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A hybrid biogeography-based optimization method for the inverse kinematics problem of an 8-DOF redundant humanoid manipulator

Key words: Inverse kinematics problem, 8-DOF humanoid manipulator, Biogeography-based optimization (BBO), Differential evolution (DE)

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Introduction

- The redundant humanoid manipulator has characteristics of multiple degrees of freedom and complex joint structure, and it is not easy to obtain its inverse kinematics solution.
- To solve the inverse kinematics problem of the redundant humanoid manipulator, a hybrid biogeography-based optimization (HBBO) algorithm, which is based on BBO and differential evolution (DE), is presented.
- An 8-DOF redundant humanoid manipulator is employed as an example, and the HBBO is applied to solve its inverse kinematics problem.

Hybrid biogeography-based optimization algorithm (HBBO)

Algorithm 1 Hybrid migration operator in HBBO

```
1: for each habitat  $i \in \{1, 2, \dots, m\}$  do
2:   Randomly choose index  $r_1 \neq r_2 \neq i$ 
3:   Normalize the immigration rate  $\lambda$ 
4:   for each SIV  $k \in \{1, 2, \dots, n\}$  do
5:     Select habitat  $H_i$  with probability  $\propto \lambda_i$ 
6:     if  $H_i$  is selected then
7:       Select habitat  $H_j$  with probability  $\propto \vartheta_j$ 
8:        $H_i(k) \leftarrow H_j(k)$ 
9:     else
10:      if  $\text{rand}(\cdot) \leq \text{CR} | k = \text{randn}$  then
11:         $H_i(k) \leftarrow H_{\text{gbest}}(k) + F (H_{r_1}(k) - H_{r_2}(k))$ 
12:      else
13:         $H_i(k) \leftarrow H_i(k)$ 
14:      end if
15:    end if
16:  end for
17: end for
```

Algorithm 2 Gaussian mutation operator in HBBO

```
1: for each habitat  $i \in \{1, 2, \dots, m\}$  do
2:   Compute the probability  $P_i$ 
3:   Select SIV  $H_i(k)$  with mutation probability  $m_i$ 
4:   if  $H_i(k)$  is selected then
5:     Update  $H_i(k)$  with the Gaussian mutation
       operator
6:   end if
7: end for
```

Algorithm 3 HBBO algorithm

```
1: Generate a random set of habitats  $H_1, H_2, \dots, H_m$ 
2: Evaluate the HSI value for each habitat
3: Initialize the generation counter  $t = 1$ 
4: while the halting criterion is not satisfied do
5:   Sort the population from best to worst
6:   Compute the migration rates  $\lambda$  and  $\vartheta$  for each
       habitat based on HSI
7:   Implement the hybrid migration operator shown
       in Algorithm 1
8:   Update the probability for each habitat
9:   Perform the Gaussian mutation operator shown in
       Algorithm 2
10:  Recompute the HSI value for the population
11:   $t = t + 1$ 
12: end while
```

Model for the 8-DOF redundant humanoid manipulator

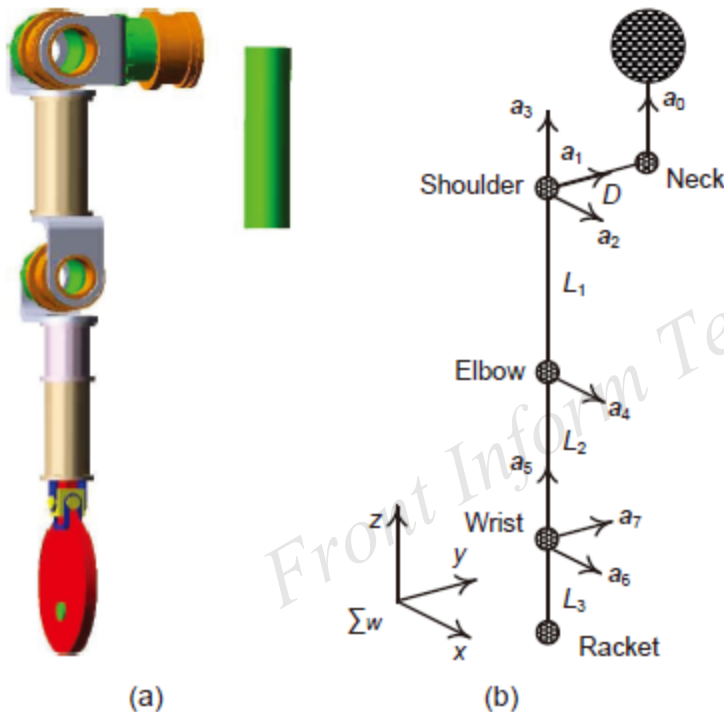


Fig. 1 Simulation model (a) and joint structure model (b) for an 8-DOF humanoid manipulator. a_0 is the rotation direction of the waist joint, and $a_1 - a_7$ are the rotation directions of the shoulder, elbow, and wrist joint, respectively. Σ_w is the world-coordinate system, D the shoulder breadth, L_1 the length from the shoulder center to the elbow center, L_2 the length from the elbow center to the wrist center, and L_3 the length from the wrist center to the racket center

Simulation results (1)

Table 2 HBBO results over 10 independent runs of the inverse problem for different α

α	Best	Average	Worst	Standard deviation	Convergence rate
10^{-4}	9.7847×10^{-5}	2.9094×10^{-2}	2.9004×10^{-2}	9.1686×10^{-2}	20%
10^{-5}	8.0733×10^{-6}	9.8058×10^{-6}	1.0437×10^{-5}	6.4780×10^{-7}	100%
10^{-6}	9.0915×10^{-7}	2.0063×10^{-2}	2.0056×10^{-1}	6.3419×10^{-2}	90%
10^{-7}	9.3703×10^{-8}	1.0826×10^{-2}	9.8405×10^{-2}	3.0845×10^{-2}	70%

Table 3 Results over 10 independent runs of the inverse kinematics problem under four different methods

Method	Best	Average	Worst	Standard deviation
Standard genetic algorithm	3.6036×10^{-2}	4.2291×10^{-2}	4.7253×10^{-2}	4.2084×10^{-3}
Differential evolution	8.4257×10^{-6}	2.0782×10^{-2}	5.7474×10^{-2}	2.8712×10^{-2}
Biogeography-based optimization	1.5693×10^{-2}	6.8393×10^{-2}	1.6791×10^{-1}	6.0563×10^{-2}
Hybrid biogeography-based optimization	8.0733×10^{-6}	9.8058×10^{-6}	1.0437×10^{-5}	6.4780×10^{-7}

Simulation results (2)

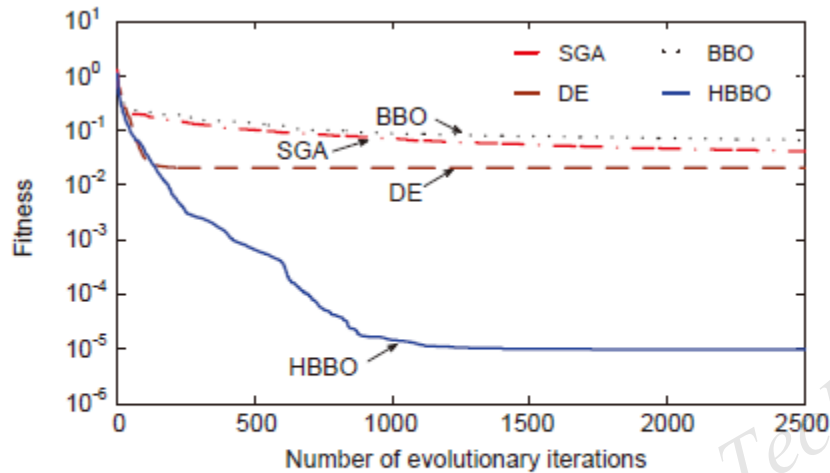


Fig. 2 Average evolution curves under different algorithms on the inverse kinematics problem of the humanoid manipulator

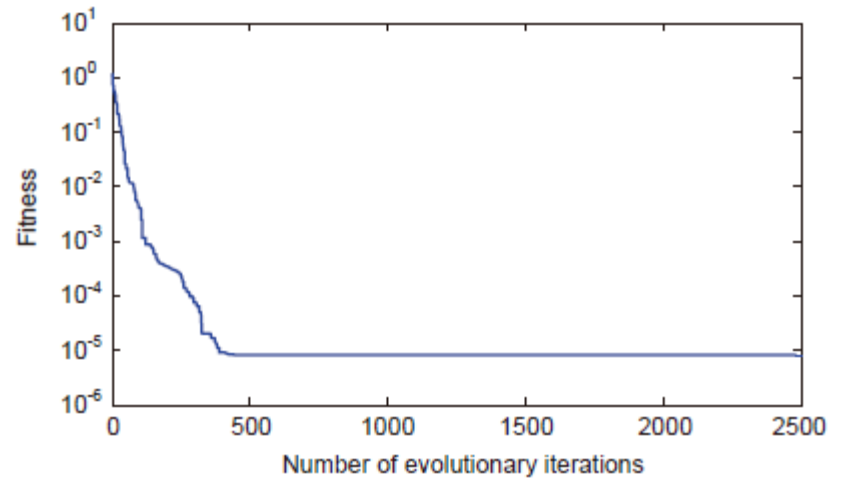


Fig. 3 A typical evolution curve of the fitness under HBBO

Position and orientation of the 8-DOF humanoid manipulator configurations



Fig. 4 Position and orientation of the 8-DOF humanoid manipulator configurations

Conclusions

- The hybrid biogeography-based optimization (HBBO) algorithm, which is a hybridization of two different evolutionary algorithms shows good optimization performance in terms of the obtained solutions.
- Compared to SGA, DE and BBO methods, the proposed HBBO algorithm is competent for the inverse kinematics problem of the 8-DOF redundant humanoid manipulator.