U}(1, \dots, n)$ ;
3   randomly choose  $k_1, k_2, k_3 \in \{1, \dots, p\} \setminus \{i\}$ ;
4   let  $Trial := x_{k_1} + F(x_{k_2} - x_{k_3})$ ;
5   for  $j \in 1, \dots, n : j \neq ii$  do
6     if  $\mathcal{U}(0, 1) < CR$  then
7       let  $Trial^{(j)} := x_i^{(j)}$ ;
8     end
9   end
10  // let  $Trial := \mathcal{L}(f, Trial)$ 
11  if  $f(Trial) < f(x_i)$  then
12    let  $x_i := Trial$ ;
13  end
14 end
```

Algorithm 2: Differential Evolution

Animations - DE vs MDE on Ackley

- Left example: DE reaches convergence in 38 iterations
- Right example: MDE reaches convergence in 3 iterations

DE on Ackley

MDE on Ackley

Animations - DE vs MDE on Quadratic

- Left example: DE reaches convergence in 19 iterations
- Right example: MDE reaches convergence in 1 iteration

DE on Quadratic

MDE on Quadratic

Command line interface (CLI)

goa provides a CLI that allows to choose:

- 1 problem
- 2 problem bounds (optional)
- 3 optimization algorithm
- 4 local optimizer (required only for memetic global optimizer)
- 5 animation filepath (optional)

```
(goa) → presentation git:(main) x python -m goa
Select a problem (Ackley, Rastrigin, quadratic) [Ackley]:
[OPTIONAL] Change the problem bounds [(-2.5, 2.5)]:
Select an optimization algorithm (MDE, DE, CM) [DE]:
[REQUIRED only with MDE] Select a local search algorithm (CM, None) [None]:
[OPTIONAL] Want an animation? Provide a filepath .gif []: 'example.gif'
Iteration:    5 | RMSE: 0.91623379
Iteration:   10 | RMSE: 0.84967181
Iteration:   15 | RMSE: 0.74277798
Iteration:   20 | RMSE: 0.06209373
Iteration:   25 | RMSE: 0.00837453
Iteration:   30 | RMSE: 0.00156031
Iteration:   35 | RMSE: 0.00026433
Terminated at Iteration: 37
```


Conclusion

Who would need **goa**?

- anyone that would like to support an explanation
- anyone that would like to have another point of view

What could be improved?

- scalability (abstract classes for both global and local optimization algorithms)
- reproducibility (all RNGs should depend on a single random seed)

Where can the implementation be found?

- <https://github.com/deeplego/goa>

Thanks for your attention!

Do you have any questions?



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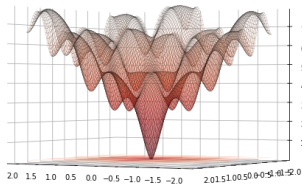
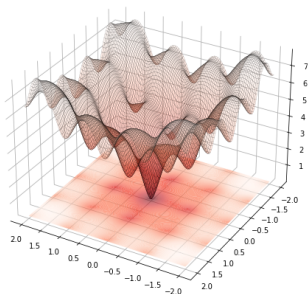
Backup slides

Definition (Ackley function)

$$f(x_1 \cdots x_n) = -20 \exp\left(-0.2 \sqrt{\frac{1}{n} \sum_{i=1}^n x_i^2}\right) - \exp\left(\frac{1}{n} \sum_{i=1}^n \cos(2\pi x_i)\right) + 20 + e$$

$$-32 \leq x_i \leq 32$$

minimum at $f(x^*) = 0$, with $x^* = (0, \dots, 0)$



Definition (Rastrigin function)

$$f(x_1 \cdots x_n) = 10n + \sum_{i=1}^n (x_i^2 - 10 \cos(2\pi x_i))$$

$$-5.12 \leq x_i \leq 5.12$$

minimum at $f(x^*) = 0$, with $x^* = (0, \dots, 0)$

