

#### **Program 4 - Cuckoo Search for Optimizing a Mathematical Function :**

Design and implement a Cuckoo Search (CS) algorithm in Python to find the global maximum of a given mathematical function. The CS algorithm will mimic the brood parasitism behavior of cuckoo birds, using Lévy flights to explore the solution space and discover the optimal solution.

Algorithm :

1. **Define the Problem:** Use  $f(x) = -x^2 + 7x + 10$  as the objective function.
2. **Initialize Parameters:**
  - Number of nests  $NN$
  - Probability of discovery  $pap\_apa$
  - Number of iterations  $III$
3. **Initialize Population:** Randomly generate initial positions of nests within the search space.
4. **Evaluate Fitness:**
  - Calculate  $f(x)$  for each nest to evaluate its fitness.
5. **Generate New Solutions:**
  - Use Lévy flights to create new solutions for each nest.
6. **Abandon Worst Nests:**
  - With probability  $pap\_apa$ , replace the worst-performing nests with random solutions.
7. **Iterate:**
  - Repeat the evaluation, updating, and replacement process for  $III$  iterations.
8. **Output the Best Solution:**
  - Track and output the nest position with the highest fitness.

code:

```
import numpy as np

# Define the fitness function
def fitness_function(x):
    return -x**2 + 7*x + 10

# Lévy flight function
def levy_flight(Lambda=1.5):
    sigma = (np.math.gamma(1 + Lambda) * np.sin(np.pi * Lambda / 2) /
              (np.math.gamma((1 + Lambda) / 2) * Lambda * 2**((Lambda - 1) /
```

```

2)))**(1 / Lambda)

    u = np.random.normal(0, sigma, 1)
    v = np.random.normal(0, 1, 1)
    step = u / abs(v)**(1 / Lambda)
    return step

# Cuckoo Search implementation
def cuckoo_search():
    # Input parameters
    num_nests = int(input("Enter number of nests: "))
    probab_discovery = float(input("Enter probability of discovery (p_a): "))
    num_iterations = int(input("Enter number of iterations: "))
    x_min = float(input("Enter minimum value of x: "))
    x_max = float(input("Enter maximum value of x: "))

    # Initialize nests
    nests = np.random.uniform(x_min, x_max, num_nests)
    fitness = np.array([fitness_function(x) for x in nests])

    best_nest = nests[np.argmax(fitness)]
    best_fitness = max(fitness)

    for _ in range(num_iterations):
        # Generate new solutions using Lévy flights
        for i in range(num_nests):
            step_size = levy_flight()
            nests[i] += step_size
            nests[i] = np.clip(nests[i], x_min, x_max)

        # Evaluate new fitness
        new_fitness = np.array([fitness_function(x) for x in nests])

        # Update nests with better solutions
        for i in range(num_nests):
            if new_fitness[i] > fitness[i]:
                fitness[i] = new_fitness[i]
                nests[i] = nests[i]

        # Abandon a fraction of the worst nests
        abandon_count = int(probab_discovery * num_nests)
        worst_indices = np.argsort(fitness)[:abandon_count]

```

```

    for idx in worst_indices:
        nests[idx] = np.random.uniform(x_min, x_max)
        fitness[idx] = fitness_function(nests[idx])

    # Update best solution
    current_best_idx = np.argmax(fitness)
    if fitness[current_best_idx] > best_fitness:
        best_fitness = fitness[current_best_idx]
        best_nest = nests[current_best_idx]

    return best_nest, best_fitness

# Run the Cuckoo Search algorithm
best_solution, best_value = cuckoo_search()
print(f"Best Solution: {best_solution}, Fitness: {best_value}")
print("Name-pooja Gaikwad(1BM22CS194)")

```

Output:

```

Enter number of nests: 10
Enter probability of discovery (p_a): 0.25
Enter number of iterations: 100
Enter minimum value of x: -10
Enter maximum value of x: 10
Best Solution: 3.506038131134606, Fitness: 22.249963540972402
Name-pooja Gaikwad(1BM22CS194)

```

Observation:

## ④ Cuckoo Search Algo

1. Start

2. Initialize parameters :

 $n \leftarrow$  no. of nests $\text{max-iter} \leftarrow$  max. no. of iterinitialise population of  $n$  nests3. nests = Random nests ( $n$ )

fitness = evaluate fitness (nests)

While termination-cond not met :

for each nest in nests :

new-nest = Generate sol (nest)

[Generate new sol. using Levy flight]

 $\text{new-nests} = \text{nest} + \text{step-size} * \text{Levy}(\beta)$  $\text{new-fitness} = \text{evaluate fitness}(\text{new-nests})$ if  $\text{new-fitness} > \text{fitness}$  $\text{nests}(\text{nest}) = \text{new-nests}(\text{nest})$  $\text{fitness}(\text{nest}) = \text{new-fitness}$ 

4. Replace worst nests

nests = Replace (nests, fitness)

best-nest = Best sol (nest, fitness)