

Improving the system using the bee algorithm and the ant algorithm and comparing them

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ABSTRACT

In The allocation of reliability and optimization for each part of the complex system has been determined in this paper, in order to solve the problem of allocation and to improve network stability we will use Bees Colony Optimization (BCO) and the ant colony optimization (ACO) algorithm. The three cost functions are also introduced (exponential behavior with feasibility factor model, exponential behavior model and logarithmic model). After solving the allocation problem. The aim of this paper is to compare the effects of the three cost functions using Bees Colony algorithm and the ant colony optimization algorithm in terms of reliability allocation and optimization to determine which is more efficient.

Keywords:

1. Introduction

The present article, we watched the installed complex system's efficiency. By using limited paths across relation matrices, this system was found to be efficient. (To create minimum paths, remove nodes) as well as Boolean algebra. The goal of evaluating to be learn about the protection of the sophisticated unit that has been built [1-3]. Despite the networks' foundations. the optimum distribution dependability is treated as a mathematical task in this research. Each and every dynamic component Each device offers a unique set of features. dependability standards that are optimized and reliant on its importance [2, 4-6]. In order to achieve the highest level of efficiency in general, depending on the allocation of every element in the System, which varies between the component and the component, such components may require allocations. Engineers working large on improving mechanical and electrical systems encounter a variety of challenges [7]. The focus

of this work is on complicated structure allocation and improvement, in addition to the cost system, which can be assessed in terms of size, and proportion, or other criteria. The dependability of this component is based on two keys needs. To begin, the model should be used to determine the input variable reliability. You will change the parameters of the suggested cost parameter [3, 8-10]. Engineers may now upgrade their duties for each machine and prepare for the utmost Each device's productivity, second, the model's analytical performance should be balanced of the input system. This can be a huge difficulty in simple structures, but it can be In complex systems, this is a significant issue [11]. The expenses are calculated using the Bees Colony Optimization algorithm (BCO) and the ant colony optimization (ACO) algorithm which uses [2] complex systems to solve optimization problems. The algorithm uses three cost functions " Modeling activity exponentially with the viability factor, The exponential

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behavior paradigm and a logarithmic model". The goal is to improve device reliability while lowering total expenses.

2. Optimization of the complex network

Think about a sophisticated network made up of parts. [2-4]. We are reliant on statements.: $0 \le R_i \le 1$ dependability of a component i; $C_i(R_i)$ the Individual Costs, Components i , C $(R_{1},...,R_{n}) = \sum_{i=1}^{n} a_{i} c_{i}$ (Ri) Parts' Prices $a_{i} > 0$; The effectiveness of the system is measured in R_s, the device's reliability objective is denoted by the letter RG. Each component of the scheme has a distinct goal, and there are several options. Device modules have their own set of features. varying levels of assurance in the same characteristics. The objective is to ensure that any or all gadget components are trustworthy. The Q issue is a non-linear threshold with an evaluable cost and function [12-15]. Seek out

 $\begin{array}{l} \text{MinimizeC}(R_{i}, ..., R_{n}) = \\ \sum_{i=1}^{n} a_{i} C_{i}(R_{i}), \ a_{i} > 0, \end{array} \tag{1}$

within the terms of

$$R_G \leq R_S$$

 $0 \le R_i < 1, i = 1, ..., n.$ Let's pretend the cost function is appropriate.. $C_i(R_i)$ satisfies those requirements [12, 16]. The favorable, unique role is being improved.

$$\left[\Longrightarrow \frac{\mathrm{dC}_{\mathrm{i}}}{\mathrm{dR}_{\mathrm{i}}} \ge 0\right].$$

The goal of the prior strategy was to achieve a cost basis that included everything [2, 4]. The stability limit of the system has been lowered. However It's in the vicinity the scope of R_G.

3. Application to complex network

All components in the complex network illustrated have the same as in fig. (1) primary trust level of 90% at the specified periods [11]. The system's aim for stability is 90 percent at

any given time. The dependability polynomial for the proposed technique was found using the approach to minimal path [14].

 $R_{s} = R_{1}R_{9} R_{10} + R_{2} R_{8} R_{10} + R_{2}R_{5}R_{7}R_{10} +$ $R_3R_4R_7R_{10} + R_1R_6R_8R_{10} + R_2R_6R_9R_{10} - R_2R_6$ R8 R10 - R1 R2 R6 R9 R10 + R1 R5 R6 R7 R10 -R1 R2 R8 R9 R10 + R3 R4 R5 R8 R10 - R2 R5 R7 Rg R10 - R1 R6 R8 R9 R10 - R2 R6 R8 R9 R10 - R1 R2 R₅ R₆ R₇ R₁₀ - R₂ R₃ R₄ R₅ R₇ R₁₀ - R₂ R₃ R₄ R5 R8 R10 - R1 R2 R5 R7 R9 R10 - R1 R3 R4 R7 R9 R10 - R2 R3 R4 R7 R8 R10 + 2R1 R2 R6 R8 R9 R10 - R1 R5 R6 R7 R8 R10 + R3 R4 R5 R6 R9 R10 -R3 R4 R5 R7 R8 R10 - R1 R5 R6 R7 R9 R10 -R2 R5 R6 R7 R9 R10 - R1 R3 R4 R5 R6 R7 R10 -R1 R3 R4 R5 R6 R8 R10 - R1 R3 R4 R5 R6 R9 R10 + R1 R2 R5 R6 R7 R8 R10 - R1 R3 R4 R6 R7 R8 R10 - R₂ R₃ R₄ R₅ R₆ R₉ R₁₀ + 2R₂ R₃ R₄ R₅R₇ R₈ R₁₀ + 2R₁ R₂ R₅ R₆R₇ R₉ R₁₀ - R₁ R₃ R₄ R₅ R₈ R9 R10 - R2 R3 R4 R6 R7 R9 R10 + R1 R2 R5 R7 R8 R9 R10 - R3 R4 R5 R6 R7 R9 R10 - R3 R4 R5 R6 R8 R9 R10 + R1 R5 R6 R7 R8 R9 R10 + R2 R5 R₆R₇ R₈ R₉ R₁₀ + R₁ R₂ R₃ R₄ R₅ R₆ R₇ R₁₀ + R₁ R₂ R₃ R₄ R₅ R₆ R₈ R₁₀ + R₁ R₂ R₃ R₄ R₅ R₆ R9R10 + R1 R2 R3 R4 R5 R7 R9 R10 + R1 R2 R3 R4 R6 R7 R8 R10 + R1 R2 R3 R4 R5 R8 R9 R10 + R1 R2 R3 R4 R6 R7 R9 R10 + R1 R2 R3 R4 R7 R8 R9 R10 + 2 R1 R3 R4 R5 R6 R7 R8 R10 + 2 R1 R3 R4 R5 R6 R7 R9 R10 + 2 R1 R3 R4 R5 R6 R8 R9 R10 + 2 R2 R3 R4 R5 R6 R7 R9 R10 + R1 R3 R4 R5 R7 R8 R9 R10 + R2 R3 R4 R5 R6 R8 R9 R10 - 2 R1 R2 R5 R6 R7 R8 R9 R10 + R₁ R₃ R₄ R₆ R₇ R₈ R₉ R₁₀ + R₂ R₃ R₄ R₆ R₇ R₈ R9 R10 + R3 R4 R5 R6 R7 R8 R9 R10 -2 R1 R2 R3 R4 R5 R6 R7 R8 R10 - 3 R1 R2 R3 R4 R5 R6 R7 R9 R10 - 2 R1 R2 R3 R4 R5 R6 R8 R9 R₁₀ - 2 R₁ R₂ R₃ R₄ R₅ R₇ R₈ R₉ R₁₀ - 2 R₁ R₂ R3 R4 R6 R7 R8 R9 R10 - 3 R1 R3 R4 R5 R6 R7 R₈ R9 R10 - 2 R2 R3 R4 R5 R6 R7 R8 R9 R10 + 4 R1 R2 R3 R4 R5 R6 R7 R8 R9 R10



Figure 1: Complex Network

significant cost models for Three 4. reliability

4.1 Model of exponential behavior with a feasibility parameter

Suppose $0 < f_i < 1$ be the feasibility factor, $R_{i,min}$ be minimum reliability[5,17] and $R_{i,max}$ be maximum reliability[3,4]. Exponential behavior is another important cost function.

$$C_i(R_i) = \exp\left[(1 - f_i)\frac{R_i - R_{i,min}}{R_{i,max} - R_i}\right], R_{i,min} \le R_i \le R_{i,max}, i = 1, 2, 3, \dots, n.$$
(4.1)
The issue of optimization then:

Minimize
$$C(R_i, ..., R_n) = \sum_{i=1}^n a_i \exp[(1 - f_i) \frac{R_i - R_{i,min}}{R_{i,max} - R_i}], i = 1, 2, ..., n.$$

Subject to:

 $Rs \ge R_G$ $R_{i.min} \le R_i < R_{i.max}$, i = 1, ..., n.

Table 1: The Optimal Reliability Allocation (ACO, BCO).				
Components	BCO	ACO		
R1	0.85675	0.88761		
R ₂	0.91404	0.88761		
R ₃	0.805	0.81776		
R4	0.86674	0.87665		
R5	0.77627	0.81568		
R6	0.78884	0.61178		
R7	086263	0.84618		
R ₈	0775	0.75824		
R9	0.91404	0.87665		
R ₁₀	0.91404	0.9133		
Rs	090555	0.90166		



Figure 2: Allocation of reliability utilizing the given feasibility factor model and ACO and BCO.

4.2 Exponential behavior model

Suppose $0 \le R_i$, and $R_i \le 1$ and, [9,15] Constants a_i and b_i , where $i = 1, ..., n_i$. The exponential procedure is the most important cost function. Suggested it in the following format:

$$C_{i}(R_{i}) = a_{i}e^{\left(\frac{z_{i}}{1-R_{i}}\right)}, a_{i} > 0, b_{i} > 0, \text{ where } i = 1, ..., n.$$
(4.2)
The issue of optimization then:

Minimize $C(R_i, ..., R_n) = \sum_{i=1}^n a_i e^{\left(\frac{D_i}{1-R_i}\right)}$, where i = 1, ..., n. Subject to:

 $\begin{array}{l} R_G \leq R_S \\ 0 \leq R_i < 1, i=1, \ldots, n. \\ \mbox{Table 2: The Optimal Reliability Allocation (ACO, BCO).} \end{array}$

Components	BCO	ACO
R ₁	0.863	0.88761
R ₂	0.877	0.88761
R ₃	0.87131	0.81776
R4	0.87117	0.87665
R5	0.825	0.81568
R ₆	0.912	0.61178
R ₇	0.912	0.84618
R ₈	0.876	0.75824
R9	0.79158	0.88761
R10	0.912	0.9133
Rs	0.90461	0.90222



Figure 3: Utilizing ACO and BCO to allocate reliability using the exponential behavior model.

4.3 Logarithmic model

Let $0 \le R_i$, and $R_i \le 1$ and, Constants a_i where i = 1, ..., n. Let's suppose it in the form $C_i(R_i) = a_i \ln\left(\frac{1}{1-R_i}\right), a_i > 0$, i = 1, ..., n(4.3)

The issue of optimization then.:

MinimizeC(R_i,...,R_n) = $\sum_{i=1}^{n} a_i \ln\left(\frac{1}{1-R_i}\right)$, where i = 1, ..., n.

Subject to:

$$\begin{array}{l} R_G \leq R_S \\ 0 \leq R_i < 1, i=1, \ldots, n. \ . \end{array}$$

Table 3: Table for (ACO	, BCO) -Based O	ptimal Reliability	y Allocation
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Components	BCO	ACO
R ₁	0.997	0.94267
R ₂	0.852	0.994
R ₃	0.55	0.54326
R4	0.586	0.61
R5	0.57	0.56
R ₆	0.592	0.86697
R ₇	0.637	0.738
R8	0.997	0.912
R9	0.977	0.994
R ₁₀	0.997	0.999
Rs	0.99564	0.99818



Figure 4: Utilizing ACO and BCO to allocate reliability using the provided Logarithmic model

5. Conclusions

In this paper, the optimal reliability allocation by to optimally distribute reliability for each component of the system was calculated using algorithms (ACO and BCO), in addition to calculating the exact reliability Rs. After solving problem the of system's component assignment and optimization, the results are generally better in terms of BCO assignment than ACO. The highest component allocation was through the use of the cost function (logarithmic model) where Rs in the ant algorithm it was equal to (0.99818), while in the Bees Colony Optimization algorithm it was to (0.99564),The cost equal function exponential behavior model was used to determine the lowest allocation to the system's parts; Rs in the ant colony optimization was equal to (0.90461)algorithm and (0.90222) in the Bees Colony Optimization algorithm technique, as indicated in Table (1, 2, 3).

References

 S. A. K. Abbas and Z. A. Haddi Hassan, Increase the Reliability of Critical Units by Using Redundant Technologies, *Journal of Physics: Conference Series*, vol. 1999, no. 1, p. 012107, 2021/09/01 2021, doi: 10.1088/1742-6596/1999/1/012107.

- S. A. K. Abbas and Z. A. H. Hassan, Use of ARINC Approach method to evaluate the reliability assignment for mixed system, *Journal of Physics: Conference Series*, vol. 1999, no. 1, p. 012102, 2021/09/01 2021, doi: 10.1088/1742-6596/1999/1/012102.
- [3] F. H. Abd Alsharify and Z. A. Haddi Hassan, Computing the reliability of a complex network using two techniques, *Journal of Physics: Conference Series*, vol. 1963, no. 1, p. 012016, 2021/07/01 2021, doi: 10.1088/1742-6596/1963/1/012016.
- [4] F. H. Abd Alsharify, G. A. Mudhar, and Z. A. Haddi Hassan, A modified technique to compute the minimal path sets for the reliability of complex network, *Journal of Physics: Conference Series*, vol. 1999, no. 1, p. 012083, 2021/09/01 2021, doi: 10.1088/1742-6596/1999/1/012083.
- [5] G. Abdullah and Z. A. H. Hassan, Use of Bees Colony algorithm to allocate and improve reliability of complex network, *Journal of Physics: Conference Series*, vol. 1999, no. 1, p. 012081, 2021/09/01

2021, doi: 10.1088/1742-6596/1999/1/012081.

- [6] L. A. A. Ameer issa and Z. A. Haddi Hassan, Use of a modified Markov models for parallel reliability systems that are subject to maintenance, *Journal of Physics: Conference Series*, vol. 1999, no. 1, p. 012087, 2021/09/01 2021, doi: 10.1088/1742-6596/1999/1/012087.
- [7] Abdullah, G., Haddi Hassan, Z. A., 2020, Using of Genetic Algorithm to Evaluate Reliability Allocation and Optimization of Complex Network, IOP Conf. Ser.: Mater. Sci. Eng. 928(4) 0420333
- [8] Abdullah, G., Haddi Hassan, Z. A., 2020, Using of particle swarm optimization (PSO) to addressed reliability allocation of complex network, J. Phys.: Conf. Ser. 1664 (1) 012125.
- [9] Abdullah, G., Haddi Hassan, Z. A., 2021, A Comparison Between Genetic Algorithm and Practical Swarm to Investigate the Reliability Allocation of Complex Network, J. Phys.: Conf. Ser. 1818 (1) 012163.
- [10] Mutar, E. K. and Hassan Z. A. H., 2022, New properties of the equivalent reliability polynomial through the geometric representation, International Conference on Electrical, Computer and Energy Technologies (ICECET 2022), Prague-Czech Republic.
- [11] Abd Alsharify, F. H., & Hassan, Z. A. H.
 (2022). Optimization of complex system reliability: Bat algorithm based approach, International Journal of Health Sciences,6(S1), 14226–14232. https://doi.org/10.53730/ijhs.v6nS1.86 37.j
- [12] Hassan, Z. A. H. and Mutar, E. K., (2017), Geometry of reliability models of electrical system used inside spacecraft, 2017 Second Al-Sadiq International Conference on Multidisciplinary in IT and Communication Science and Applications (AIC-MITCSA), pp. 301-306.
- [13] Hassan, Z. A. H. and Balan, V., (2017), Fuzzy T-map estimates of complex circuit reliability, International

Conference on Current Research in Computer Science and Information Technology (ICCIT-2017), IEEE, Special issue, pp.136-139.

- [14] Hassan, Z. A. H. and Balan, V. (2015), Reliability extrema of a complex circuit on bi-variate slice classes, Karbala International Journal of Modern Science, vol. 1, no. 1, pp. 1-8.
- [15] Hassan, Z. A. H., Udriste, C. and Balan, V., (2016), Geometric properties of reliability polynomials, U.P.B. Sci. Bull., vol. 78, no. 1, pp. 3-12.
- [16] Saad Abbas Abed et al, 2019, Reliability Allocation and Optimization for (ROSS) of a Spacecraft by using Genetic Algorithm, J. Phys.: Conf. Ser. 1294 032034 10.1088/1742-6596/1294/3/032034
- [17] Hameed Saleh, A. A., & Haddi Hassan, Z. A.
 (2022). Addressing the problem of increasing the reliability of a mixed system. International Journal of Health Sciences,6(S5), 1013–1018. https://doi.org/10.53730/ijhs.v
 6nS5.8802