Efficient path planning for UAV formation via comprehensively improved PSO

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Table of contents

01.	02 .	03.
Problem	Environment	PSO

04.	05.	06 .
Improved PSO	Algorithm	Performance

.

01. Problem







UAV Formation

- Military purposes
- Entertainment purposes

Goal





- Plan routes for a formation of UAVs
- Blue dost starting positions
- Red dots final positions

Goal





- Plan routes for a formation of UAVs
- Blue dost starting positions
- Red dots final positions

O2. Environment



Environment





• 3D space divided to waypoints

• Threats:

Mountains

Radars



Environment - Mountains



Environment - Radars



$$T_k = \left(x_k^r, y_k^r, z_k^r, R_k\right)$$

• intensity δ

Objective function



$f = f_L + f_T + f_R + f_C$

- path cost
- terrain cost
- radar cost
- collision cost



$$f_{L} = \frac{\sum_{i=2}^{N_{w}} \sqrt{(x_{i} - x_{i-1})^{2} + (y_{i} - y_{i-1})^{2} + (z_{i} - z_{i-1})^{2}}}{\sqrt{(x_{N_{w}} - x_{1})^{2} + (y_{N_{w}} - y_{1})^{2} + (z_{N_{w}} - z_{1})^{2}}}$$

Objective function - Terrain cost

$$f_T = \sum_{i=2}^{N_w} \sum_{j=1}^n A_{i,j} \quad \text{with} \quad A_{i,j} = \begin{cases} 1, & \text{if } z_{i,j} \leq z_{k,i,j}^m \\ 0, & \text{otherwise} \end{cases}$$

• segments between waypoints divided into n parts



Objective function - Radar cost

$$f_{R} = \sum_{i=2}^{N_{w}} \sum_{j=1}^{n} B_{i,j} \quad \text{with} \quad B_{i,j} = \begin{cases} \left(\delta/D_{i,j}\right)^{4}, & \text{if } D_{i,j} \leq R_{k} \\ 0, & \text{otherwise} \end{cases}$$
$$D_{i,j} = \sqrt{\left(x_{i,j} - x_{k}^{r}\right)^{2} + \left(y_{i,j} - y_{k}^{r}\right)^{2} + \left(z_{i,j} - z_{k}^{r}\right)^{2}}$$

• segments between waypoints divided into n parts

Objective function - Collision cost

$$f_{C} = \sum_{i=2}^{N_{w}} \sum_{j=1}^{n} C_{i,j} \quad \text{with} \quad C_{i,j} = \begin{cases} 1, & \text{if } d_{i,j} \leq \overline{d} \\ 0, & \text{otherwise} \end{cases}$$

- segments between waypoints divided into n parts
- d_{i,j} = shortest distance to other path points of different UAV
 d safe distance between UAV

03. PS0



Particle Swarm Optimization



- ······
- population based
- swarm of candidate solutions
- each candidate has its velocity
- velocity = search direction



Particle Swarm Optimization



- personal best positions
- global best position

Particle Swarm Optimization Position and Velocity Updates

$$\begin{cases} \mathbf{v}_{i,j} \left(t+1\right) = w \cdot \mathbf{v}_{i,j} \left(t\right) + c_1 \cdot r_1 \cdot \left(\mathbf{p}_{i,j,best} \left(t\right) - \mathbf{p}_{i,j} \left(t\right)\right) \\ + c_2 \cdot r_2 \cdot \left(\mathbf{g}_{j,best} \left(t\right) - \mathbf{p}_{i,j} \left(t\right)\right) \\ \mathbf{p}_{i,j} \left(t+1\right) = \mathbf{p}_{i,j} \left(t\right) + \mathbf{v}_{i,j} \left(t+1\right) \end{cases}$$

$$w = w_{\max} - rac{(w_{\max} - w_{\min})t}{T}$$

- w inertia weight
- c_1, c_2 acceleration coefficients
- r_1, r_2 random numbers

Particle Swarm Optimization in 3D pathfinding



- particle = one whole path from start to finish
- D number of waypoints within particle

$$\mathbf{p}_{i} = \left(\mathbf{p}_{i,1}, \ldots, \mathbf{p}_{i,D}\right)^{T} = \left(\left(p_{i,1}^{x}, p_{i,1}^{y}, p_{i,1}^{z}\right), \ldots, \left(p_{i,D}^{x}, p_{i,D}^{y}, p_{i,D}^{z}\right)\right)^{T}$$

$$\mathbf{v}_{i} = \left(\mathbf{v}_{i,1}, \ldots, \mathbf{v}_{i,D}\right)^{T} = \left(\left(v_{i,1}^{x}, v_{i,1}^{y}, v_{i,1}^{z}\right), \ldots, \left(v_{i,D}^{x}, v_{i,D}^{y}, v_{i,D}^{z}\right)\right)^{T}$$

04. Improved PSO



Chaos-based particle initialization

- more uniform initial distribution is better
- Logistic map $x_{n+1} = \mu x_n (1 x_n)$
- $\mu = 4 =>$ system produces chaotic signal



Adaptive parameter adjustment

- c_1, c_2 determine exploration/exploitation
- explore at start, exploit in the end

$$c_1 = c_{\max} - \frac{(c_{\max} - c_{\min})t}{T}$$
$$c_2 = c_{\min} + \frac{(c_{\max} - c_{\min})t}{T}$$

$$\begin{cases} \mathbf{v}_{i,j} (t+1) = w \cdot \mathbf{v}_{i,j} (t) + c_1 \cdot r_1 \cdot (\mathbf{p}_{i,j,best} (t) - \mathbf{p}_{i,j} (t)) \\ + c_2 \cdot r_2 \cdot (\mathbf{g}_{j,best} (t) - \mathbf{p}_{i,j} (t)) \\ \mathbf{p}_{i,j} (t+1) = \mathbf{p}_{i,j} (t) + \mathbf{v}_{i,j} (t+1) \end{cases}$$

Maximum velocity design



- V_1 , V_2 start/end maximum velocities
- bigger in the beginning to allow for wider exploration
- smaller in the end to refine existing solution

$$V_{\max} = V_1 - \frac{(V_1 - V_2) t}{T}$$

Position updating strategy

- sort particles by fitness
- divide sorted particles into groups p_{large} and p_{small}
- replace particles with large fitness by mutating particles with small fitness
- r3 random number
- a position offset constant

$$\begin{cases} \mathbf{p}_{l\,\mathrm{arg}\,e} = \mathbf{p}_{small} + a \cdot r_3 \\ \mathbf{v}_{l\,\mathrm{arg}\,e} = \mathbf{v}_{small} \end{cases}$$

O5. Algorithm



Algorithm - Environment Construction



- 1. Set flying space
- 2. Set mountains
- 3. Set radars
- 4. Set starting and destination positions

Algorithm - Main Loop

for i = 1 to UAVnumber

Initialize PSO for given UAV

for j = 1 to T

Update PSO parameters (inertia weight...)

Generate new position and velocity

Calculate fitness

Select and remember gbest and pbest

Adopt mutation strategy and generate new position



06. Performance



Solution optimality





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Solution optimality





Convergence Speed



Convergence Speed



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Convergence Speed



Success Rate





Running Time





Comparison with MGA - CIPSO Paths



Comparison with MGA - MGA Paths



Comparison with MGA



(b) AFV comparison between MGA and CIPSO



Comparison with MGA



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Thank you

Do you have any questions?

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