

Lab 5 - Implement and Understand PSO algorithm

CSE, SUSTech

Outline of This Lab

- Teaching Assistant
- How Does The PSO with Inertia Weight Work
- Does The Inertia Weight Matter?
- Test Functions
- Illustrate The Results!

Teaching assistant

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How Does The PSO with Inertia Weight Work

-Part I

The PSO Algorithm (Main Loop):

1. Initialize a swarm of particles with random positions and velocities.
2. Evaluate each particle's fitness.
3. Update each particle's pbest and the swarm's gbest.
4. For each particle:
 - a. Calculate its new velocity.
 - b. Update its position.
5. Repeat from Step 2 until a stopping criterion is met (e.g., max iterations, solution found).

How Does The PSO with Inertia Weight Work -Part II

$$\begin{aligned} v_{id} &= \textcolor{blue}{v_{id}} + c_1 \text{rand}() (p_{id} - x_{id}) + c_2 \text{Rand}() (p_{gd} - x_{id}) \\ x_{id} &= x_{id} + v_{id} \end{aligned}$$

add inertia weight

\downarrow

$$v_{id} = w v_{id} + c_1 \text{rand}() (p_{id} - x_{id}) + c_2 \text{Rand}() (p_{gd} - x_{id})$$
$$x_{id} = x_{id} + v_{id}$$

Does The Inertia Weight Matter?

- Implement the above algorithm with the real-valued representation with different inertia weight setting

Test Functions

-7 unimodal benchmark functions

- Unimodal functions: f_1-f_5
- f_6 is the step function (one minimum, discontinuous).
- f_7 is a noisy quartic function, where $\text{random}[0, 1]$ is a uniformly distributed random variable in $[0, 1)$.

Test function	n	S	f_{min}
$f_1(\mathbf{x}) = \sum_{i=1}^n x_i^2$	30	$[-100, 100]^n$	0
$f_2(\mathbf{x}) = \sum_{i=1}^n x_i + \prod_{i=1}^n x_i $	30	$[-100, 100]^n$	0
$f_3(\mathbf{x}) = \sum_{i=1}^n (\sum_{j=1}^i x_j)^2$	30	$[-10, 10]^n$	0
$f_4(\mathbf{x}) = \max_i\{ x_i , 1 \leq i \leq n\}$	30	$[-100, 100]^n$	0
$f_5(\mathbf{x}) = \sum_{i=1}^n [100(x_{i+1} - x_i^2)^2 + (x_i - 1)^2]$	30	$[-100, 100]^n$	0
$f_6(\mathbf{x}) = \sum_{i=1}^n (x_i + 0.5)^2$	30	$[-30, 30]^n$	0
$f_7(\mathbf{x}) = \sum_{i=1}^n ix_i^4 + \text{random}[0, 1)$	30	$[-1.28, 1.28]^n$	0

Test Functions

-8 multimodal benchmark functions

Test function	n	S	f_{min}
$f_8(\mathbf{x}) = \sum_{i=1}^n -x_i \sin(\sqrt{ x_i })$	30	$[-500, 500]^n$	-12569.5
$f_9(\mathbf{x}) = \sum_{i=1}^n [x_i^2 - 10 \cos(2\pi x_i) + 10]$	30	$[-5.12, 5.12]^n$	0
$f_{10}(\mathbf{x}) = -20 \exp\left(-0.2\sqrt{\frac{1}{n} \sum_{i=1}^n x_i^2}\right) - \exp\left(\frac{1}{n} \sum_{i=1}^n \cos 2\pi x_i\right) + 20 + e$	30	$[-32, 32]^n$	0
$f_{11}(\mathbf{x}) = \frac{1}{4000} \sum_{i=1}^n x_i^2 - \prod_{i=1}^n \cos\left(\frac{x_i}{\sqrt{i}}\right) + 1$	30	$[-600, 600]^n$	0
$f_{12}(\mathbf{x}) = \frac{\pi}{n} \left\{ 10 \sin^2(\pi y_1) + \sum_{i=1}^{n-1} (y_i - 1)^2 [1 + 10 \sin^2(\pi y_{i+1})] + (y_n - 1)^2 \right\} + \sum_{i=1}^n u(x_i, 10, 100, 4),$ $y_i = 1 + \frac{1}{4}(x_i + 1)$ $u(x_i, a, k, m) = \begin{cases} k(x_i - a)^m, & x_i > a, \\ 0, & -a \leq x_i \leq a, \\ k(-x_i - a)^m, & x_i < -a. \end{cases}$	30	$[-50, 50]^n$	0
$f_{13}(\mathbf{x}) = 0.1 \left\{ \sin^2(3\pi x_1) + \sum_{i=1}^{n-1} (x_i - 1)^2 [1 + \sin^2(3\pi x_{i+1})] + (x_n - 1)^2 [1 + \sin^2(2\pi x_n)] \right\} + \sum_{i=1}^n u(x_i, 5, 100, 4)$	30	$[-50, 50]^n$	0
$f_{14}(\mathbf{x}) = \left[\frac{1}{500} + \sum_{j=1}^{25} \frac{1}{j + \sum_{i=1}^2 (x_i - a_{ij})^6} \right]^{-1}$	2	$[-65.536, 65.536]^n$	1
$f_{15}(\mathbf{x}) = \sum_{i=1}^{11} \left[a_i - \frac{x_1(b_i^2 + b_i x_2)}{b_i^2 + b_i x_3 + x_4} \right]^2$	4	$[-5, 5]^n$	0.0003075

Experimental Setup

- 15 Test Minimization Functions (7 unimodal + 8 multimodal benchmark functions).
- Population size 50.
- Maximum function evaluation 500, 000.
- Set inertia weight as 0.5, 0.0, and 1.2, respectively
- 30 independent runs for each function and each different inertia weight setting.

Illustrate The Results!

Plot 45 figures with one for each test function and each inertia weight setting

- ✓ x-axis: current generation number.
- ✓ y-axis: average fitness value of the best individual of current population over 30 runs.