

A
Synopsis
on

Design and Implementation of Cheetah Optimizer Algorithm For Solving Optimization Problem

Submitted in partial fulfillment for

Final Year Project

Submitted by

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1. Name of students with PRN No :

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2. Title of Project :

Design and Implementation of Cheetah Optimizer Algorithm For Solving Optimization Problem

3. Introduction:

Recently, solving optimization problems have become a challenging and exciting topic In most research area. Decision-making problems that are overgrowing Can be defined as optimization problems. An optimization problem includes one or more objective functions, decision variables, and constraints to be minimized or maximized. Many deterministic approaches, such as linear programming, Newton methods, quadratic programming, dynamic programming, simplex method, gradient method, etc., are the well-known Classical methods to solve optimization problems These algorithms robustly Result in the same solution for a given optimization problem with an identical initial starting point. Although such techniques can find optimal solutions in a reasonable time, they need the objective function and constraints to be convex and derivable. These cause the deterministic algorithms to fall into locally optimal solutions, which is the main shortcoming of such methods in solving real-world problems. This defect becomes more prominent as the dimension Of the problem increases. Therefore, stochastic methods for dealing with it have been developed. These algorithms intrinsically neglect the characteristics of the objective functions and constraints, so they treat the problem as a black box. Another advantage of the most metaheuristic algorithms is their simplicity

4. Literature review/Existing Work :

Metaheuristic algorithms can be categorized into two main classes, single-solutionbased and multiple-solution-based (or population-based). Te mostpopular single-solution-based metaheuristic algorithm is simulated annealing. Tis algorithm's process starts with a random candidate solution (a population) and then moves and improves it in the promising search space in aniterative manner to fnd the superior solution. However, multiple-solution-based algorithms implement more than one random candidate solution to enhance the speed and the chance to avoid local optima entrapment.

5. Problem statement :

Design and Implementation of Cheetah Optimizer Algorithm For Solving Optimization Problem

6. Objectives :

- i. To design Cheetah Optimization Algorithm(COA).
- ii. To implement COA for Optimization
- iii. To compare and analyze the performance of COA with other algorithm.

7. Methodology :

1. Start
2. Prepare for new hunt.
3. Initialize the population size and dimensions.
4. Generate the initial population of cheetahs and fitness of cheetahs
5. Start searching and move to next and update the solutions of members and lead
6. Search strategy equation :

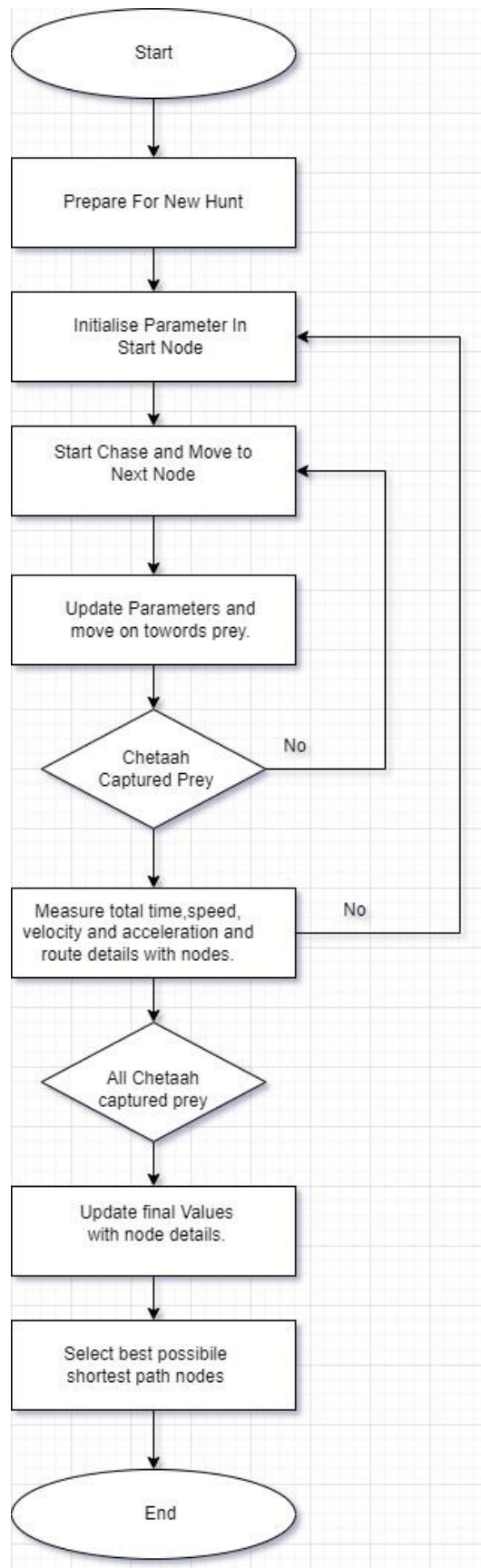
$$X_{i,j}^{t+1} = X_{i,j}^t + \hat{r}_{i,j}^{-1} \cdot \alpha_{i,j}^t$$

7. Sit-and-wait strategy equation :

$$X_{i,j}^{t+1} = X_{i,j}^t$$

8. Attack equation :

$$X_{i,j}^{t+1} = X_{B,j}^t + \check{r}_{i,j} \cdot \beta_{i,j}^t$$



8. Technology Required:

Language : Python

Libraries : Matplotlib(For Data Visualization)

9. References:

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3. Lu, T. & Liu, S.-T. Fuzzy nonlinear programming approach to the evaluation of manufacturing processes. Eng. Appl. Artif. Intell. 72, 183–189 (2018).
4. Koc, I., Atay, Y. & Babaoglu, I. Discrete tree seed algorithm for urban land readjustment. Eng. Appl. Artif. Intell. 112, 104783 (2022).
5. Introduction to Stochastic Search and Optimization: Estimation, Simulation, and Control Vol. 65 (Wiley, 2005).

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