

Lab 5

Cuckoo Search Optimization

17/10/25 LAB - V
 CUCKOO SEARCH OPTIMIZATION

Pseudocode:
 Initialize value is set to no. of schedules n , probability $E(0,1)$ and maximum iterations MaxIter
 Iteration counter $t = 0$
 for $i = 1$ to n do
 generate an initial schedule S_i randomly
 evaluate the fitness func $f(S_i)$ + $f(S_i) = \text{total completion time of schedule } S_i$
 End for
 while $t < \text{MaxIter}$ do
 Generate a new schedule S'_i from S_i using Levy Flight (small random changes in task order)
 Evaluate the fitness $f(S'_i)$
 Randomly select a schedule S_j among n schedules
 If $f(S'_i) < f(S_j)$ then
 Replace S_j with new schedule S'_i
 End if
 Abandon Δ a fraction P_a of worst schedules and generate new schedules randomly
 Keep the best schedules found so far
 Rank all schedules by fitness & update the current best
 Increment iteration $t = t + 1$
 End while
 Output the best schedule S_{best}
 Output:
 Iteration 1 : Best Fitness = 229
 Iteration 2 : Best Fitness = 228
 Iteration 100 : Best Fitness = 211
 Best schedule : [6, 1, 3, 5, 0, 4, 7, 2]
 Best Fitness : 211

Code:

```
import numpy as np
import math
print('Shreya Raj 1BM23CS317')

def objective_function(x):
    return np.sum(x**2)

def initialize_nests(num_nests, dim, lower_bound, upper_bound):
    return np.random.uniform(lower_bound, upper_bound, size=(num_nests, dim))

def levy_flight(Lambda, size):
    sigma = (math.gamma(1 + Lambda) * math.sin(math.pi * Lambda / 2) /
             (math.gamma((1 + Lambda) / 2) * Lambda * 2**((Lambda - 1) / 2))) ** (1 / Lambda)
    u = np.random.randn(*size) * sigma
    v = np.random.randn(*size)
    step = u / np.abs(v) ** (1 / Lambda)
    return step

def cuckoo_search(num_nests=25, dim=2, lower_bound=-10, upper_bound=10,
                  pa=0.25, max_iter=100):

    nests = initialize_nests(num_nests, dim, lower_bound, upper_bound)
    fitness = np.apply_along_axis(objective_function, 1, nests)

    best_nest = nests[np.argmin(fitness)].copy()
    best_fitness = np.min(fitness)
```

```

for t in range(max_iter):

    new_nests = nests + 0.01 * levy_flight(1.5, nests.shape) * (nests - best_nest)
    new_nests = np.clip(new_nests, lower_bound, upper_bound)

    new_fitness = np.apply_along_axis(objective_function, 1, new_nests)

    mask = new_fitness < fitness
    nests[mask] = new_nests[mask]
    fitness[mask] = new_fitness[mask]

    rand = np.random.rand(num_nests, dim)
    new_nests = np.where(rand > pa, nests,
                         initialize_nests(num_nests, dim, lower_bound, upper_bound))

    new_fitness = np.apply_along_axis(objective_function, 1, new_nests)

    mask = new_fitness < fitness
    nests[mask] = new_nests[mask]
    fitness[mask] = new_fitness[mask]

if np.min(fitness) < best_fitness:
    best_nest = nests[np.argmin(fitness)].copy()
    best_fitness = np.min(fitness)

print(f'Iteration {t+1}/{max_iter} | Best Fitness: {best_fitness:.6f}')

return best_nest, best_fitness

```

```

best_solution, best_value = cuckoo_search()
print("\nBest solution found:", best_solution)
print("Best fitness value:", best_value)

```

Output:

*** Shreya Raj 1BM23CS317	
Iteration 1/100 Best Fitness: 14.471185	Iteration 80/100 Best Fitness: 0.000133
Iteration 2/100 Best Fitness: 14.471185	Iteration 81/100 Best Fitness: 0.000133
Iteration 3/100 Best Fitness: 14.471185	Iteration 82/100 Best Fitness: 0.000133
Iteration 4/100 Best Fitness: 0.894298	Iteration 83/100 Best Fitness: 0.000133
Iteration 5/100 Best Fitness: 0.894298	Iteration 84/100 Best Fitness: 0.000133
Iteration 6/100 Best Fitness: 0.894298	Iteration 85/100 Best Fitness: 0.000133
Iteration 7/100 Best Fitness: 0.894298	Iteration 86/100 Best Fitness: 0.000133
Iteration 8/100 Best Fitness: 0.894298	Iteration 87/100 Best Fitness: 0.000133
Iteration 9/100 Best Fitness: 0.564269	Iteration 88/100 Best Fitness: 0.000133
Iteration 10/100 Best Fitness: 0.564269	Iteration 89/100 Best Fitness: 0.000133
Iteration 11/100 Best Fitness: 0.564269	Iteration 90/100 Best Fitness: 0.000133
Iteration 12/100 Best Fitness: 0.457079	Iteration 91/100 Best Fitness: 0.000133
Iteration 13/100 Best Fitness: 0.457079	Iteration 92/100 Best Fitness: 0.000133
Iteration 14/100 Best Fitness: 0.457079	Iteration 93/100 Best Fitness: 0.000133
Iteration 15/100 Best Fitness: 0.457079	Iteration 94/100 Best Fitness: 0.000133
Iteration 16/100 Best Fitness: 0.457079	Iteration 95/100 Best Fitness: 0.000133
Iteration 17/100 Best Fitness: 0.457079	Iteration 96/100 Best Fitness: 0.000133
Iteration 18/100 Best Fitness: 0.457079	Iteration 97/100 Best Fitness: 0.000133
Iteration 19/100 Best Fitness: 0.457079	Iteration 98/100 Best Fitness: 0.000133
Iteration 20/100 Best Fitness: 0.457079	Iteration 99/100 Best Fitness: 0.000133
Iteration 21/100 Best Fitness: 0.457079	Iteration 100/100 Best Fitness: 0.000133
Iteration 22/100 Best Fitness: 0.457079	
Iteration 23/100 Best Fitness: 0.457079	
Iteration 24/100 Best Fitness: 0.457079	
Iteration 25/100 Best Fitness: 0.457079	
Iteration 26/100 Best Fitness: 0.457079	
Iteration 27/100 Best Fitness: 0.457079	

Best solution found: [-0.01150773 0.00076445]
 Best fitness value: 0.0001330122491796386