

Solving quadratic assignment problem using water cycle optimization algorithm

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Abstract: The Quadratic Assignment Problem (QAP) is one of combinatorial optimization problems which devote some facilities to some locations. The aim of this problem is assignment of each facility to a location which minimizes total cost. Because the QAP is NP-hard, so it couldn't be solved by exact methods. In recent years, meta-heuristic algorithms are used in solving NP-hard optimization problems increasingly. In this article Water Cycle Optimization Algorithms (WCO) is used to solve QAP. The implementation of proposed algorithms on standard test functions and also its result comparison with other meta-heuristics algorithms express algorithm's desirable quality and its prominence to other meta-heuristics algorithms.

Keywords: Quadratic Assignment Problem, Combinatorial Optimization Problems, Water Cycle Optimization Algorithms, Meta-Heuristics Algorithms

1. Introduction

The Quadratic Assignment Problem (QAP) is one of combinatorial optimization problems that, in many firms have done various researches in order to devote some facilities to some locations. The most important thing in this process is assignment costs, which this type of costs become minimized. The computation of optimized solutions is really hard and difficult for most of optimization problems observed in many operational and practical fields. The solutions for optimization problems include two groups of exact and heuristics methods. The exact methods gain optimized solution and guarantee optimizing condition [1]. Heuristics methods produce high qualified solutions in sensible time, but didn't guarantee to find the thorough optimized solution [2]. The QAP is a problem with exponential complexity. Totally, problems with larger than 20 dimensions aren't solvable through exact methods otherwise from the view point of time they aren't economical. Therefore, meta-heuristics methods are used in these cases. Indeed meta-heuristics algorithms is one of heuristics optimization algorithms types which have outgo mechanisms from local optimization and are applicable in many different problems [3]. Various types of these algorithms are developed during recent decades. The meta-heuristics algorithms which used in solving QAP are Annealing algorithm [4], Tabu Search

[5, 6], Genetics algorithm [7], Ant Colony Optimization algorithm [8, 9], Harmony Search algorithm [10], Imperialist competitive algorithm [11], Artificial Bee Colony algorithm [12], and Cuckoo algorithm [13].

In presented design, the syntax of WCO meta-heuristics algorithms are used to solve QAP. Implementation of proposed algorithm on standard test functions and also its comparison with other meta-heuristics algorithms express desirable quality and its prominence to other meta-heuristics algorithms. The structure of the reminder of this paper is as follows: Section 2 is an introduction to QAP. In section 3 an introduction to WCO algorithm is given. In section 4, we describe the proposed algorithm based on WCO algorithm to solve the problem. Section 5 and 6 are dedicated to describe experimental results and paper conclusion, respectively.

2. Definition of Quadratic Assignment Problem

The QAP is one of combinatorial optimization problem which devote a category of facilities to a category of locations, along with cost that is function of distance and the

flow among facilities in addition to relevant costs to facilities which located in specified location. The aim of this problem is assignment of each facility to each location in way that total cost becomes minimized. General model of QAP is presented below.

$$\min \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n \sum_{l=1}^n f_{ij} d_{kl} x_{ik} x_{jl} \quad (1)$$

$$\text{s.t. } \sum_{i=1}^n x_{ij} = 1 \quad 1 \leq j \leq n \quad (2)$$

$$\sum_{j=1}^n x_{ij} = 1 \quad 1 \leq i \leq n \quad (3)$$

Where f_{ij} is the flow between I and j facilities, and d_{kl} is the distance among k and l locations [14]. If pay attention to activities assignment to locations, general model of QAP is like below.

$$\min \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n \sum_{l=1}^n f_{ij} d_{kl} x_{ik} x_{jl} + \sum_i \sum_k C_{ik} x_{ik} \quad (4)$$

Kopmans and Beckman introduced QAP for the first time in 1957 as a mathematic model which is relevant to economic activities [15]. This problem has many different applications from that time up to now. One of the main usages of QAP is in Placement theory, which $F = (f_{ij})$ is a flow matrix. The aim of this problem is finding an assignment of all facilities to all locations with minimized cost. Other usages of it is sorting constructions in hospitals, warehouses management and distribution strategies, minimizing wire's cables length in electronic circuits, assignment of some factories to some locations, wiring the circuit problem, controlling panels and typist keyboards designing.

3. Water Cycle Optimization Algorithm

WCO Algorithm [16] is a meta-heuristics algorithm which is inspired from water cycle in nature. In presented scheme, there have been attempt to introduce a new algorithm using water cycle in nature. The aim of this problem is benchmarking WCO like raining, the population of each generation has suitable diversity and scattering, and also by simulation of falling water drops motion process toward river, algorithm has a high convergence speed to gain thorough optimum. In addition to benchmark saturated pits during raining, escaping designed local optimum make algorithm to have suitable search in problem space and doesn't trap into local optimum. This algorithm includes a set of primary values for each parameters of problem. In each cycle of performing this algorithm, some values are selected based on a priority for each parameter of problem and make a smaller category, then by considering values competence, several solution will be generated by incorporating values of each parameter. After that each solution of problem located in its local optimum using local search algorithm. Each local optimum according to its propriety receives a capacity that explains the survival of that answer in consecutive performances and also the probability of being candidate to move other answers towards itself. In each cycle of performance, the answers which should be exited out of

problem move towards answers with better competence and reach to local optimum that chose by themselves, if they didn't meet any more competence answer during path. Each cycle one of problem solutions with the most competence reminds constant and at the end of each cycle, values of each parameter of problem solution enter to set of primary values. After ending n cycle, the best solution of nth cycle is optimum solution of problem.

4. Water Cycle Algorithm Steps to Solve QAP

Water-cycle algorithm has 5steps which include:

4.1. Initialization

Such as many evolutionary algorithms, this algorithm also initializes with set of random primary values. In first performance of it, ocean initialized by random primary values. It assumed for discrete problem solution, the answer is a permutation of 1 to n (problems with n variables).

Ocean and initialization: it is a $n*n$ matrix that its initial value will be $(Ocean\ Size)/n$, that Ocean size expresses size of ocean. In an example with 5 variables, ocean is a $5*5$ matrix and initial values of all its elements will be $200/5 = 40$. this values will change in river's water entrance into ocean. In this matrix, columns present a variable (tiny drop) of problem. Rows show the probability of locating one variable in location of 1 to n.

4.2. Evaporation Step Out of Ocean and Formation of Cloud Droplets

In this step some droplets should be selected for evaporation or formation of drop. This means that according to figures in each column, a droplet is selected in a way that more figure for a droplet, has more probability to choose. A simple solution can be in this way that sum first column droplets values, then generate a random value between 1 and sum of droplets value. This random value placed in intervals of each droplet, the droplet will be selected. Finally with n droplet, one drop will be formed.

4.3. Local Search

In local search a drop should immigrate to its better neighbor. Three types of neighborhood are defined for permutation problems.

4.3.1. Replacing Two Quantities

Two locations selected randomly and their quantities replace with each other.

Example: 1 5 4 2 3, elements 1 and 4 are selected and result will be 1 3 4 2 5

4.3.2. Inverting a Limited Area

Two randomly quantities are generated and the limited area between them will be inverted.

Example: 1 5 4 2 3 numbers of 1 and 4 are selected and

result will be 1 3 2 4 5

Tow random quantities are generated and the first one converts to next element after second quantity.

Example: 1 5 4 2 3 numbers of 1 and 4 are selected and result will be 1 5 3 2

4.4. Surface Runoff

The way of falling water motion from the origin pit to destination pit is in this way that a permutation of numbers from 1 to n are generate randomly and based on that permutation, the droplet equal with the droplet on destination pit on that location, founded in origin pit and replaces with droplet in origin pit on that location.

Origin pit: 2 4 5 3 1
 Destination pit: 4 3 2 1 5
 Generated permutation: 3 2 4 5 1

The origin pit after one step movement: 1 3 2 4 5 the first number which generated by permutation is 3, the third number of destination pit is 2, that should be located in its main location in origin pit, it means that 2 replaces with 5.

The origin pit after one step movement: 5 3 2 4 1
 The origin pit after one step movement: 5 3 2 1 4
 The origin pit after one step movement: 4 3 2 1 5
 Surface runoff to n-1 steps are needed.

4.5. River Water Entry to Ocean

The numbers inside ocean change like below in each step:

Table 1. Results of running the program with 200 repeats for the problems chosen from qaplib. errors are in percent.

Problem Name	Problem Size	Best Solution Found So Far	Solution Method	Errors of Previous Methods	HBMO Error	COA Error	WCO Error
Lipa30a	30	13178	Exact		3.74	2.01	1.31
Lipa60a	60	107218	Exact	-	2.25	0.95	0.58
Lipa90a	90	360630	Exact	-	1.67	0.39	0.28
Sko49	49	23386	RO-TS	5.91	16.11	3.56	0.10
Sko56	56	34458	RO-TS	5.37	18.49	4.39	1.04
Sko64	64	48498	RO-TS	5.7	16.91	4.47	1.41
Sko72	72	66256	RO-TS	5.38	14.34	4.85	1.34
Sko81	81	90998	GA	5/41	13.57	4.332	1.38
Sko100a	100	152002	GA	6/41	12.74	12.95	1.48

Table 2. Results of running the program with 500 repeats for the problems chosen from qaplib. errors are in percent.

Problem Name	Problem Size	Best Solution Found So Far	Solution Method	Errors of Previous Methods	HBMO Error	COA Error	WCO Error
Lipa30a	30	13178	Exact		3.78	2.00	1.08
Lipa60a	60	107218	Exact	-	2.3	0.96	0.48
Lipa90a	90	360630	Exact	-	1.65	0.38	0.19
Sko49	49	23386	RO-TS	5.91	18.82	3.55	0.14
Sko56	56	34458	RO-TS	5.37	15.88	4.38	0.05
Sko64	64	48498	RO-TS	5.7	14.36	4.45	0.86
Sko72	72	66256	RO-TS	5.38	13.78	4.78	0.85

$$s_{ij} = s_{ij} - (\alpha \times s_{ij}) / (O_{sean} \text{ Sizae}) \tag{5}$$

$$s_{r_i} = s_{r_i} + \alpha \tag{6}$$

Where S is ocean itself, α is a coefficient which delivered after entry and r is river.

5. Experiments and Simulation Results

This study addresses the general form of QAP and its purpose is to evaluate the WCO in solving QAP so that its applicability is confirmed and it can then be used in solving specific real cases in next studies. For this reason, those problems are chosen which are more famous and have been used for testing other algorithms. Therefore, the most credible reference of QAPs, QAPLIB [17], was used which is prepared by Peter Hann, Berkard, Chella, Randal, and Karisch who are mathematics professors that specialize in QAP. In QAPLIB, different QAPs of different sizes are defined and solved by scientists such as Berkard, Al-Shaafi, Steinburg, etc. using exact, heuristic, and meta-heuristic methods.

The results of experiments and simulations are presented in Tables 1, 2 and 3 and graphically in Figures 1 and 2. First, we determine the results for a small population. Then, the resulting error percent is calculated and compared to the genetic and honeybee and water cycle improved algorithms.

Table 3. Results of running the program for the problems chosen from qaplib with large population.

Problem Name	Problem Size	Optimized Solution	Error of the Best GA Solution (%)	Error the Best HBMO Solution (%)	Error of the Best COA Solution (%)	Error of the Best WCO Solution (%)
Esc32a	32	130	21.69	54.86	10.769	3.70
Esc32b	32	168	20.75	50.56	16.66	1.75
Esc32c	32	642	0	9.7	0	0
Esc32d	32	200	0	29.57	0	0
Esc32e	32	2	0	0	0	0
Esc32h	32	438	1.79	22.06	0.9132	0.4235

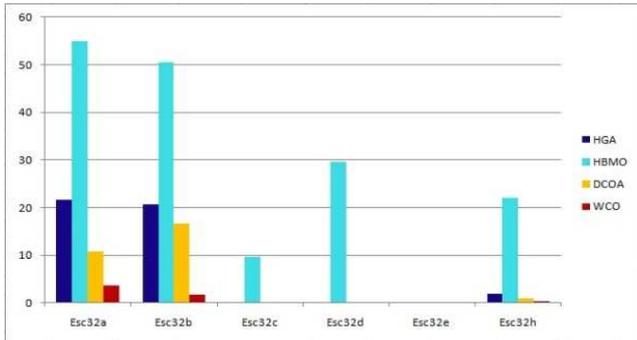


Figure 1. Comparison of Algorithms results for low-level.

WCO the problems and the weak points of other optimization algorithms (such as GA, HBMO and the other new methods like colonial completion algorithm) are not seen in this method, therefore WCO method has high capability to converge to the optimized solution faster than the other algorithms and also has this ability to find overall optimized points with higher accuracy. The ability of WCO method was examined for problems with high dimensions and the results showed that this method works very well in this issue.

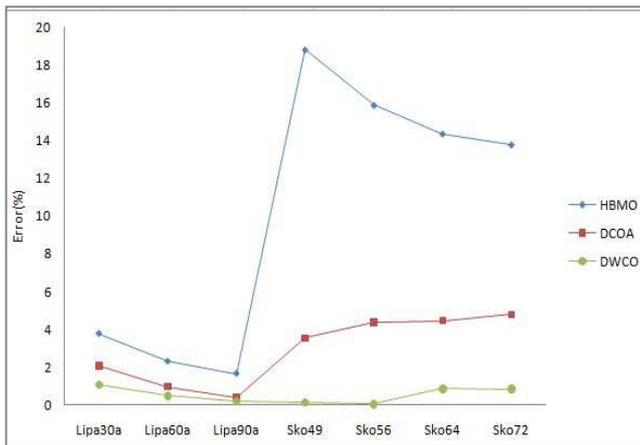


Figure 2. WCO algorithm accuracy for sko* problems.

6. Discussion and Conclusion

In this article the idea of WCO algorithm to solve QAP problem was studied. Results were gained for low, medium and large population. The effects of increasing optimization

repeats in enhancement of algorithm's answering precision are also investigated. The results indicate that water-cycle optimization algorithm performs better than Genetics, Cuckoo and Bee Colony algorithms, and in most comparisons, has better performance to other evolutionary algorithms in different test functions to find the best solution. Therefore, it has very fast convergence capability and the power of finding optimum points will be more precise. This algorithm generates appropriate solutions for problems with large dimensions. While the error of Genetics and Bee Colony algorithms are increased by problem with medium dimensions.

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