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import numpy as np
def objective_function(x):
    return np.sum(x**2)
class GreyWolfOptimizer:
    def __init__(self, n_wolves, n_dimensions, max_iter, lower_bound, upper_bound, objective_function, convergence_threshold=1e-6):
        self.n_wolves = n_wolves # Number of wolves
        self.n_dimensions = n_dimensions # Number of dimensions
        self.max_iter = max_iter # Maximum iterations
        self.lower bound = lower bound # Lower bound of the search space
        self.upper_bound = upper_bound # Upper bound of the search space
        self.objective_function = objective_function # Objective function to optimize
        self.convergence_threshold = convergence_threshold # Convergence threshold for stopping early
        # Initialize the wolves' positions randomly within the bounds
        self.positions = np.random.uniform(self.lower_bound, self.upper_bound, (self.n_wolves, self.n_dimensions))
        self.fitness = np.array([self.objective_function(self.positions[i]) for i in range(self.n_wolves)])
        # Initialize alpha, beta, and delta wolves
        self.alpha_position = np.zeros(self.n_dimensions)
        self.beta_position = np.zeros(self.n_dimensions)
        self.delta_position = np.zeros(self.n_dimensions)
        self.alpha score = float('inf')
        self.beta_score = float('inf')
        self.delta_score = float('inf')
    def update_wolves(self, a):
        for i in range(self.n_wolves):
            # Calculate the distance of the current wolf from the alpha, beta, and delta wolves
            A1 = 2 * a * np.random.rand(self.n_dimensions) - a
            C1 = 2 * np.random.rand(self.n dimensions)
            D_alpha = np.abs(C1 * self.alpha_position - self.positions[i])
            X1 = self.alpha position - A1 * D alpha
            A2 = 2 * a * np.random.rand(self.n_dimensions) - a
            C2 = 2 * np.random.rand(self.n dimensions)
            D_beta = np.abs(C2 * self.beta_position - self.positions[i])
            X2 = self.beta_position - A2 * D_beta
            A3 = 2 * a * np.random.rand(self.n_dimensions) - a
            C3 = 2 * np.random.rand(self.n_dimensions)
            D_delta = np.abs(C3 * self.delta_position - self.positions[i])
            X3 = self.delta_position - A3 * D_delta
            # Update the position of the wolf
            self.positions[i] = (X1 + X2 + X3) / 3
            # Apply bounds to the new position
            self.positions[i] = np.clip(self.positions[i], self.lower_bound, self.upper_bound)
            # Calculate fitness for the new position
            fitness = self.objective_function(self.positions[i])
            # Update alpha, beta, and delta if necessary
            if fitness < self.alpha score:
                self.alpha_score = fitness
                self.alpha position = self.positions[i]
            elif fitness < self.beta_score:</pre>
               self.beta_score = fitness
                self.beta_position = self.positions[i]
            elif fitness < self.delta_score:</pre>
                self.delta score = fitness
                self.delta_position = self.positions[i]
    def optimize(self):
        # Iterate for the given number of iterations
        for t in range(self.max iter):
            # Calculate the coefficient a
            a = 2 - t * (2 / self.max_iter)
            # Update the wolves' positions
            self.update_wolves(a)
            # Check for convergence based on the alpha_score (best fitness)
            if self.alpha_score < self.convergence_threshold:</pre>
                print(f"Converged early at iteration {t+1} with best fitness = {self.alpha_score}")
                break # Exit early if the convergence threshold is reached
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print(f"Iteration {t+1}/{self.max_iter}: Best Fitness = {self.alpha_score}")
        # Return the best position and the corresponding fitness score
        return self.alpha_position, self.alpha_score
# Parameters for optimization
n wolves = 30 # Number of wolves
n_dimensions = 2 # Number of dimensions (problem size)
max_iter = 100  # Maximum number of iterations
lower bound = -10 # Lower bound of search space
upper_bound = 10  # Upper bound of search space
# Initialize and run the Grey Wolf Optimization
{\tt gwo} = {\tt GreyWolfOptimizer}({\tt n\_wolves}, \ {\tt n\_dimensions}, \ {\tt max\_iter}, \ {\tt lower\_bound}, \ {\tt upper\_bound}, \ {\tt objective\_function})
best_position, best_fitness = gwo.optimize()
print(f"Best Solution: {best position}")
print(f"Best Fitness: {best_fitness}")
→ Iteration 1/100: Best Fitness = 11.445334546946041
     Iteration 2/100: Best Fitness = 0.20456898464499124
     Iteration 3/100: Best Fitness = 0.11100067443115716
     Iteration 4/100: Best Fitness = 0.03246062571756299
     Iteration 5/100: Best Fitness = 0.02633145150915702
     Iteration 6/100: Best Fitness = 0.0010657909801555366
     Iteration 7/100: Best Fitness = 0.0007974787349212369
     Iteration 8/100: Best Fitness = 2.4891278605268483e-05
     Iteration 9/100: Best Fitness = 2.3822376131733303e-06
     Converged early at iteration 10 with best fitness = 1.0274802206744994e-07
     Best Solution: [-2.46266618e-05 -3.19595916e-04]
     Best Fitness: 1.0274802206744994e-07
```