

Unmanned Aerial Vehicle Path Planning using Bat Algorithm



S. Aicevarya Devi, C. Vijayalakshmi

Abstract: *Unmanned Aerial Vehicles (UAV) was introduced after World War II. In 1980's UAV consider as important weapon system. Initially UAV needs initial position and target position. In this paper bat algorithm is proposed with mixed objective constraints which helps in directing the UAV. The process is initialized by generating the initial population of bat. Then by updating the population size and generation of bat the fitness value with minimum frequency is found that helps to avoid convergence among UAV. Finally the evaluation which gives minimum frequency is considered as optimal solution.*

Keywords: *Bat algorithm, Unmanned Aerial Vehicle, Population, Path planning, Frequency, Position.*

I. INTRODUCTION

Unmanned Aerial Vehicle (UAV) is shortly called military aircraft. This mainly used in war and during disaster situation and also helps in weather monitoring [1,2]. UAV system important requirements are autonomous exploration and direction [3]. Generally the initial and destination position is needed to plan the path[4] and also it helps to avoid crash and obstacles. Various algorithm are used to solve UAV problem[5,6] such as particle swarm optimization algorithm [7], ant colony optimization algorithm [8], pigeon-inspired optimization algorithm (PIO) [9], fruit fly optimization algorithm (FOA) [10].

The PSO is otherwise called as random search algorithm which helps in solving UAV path planning [11]. By updating the parameters in hybrid PSO, the accuracy rate in UAV path planning is defined. The ACO algorithm is based on ant colony behaviour to optimize path planning problems [12]. ACO commonly needs number of calculations. The direction and chaotic theory are practiced in ACO for improvement. ADD ON algorithm also applied to solve multi objective problems [13]. PIO is swarm intelligence optimization algorithm. SIOA is motivated by pigeon behaviour. This PIO has two operators one is design and circuit operator and second is memorial operator [14]. PIO WITH QUANTUM BEHAVED

THEORY PERFORM IN UAV PATH PLANNING. THIS PAPER INCLUDES THE UAV PATH PLANNING PROBLEM MANAGEMENT AND PROPOSED BAT ALGORITHM IN UAV PATH PLANNING. with quantum behaved theory perform in UAV path planning.

This paper includes the UAV path planning problem management and proposed bat algorithm in UAV path planning.

II. MATHEMATICAL MODEL FORMULATION

- α Euler angles
- β Angular angles
- X Position
- Y Linear velocity
- x_t, y_t, z_t Position coordinates
- r_t Radius
- k_t Threat level
- J Cost function
- J_i Path length
- J_a Altitude
- J_c Threat cost
- l_i Length above sea level
- h_i Height above sea level
- Q Frequency of bat
- Q_i Current frequency of bat
- Q_{min} Minimum frequency of bat
- Q_{max} Maximum frequency of bat
- S Velocity of bat
- s_i^t Current Velocity of bat
- s_i^{t-1} Previous Velocity of bat
- g Position of bat
- g_i Current position of bat
- g_i^{t-1} Previous position of bat
- N_iter Number of evaluation
- n Population size
- A Loudness
- d Dimension
- C Random value

III. UNMANNED AERIAL VEHICLE

As per the record in July 1849 UAV is used for war fighting[15]. The UAV arose because of U.S Airforce efforts[16]. Unmanned Aerial Vehicle need short and secure path planning from initial to destination position. All obstacles and danger zone are taken into account and also

Revised Manuscript Received on May 15, 2020.

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altitude changes and fly speed are needed.

UAV dynamic and UAV threats are constraints with mixed objective optimization problem used for path planning. Consider the UAVs satisfy upcoming dynamics.

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix} = \begin{bmatrix} v \cos \phi - w \sin \phi \\ u + (w + \cos \phi + v \sin \phi) \tan \theta \\ \frac{1}{\cos \theta} (w \cos \phi + v \sin \phi) \end{bmatrix}$$

$$\begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} a \cos \phi \cos \psi + b(\sin \phi \sin \theta \cos \psi - \cos \phi \sin \psi) + c(\sin \phi \sin \psi + \cos \phi \sin \theta \cos \psi) \\ a \cos \phi \cos \psi + b(\sin \phi \sin \theta \cos \psi + \cos \phi \sin \psi) + c(-\sin \phi \sin \psi + \cos \phi \sin \theta \cos \psi) \\ a \sin \theta - b \sin \theta \cos \theta - c \cos \phi \cos \theta \end{bmatrix}$$

Then $\alpha = [xyz]$ are Euler angles,
 $\beta = [uvw]$ are Zigzag angles,
 $X = [pqr]$ are posture,
 $Y = [abc]$ are linear velocity.

Next to calculate the threats in the circumstances the radius r_t , position coordinates (x_t, y_t, z_t) and threat level k_t . It includes length, time and security element. The successive optimization problem helps to solve UAV path planning.

$$\min I = i_1 I_l + i_2 I_k + i_3 I_m i_1 + i_2 + i_3 = 1$$

where I_m, I_l and I_k are the cost function which indicates altitude cost, path length cost and threat cost.

The altitude cost is measured by $I_m = \sum_{j=1}^n h_j$

The path length cost is measured by $I_l = \sum_{j=1}^n l_j^2$

where h_j and l_j are height and length over the sea level.

The threat cost is measured by 4 points and these points have connecting edges between two points.

$$I_k = \frac{A_{mn}}{4} \sum_{t=1}^{N_t} k_t \left(\frac{1}{f_{0.1,t}^4} + \frac{1}{f_{0.3,t}^4} + \frac{1}{f_{0.5,t}^4} + \frac{1}{f_{0.7,t}^4} \right)$$

where $f_{(0.1,t)}$ is the breadth between 1/10 and t – threat core.

IV. BAT ALGORITHM

Bat algorithm was first proposed by Xin-She Yang [17][18]. This algorithm is based on bats behaviour[19]. The main behaviour of bat is echolocation that helps to predict the difference between prey and obstacles. It has 3 rules.

Rule 1: Echolocation used to sense the distance and helps to identify difference between food and prey.

Rule 2: Bat fly at random with velocity v_i at position x_i with different frequency ranges $f[\min, \max]$ varying wavelength λ and loudness A to look for prey. It naturally contain the emitted pulses.

Rule 3: Loudness differ from large value to minimum constant value $[A_0 \text{ to } A_{\min}]$.

Frequency ranges between maximum to minimum or it is called as frequency-tuning algorithm. Every bat correlate with velocity s_i^t and position g_i^t in search space at each iteration t with respect to frequency Q_i . So in every iteration the frequency, the position and the speed updating is needed.

TO CALCULATE THE FREQUENCY OF BAT

$$Q_i = Q_{\min} + (Q_{\max} - Q_{\min})C$$

TO CALCULATE VELOCITY OF BAT

$$s_i^t = s_i^{(t-1)} + (g_i^{(t-1)} - g_*)Q_i$$

TO CALCULATE POSITION OF BAT

$$g_i = g_i^{(t-1)} + s_i^t$$

A. Procedure to Plan Route for UAV Using Bat Algorithm

Step 1: Initialize bat population, its position, velocities and frequency of all bats.

Step 2: Initialize Echolocation of all bats.

Step 3: Evaluating micro-bats in initial position, the calculated solutions are stored in resource log.

Step 4: Arrange the current population in downward order.

Step 5: Develop candidate micro-bats.

Step 6: Irregular flying by adjusting new position, velocity and direction.

Step 7: Inconstantly choose a micro-bat from population.

Step 8: Evaluate micro-bat by fitness function.

Step 9: For upcoming position echolocation parameters is updated.

Step 10: Select the fittest micro-bat for next iteration.

Step 11: Until reach the expectation criterion rerun the process from step 4 to step 10.

B. General Algorithm

{initialize bat population, its position, velocity and frequency}

Initialize population of bat $x_i (i = 1, 2, \dots, n)$ and s_i

Determine pulse frequency f_i at x_i

Define heart rate r_i and loudness A_i

while($t <$ Maximum number of repetition)

 Provoke new explanation by modifying frequency and updating velocities and position

 if(rand r)

 Single out the solution amid finest solution

 Generate local



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solution over the elected best
    solution
end if
Develop new solution randomly
Test the fitness of new solution w.r.t  $r$  and  $A$ 
    Obtain the new solution
    Extend  $r_i$  and cut down  $A_i$ 
end if
Grade the bats and find the current  $x_*$ 
end while
To calculate the minimum frequency MATLAB software is
used and the number of evaluation taken from 100 to 4000
under dimension one.
    
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V. RESULT

From the above calculation the number of bats are evaluated from 100 to 4000 and in 1000th evaluation best fitness value is found with the minimum frequency which gives best path for UAV. In this paper one dimension calculation is manipulated even it can extend to n dimensions whereas this extension of dimension is not applicable in other algorithms.

VI. CONCLUSION

To forbid quick loss and numerical incomplete concurrence at few native optimum, a replacement community change approach was bestowed concerning these problems. Comparing with other algorithms bat algorithm is flexible, easy to understand and gives best solution in minimum time.

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| Number of Evaluation | BEST | fmin |
|----------------------|-------------|------------|
| 100 | -0.1115 | 0.012433 |
| 200 | -1.3354e-05 | 1.7834e-10 |
| 300 | -6.1153e-06 | 3.7396e-11 |
| 400 | -7.2749e-06 | 5.2923e-11 |
| 500 | 2.2173e-06 | 4.9165e-12 |
| 600 | 1.2309e-06 | 1.5152e-12 |
| 700 | 1.8205e-06 | 3.3143e-12 |
| 800 | -1.3192e-06 | 1.7403e-12 |
| 900 | -3.883e-06 | 1.5077e-11 |
| 1000 | 1.1005e-07 | 1.211e-14 |
| 1100 | 4.0589e-06 | 1.6475e-11 |
| 1200 | -2.9884e-05 | 8.9304e-10 |
| 1300 | -2.4258e-06 | 5.8844e-12 |
| 1400 | 7.8483e-07 | 6.1596e-13 |
| 1500 | -3.0766e-06 | 9.4657e-12 |
| 1600 | -2.0164e-06 | 4.066e-12 |
| 1700 | -7.0061e-07 | 4.9085e-13 |
| 1800 | -1.3412e-07 | 1.7989e-14 |
| 1900 | -1.0088e-06 | 1.0176e-12 |
| 2000 | -3.7082e-07 | 1.3751e-13 |
| 2100 | -4.6404e-07 | 2.1533e-13 |
| 2200 | -2.2475e-06 | 5.0513e-12 |
| 2300 | 2.4211e-06 | 5.8617e-12 |
| 2400 | 1.5787e-06 | 2.4922e-12 |
| 2500 | 2.0714e-06 | 4.2905e-12 |
| 2600 | 2.3484e-06 | 5.5151e-12 |
| 2700 | 1.4283e-06 | 2.0402e-12 |
| 2800 | 8.1923e-07 | 6.7114e-13 |
| 2900 | -8.4173e-07 | 7.0851e-13 |
| 3000 | 4.5351e-07 | 2.0567e-13 |
| 3100 | -7.0359e-07 | 4.9503e-13 |
| 3200 | 7.0965e-07 | 5.036e-13 |
| 3300 | 2.3062e-07 | 5.3184e-14 |
| 3400 | -2.0333e-06 | 4.1343e-12 |
| 3500 | 2.06e-07 | 4.2436e-14 |
| 3600 | -9.9604e-07 | 9.921e-13 |
| 3700 | 1.642e-07 | 2.6961e-14 |
| 3800 | -2.9269e-07 | 8.5668e-14 |
| Number of Evaluation | BEST | fmin |
| 3900 | -3.4921e-07 | 1.2195e-13 |
| 4000 | -5.0015e-07 | 2.5015e-13 |



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