COMMENTARY

Comments on "Particle swarm optimization with fractional-order velocity"

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Abstract In this paper, some comments on the paper Solteiro Pires et al. (Nonlinear Dyn. 67:893–901, 2010) are presented. We demonstrate that the authors of the above paper have deduced the incorrect formula about the velocity updating strategy of the fractional-order particle swarm optimization algorithm. This paper deduces the modified updating formula, and verified experiments are also conducted.

Keywords Fractional-order calculus · Particle swarm optimization

In Ref. [1], the authors have proposed an improved version of particle swarm optimization (PSO) using fractional-order calculus concepts, in which fractional calculus is used to control its convergence. During the past several years, the fractional-order PSO algorithm has attracted the attention of several researchers [2–4], and many new researches on PSO model improvement have also conducted [5–7].

One error occurred in their designed fractional-order PSO model, which is described in the following section.

As for the original PSO algorithm, the particle movement is characterized by two vectors, namely the current position x and the velocity v. At time t, each particle updates its velocity by the following equation:

$$v_{t+1} - v_t = \phi_1(b - x) + \phi_2(g - x), \tag{1}$$

where *b* denotes the best position found by the particle so far, and *g* denotes the global best position achieved by the whole swarm so far. ϕ_1 and ϕ_2 are the randomly uniformly generated terms. For simplicity, symbols and notation are employed with the same meanings as those in Ref. [1]. From the classical integer-order area, the fractional-order PSO algorithm extends the velocity derivative to the fractional-order area, yielding

$$D^{\alpha}[v_{t+1}] = \phi_1(b-x) + \phi_2(g-x).$$
(2)

Thus, Pires et al. [1] derive the new velocity updating strategy as shown below:

$$v_{t+1} - \alpha v_t - \frac{1}{2} \alpha v_{t-1} - \frac{1}{6} \alpha (1 - \alpha) v_{t-2} - \frac{1}{24} \alpha (1 - \alpha) (2 - \alpha) v_{t-3} = \phi_1 (b - x) + \phi_2 (g - x).$$
(3)

According to Ref. [1], (1) is the special case of (3) when $\alpha = 1$. However, it can be observed that (1) cannot be deduced from (3). Therefore, the key formula of the fractional-order PSO algorithm in Ref. [1] is wrong.

To correct the formula, we substitute the following discrete time implementation expression of fractional differential into (2) again.

$$D^{\alpha}[x(t)] = \frac{1}{T^{\alpha}} \sum_{k=0}^{r} \frac{(-1)^{k} \Gamma(\alpha+1) x(t-kT)}{\Gamma(k+1) \Gamma(\alpha-k+1)}, \quad (4)$$

where T is the sampling period and r is the truncation order. r = 4 is used in this paper, which is in agreement with Ref. [1].

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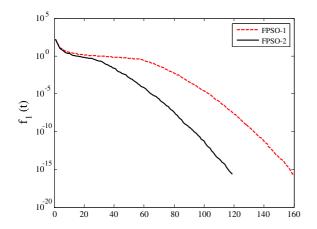


Fig.1 Evolution of the Bohachevsky 1 function using the FPSOs

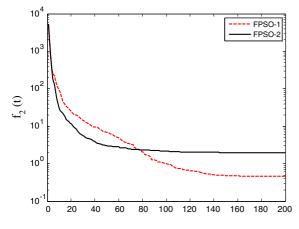


Fig. 2 Evolution of the Colville function using the FPSOs

Hence, the fractional-order behavior of PSO can be written as

$$v_{t+1} - \alpha v_t - \frac{1}{2} \alpha (1 - \alpha) v_{t-1} - \frac{1}{6} \alpha (1 - \alpha) (2 - \alpha) v_{t-2} - \frac{1}{24} \alpha (1 - \alpha) (2 - \alpha) (3 - \alpha) v_{t-3} = \phi_1 (b - x) + \phi_2 (g - x)$$
(5)

That is

$$v_{t+1} = \alpha v_t + \frac{1}{2} \alpha (1-\alpha) v_{t-1} + \frac{1}{6} \alpha (1-\alpha) (2-\alpha) v_{t-2} + \frac{1}{24} \alpha (1-\alpha) (2-\alpha) (3-\alpha) v_{t-3} + \phi_1 (b-x) + \phi_2 (g-x)$$
(6)

In the following sections, we revalidate the performance of the fractional-order PSO algorithm using (6), and the results are compared with those obtained by

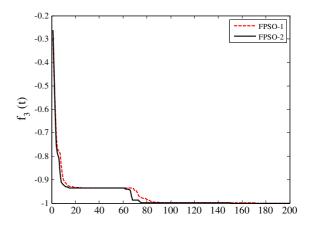


Fig. 3 Evolution of the Drop wave function using the FPSOs

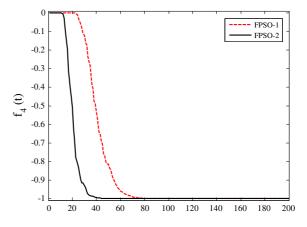


Fig. 4 Evolution of the Easom function using the FPSOs

Pires et al. in Ref. [1]. To indicate the difference, we denote the PSO algorithm described in this comment as FPSO-2, while the algorithm presented in Ref. [1] is denoted by FPSO-1. The test functions adopted herein are the five well-known functions namely Bohachevsky 1, Colville, Drop wave, Easom, and Rastrigin, which are the same expressions as presented in Ref. [1]. Parameters of the FPSO algorithms are also in agreement with Ref. [1] as well, which are set as follows: the population size is 10, the maximum number of iteration is 200, and ϕ_1 and ϕ_2 are randomly uniformly generated in [0, 1]. Moreover, the value of α reduces according to $\alpha(t) = 0.9 - 0.6t/200, t = 0, 1, \dots, 200.$ For the purpose of reducing statistical errors, each algorithm is tested 201 times independently for every function and the median results are used in the comparison. Figures 1, 2, 3, 4 and 5 demonstrate the iteration evolutionary progresses. The correct results for the PSO

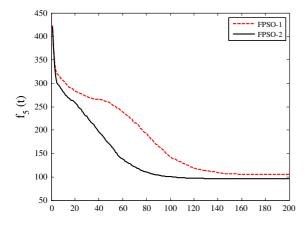


Fig. 5 Evolution of the Rastrigin function using the FPSOs

with fractional-order velocity are indicated by black solid lines.

Moreover, the global minimum value of the Droop wave function is $f^*(x) = -1.0$, rather than $f^*(x) = 0.0$ given by Ref. [1].

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