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A fast and efficient multi-objective optimization algorithm

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Abstract—NSGA-II is very popular for solving multi-objective optimization problems. But its efficiency decreases gradually as the number of variables increases. In this work, we propose a novel multi-objective optimization algorithm to solve this problem. Numerical experiments show that our algorithm performs significantly better than NSGA-II on the test problems.

I. Introduction

这是中文 这是楷体,目标函数 f(x) 其中 x 是设计变量。

NSGA-II [1] is very popular for solving multi-objective optimization problems. Popular algorithms for multi-objective optimization problems are [1]–[3].

Section I gives the introduction. Section II introduces the backgrounds of this work.

II. BACKGROUNDS

We have a scalar x, a vector x, a random variable X, a matrix X, where $x = [x_1, x_2, \dots, x_d]$, $X = \{x^{(1)}, x^{(2)}, \dots, x^{(n)}\}$, where n n is the number of points. we try to solve the following optimization problem in (1).

minimize
$$f(x)$$
 (1)

The algorithm is given in Algorithm 1.

Algorithm 1 Algorithm of computing the maximum of two variables

Require: a, bEnsure: $c = \max(a, b)$ if a > b then c = aelse c = bend if

- 1) Step 1: see whether a is greater than b.
- 2) Step 2: if it is true, return a.
- 3) Step 3: if not, return b.
- A. Multi-objective optimization problem
- B. NSGA-II
 - 1) Non-dominated Sorting:
 - 2) Crowding Distance:

III. Proposed Algorithm

IV. NUMERICAL EXPERIMENTS

The experiment results are given in Table I. The convergence curves are given in Figure 1.

TABLE I: Results of NSGA-II and X

problem	NSGA-II	X
f_1	7.8 (6.1)	10.8 (5.4)
f_2	8.9 (7.2)	11.9 (6.5)
f_3	9.0 (8.3)	12.0 (6.6)

TABLE II: Results of NSGA-II and X

problem	NSGA-II		X	
	mean	std	mean	std
f_1	7.8	(6.1)	10.8	(5.4)
f_2	8.9	(7.2)	11.9	(6.5)
f_3	9.0	(8.3)	12.0	(6.6)

V. Conclusion

REFERENCES

- [1] K. Deb, A. Pratap, S. Agarwal, and T. Meyarivan, "A fast and elitist multiobjective genetic algorithm: Nsga-ii," *IEEE Transactions on Evolutionary Computation*, vol. 6, no. 2, pp. 182–197, 2002.
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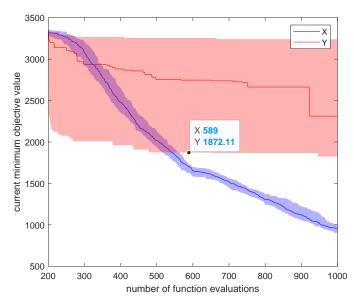


Fig. 1: Convergence curves

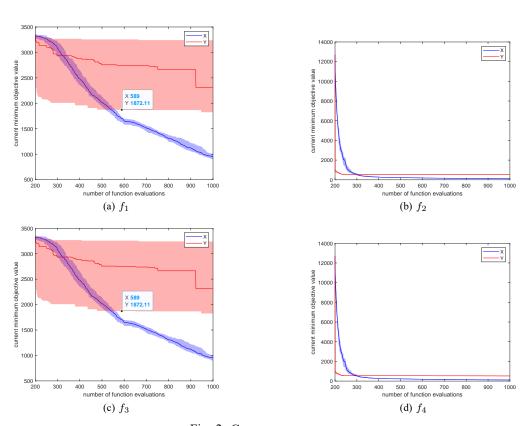


Fig. 2: Convergence curves