Lab-5-Grey Wolf Optimizer (GWO)

Code:

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import numpy as np
# Define the Grey Wolf Optimizer (GWO) algorithm
class GreyWolfOptimizer:
  def init (self, func, dim, num wolves, max iter, lb, ub):
     :param func: The objective function to optimize (should take a numpy array as input)
     :param dim: The dimension of the search space
     :param num wolves: The number of wolves (population size)
     :param max iter: The maximum number of iterations
     :param lb: Lower bound of the search space
     :param ub: Upper bound of the search space
     self.func = func
     self.dim = dim
     self.num wolves = num wolves
     self.max iter = max iter
     self.lb = lb
     self.ub = ub
     self.alpha pos = np.zeros(dim)
     self.beta pos = np.zeros(dim)
     self.delta pos = np.zeros(dim)
     self.alpha score = float("inf")
     self.beta score = float("inf")
     self.delta score = float("inf")
    # Initialize the positions of the wolves randomly
     self.positions = np.random.rand(self.num wolves, self.dim) * (self.ub - self.lb) + self.lb
  def fitness(self, position):
     """Calculate the fitness value of a wolf"""
    return self.func(position)
  def update position(self, wolf, alpha pos, beta pos, delta pos, a, A, C):
     """Update the position of a wolf based on the positions of alpha, beta, and delta wolves"""
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# Update for alpha, beta, and delta wolves, using the respective A and C values
     D alpha = np.abs(C[0] * alpha pos - wolf)
     D beta = np.abs(C[1] * beta pos - wolf)
     D delta = np.abs(C[2] * delta pos - wolf)
     # Updated positions
     X1 = alpha pos - A[0] * D_alpha
     X2 = beta pos - A[1] * D beta
     X3 = delta pos - A[2] * D delta
     # New position is the average of the three updated positions
     new position = (X1 + X2 + X3) / 3
     return new position
  def optimize(self):
     """Run the optimization process"""
     for t in range(self.max iter):
       a = 2 - t * (2 / self.max iter) # Declining a over iterations
       A = 2 * a * np.random.rand(self.num wolves, self.dim) - a # Random vector for wolves'
A values
       C = 2 * np.random.rand(self.num wolves, self.dim) # Random vector for wolves' C
values
       for i in range(self.num wolves):
          # Evaluate fitness of each wolf
          fitness_value = self.fitness(self.positions[i])
         # Update alpha, beta, and delta wolves
         if fitness value < self.alpha score:
            self.alpha score = fitness value
            self.alpha pos = self.positions[i]
          elif fitness value < self.beta score:
            self.beta score = fitness value
            self.beta pos = self.positions[i]
          elif fitness value < self.delta score:
            self.delta score = fitness value
            self.delta pos = self.positions[i]
       # Update the positions of all wolves
       for i in range(self.num wolves):
```

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self.positions[i] = self.update position(self.positions[i], self.alpha_pos, self.beta_pos,
self.delta pos, a, A[i], C[i])
       # Optionally, print progress
       if t \% 1 == 0: # Adjust the interval for printing
         print(f"Iteration {t}/{self.max iter}, Best fitness: {self.alpha score}")
     return self.alpha pos, self.alpha score
# Example usage:
def objective function(x):
  """A simple objective function: Sphere function"""
  return np.sum(x**2)
# User input for parameters
dim = int(input("Enter the number of dimensions: "))
num wolves = int(input("Enter the number of wolves: "))
max iter = int(input("Enter the maximum number of iterations: "))
lb = float(input("Enter the lower bound of the search space: "))
ub = float(input("Enter the upper bound of the search space: "))
# Initialize and run the GWO algorithm
optimizer = GreyWolfOptimizer(func=objective function, dim=dim, num wolves=num wolves,
max iter=max iter, lb=lb, ub=ub)
best position, best score = optimizer.optimize()
print(f"\nBest Position: {best position}")
print(f"Best Score: {best score}")
```

Output:

```
Free the number of dimensions: 3
    Enter the number of wolves: 35
    Enter the maximum number of iterations: 9
    Enter the lower bound of the search space: -5
    Enter the upper bound of the search space: 5
    Iteration 0/9, Best fitness: 5.1261148298595725
    Iteration 1/9, Best fitness: 1.7348203391018997
    Iteration 2/9, Best fitness: 1.350210015340434
    Iteration 3/9, Best fitness: 0.30549540443394696
    Iteration 4/9, Best fitness: 0.19136125125293585
    Iteration 5/9, Best fitness: 0.15841368969575084
    Iteration 6/9, Best fitness: 0.11811841118705839
    Iteration 7/9, Best fitness: 0.10128658372111207
    Iteration 8/9, Best fitness: 0.091277213497343
    Best Position: [-0.21950171 0.16166378 0.14070943]
    Best Score: 0.091277213497343
```