Artificial Bee Colony Algorithm

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Outline

- Description
- The algorithm
- Implementation
- 4 ABC in Software Engineering

What is it?

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- what is foraging?
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Many variations

There are many changed versions of the algorithm for specific needs. It all depends on the problem you are willing to solve. The survey in [2] shows different variations and applications.

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- Employed bees: memorize and exploit specific food sources;
- Onlookers: exploit already known good food sources;
- Scouts: search for new food sources around without prior information.

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- The nectar amount of a food source represents its fitness value.
- Each **type of bee** represents a **step** of the algorithm.

Pseudocode

```
procedure ABC(population_size, maximum_cycle_number, limit)
   initialize population_size food_sources
   generate random position for food_sources
   while maximum_cycle_number is not met do
      send employed bees
      calculate the probabilities
      send onlookers
      send scouts
      memorize the best food source
   end while
end procedure
```

Employed bees phase

- Each food source will be exploited once. The exploitation process generates modifications in the parameters of the food source, hoping that a better solution will be found.
- If a better food source is found, the old one is forgotten.

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Equation 1

Given the number of food sources F and the dimension of the problem D, each food source f_i will generate a possible new food source v_i following the formula:

$$\mathsf{v}_{ij} = \mathsf{f}_{ij} + \phi_{ij}(\mathsf{f}_{ij} - \mathsf{f}_{kj})$$

where $i \in [1, 2, 3, ..., F]$, $j \in [1, 2, 3, ..., D]$, $k \in [1, 2, 3, ..., F]$ and ϕ_{ij} is a random number $\in [-1, 1]$.

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 The algorithm iterates through the food sources array and selects food sources based on probability until the established amount of onlookers is reached.

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Equation 2

Given the number of food sources F, the probability p_i for a food source i will be calculated by the formula:

$$p_i = \frac{fit_i}{\sum\limits_{n=1}^{F} fit_n}$$

where $i \in [1, 2, 3, ..., F]$.

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• The chosen food sources will be exploited. If a new food source with better fitness is found, the old one is forgotten.

Scouts phase

- If a food source was not changed in a cycle, its trial counter will rise.
- At the end of every cycle, the food source with the biggest trial counter will receive a random solution if its trial counter is bigger than the parameter *limit*.

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Why does it work?

- As said by [3] the ABC algorithm performs different selection process:
 - a global selection process performed by the onlookers;
 - a local selection process performed by onlookers and employed bees;
 - a random selection process performed by scouts.

Why does it work?

- As said by [3] the ABC algorithm performs different selection process:
 - a global selection process performed by the onlookers;
 - a local selection process performed by onlookers and employed bees;
 - a random selection process performed by scouts.
- The employed bees and onlookers are responsible for the exploitation process in the search space, while the scouts are responsible for the exploration process.

Implementation

• You can find the python code on https://goo.gl/B6tTgv.



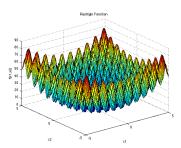
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- To validate my implementation, I will test the algorithm with a function from [3] using the same parameters and compare the results.

Rastrigin function

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

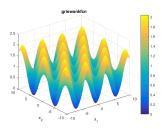
• It is a multimodal function with locations of the minima regularly distributed. The interval is $-15 \le x_i \le 15$ and the function has its global minimum at $f(0, \dots, 0) = 0$.



Griewank function

$$f(x_1 \cdots x_n) = 1 + \frac{1}{4000} \sum_{i=1}^n x_i^2 - \prod_{i=1}^n \cos(\frac{x_i}{\sqrt{i}})$$

• It is a multimodal function with locations of the minima regularly distributed. The function has a interdependence between variables. The interval is $-600 \le x_i \le 600$ and the function has its global minimum at $f(0, \cdots, 0) = 0$.





Parameters

- 50% of employed bees, 50% of onlookers, one scout.
- population_size: 125.
- maximum_cycle_number: 500.
- dimension: 10.
- limit: 50.
 - The author did not write about this parameter in [3].
- Each experiment was repeated 30 times.

Results

Table: Comparison of results

	Results from [3]		My results	
	Mean	SD	Mean	SD
Rastrigin	0	0	$1.76 * 10^{-12}$	$2.83367 * 10^{-12}$
Griewank	0.000870	0.002535	0.000142	0.000497

Examples of applications

As showed by [2], the ABC algorithm has already been successfully applied to Search-based Software Engineering problems. Some examples are:

- software test case optimization;
- automated generation of structural tests;
- automated software refactoring;

References

- [1] Dervis Karaboga. An idea based on honey bee swarm for numerical optimization. Technical report, 2005.
- [2] Dervis Karaboga, Beyza Gorkemli, Celal Ozturk, and Nurhan Karaboga. A comprehensive survey: Artificial bee colony (abc) algorithm and applications. Artif. Intell. Rev., 42(1):21–57, June 2014.
- [3] Dervis Karaboga and Bahriye Basturk. A powerful and efficient algorithm for numerical function optimization: Artificial bee colony (abc) algorithm. *J. of Global Optimization*, 39(3):459–471, November 2007.

Thank you for your attention! https://goo.gl/B6tTgv almeida.warley@outlook.com