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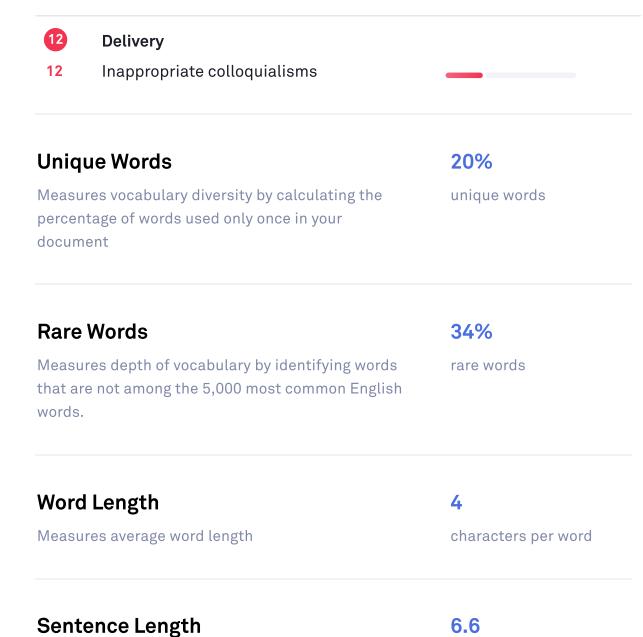
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Solving fuzzy shortest path problem by genetic algorithm

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Abstract: Shortest Path Problem (SPP) is known as one of well-studied fields in the area Operations Research and Mathematical Optimization. It has been applied for many engineering and management designs. The objective is usually to determine path(s) in the network with minimum total cost or traveling time. In the past, the cost value for each arc was usually assigned or estimated as a deteministic value. For some specific real world applications, however, it is often difficult to determine the cost value properly. One way of handling such uncertainty in decision making is by introducing fuzzy approach. With this situation, it will become difficult to solve the problem optimally. This paper presents the investigations on the application of Genetic Algorithm (GA) to a new SPP model in which the cost values are represented as Triangular Fuzzy Number (TFN). We adopts the concept of ranking fuzzy numbers to determine how good the solutions. Here, by giving his/her degree value, the decision maker can determine the range of objective value. This would be very valuable for decision support system in the real world applications. Simulation experiments were carried out by modifying several test problems with 10-25 nodes. It is noted that the proposed approach is capable attaining a good solution with different degree of optimism for the tested problems.

Keywords: Shortest Path Problem, Network Design Problem, Genetic Algorithm, Ranking Fuzzy Number, Decision Support System

1. Introduction

In business management, many important problems are expressed as network design problems. Among all classes of problems in network optimization, Shortest Path Problem (SPP) has been one of the most extensively disscused in computer science literature. It is commonly encountered in wide array of practical applications including transportation planning, communication, and production applications, salesperson routing and often imbedded within other types of network optimization problems [1], [2].

Given a weighed ³¹ network graph, the SPP is to obtain the minimum total weight path(s) between nodes in the network. In the real world applications, the weight value represents the average delivery time of the commodities, reliability of transportation, and accessibility to the users, product deterioration, and so on. ³² Thus, finding a good ³³ solution of SPP would be very important ³⁴ and could save millions of transportation and/or ³⁵ industrial cost, yet computationally difficult task. There have been several methods for solving traditional SPP including ³⁷ Dijkstra ³⁸ method [3]. However, due to the intrinsic difficulty, many authors introduced search methods based upon heuristics to give solutions within reasonable computational time [4], [5].

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Traditionally, SPP was investigated ⁴⁴ in the area of Operations Research and Artificial Intelligence that is modelled ⁴⁵ by certain/deterministic cost values. However, for many real world ⁴⁶ applications, the deterministic model can be far from sufficient. For instance, in a transportation network, the path that has the least travel <u>time</u>, ⁴⁷ during the rush hours would have traffic jams that may lead to a drastic increase in travel time. One way of handling such uncertainty in decision making is by introducing fuzzy programming approach [6].

Artificial Intelligence (AI) has been discussed in many recent research fields. It has been successfully used to solve many difficult-real-world optimization problems. Among them, Genetic ⁵⁰ algorithm (GA) has been known ⁵¹ as one of effective ⁵² and popular methods. Since introduced by Holland [7], it has been implemented to give optimal ⁵³ near optimal ⁵⁴ solution ⁵⁵ for many NP-hard optimization problems with less computational cost [8]. There are several advantages of using GA as follows: It deals with coding instead of decision variables. It requires no domain knowledge - only the payoff or objectives. In addition, ⁵⁶ traditional methods use deterministic transition rules to guide the

search, such as hill-climbing, and neighbourhood search. Another difference between traditional methods and genetic algorithms is: the latter searches from a set of points, while the former from a single point. This makes GA more robust than traditional methods regarding their potential as optimization techniques [9]. In our previous works, we also have successfully implemented GA to solve various combinatorial optimization problems and found that GA can give very good solutions to the problems within reasonable computational time [10], [11], [12], [13].

In this paper, we develop new model of SPP called fuzzy Shortest Path Problem (f-SPP). Here, the cost values of the f-SPP are represented by using Triangular Fuzzy numbers (TFN). Our major efforts in this work include the adaption of the the rangking method for handing the fuzziness. We adopt the concept of the ranking Fuzzy numbers with integral value for the evaluation and selection. In this technique, the decision maker can determine his decision by giving flexible value for the degree of optimism [14].

- 248 The remainder of this paper is organized as follows: The description of the mathematical formulation of the f-SPP is given in Section 2. Section 3 describes a brief discussion design of algorithm. It includes ranking Fuzzy numbers, representation of chromosome, genetic operations and selection method. Section 4 presents the numerical experiments and results obtained by the proposed method. Finally, some concluding remarks are given in Section 5.

2. Mathematical Model

250

In this section, we shall present a comprehensive mathematical model of the problem. Suppose that we are given ⁷⁹ a graph G in which thr ⁸⁰lenght ⁸¹ of each arc has associated with a TFN. In some applications, that TFN may actually ⁸²

represent traveling time or some other values. For any two vertices, s and t in G, it is possible that there exist several paths from s to t. The problem involves finding path(s) from s to t that has the shortest traveling time.

The model of $f^{\frac{84}{5}}$ SPP can mathematically be stated as follows:

min m n

ååcij xij

.....



i=1 j=1

n n

s.t.

åxij

– åxki

j=1 k =1

ì 1 (i =1) ï

= í 0 (i = 2, 3,..., n – 1) ï (i = n) î– 1



(1)
(2)
(3)
(4)

2

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ì1,



if $link \overset{86}{(i,j)}$ is included in the path

xij = í

î0,

otherwise

In the above model, directed graph G=(V, A) where V is a set of nodes, A is a set of links.

~

is a TFN

c ij⁸⁹

representing ⁹⁰ fuzzy cost associated with each arc(i, j). Source node and destination node are node 1 ⁹¹ and node n respectively. The length of a path is



defined as the sum of the lengths of all individual arcs comprising the path.⁹⁴

- 3. Design of the Proposed Algorithm
- 3.1. Ranking Fuzzy Number (TFN)

When considering an optimization problem that its coefficients are represented with TFNs, the objective values of the problem will also be TFNs. In a fuzzy environment, an effective, efficient, and accurate ranking method becomes a very important decision making procedure. Many authors have investigated various ranking methods for fuzzy numbers [15]. The ranking fuzzy number with integral value technique was proposed by Liou & Wang [14]. The basic concept of this technique is to rank the TFN based on its total integral value that represents a "mean value".⁹⁹ The left and right integral values are used to reflect the pessimistic and optimistic viewpoint of decision maker respectively.¹⁰¹ The total integral value is then computed based on this degree of optimism and each objective function values. A parameter α is given to adjust the degree of optimism.

```
Definition 1 [14]
```

```
~
```

Let A be a TFN with its membership function $\mu \sim as^{102}$ shown in Figure 1. <u>A TFN is</u> denoted by a triplet

А

(a1, a2, a3) where a1, a2, a3 are real numbers. i.e. a minimum value, a modal value and a maximum

~



value.¹⁰⁸ The membership function of a TFN <u>A</u>, ¹⁰⁹ A

A □ 0 (x - a1) (a2 - a1) (x) = í (a3 - x) (a3 - a2) ïî0ï □

~

,x<a1



, a1
ר א בי a2
22
, a2 ♪ x ♪ a3
(5)

,x>a3

Figure 1. The membership function of TFN

3

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From the definition of the above membership function, we can see that the left membership function

 μ ~ (x) L is continuous and strictly increasing. Its inverse function g ~ (x) L would exist and continuous 111



А

А

on interval [0, 1]. So we can compute its integral value on that interval. Similarly for the right

membership function μ ~ (x) R

. We can compute both left and right integral values as follows:







1



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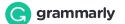
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l	L ¹¹⁷)	
	= ò0	
((x)	
=	=	



(a 2+a3)



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~ is

Thus, the total integral value for triangular fuzzy number A

~ R (1

~

~ L $\overset{^{119}}{\overset{}{}_{-}} \overset{^{119}}{(A)} = \alpha I (A)$ +– α)I (A) = 1 [α a 3 + a 2 $+ (1 - \alpha) a1$]..... (8)



2

The following definition is used $\frac{120}{10}$ relation to the ranking of the TFN:

Definition 2: Suppose that

~



~ ~ is a set of convex fuzzy numbers and that ranking fuzzy

S = {A1

•••

, An }

,A2,

~

~



number R is a mapping from S to the real numbers. For any distinct

Ai

, A j \in S the ranking has the

following properties.

~) < ~) then

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1. if R (Ai R (A j Ai

< A j

~) = ~



) then

- ~
- ~

- 2. if
- R (Ai
- R (A j
- Ai

~

= A j



) >		
~		
) then		
~		

~

3. if R (Ai R (A j Ai

> A j



Based on the above definition, thus for any fuzzy numbers Ai and A_j^{124}

we shall have:

~

if α ~ α



)		
then		
~		

~

L¹²⁵ (Ai

) < I T (<u>A j</u>¹²⁶

Ai

< A j

if



- α _____α
- ~) then
- ~
- ~

- I T¹²⁷
- (Ai
- $) = I T^{128}$ (A j¹²⁹
- Ai
- = A j





- α
- ~
- α
- ~

) then

- ~
- ~



> A <u>j</u>.¹³³

3.2. Design of Genetic Algorithm (GA)

Generally, the behaviour of GAs depends on many uncertain factors. One of the ¹³⁵ important factors is the balance between exploitation and exploration in the search space. The balance is strongly affected by design strategy for GA and GA parameter values. Here, we would like to describe the consideration underlying the design of GA ¹³⁷ approach including ¹³⁸ representation method, genetic operators and ¹³⁹ and ¹³⁹

3.2.1. Chromosome Representation and Evaluation

It is well known that the first step in applying GA is to find a representation for a possible solution. An efficient design of the chromosome would lead to $\underline{a \text{ good}}^{14^{\circ}}$ performance of GA.

For this problem, we adopt permutation-based representation. In this representation, the value of gene represents the priority of the node number for being included in the arc. ¹⁴³ There are several advantages of using this representation for network problems [9]. Those include, first, any permutation¹⁴

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of the encoding corresponds to a path (legality). Most existing genetic operators can be easily applied to the encoding. Any path has a corresponding encoding (completeness). And, any point in solution space is accessible for genetic search.

The following Figure 2 shows the illustration of priority based representation.

Figure 2. Priority-based chromosome

In order to compute the fitness value, we decode the above chromosome using the decoding procedure as follows:

Procedure: Decoding Step 1: number of nodes n, the alpha value the set of nodes <u>Si with</u>¹⁴⁹ all nodes adjacent to node i.

the set of the TFN weight values denoted by (a1, a2, a3)

Step 2 Determine path corresponding to the chromosome as follows: initial source node $i \leftarrow \overset{153}{1}$, Pk $\leftarrow \phi$; while Si $\neq \phi$ do select l from Si with the highest priority; if $vk(l) \neq 0$ then vk(l) = 0; Pk $\leftarrow PkU{xil}$;



157,<mark>158</mark> i←l;

else Si \leftarrow Si \{l}

end

output the complete path Pk ;

end

Step 3 Compute the fuzzy objective values

Step 4

Determine the integral values of TFNs

3.2.2. Crossover and Mutation

The next step of GA is to select chromosome for replication by using genetic operator, in order to have new chromosome to be evaluated further. Basically, GA use two genetic operators, crossover and mutation.

²⁵¹ Crossover is known as the most important recombination operator in GA. There have been many variants of crossover operations given in the literature for permutation based ¹⁶⁶ representation such as: ¹⁶⁷ PMX method (Partially Matched Crossover) [16], PX (Position-based crossover) [17] and WMX (Weight mapping crossover) [18].

In this paper, we adopt PMX crossover operations as follows:



Procedure : PMX

Step 1: Select a section of chromosome randomly

Selected sub string

5

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Parent 2

- ,



Step 2: Exchanged each sub string¹⁷¹

Parent 1

Parent 2



Step 3: Determine the mapping of genes in each substring¹⁷²

Step 4: Update chromosome with information given by Step 3

Offspring 2

- 7 6
- 2
- 3
- 4
- -
- 1
- 8
- 5

Mutation¹⁷⁴ is usually done to prevent premature loss of information by exchanging the information within a chromosome. In this paper, we adopt the

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inversion mutation. It is done by selecting two positions within a chromosome at random and then inverts the sub-string between these two positions. We show the illustration of this mutation operation in Figure 3

Parent
4
6
3
5
7
1
8
2

offspring

- 4
- 6
- 7



- 5 3 1 8
- 2

Figure 3. Illustration of inversion mutation

3.2.3. Selection

When implementing GA, the way to select chromosome for the next generation ¹⁷⁶ ¹⁷⁷ is also very important issue. It should provide a balance between exploitation and exploration to search solution ¹⁷⁸ fitness-proportional methods called roulette wheel selection [2]. If the best chromosome is not selected for the next generation, we select one of the ¹⁸⁰ chromosome randomly and replace it with the best chromosome. We express our selection method as follows:

Procedure: Selection

Step 1: Calculate a cumulative probability ap for each chromosome X \underline{p} , $(p = 1, 2, \dots, pop _ size)$.

Step 2: Generate a random real number r in [0,1].

Step 3: If r = a1, then select the first chromosome X1; otherwise select the pth chromosome X p, (2 r = p) such that a p - 1 < r < a p.

²⁵³ Step 4: Repeat Steps 2 and 3 pop_size times and obtain pop_size copies of chromosome.

Step 5: If the best chromosome is not selected ¹⁹⁰ in the next generation, replace one randomly from the new population by the best one.

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3.2.4. Overall Procedure

The overall GA procedure used in this paper is illustrated 192 as in Figure 4.

Figure 4. Overall procedure of proposed method

4. Experimental Design and Results

To explore how well the proposed GA can solve f-SPP. we have designed several numerical experiments. The proposed method was implemented by Using C language and run on PC Core i3. For the test problems, we modified the SPP test problems given [19], [2] and [20]. The size of test problems and GA 198 parameters used for these experiments are set as presented in Table 1:

Table 1. Problem size and GA parameter values

Test Number of Crossover Mutation Population 199 Maximimum Problem Node Probability Probability Size

G grammarly

generation
1
20
0.4
0.2
20
100
2
10
0.4
0.2
10
100
3
25
0.4
0.2
15
100

The first experiments were cunducted to see the effectiveness of GA with respect to the heterogeneity of the degree of optimism. To investigate this issue, for all test problems, we take a look at the variation of the degree of optimisms (0, 0.1, 0.2, ..., 1). For each different degree of optimism, the experiment was done 10 (ten) times. We note the best, the average and the worst results. The experimental results are summarized in Table 2.



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Table 2. Computational results for different alpha values

Alpha

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G grammarly Report: admi

Problem 1
0.0
0.1
0.2
0.3
0.4
0.5
0.6
0.7
0.8
0.9
1.0
Maximum
152,5
164,6
176,7
188,8
200,9
213
232,6
237,2
249,3
261,4
273,5
Minimum



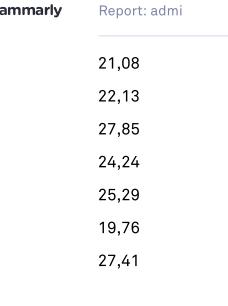
122,1
131,7
141,3
150,9
160,5
170,1
179,7
189,3
198,9
208,5
Average
120,5
130,6
140,7
155,55
160,9
171
187,35
191,2
201,3
205,2
221,5
STDEV

221,5 STDEV 16,86 17,91

18,97

22,94





Problem 2

- 0.0
- 0.1
- 0.2
- 0.3
- 0.4
- 0.5
- 0.6
- 0.7
- 0.8
- 0.9
- 1.0

Maximum

- 94
- 95,4
- 96,8
- 98,2
- 99,6
- 101



grammarly	Report: ad
	102,4
	103,9

107,3

107,3

109

Minimum

92

93,7

95,4

97,1

98,8

100,5

102,2

103,8

105,2

106,6

108

Average

92,6

94,21

95,96

97,21

99,04

100,65

102,24

103,8

105,44



106,7
108,3
STDEV
0,96
0,82
0,72
0,34
0,38
0,24
0,08
0,042
0,66
0,29
0,483

Problem 3

- 0.0
- 0.1
- 0.2
- 0.3
- 0.4
- ••••
- 0.5
- 0.6
- 0.7
- 0.8
- 0.9

G grammarly

1.0

Maximum

89

88,35

90,7

92,8

95,4

98

100,6

103,2

104,8

107,2

110

Minimum

85

87,6

90

92,5

94,5

96,75

99

101,3

103,5

105,8

108

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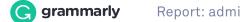


Average
85,85
87,825
90,07
92,56
94,59
97
99,32
101,4
103,63
106,2
108,2
STDEV
1,31
0,36
0,22
0,12
0,28
0,52
0,67
0.047

- 0,617
- 0,41
- 0,67
- 0,632

To observe the speed of convergence of the proposed GA, we noted the objective value in each generation for the test problem 3. We compared the results for three different alpha values (0, 0.5 and 208 1). The results are shown in Figure 5.

Objectiveval	
200	
180	
160	
140	



alpha =0,5

Aplha = 1

²⁵⁴ 1 11 21 31 41 51 61 71 81 91

Generation

Figure 5. The comparative convergence behavior for different alpha values

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We noted from the above results that, GA is very sensitive for overall optimization degrees. For different degree of optimism (alpha), GA can give different objective value all of the times. Thus by giving different alpha value, GA would be able to give decision maker solution for the problem. The above results also show that GA can convergence to the best solution very fast.

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5. Conclusion

The contribution of this paper consists of introducing the model of shortest path problem with fuzzy coefficient (fSPP) and providing GA-based algorithm to solve it. Here, the proposed method use the fuzzy ranking strategy based on the alpha value/degree of optimism. The performance of the proposed method was evaluated by using modified several test problems given in the literature. The findings reported in this study show that, GA is very sensitive for overall degrees of optimism. Thus, the decision makers can have their solution based on their degree of optimism. Moreover, the results also show that GA can²²⁷ convergence to the best solution very fast (less than 50 generations).

Acknowledgments

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Appendix: Test Problems



Figure 6. Test Problem 1 (modified from [19])

Figure 7 Test Problem 2 (modified from [2])



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Figure 8 Test Problem 3 (modified from [20])









1.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
2.	the well-studied	Determiner Use (a/an/the/this, etc.)	Correctness
3.	of Operations	Wrong or Missing Prepositions	Correctness
4.	been applied	Passive Voice Misuse	Clarity
5.	for → to	Wrong or Missing Prepositions	Correctness
6.	the path	Determiner Use (a/an/the/this, etc.)	Correctness
7.	deteministic → deterministic	Misspelled Words	Correctness
8.	real world → real-world	Misspelled Words	Correctness
9.	properly → accurately	Word Choice	Engagement
10.	a fuzzy	Determiner Use (a/an/the/this, etc.)	Correctness
11.	the Genetic	Determiner Use (a/an/the/this, etc.)	Correctness
12.	are represented	Passive Voice Misuse	Clarity
13.	<mark>adopts</mark> → adopt	Faulty Subject-Verb Agreement	Correctness
14.	decision maker → decision-maker	Misspelled Words	Correctness
15.	This	Intricate Text	Clarity
16.	a decision	Determiner Use (a/an/the/this, etc.)	Correctness
17.	the real	Determiner Use (a/an/the/this, etc.)	Correctness
18.	real world → real-world	Misspelled Words	Correctness

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19.	. Simulation	Improper Formatting	Correctness
20.	is noted	Passive Voice Misuse	Clarity
21.	capable → capably	Misuse of Modifiers	Correctness
22.	capable of	Wrong or Missing Prepositions	Correctness
23.	good → right	Word Choice	Engagement
24.	degree → degrees	Incorrect Noun Number	Correctness
25.	important → vital	Word Choice	Engagement
26.	are expressed	Passive Voice Misuse	Clarity
27.	disscused → discussed	Misspelled Words	Correctness
28.	a wide	Determiner Use (a/an/the/this, etc.)	Correctness
29.	, including	Punctuation in Compound/Complex Sentences	Correctness
30.	, and	Punctuation in Compound/Complex Sentences	Correctness
31.	weighed → weighted	Confused Words	Correctness
32.	and so on	Inappropriate Colloquialisms	Delivery
33.	good → right	Word Choice	Engagement
34.	very important → significant	Word Choice	Engagement
35.	and/or → and, or	Inappropriate Colloquialisms	Delivery
36.	difficult → tricky	Word Choice	Engagement
37.	, including	Punctuation in Compound/Complex Sentences	Correctness

38.	the Dijkstra	Determiner Use (a/an/the/this, etc.)	Correctness
39.	be used	Passive Voice Misuse	Clarity
40.	licence → license	Mixed Dialects of English	Correctness
41.	, and	Comma Misuse within Clauses	Correctness
42.	licence → license	Mixed Dialects of English	Correctness
43.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
44.	was investigated	Passive Voice Misuse	Clarity
45.	modelled → modeled	Mixed Dialects of English	Correctness
46.	real world → real-world	Misspelled Words	Correctness
47.	time,	Punctuation in Compound/Complex Sentences	Correctness
48.	been discussed	Passive Voice Misuse	Clarity
49.	recent → new	Word Choice	Engagement
50.	the Genetic	Determiner Use (a/an/the/this, etc.)	Correctness
51.	been known	Passive Voice Misuse	Clarity
52.	the effective	Determiner Use (a/an/the/this, etc.)	Correctness
53.	an optimal	Determiner Use (a/an/the/this, etc.)	Correctness
54.	near optimal → near-optimal	Misspelled Words	Correctness
55.	solution → solutions	Incorrect Noun Number	Correctness
56.	In addition → Also, Besides	Wordy Sentences	Clarity



57.	hill-climbing,	Punctuation in Compound/Complex Sentences	Correctness
58.	neighbourhood → neighborhood	Mixed Dialects of English	Correctness
59.	is:	Misuse of Semicolons, Quotation Marks, etc.	Correctness
60.	latter → recent, last	Word Choice	Engagement
61.	This	Intricate Text	Clarity
62.	very good → perfect, excellent, outstanding	Word Choice	Engagement
63.	a new	Determiner Use (a/an/the/this, etc.)	Correctness
64.	the fuzzy	Determiner Use (a/an/the/this, etc.)	Correctness
65.	are represented	Passive Voice Misuse	Clarity
66.	major → significant	Word Choice	Engagement
67.	the the rangking	Misspelled Words	Correctness
68.	rangking → ranking	Misspelled Words	Correctness
69.	ranking Fuzzy → Fuzzy ranking	Misplaced Words or Phrases	Correctness
70.	decision maker → decision-maker	Misspelled Words	Correctness
71.	is organized	Passive Voice Misuse	Clarity
72.	is given	Passive Voice Misuse	Clarity
73.	The remainder of this paper is organized as follows: The description of the mathematical formulation of the f-SPP is given in Section 2.	Wordy Sentences	Clarity

74.	the algorithm	Determiner Use (a/an/the/this, etc.)	Correctness
75.	ranking Fuzzy → Fuzzy ranking	Misplaced Words or Phrases	Correctness
76.	, and	Punctuation in Compound/Complex Sentences	Correctness
77.	Section 3 describes a brief discussion design of algorithm. It includes ranking Fuzzy numbers, representation of chromosome, genetic operations and selection method. Section 4 presents the numerical experiments and results obtained by the proposed method.	Monotonous Sentences	Engagement
78.	are given	Passive Voice Misuse	Clarity
79.	are given	Passive Voice Misuse	Clarity
80.	$\frac{\text{thr}}{\text{thr}} \rightarrow \text{the}$	Misspelled Words	Correctness
81.	<mark>lenght</mark> → length	Misspelled Words	Correctness
82.	actually	Wordy Sentences	Clarity
83.	a path, or the path	Determiner Use (a/an/the/this, etc.)	Correctness
84.	off → off	Confused Words	Correctness
85.	be stated	Passive Voice Misuse	Clarity
86.	a link, or the link	Determiner Use (a/an/the/this, etc.)	Correctness
87.	i, i ,	Improper Formatting	Correctness
88.	is included	Passive Voice Misuse	Clarity
89.	ij	Unknown Words	Correctness

90.	$\frac{1}{1}$ representing \rightarrow Representing	Improper Formatting	Correctness
91.	1 → one	Improper Formatting	Correctness
92.	, respectively	Punctuation in Compound/Complex Sentences	Correctness
93.	is defined	Passive Voice Misuse	Clarity
94.	The length of a path is defined as the sum of the lengths of all individual arcs comprising the path.	Wordy Sentences	Clarity
95.	are represented	Passive Voice Misuse	Clarity
96.	a very important → a critical, a crucial, a significant, a vital	Word Choice	Engagement
97.	ranking fuzzy → fuzzy ranking	Misplaced Words or Phrases	Correctness
98.	an integral	Determiner Use (a/an/the/this, etc.)	Correctness
99.	$\stackrel{"}{\rightarrow}$."	Misuse of Semicolons, Quotation Marks, etc.	Correctness
100.	decision maker → decision-maker	Misspelled Words	Correctness
101.	, respectively	Punctuation in Compound/Complex Sentences	Correctness
102.	, as	Punctuation in Compound/Complex Sentences	Correctness
103.	A triplet denotes a TFN	Passive Voice Misuse	Clarity
104.	where → Where	Improper Formatting	Correctness
105.	a1, →a1,	Improper Formatting	Correctness
106.	$2 \rightarrow 2$,	Improper Formatting	Correctness

i.e.,	Comma Misuse within Clauses	Correctness
valuo → Value	Improper Formatting	Correctness
$A, \rightarrow A,$	Improper Formatting	Correctness
<mark>doi</mark> → DOI	Misspelled Words	Correctness
continuous.	Closing Punctuation	Correctness
So wo → So we	Improper Formatting	Correctness
<mark>can compute</mark> → can compute	Improper Formatting	Correctness
compute its → compute its	Improper Formatting	Correctness
Ι	Inappropriate Colloquialisms	Delivery
dy → by	Misspelled Words	Correctness
I	Inappropriate Colloquialisms	Delivery
$dy \rightarrow by$	Misspelled Words	Correctness
1	Inappropriate Colloquialisms	Delivery
is used	Passive Voice Misuse	Clarity
in relation to → about, to, with, concerning	Wordy Sentences	Clarity
<mark>number</mark> → Number	Improper Formatting	Correctness
following → I am following, you are following	Incomplete Sentences	Correctness
<mark>A-j</mark> → Aj	Confused Words	Correctness
I	Inappropriate Colloquialisms	Delivery
<mark>A-j</mark> → Aj	Confused Words	Correctness

127.	1	Inappropriate Colloquialisms	Delivery
128.	1	Inappropriate Colloquialisms	Delivery
129.	<mark>A-j</mark> → Aj	Confused Words	Correctness
130.	Ι	Inappropriate Colloquialisms	Delivery
131.	Ι	Inappropriate Colloquialisms	Delivery
132.	<mark>A-j</mark> → Aj	Confused Words	Correctness
133.	j. → j.	Improper Formatting	Correctness
134.	behaviour → behavior	Mixed Dialects of English	Correctness
135.	important → critical, crucial, essential	Word Choice	Engagement
136.	the design	Determiner Use (a/an/the/this, etc.)	Correctness
137.	the GA	Determiner Use (a/an/the/this, etc.)	Correctness
138.	, including	Punctuation in Compound/Complex Sentences	Correctness
139.	, and	Punctuation in Compound/Complex Sentences	Correctness
140.	is well known	Passive Voice Misuse	Clarity
141.	a good → an excellent	Word Choice	Engagement
142.	the gene	Determiner Use (a/an/the/this, etc.)	Correctness
143.	In this representation, the value of gene represents the priority of the node number for being included in the arc.	Wordy Sentences	Clarity

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144.	permutation.	Closing Punctuation	Correctness
145.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
146.	<mark>ef</mark> → 0f	Improper Formatting	Correctness
147.	be easily applied	Passive Voice Misuse	Clarity
148.	And → Moreover, Furthermore	Inappropriate Colloquialisms	Delivery
149.	Si with \rightarrow Si with	Improper Formatting	Correctness
150.	<mark>a1,</mark> →a1,	Improper Formatting	Correctness
151.	a2 , → a2,	Improper Formatting	Correctness
152.	the path	Determiner Use (a/an/the/this, etc.)	Correctness
153.	∔ → I	Misspelled Words	Correctness
154.	<mark>vk</mark> → ok	Misspelled Words	Correctness
155.	vk	Unknown Words	Correctness
156.	Pk ← Pk	Misspelled Words	Correctness
157.	i	Inappropriate Colloquialisms	Delivery
158.	i → I	Misspelled Words	Correctness
159.	a chromosome	Determiner Use (a/an/the/this, etc.)	Correctness
160.	a genetic	Determiner Use (a/an/the/this, etc.)	Correctness
161.	operator,	Punctuation in Compound/Complex Sentences	Correctness
162.	a new	Determiner Use (a/an/the/this,	Correctness

	etc.)	
Basically,	Wordy Sentences	Clarity
uses → uses	Faulty Subject-Verb Agreement	Correctness
mportant → critical, crucial	Word Choice	Engagement
permutation-based	Misspelled Words	Correctness
as:	Misuse of Semicolons, Quotation Marks, etc.	Correctness
Procedure : → Procedure:	Improper Formatting	Correctness
sub string \rightarrow substring	Confused Words	Correctness
loi → DOI	Misspelled Words	Correctness
sub string → substring	Confused Words	Correctness
substring.	Closing Punctuation	Correctness
he information	Determiner Use (a/an/the/this, etc.)	Correctness
The mutation, or A mutation	Determiner Use (a/an/the/this, etc.)	Correctness
a chromosome	Determiner Use (a/an/the/this, etc.)	Correctness
a very	Determiner Use (a/an/the/this, etc.)	Correctness
very important → critical, significant, fundamental, crucial	Word Choice	Engagement
solution → solutions	Incorrect Noun Number	Correctness

is not selected	Passive Voice Misuse	Clarity
shromosome → chromosomes	Incorrect Noun Number	Correctnes
∋, → p,	Improper Formatting	Correctnes
21, → a1,	Improper Formatting	Correctnes
(1; →X1;	Improper Formatting	Correctnes
otherwise,	Punctuation in Compound/Complex Sentences	Correctnes
o th → path	Misspelled Words	Correctnes
р,	Punctuation in Compound/Complex Sentences	Correctnes
, → p,	Improper Formatting	Correctnes
 → p.	Improper Formatting	Correctnes
a chromosome, or the chromosome	Determiner Use (a/an/the/this, etc.)	Correctnes
s not selected	Passive Voice Misuse	Clarity
loi → DOI	Misspelled Words	Correctne
s illustrated	Passive Voice Misuse	Clarity
86	Wrong or Missing Prepositions	Correctnes
The overall	Determiner Use (a/an/the/this, etc.)	Correctnes
he proposed	Determiner Use (a/an/the/this, etc.)	Correctnes
	Improper Formatting	Correctnes
ve → We		

		Compound/Complex Sentences	
198.	are set	Passive Voice Misuse	Clarity
199.	Maximimum → Maximum	Misspelled Words	Correctness
200.	cunducted → conducted	Misspelled Words	Correctness
201.	with respect to \rightarrow concerning, for, to	Wordy Sentences	Clarity
202.	The first experiments were cunducted to see the effectiveness of GA with respect to the heterogeneity of the degree of optimism.	Wordy Sentences	Clarity
203.	optimisms → optimism	Confused Words	Correctness
204.	was done	Passive Voice Misuse	Clarity
205.	, and	Punctuation in Compound/Complex Sentences	Correctness
206.	are summarized	Passive Voice Misuse	Clarity
207.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
208.	, and	Punctuation in Compound/Complex Sentences	Correctness
209.	are shown	Passive Voice Misuse	Clarity
210.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
211.	a different	Determiner Use (a/an/the/this, etc.)	Correctness
212.	degree → degrees	Incorrect Noun Number	Correctness
213.	value → values	Incorrect Noun Number	Correctness
214.	times → time	Incorrect Noun Number	Correctness

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/alue → values	Incorrect Noun Number	Correctness
a decision	Determiner Use (a/an/the/this, etc.)	Correctness
lecision maker → decision-maker	Misspelled Words	Correctness
an provide, can achieve, can see, an facilitate	Incomplete Sentences	Correctness
he shortest	Determiner Use (a/an/the/this, etc.)	Correctness
fuzzy	Determiner Use (a/an/the/this, etc.)	Correctness
SPP → SPP	Misspelled Words	Correctness
GA-based, or the GA-based	Determiner Use (a/an/the/this, etc.)	Correctness
<mark>lse</mark> → uses	Faulty Subject-Verb Agreement	Correctness
vas evaluated	Passive Voice Misuse	Clarity
that,	Punctuation in Compound/Complex Sentences	Correctness
lecision makers → decision-makers	Misspelled Words	Correctness
an provide, can achieve, can see, an facilitate	Incomplete Sentences	Correctness
acknowledge → acknowledges	Faulty Subject-Verb Agreement	Correctness
<mark>The</mark> → the	Confused Words	Correctness
<mark>M-S</mark> → MS	Confused Words	Correctness
Gen M and Cheng R 1997 Genetic Algorithm and Engineering Design, 2nd ed. (New York: John Willey and Sons).	Incomplete Sentences	Correctness



232.	E-₩ → EW	Confused Words	Correctness
233.	, and	Comma Misuse within Clauses	Correctness
234.	Society → Society	Misspelled Words	Correctness
235.	<mark>¥-S</mark> → YS	Confused Words	Correctness
236.	, and	Punctuation in Compound/Complex Sentences	Correctness
237.	Two Stage → Two-Stage	Misspelled Words	Correctness
238.	Based on	Wrong or Missing Prepositions	Correctness
239.	, and	Punctuation in Compound/Complex Sentences	Correctness
240.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
241.	Concernig → Concerning	Misspelled Words	Correctness
242.	<mark>Koufman</mark> → Kaufman	Misspelled Words	Correctness
243.	, and	Punctuation in Compound/Complex Sentences	Correctness
244.	Hashier → Cashier	Misspelled Words	Correctness
245.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
246.	<mark>doi</mark> → DOI	Misspelled Words	Correctness
247.	Content from this work may be used under the terms of the Creative Commons Attribution 3.0 licence. Any further distribution of this work must maintain attribution to the author(s) and the title of the work, journal citation and DOI.	Quantifying the consensus on anthropogenic global warming <u>https://iopscience.iop.org/article/</u> <u>10.1088/1748-</u> <u>9326/8/2/024024/pdf</u>	Originality
248.	The remainder of this paper is	Information Free Full-Text	Originality



given in Section 5. 2. Mathematical ModelVARIOUS GENETIC ALGORITHM http://www.arpnjournals.org/jeas/ research_papers/rp_2016/jeas_0 416_4013.pdf0.For any two vertices, s and t inRemovable paths and cycles with parity constraints https://www.sciencedirect.com/s cience/article/pii/S009589561400 0069	Originality
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<i>important recombination operator in</i> <i>GA.</i> VARIOUS GENETIC ALGORITHM <u>http://www.arpnjournals.org/jeas/</u> <u>research_papers/rp_2016/jeas_0</u>	Originality
	Originality
 2. pop _ size). Step 2: Generate a random real number r in [0, A robust optimization model for green regional logistics https://journals.sagepub.com/doi/full/10.1177/1687814015620518 	Originality
 Step 4: Repeat Steps 2 and 3 pop_size times and obtain pop_size copies of A robust optimization model for green regional logistics <u>https://journals.sagepub.com/doi/</u> full/10.1177/1687814015620518 	Originality
. 1112131415161718191 Kingston Waterfront Real Estate - Neil Swanson Properties <u>http://neilswansonproperties.com</u> <u>/</u>	Originality
The performance of the proposed method was evaluated byA Functional Specialization-Based Multiobjective https://onlinelibrary.wiley.com/doi /abs/10.1002/ecj.12030	Originality
The findings reported in this study show that,Gender, Spousal Caregiving, and Depression: Does Paid Work Matter?	

257.	2nd ed. (New York: John Wiley and Sons	Airborne Transmission of Viral Respiratory Pathogens: Don't Stand So Close to Me?	Originality
258.	Holland J 1975 Adaptation in Natural and Artificial Systems	Emergence of Swarming Behavior: Foraging Agents Evolve Collective Motion Based on Signaling	Originality
259.	Syarif A, Yun Y S and Gen M 2002 Study on Multi-stage	PERFORMANCE EVALUATION OF VARIOUS GENETIC ALGORITHM http://www.arpnjournals.org/jeas/ research_papers/rp_2016/jeas_0 416_4013.pdf	Originality
260.	Syarif A and Gen M 2003 Solving Exclusionary Side	PERFORMANCE EVALUATION OF VARIOUS GENETIC ALGORITHM http://www.arpnjournals.org/jeas/ research_papers/rp_2016/jeas_0 416_4013.pdf	Originality
261.	Syarif A and Gen M 2003 Double Spanning	PERFORMANCE EVALUATION OF VARIOUS GENETIC ALGORITHM http://www.arpnjournals.org/jeas/ research_papers/rp_2016/jeas_0 416_4013.pdf	Originality