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APPLICATION OF THE METAHEURISTIC TUNICATE SWARM ALGORITHM IN SOLVING APPLIED MECHANICS PROBLEMS

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Abstract: In this paper, the principles of a metaheuristic algorithm based on tunicate swarm behavior are shown. The Tunicate Swarm Algorithm (TSA for short) was used for solving problems in applied mechanics (speed reducer, cantilever beam and three-dimensional beam optimization). In the end, a comparison of results obtained by TSA and results obtained by other methods is given.

Key words: metaheuristics, tunicate, algorithm, mechanics

Primena metaheurističkog algoritma roja plastaša za rešavanje problema primenjene mehanike. U ovom radu su izloženi principi rešavanja metaheurističkog algoritma zasnovanog na ponašanju roja plaštaša.Primenom algoritma roja plaštaša (TSA) rešeni su problemi u okviru primenjene mehanike (optimizacija jednostepenog reduktora, konzolne grede i trodimenzionalnog nosača). Na kraju rada će biti prikazani rezultati dobijeni metodom TSA u odnosu na prethodno dobijene rezultate sa drugim metodama. **Ključne reči:** metaheuristika, plaštaši, algoritam, mehanika.

1. INTRODUCTION

Most engineering problems can be formulated as optimization problems. To solve optimization problems, different methods have been studied in mathematical programming, operations research, etc.

The classification of metaheuristic algorithms [4, 6, 7, 8, 9, 10] is shown in figure 1.



Fig. 1. Classification of metaheuristic algorithms

In this paper, the Tunicate swarm algorithm is used to solve the following engineering problems.

The first problem to be solved is speed reducer optimization, having the goal of minimizing reducer weight in accordance with bending stress constraints of gear teeth, surface stresses, transverse deflections of shafts and stresses in shafts. This problem was first analyzed and solved by Coello using GA [1].

The second engineering problem is cantilever beam optmization, where minimal weight that fulfills the constraints is sought after. Gandomi has solved this problem using the Cuckoo Search Algorithm (CSA) [2].

The last engineering problem that will be considered in this paper is optimization of a cantilever beam. The goal of this optimization is to minimize cross-section heights for each beam element. Miodragovic used FA algorithm to solve this problem [3].

2. TUNICATE SWARM ALGORITHM

Tunicates, commonly called sea squirts, are a group of marine animals that spend most of their lives attached to docks, rocks or the undersides of boats. The name, "tunicate" comes from the firm, but flexible body covering, called a tunic. They have the shape of a small barrel, and have two openings, called siphons, enabling water to stream through them so that they can feed on planktons. The tunicate moves through the ocean by jet propulsion of water through the siphons. Tunicate is only animal to move around the ocean with such fluid jet like propulsion. Tunicates are often found at depth of 500–800 m and migrate upwards in the upper layer of surface water at night.

In the Tunicate Swarm Algorithm, tunicates are in search of food source. The food source is indicated by the best search agent, while the tunicates seeking the food source are all other search agents. The jet propulsion behavior must satisfy three conditions:

- 1. Avoiding conflicts between search agents
- 2. Movement towards the best neighbor
- 3. Converging towards the best search agent

In order to avoid conflicts between search agents, the following model is applied:

$$\vec{A} = \frac{G}{\overline{M}} \tag{1}$$

$$\vec{G} = c_2 + c_3 - \vec{F} \tag{2}$$

$$\vec{F} = 2 * c_1 \tag{3}$$
$$\vec{M} = \begin{bmatrix} P & + c & * P & P \end{bmatrix}$$

$$M = \left[P_{\min} + c_1 * P_{\max} - P_{\min} \right]$$
(4)

where A -new agent position, G - gravity force, M - social forces, \vec{F} - water flow advection, c_1, c_2, c_3 - random numbers in range [0,1], P_{\min}, P_{\max} - initial and subordinate speeds for social interaction.

Usually, the values for P_{\min} , P_{\max} are 1 and 4, respectively.

To avoid the conflict between neighbors, the search agents move towards the direction of the best neighbor.

$$\overrightarrow{PD} = \left| \overrightarrow{FS} - rand * \overrightarrow{P_p(x)} \right|$$
(5)

where PD- distance between the food source and search agent, \overrightarrow{FS} - position of food source, *rand* - random number in range [0,1], $\overrightarrow{P_p(x)}$ - position of tunicate.

The search agents move towards the food source using the following equation:

$$\overrightarrow{P_{p}(x')} = \begin{cases} \overrightarrow{FS} + \overrightarrow{A} * \overrightarrow{PD}, & \text{if } rand \ge 0.5 \\ \overrightarrow{FS} - \overrightarrow{A} * \overrightarrow{PD}, & \text{if } rand < 0.5 \end{cases}$$
(6)

Where $P_p(x)$ is the updated position of the search agent with respect to the food source. In order to

simulate the swarm behavior, the first two optimal best solutions are recorded, and all the other solutions update their position with respect to those positions. This is done using the following equation:

$$\overrightarrow{P_p(x+1)} = \frac{\overrightarrow{P_p(x)} + \overrightarrow{P_p(x+1)}}{2 + c_1}$$
(7)

The TSA consists of following steps:

Step 1: Initialize the tunicate population P_n .

Step 2: Choose the initial parameters and maximum number of iterations.

Step 3: Calculate the fitness value of each search agent. **Step 4**: After computing the fitness value, the best search agent is explored in the given search space.

Step 5: Update the position of each search agent using the equation for $\overline{P_p(x+1)}$.

Step 6: Adjust the updated search agent which goes beyond the boundary in a given search space.

Step 7: Compute the updated search agent fitness value. If there is a better solution than the previous optimal solution, then update P_n .

Step 8: If the stopping criterion is satisfied, then the algorithm stops. Otherwise, repeat the Steps 5–8.

Step 9: Return the best optimal solution which is obtained so far.

3. EXPERIMENTAL ENGINEERING EXAMPLES FOR OPTIMIZATION

This chapter will present certain examples of engineering problems, such as: optimization of speed reducer, cantilever beam and three dimensional beam. The basis of the problem, the objective function, variable parameters that should be found as well as the constraints that should be respected will be shown.

The goal of speed reducer optimization is minimizing the reducer weight whilst fulfilling all the defined constraints.

In Figure 2 a schematic view of speed reducer is shown.



Fig. 2. Speed reducer design problem

Project variables for the speed reducer problem are: the width between the shafts (x_1) , the module of the teeth (x_2) , the number of teeth in the pinion (x_3) , the length of the first shaft between the bearings (x_4) , the length of the second shaft between the bearings (x_5) , the diameter of the first shaft (x_6) and the diameter of the second shaft (x_7) .

Goal function to be minimized is defined as:

$$f(x) = 0,7854x_1x_2^2(3,3333x_3^2 + 14,933x_3 - 43,0934) - 1,508x_1(x_6^2 + x_7^2) + +7,4777(x_6^3 + x_7^3) + 0,7854(x_4x_6^2 + x_5x_7^2)$$
(8)

subject to the following constraints:

$$g_1(x) = \frac{27}{x_1 x_2^2 x_3} - 1 \le 0 \tag{9}$$

$$g_2(x) = \frac{397.5}{x_1 x_2^2 x_3^2} - 1 \le 0 \tag{10}$$

$$g_3(x) = \frac{1.93x_5^3}{x_2x_3x_7^4} - 1 \le 0 \tag{11}$$

$$g_4(x) = \frac{1.93x_5^3}{x_2x_3x_7^4} - 1 \le 0 \tag{12}$$

$$g_5(x) = \frac{\left(\left(\frac{475x_4}{x_2x_3}\right)^2 + 16.9 \times 10^6\right)^{1/2}}{110x_6^3} - 1 \le 0$$
(13)

$$g_6(x) = \frac{\left(\left(\frac{475x_4}{x_2x_3}\right)^2 + 157.5 \times 10^6\right)^{1/2}}{85x_7^3} - 1 \le 0$$

$$g_7(x) = \frac{x_2 x_3}{40} - 1 \le 0 \tag{15}$$

(14)

$$g_8(x) = \frac{5x_2}{x_1} - 1 \le 0 \tag{16}$$

$$g_9(x) = \frac{x_1}{12x_2} - 1 \le 0 \tag{17}$$

$$g_{10}(x) = \frac{15x_6 + 1.9}{x_4} - 1 \le 0 \tag{18}$$

$$g_{11}(x) = \frac{1.1x_7 + 1.9}{x_5} - 1 \le 0$$

$$2,6 \le x_1 \le 3,6; 0,7 \le x_2 \le 0,8; 17 \le x_3 \le 28$$

$$7,3 \le x_4 \le 8,3; 7,3 \le x_5 \le 8,3; 2,9 \le x_6 \le 3,9$$

$$5,0 \le x_7 \le 5,5$$
(20)

In Figure 4 a schematic view of cantilever beam is shown.



Fig. 4 . Cantilever beam design problem

As seen in Figure , the cantilever beam consists of five hollow, box shaped bearings with a square shaped frame. Project variables are lengths of the five squares($x_{1,x_{2,}x_{3.}x_{4,}x_{5,}$) which make up the cantilever beam.

Objective function to be minimized, as defined in [5]:

$$f(\vec{x}) = 0,6224(x_1 + x_2 + x_3 + x_4 + x_5), \quad (21)$$

Whilst the only constraint for this problem being:

$$g\left(\vec{x}\right) = \frac{61}{x_1^3} + \frac{27}{x_2^3} + \frac{19}{x_3^3} + \frac{7}{x_4^3} + \frac{1}{x_5^3} - 1 \le 0, \quad (22)$$

$$0,01 \le x_1, x_2, x_3, x_4, x_5 \le 100, \tag{23}$$

The last problem consists of minimizing crosssection heights of all elements of a cantilever beam, which is shown in Figure 3.



Fig. 3. Three dimensional beam design problem

A vertical shift of point A is defined in advance, having a specified upper limit. The beam is under continual load (q_1, q_2) on horizontal parts of the beam, as well as horizontal force F, which affects the vertical part of the beam.

Goal function to be minimized is defined as:

$$f(\mathbf{X}) = 0.8x_1 + x_2 + 0.8x_3, \qquad (24)$$

subject to the following constraints:

$$u_{A}(X) = \left[\frac{11.2480 \cdot 10^{-3}}{x_{1}^{3}} + \frac{3.539\bar{9} \cdot 10^{-3}}{x_{2}^{3}} + \frac{0.3840 \cdot 10^{-3}}{x_{3}^{3}}\right] \le 0.05[m], \quad (25)$$

$$0.1 \le x_{1} \le 0.9,$$

$$0.1 \le x_{2} \le 0.9,$$

$$0.1 \le x_{3} \le 0.9,$$

$$(26)$$

4. OPTIMIZATION RESULTS

In this section, the results obtained by using TSA algorithm on previously defined engineering problems is given.

A detailed display of the results obtained by TSA and a comparison with several results obtained by other methods, for the problem of speed reducer, three dimensional beam and cantilever beam are shown in Table 1,2,3.

Variables	GWO[4]	GOA[5]	WCA[6]	TSA
x_{l}	3.502	3.5	3.5	3.505
x_2	0.7	0.7	0.7	0.7
x_3	17	17	17	17
x_4	7.333	7.3	7.3	7.3
x_5	7.8	7.8	7.715	7.8
x_6	3.351	3.350	3.350	3.352
x_7	5.288	5.287	5.286	5.301
f(x)	2998.299	2996.964	2994.471	3007.327

Table 1. Comparison of results for the speed reducer

Variables	ALO[7]	GOA[5]	MMA[8]	TSA
x ₁	6.018	6.011	6.010	5.973
X2	5.311	5.312	5.300	5.258
X3	4.488	4.483	4.490	4.514
x_4	3.497	3.502	3.490	3.529
x_5	2.158	2.163	2.150	2.203
f(x)	1.339	1.339	1.340	1.340

Table 2. Comparison of results for the cantilever beam

Variables	ANSYS[9]	GOA[7]	FA[3]	TSA
X ₁	0.80458	0.80453	0.89071	0.8041
x ₂	0.44054	0.56995	0.56197	0.5686
X3	0.51398	0.34588	0.32444	0.3480
f(x)	1.4618	1.4903	1.4900	1.4903

Table 3. Comparison of results for the three dimensional beam

5. CONCLUSION

This paper deals with the TSA optimization algorithm and applies it to a few engineering design examples: speed reducer, cantilever beam, and three dimensional beam.

The obtained results were compared to latest papers published in SCI list journals.

This algorithm was used to obtain optimal or near-optimal results, as shown in the examples. Further developing of this algorithm can be used to redefine and upgrade this method, as to gain better results.

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