

# Differential Evolution – A diversity approach

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**Abstract** Insert your abstract here. Include keywords, PACS and mathematical subject classification numbers as needed.

**Keywords** Diversity · Differential Evolution · Evolutionary

## 1 Introduction

Evolutionary Algorithms (EAs) are built to deal with optimization problems, which are designed from many scientific and application fields, such as science, economic and engineering [1,2]. Principally, EAs can be classified into following categories, such as Genetic Algorithms (GAs) [3,4], Evolutionary Strategies (ESs) [5], Genetic Programming (GP) [6], Evolutionary Programming (EP) [7], Differential Evolution (DE) [8] and other natural-inspired algorithms [9]. DE was introduced by Storn and Price [8], also is considered as one of the most effective EAs used to deal with real-world optimization problems, mainly for its convergency properties. Similarly than with other EAs, DE follows the natural evolution process which involves mutation, recombination and selection to evolve a population through an iterative progress until the criteria stop is reached. However, the peculiarity of DE is that employs the difference of vectors parameters to explore the search space, being very similar than its precursor algorithms namely the Nelder-Mead [10] and the Controlled Random Search (CRS) [11]. In spite of the popularity and effectiveness of DE, there exists several weakness that had been partially solved through learning techniques. One of the first weakness and possibly one of the most important

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is that the performance of DE is very sensitive to choice of the strategy parameters depending in the objective function [12]. Several strategies to alleviate the sensitivite have been proposed. Although that several strategies have been proposed ref 17 and 37,

**\*\*Weakness discovered in DE.** **\*\*Dependence between parameters and the quality solutions** ref 15, 60. **\*\*The strength is internally induces, the mutation depends on the content of the population, due the limited number of different trial slutions within one generation, producing a stagnatin** ref 19. **\*\*Influence of the population size and stagnation** ref20 **\*\*Premature loss of diversity** ref 2. **\*\*Techniques to deal with this hybridization with anneling procedures to reduce the selection pressure** ref37 **\*\*Generational replacement** ref3. **\*\*Incrementing the population size** ref19. **\*\*Organization of the paper.**

## 2 Differential Evolution

\*Fundamentals. - Fully description of Differential Evolution. — Description of differential evolution. \*Decription of the single-objective problem. \*Explain each stage in DE. \* \*

—Diversity Revision \*Explain the influence of the paramters. \*Show the implication of these operators with the population diversity. \*Talk about hybrid and adaptive strategies.

## 3 Differential Evolution Trends

## 4 Proposal

## 5 Experimental Study

## 6 Conclusion

### 6.1 Subsection title

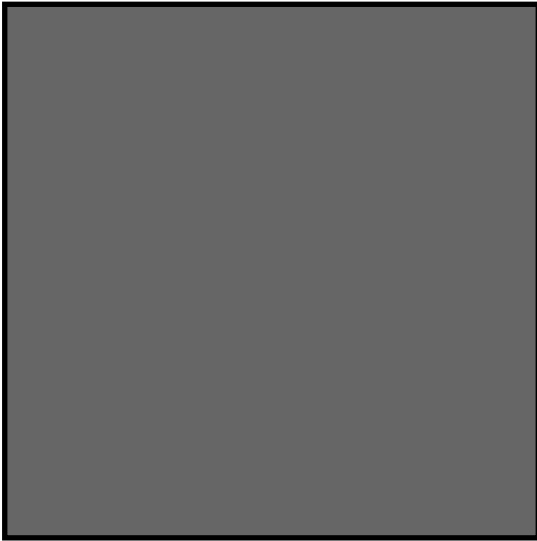
as required. Don't forget to give each section and subsection a unique label (see Sect. ??).

*Paragraph headings* Use paragraph headings as needed.

$$a^2 + b^2 = c^2 \quad (1)$$

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**Fig. 1** Please write your figure caption here



**Fig. 2** Please write your figure caption here

**Table 1** Please write your table caption here

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