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import numpy as np
import matplotlib.pyplot as plt
# Sphere function (fitness function)
def sphere_function(position):
    return np.sum(position**2)
# Parameters
num_particles = 30 # Number of particles
num_iterations = 100 # Number of iterations
dim = 2 # Dimensionality of the problem
bounds = [-10, 10] # Search space bounds
inertia_weight = 0.7 # w
cognitive_coefficient = 1.5 # c1
social_coefficient = 1.5 # c2
tolerance = 1e-6 # Stopping tolerance for fitness
# Initialize particles
class Particle:
    def __init__(self):
        self.position = np.random.uniform(bounds[0], bounds[1], dim) # Random
position
        self.velocity = np.random.uniform(-1, 1, dim) # Random velocity
        self.best_position = np.copy(self.position) # Personal best position
        self.best_fitness = sphere_function(self.position) # Personal best
fitness
        self.fitness = self.best fitness # Current fitness
    def update_velocity(self, global_best_position):
        r1, r2 = np.random.rand(dim), np.random.rand(dim)
        cognitive_term = cognitive_coefficient * r1 * (self.best_position -
self.position)
        social_term = social_coefficient * r2 * (global_best_position -
self.position)
        self.velocity = inertia_weight * self.velocity + cognitive_term +
social_term
    def update_position(self):
        self.position += self.velocity
        self.position = np.clip(self.position, bounds[0], bounds[1]) # Keep
within bounds
        self.fitness = sphere_function(self.position)
        if self.fitness < self.best_fitness: # Update personal best
            self.best_fitness = self.fitness
            self.best_position = np.copy(self.position)
# PSO implementation
def particle_swarm_optimization():
    particles = [Particle() for _ in range(num_particles)]
    global_best_position = particles[0].best_position
    global_best_fitness = particles[0].best_fitness
    fitness_history = []
    # Update global best from the initial population
    for particle in particles:
        if particle.best_fitness < global_best_fitness:</pre>
            global_best_fitness = particle.best_fitness
            global_best_position = np.copy(particle.best_position)
    for iteration in range(num_iterations):
        for particle in particles:
            particle.update_velocity(global_best_position)
            particle.update_position()
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# Update global best
            if particle.best_fitness < global_best_fitness:</pre>
                global_best_fitness = particle.best_fitness
                global_best_position = np.copy(particle.best_position)
        fitness_history.append(global_best_fitness) # Track global best fitness
        # Early stopping if fitness reaches tolerance
        if global_best_fitness <= tolerance:</pre>
            print(f"Converged at iteration {iteration}")
            break
    return global_best_position, global_best_fitness, fitness_history
# Run the PSO algorithm
best_position, best_fitness, fitness_history = particle_swarm_optimization()
# Print the results
print("Best position found:", best_position)
print("Best fitness achieved:", best_fitness)
# Plot fitness over iterations
plt.plot(fitness_history)
plt.title("Fitness Over Iterations (PSO on Sphere Function)")
plt.xlabel("Iteration")
plt.ylabel("Fitness (Objective Function Value)")
plt.grid()
plt.show()
```