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28-11-24
Grey Wolf Algorithm
Algorithm:-
import numpy as np
# Sphere function: minimize f(x) = sum(x i^2)
def sphere function(position):
  return sum(x^{**}2 \text{ for } x \text{ in position})
# Grev Wolf Optimizer
def grey_wolf_optimizer(dim, wolves_count, iterations, lb, ub):
  GWO implementation to minimize a function.
  dim: Number of dimensions.
  wolves count: Number of wolves.
  iterations: Number of iterations.
  lb: Lower bound of search space.
  ub: Upper bound of search space.
  # Initialize the positions of wolves randomly
  wolves_positions = np.random.uniform(lb, ub, (wolves_count, dim))
  # Initialize alpha, beta, and delta (best, second-best, and third-best wolves)
  alpha, beta, delta = np.zeros(dim), np.zeros(dim), np.zeros(dim)
  alpha score, beta score, delta score = float("inf"), float("inf"), float("inf")
  # Iterate through generations
  for t in range(iterations):
     for i in range(wolves_count):
       fitness = sphere function(wolves positions[i])
       # Update alpha, beta, and delta based on fitness
       if fitness < alpha score:
          alpha_score, beta_score, delta_score = fitness, alpha_score, beta_score
          alpha, beta, delta = wolves positions[i], alpha, beta
       elif fitness < beta score:
          beta score, delta score = fitness, beta score
          beta, delta = wolves_positions[i], beta
       elif fitness < delta score:
          delta score = fitness
          delta = wolves_positions[i]
     # Update positions of wolves
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for j in range(dim):
          r1, r2 = np.random.rand(), np.random.rand()
          A1, C1 = 2 * r1 - 1, 2 * r2
          D_alpha = abs(C1 * alpha[j] - wolves_positions[i][j])
          X1 = alpha[j] - A1 * D alpha
          r1, r2 = np.random.rand(), np.random.rand()
          A2, C2 = 2 * r1 - 1, 2 * r2
          D beta = abs(C2 * beta[i] - wolves positions[i][j])
          X2 = beta[i] - A2 * D beta
          r1, r2 = np.random.rand(), np.random.rand()
          A3, C3 = 2 * r1 - 1, 2 * r2
          D delta = abs(C3 * delta[i] - wolves positions[i][j])
          X3 = delta[j] - A3 * D_delta
          # Update the wolf's position
          wolves_positions[i][j] = (X1 + X2 + X3) / 3
          # Boundary control
          wolves_positions[i][j] = np.clip(wolves_positions[i][j], lb, ub)
  print("\nOptimization by Devanshi Slathia: Grey Wolf Optimizer completed!")
  return alpha, alpha score
# Main function
def main():
  print("Grey Wolf Optimizer")
  dim = int(input("Enter the number of dimensions (e.g., 2, 3): "))
  wolves count = int(input("Enter the number of wolves (e.g., 5, 10): "))
  iterations = int(input("Enter the number of iterations (e.g., 50): "))
  lb, ub = map(float, input("Enter the lower and upper bounds of the search space (e.g., -10
10): ").split())
  # Run GWO
  best_position, best_score = grey_wolf_optimizer(dim, wolves_count, iterations, lb, ub)
  print("\nBest Position Found:", best_position)
  print("Best Score (Minimum Value of Function):", best score)
if __name__ == "__main__":
  main()
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for i in range(wolves count):

Output:-

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Grey Wolf Optimizer
Enter the number of dimensions (e.g., 2, 3): 2
Enter the number of wolves (e.g., 5, 10): 5
Enter the number of iterations (e.g., 50): 50
Enter the lower and upper bounds of the search space (e.g., -10 10): -10 10

Optimization by Devanshi Slathia: Grey Wolf Optimizer completed!

Best Position Found: [-0.02475322 0.0023002 ]
Best Score (Minimum Value of Function): 0.0007324328554199868
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