



## Neutrosophic approach to Dynamic Programming on group Decision Making problems

A. Kanchana<sup>1</sup>, D. Nagarajan<sup>2,\*</sup>, Broumi Said<sup>3</sup>

<sup>1,2</sup>Department of Mathematics, Rajalakshmi Institute of Technology, Chennai, India

<sup>3</sup>Faculty of Science Ben M'Sik, University of Hassan II, Casablanca, Morocco

<sup>3</sup>Regional Center for the Professions of Education and Training (CRMEF), Casablanca-Settat, Morocco

Emails: Kanchana.anbzhagan@gmail.com; dnrmu2002@yahoo.com; broumisaid78@gmail.com

Corresponding author: D.Nagarajan ; dnrmu2002@yahoo.com

### Abstract

In this article, complicated group decision-making situations where the preference data is represented by linguistic variables are addressed using the dynamic programming approach. Making conclusions clear through accurate figures is difficult for decision-makers due to the complexity and ambiguity of reality. Neutrosophic is used to encode the linguistic variables because they cannot be directly computed. Neutrosophic sets are used to manage indeterminacy in a practical situation. The relationships between single and interval Neutrosophic sets are then measured using novel distance and similarity models. The suggested dynamic programming interval-based clustering methodology is then used to group the decision-makers. Additionally, a novel method for computing the interval weights of decision-makers and clusters is described, accounting for both the cluster center and group size. A centroid-based ranking system is then used to compare and order the possibilities, and illustrated experiments are presented to demonstrate how effectively the suggested technique operates. Comparisons and discussions are also done to show its superiority.

**Keywords:** Dynamic programming; neutrosophic system; score function; clustering the decision makers.

### 1. Introduction

In order to approve the independent activities by considering the available sources, dynamic programming is the best way to take the correct decisions. Decision making problem is mainly affected by a criterion called dimension which is used to obtain the optimum solution for the problem. The decision with one criterion is called one dimensional problem otherwise it is multi-dimensional. The multi-dimensional problem was solved in [1],[2],[3].

The neutrosophic logic was founded by the American mathematician Smarandache in 1995. His interest in fuzzy will create the light to the neutrosophic logic [4],[5],[6]. The usefulness of this logic is spread fast in many researcher in various of applications [7] – [17] in the passage of time. In [18], neutrosophic fuzzy logic is used to solve the dynamic programming. The method used is to find the optimal profit of the problem. But it can be applicable for only the single phase diagram. If the decision makers were more and not in the regular form, the above method cannot be applicable. Hence our aim is to get the optimal path for the above problem which is also suitable for small number of decision makers also. In [19], we are going to take two different problems and going to clustering the decision makers in simple way.

Here our goal is use to single value neutrosophic system (SVNS) to cluster the decision makers for single diagram as well as the multiple diagram. Hence section 2 follows preliminaries followed by proposed method in section 3. In section 4 illustrative example and finally in section 5 conclusion is given.

## 2. PRELIMINARIES:

In this section, some basic definitions of Neutrosophic(NS), Single value Neutrosophic(SVNS) ,score function were defined, which will be used in the forthcoming sections.

### **Definition 2.1:**

The Neutrosophic set [20] for the universal set  $X$  is defined as  $N = \{(\hat{T}_S(x), \hat{I}_S(x), \hat{F}_S(x)), x \in X\}$  where  $\hat{T}_S(x), \hat{I}_S(x), \hat{F}_S(x) \in [0,1]$  and  $0 \leq \hat{T}_S(x) + \hat{I}_S(x) + \hat{F}_S(x) \leq 3$ .

### **Definition 2.2:**

The one dimensional logic equation for dynamic programming is defined as

$$\hat{Z}_i = \hat{F}_i(x_i) \quad i = 1, 2, \dots, n$$

### **Definition 2.3:**

The relationship [18] between the current function  $\hat{Z}_i$  and the previous function  $\hat{Z}_{i-1}$  at the stage  $i$  is given by

$$\hat{Z}_i = \hat{g}_i(x_i, z_{i-1}) \quad i = 1, 2, \dots, n$$

### **Definition 2.4:**

Let us we define the SVNS score function,

$$S^*(d_i) = S^*(u^T, v^I, w^F) = \frac{1}{2} \left( u^T - v^I + \left( \frac{w^F}{2} \right) \right) \quad (1)$$

Where  $d_i = (u^T, v^I, w^F)$ .

## 3. Proposed Neutrosopic Dynamic Programming:

To solve the complex decision problems, first we have to invite the  $p$  decision makers  $D^* = \{d_1, d_2, \dots, d_p\}$  with the similarity value is given  $D^* = [\beta_{ij}]_{n \times n}$ . In each step we have to optimize the path which is calculated from backwards and the decision makers present in the path have been clustered separately. Continue the process until all the decision makers have been alloted in any one of the clusters.

*Step 1:* Consider the problem of  $p$  decision makers  $D^* = \{d_1, d_2, \dots, d_p\}$  with the similarity value is given  $D^* = [\beta_{ij}]_{n \times n}$ .

*Step 2:* Fix the threshold for the similarity values as  $\alpha = 0.5$ . If the similarity value between  $d_i$  and  $d_j$  is greater than equal to the threshold value then the path is acceptable otherwise not. That is

$$d^*(d_i, d_j) \geq \alpha, \quad i \neq j \quad (2)$$

*Step 3:* Start from backwards consider  $d_p$  with the similarity value of remaining decision makers with the condition (2). Convert all the values to the score function using (1).

*Step 4:* Convert the problem in single value neutrosophic system.

*Step 5:* The proposed equation is

$$d_\gamma = \max\{S^*(d_i)\} \quad \text{where } \gamma, i = 1, 2, \dots, p-1, \gamma \neq i \quad (3)$$

If the value is not equal to zero. Then the path is travelling from  $d_p$  to  $d_\gamma$  otherwise not.

*Step 6:* If  $d_\gamma$  leads same value for two or more than two decision makers, both the decision makers were considered and repeat the Step 5 for both the decision makers. Among them the minimum value is considered.

*Step 7:* Consider  $d_\gamma$  with the similarity value satisfying (2) of other decision makers  $d_i$ , where  $i = 1, 2, \dots, \gamma-1$ . Then,

$$d_\delta = \max\{S^*(d_i)\} + d_\gamma \quad \text{where } \delta, i = 1, 2, \dots, \gamma - 1 \quad (4)$$

Step 8: Continue the step 5 and step 6 until the path stops with the decision makers which satisfying (2).

Step 9: Then the decision makers in this path is grouped into a new cluster. Continue the procedure until all the decision makers should present in any one of the clusters.

**4. Illustrative example:**

Here we are going to discuss two problems present in [19]

Example 1:

Suppose there are five decision makers whose similarity value is

Table 1: similarity value of five decision makers

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$
$d_1$	1	0.6	0.25	0.7	0.3
$d_2$	0.6	1	0.9	0.5	0.46
$d_3$	0.25	0.9	1	0.38	0.4
$d_4$	0.7	0.5	0.38	1	0.5
$d_5$	0.3	0.46	0.4	0.5	1

Step 1: Consider the problem of 5 decision makers  $D^* = \{d_1, d_2, \dots, d_5\}$  with the similarity value is given  $D^* = [\beta_{ij}]_{n \times n}$  which is given by table 1.

Step 2: Fix the threshold for the similarity values as  $\alpha = 0.5$  and consider only the value satisfying the condition (2). Then the table 1 becomes table 2 and the diagrammatic representation is given in fig 1.

Table 2: similarity value satisfying condition (2)

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$
$d_1$	1	0.6		0.7	
$d_2$	0.6	1	0.9	0.5	
$d_3$		0.9	1		
$d_4$	0.7	0.5		1	0.5
$d_5$				0.5	1

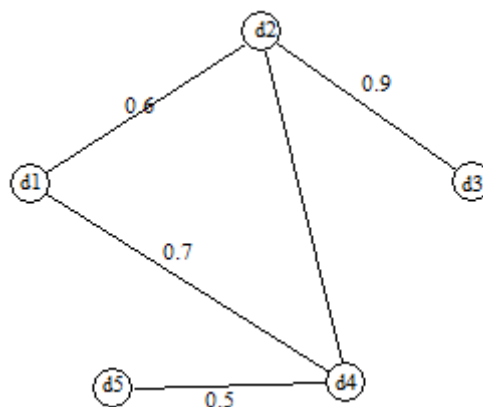


Figure 1: diagrammatic representation of table 2

Step 3,4,5: Consider the decision maker  $d_5$  with the similarity value is given in table 3

Table 3: Neutrosophic and score functional value of  $d_5$

Considered decision maker	Path moving to decision maker	Similarity value	Neutrosophic number	Score function value by (1)
$d_5$	$d_4$	0.5	(0.35,0.1,0.05)	0.2625

Hence the path is moving from  $d_5$  to  $d_4$

Step 7: Consider the decision maker  $d_4$  with the similarity value is given in table 4

Table 4: Neutrosophic and score functional value of  $d_4$

Considered decision maker	Path moving to decision maker	Similarity value	Neutrosophic number	Score function value by (1)	$d_8$ by (4) is maximum of
$d_4$	$d_1$	0.7	(0.49,0.14,0.07)	0.3675	0.63
	$d_2$	0.5	(0.35,0.1,0.05)	0.2625	0.525

The maximum value is 0.63. hence the next path leads to  $d_1$ .

Hence the path is moving from  $d_5$  to  $d_4$  to  $d_1$

Hence the cluster 1 groups  $\{d_5, d_4, d_1\}$ .

Step 8: Among the remaining decision makers, consider the next highest  $d_3$  with the similarity value is given in table 5

Table 5: Neutrosophic and score functional value of  $d_3$

Considered decision maker	Path moving to decision maker	Similarity value	Neutrosophic number	Score function value by (1)
$d_3$	$d_2$	0.9	(0.63,0.180,0.09)	0.4725

Hence the path is moving from  $d_3$  to  $d_2$ . Since there were no decision makers left over. The cluster 2 groups  $\{d_3, d_2\}$ .

Step 9: Hence the problem concludes with the solution given in table 6 and diagrammatic representation is given in fig 2.

Table 6: solution for example 1

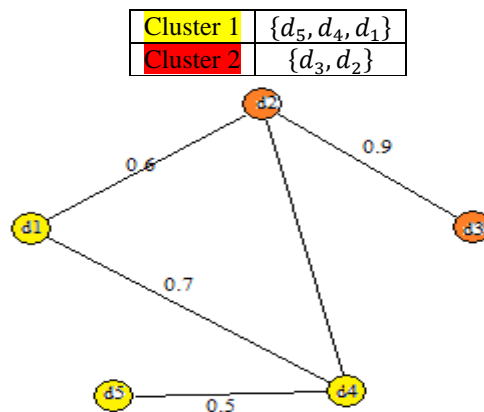


Fig 2: Diagrammatic representation of solution for example 1

Example 2:

Suppose there are twenty decision makers whose similarity value is

Table 7: similarity value of twenty decision makers

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$	$d_6$	$d_7$	$d_8$	$d_9$	$d_{10}$	$d_{11}$	$d_{12}$	$d_{13}$	$d_{14}$	$d_{15}$	$d_{16}$	$d_{17}$	$d_{18}$	$d_{19}$	$d_{20}$
$d_1$	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_2$	56	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_3$	22	22	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_4$	22	22	22	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_5$	33	33	33	33	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_6$	22	22	22	22	22	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_7$	78	78	78	78	78	78	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_8$	1	1	1	1	1	1	1	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_9$	33	33	33	33	33	33	33	33	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_{10}$	22	22	22	22	22	22	22	22	22	1	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_{11}$	22	22	22	22	22	22	22	22	22	22	1	0.	0.	0.	0.	0.	0.	0.	0.	0.
$d_{12}$	22	22	22	22	22	22	22	22	22	22	22	1	0.	0.	0.	0.	0.	0.	0.	0.
$d_{13}$	56	56	56	56	56	56	56	56	56	56	56	56	1	0.	0.	0.	0.	0.	0.	0.
$d_{14}$	1	1	1	1	1	1	1	1	1	1	1	1	1	1	0.	0.	0.	0.	0.	0.
$d_{15}$	11	11	11	11	11	11	11	11	11	11	11	11	11	11	1	0.	0.	0.	0.	0.
$d_{16}$	33	33	33	33	33	33	33	33	33	33	33	33	33	33	33	1	0.	0.	0.	0.
$d_{17}$	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	1	0.	0.	0.
$d_{18}$	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	1	0.	0.
$d_{19}$	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	11	1	0.
$d_{20}$	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	78	1

$d_2$	0.56	1	0.11	0.33	0.44	0.22	0.44	0.56	0.33	0.11	0	0.11	1	0.33	0.22	0.33	0.56	0.11	0.22	0.78
$d_3$	0.22	0.11	1	0.44	0.44	0.78	0.11	0.22	0.44	0.89	1	0.78	0.11	0.22	0.56	0.44	0.11	0.56	0.11	0.11
$d_4$	0.22	0.33	0.44	1	1	0.67	0.33	0.33	0.89	0.56	0.56	0.56	0.33	0.22	0.67	1	0.33	0.56	0.56	0.33
$d_5$	0.33	0.44	0.44	1	1	0.67	0.22	0.44	1	0.56	0.56	0.56	0.33	0.33	0.67	1	0.33	0.56	0.44	0.33
$d_6$	0.22	0.22	0.78	0.67	0.67	1	0.22	0.22	0.78	1	0.67	1	0.22	0.22	0.78	0.67	0.22	1	0.78	0.22
$d_7$	0.78	0.44	0.11	0.22	0.22	0.22	1	0.89	0.33	0.22	0.22	0.22	0.78	0.78	0.11	0.33	0.78	0.11	0.22	0.89
$d_8$	1	0.56	0.22	0.44	0.44	0.22	0.89	1	0.33	0.22	0.33	0.22	0.56	0.67	0.11	0.44	0.89	0.22	0.33	0.56
$d_9$	0.33	0.33	0.44	1	1	0.78	0.33	0.33	1	0.56	0.56	0.67	0.22	0.33	0.67	0.67	0.22	1	0.89	0.22
$d_{10}$	0.22	0.11	0.89	0.56	0.56	1	0.22	0.22	0.56	1	0.78	1	0.11	0.22	0.89	0.67	0.22	0.89	0.89	0.22
$d_{11}$	0.22	0	1	0.56	0.56	0.67	0.22	0.33	0.56	0.78	1	0.11	0	0.22	0.44	0.56	0.22	0.33	0.67	0
$d_{12}$	0.22	0.11	0.78	0.56	0.56	1	0.22	0.22	0.67	1	0.11	1	0.11	0.22	0.89	0.67	0.22	1	1	0.55
$d_{13}$	0.56	1	0.11	0.33	0.33	0.22	0.78	0.56	0.22	0.11	0	0.11	1	0.56	0.22	0.33	0.67	0.11	0.11	1
$d_{14}$	1	0.33	0.22	0.22	0.33	0.22	0.78	0.67	0.33	0.22	0.22	0.22	0.56	1	0.22	0.33	0.89	0.22	0.11	0.78
$d_{15}$	0.11	0.22	0.56	0.67	0.67	0.78	0.11	0.11	0.67	0.89	0.44	0.89	0.22	0.22	1	0.67	0.22	1	0.67	0.22
$d_{16}$	0.33	0.33	0.44	1	1	0.67	0.33	0.44	0.67	0.67	0.56	0.67	0.33	0.33	0.67	1	0.33	0.67	0.56	0.33
$d_{17}$	0.78	0.56	0.11	0.33	0.33	0.22	0.78	0.89	0.22	0.22	0.22	0.22	0.67	0.89	0.22	0.33	1	0.22	0.22	0.67
$d_{18}$	0.11	0.11	0.56	0.56	0.56	1	0.11	0.22	1	0.89	0.33	1	0.11	0.22	1	0.67	0.22	1	0.78	0.22
$d_{19}$	0.11	0.22	0.11	0.56	0.44	0.78	0.22	0.33	0.89	0.89	0.67	1	0.11	0.11	0.67	0.56	0.22	0.78	1	0.11
$d_{20}$	0.78	0.78	0.11	0.33	0.33	0.22	0.89	0.56	0.22	0.22	0	0.22	1	0.78	0.22	0.33	0.67	0.22	0.11	1

Step 1: Consider the problem of 20 decision makers  $D^* = \{d_1, d_2, \dots, d_{20}\}$  with the similarity value is given  $D^* = [\beta_{ij}]_{n \times n}$  which is given by table 7.

Step 2: Fix the threshold for the similarity values as  $\alpha = 0.5$  and consider only the value satisfying the condition (2). Then the table 7 becomes table 8 and the diagrammatic representation is given in fig 3.

Table 8: similarity value satisfying condition (2)

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$	$d_6$	$d_7$	$d_8$	$d_9$	$d_{10}$	$d_{11}$	$d_{12}$	$d_{13}$	$d_{14}$	$d_{15}$	$d_{16}$	$d_{17}$	$d_{18}$	$d_{19}$	$d_{20}$
$d_1$	1	0.56					0.78	1					0.56	1			0.78			0.78
$d_2$	0.56	1						0.56					1				0.56			0.78
$d_3$			1			0.78				0.89	1	0.78			0.56			0.56		
$d_4$				1	1	0.67			0.89	0.56	0.56	0.56			0.67	1		0.56	0.56	
$d_5$				1	1	0.67				0.56	0.56	0.56			0.67	1		0.56		
$d_6$			0.	0.	0.	1			0.	1	0.	1			0.	0.		1	0.	

			78	67	67				78		67			78	67			78	
$d_7$	0.78						1	0.89					0.78	0.78			0.78		0.89
$d_8$	1	0.56					0.89	1					0.56	0.67			0.89		0.56
$d_9$				1	1		0.78		1	0.56	0.56	0.67			0.67	0.67		1	0.89
$d_{10}$			0.89	0.56	0.56	1			0.56	1	0.78	1			0.89	0.67		0.89	0.89
$d_{11}$			1	0.56	0.56	0.67			0.56	0.78	1				0.56				0.67
$d_{12}$			0.78	0.56	0.56	1			0.67	1		1			0.89	0.67		1	1
$d_{13}$	0.56	1					0.78	0.56					1	0.56			0.67		1
$d_{14}$	1						0.78	0.67					0.56	1			0.89		0.78
$d_{15}$			0.56	0.67	0.67	0.78			0.67	0.89		0.89			1	0.67		1	0.67
$d_{16}$				1	1	0.67			0.67	0.67	0.56	0.67			0.67	1		0.67	0.56
$d_{17}$	0.78	0.56					0.78	0.89					0.67	0.89			1		0.67
$d_{18}$			0.56	0.56	0.56	1			1	0.89		1			1	0.67		1	0.78
$d_{19}$				0.56		0.78			0.89	0.89	0.67	1			0.67	0.56		0.78	1
$d_{20}$	0.78	0.78					0.89	0.56					1	0.78			0.67		1

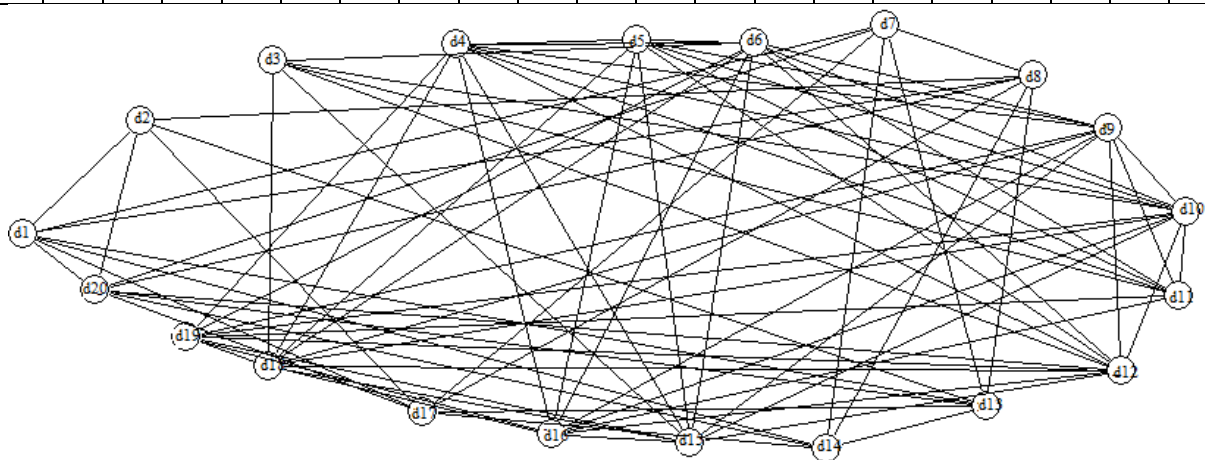


Figure 3: diagrammatic representation of table 8

Step 3,4,5,6,7,8:

**Cluster 1**

Consider the decision maker  $d_{20}$  with the similarity value is given in table 9

Table 9: Neutrosophic and score functional value of  $d_9$

Considered decision maker	Path moving to decision maker	Similarity value	Neutrosophic number	Score function value by (1)
$d_{20}$	$d_1$	0.78	(0.546,0.156,0.078)	0.4095

	$d_2$	0.78	(0.546,0.156,0.078)	0.4095
	$d_7$	0.89	(0.623,0.178,0.089)	0.46725
	$d_8$	0.56	(0.392,0.112,0.056)	0.294
	$d_{13}$	1	(0.7,0.2,0.1)	0.525
	$d_{14}$	0.78	(0.546,0.156,0.078)	0.4095
	$d_{17}$	0.67	(0.469,0.134,0.067)	0.35175

Since the maximum value is 0.525, the path is moving from  $d_{20}$  to  $d_{13}$

Consider the decision maker  $d_{13}$  with the similarity value is given in table 10

Table 10: Neutrosophic and score functional value of  $d_{13}$

Considered decision maker	Path moving to decision maker	Similarity value	Neutrosophic number	Score function value by (1)	$d_\delta$ by (4) is maximum of
$d_{13}$	$d_1$	0.56	(0.392,0.112,0.056)	0.294	0.819
	$d_2$	1	(0.7,0.2,0.1)	0.525	1.05
	$d_7$	0.78	(0.546,0.156,0.078)	0.4095	0.9345
	$d_8$	0.56	(0.392,0.112,0.056)	0.294	0.819

The maximum value is 1.05. hence the next path leads to  $d_2$ .

Hence the path is moving from  $d_{20}$  to  $d_{13}$  to  $d_2$

Consider the decision maker  $d_2$  with the similarity value is given in table 11

Table 11: Neutrosophic and score functional value of  $d_2$

Considered decision maker	Path moving to decision maker	Similarity value	Neutrosophic number	Score function value by (1)	$d_\delta$ by (4) is maximum of
$d_2$	$d_1$	0.56	(0.392,0.112,0.056)	0.294	1.344

The maximum value is 1.344. hence the next path leads to  $d_1$ .

Hence the path is moving from  $d_{20}$  to  $d_{13}$  to  $d_2$  to  $d_1$ .

Hence the cluster 1 groups  $\{d_{20}, d_{13}, d_2, d_1\}$ .

If we continue the same procedure for remaining decision makers, we get the solution for remaining clusters.

Step 9: Hence the problem concludes with the solution given in table 12 and diagrammatic representation is given in fig 4.

Table 12: solution for example 2

Cluster 1	$\{d_{20}, d_{13}, d_2, d_1\}$
Cluster 2	$\{d_{19}, d_{12}, d_6, d_3\}$
Cluster 3	$\{d_{18}, d_{15}, d_{10}, d_5, d_4\}$
Cluster 4	$\{d_{17}, d_{14}\}$
Cluster 5	$\{d_{16}, d_{11}, d_9\}$
Cluster 6	$\{d_8, d_7\}$

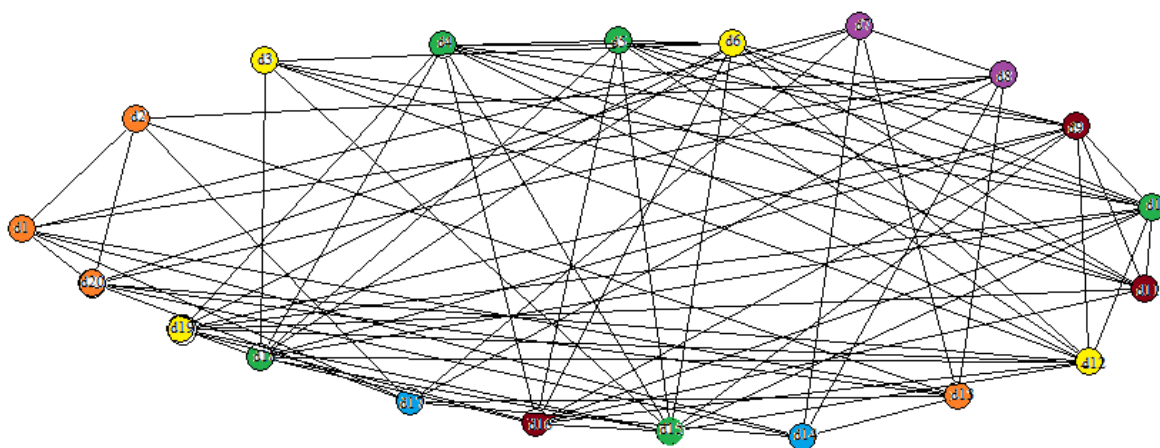


Figure 4: Diagrammatic representation of solution for example 2

## Conclusion

Since large-scale decision-makers are involved in the decision-making process, we developed a dynamic programming Neutrosophic-based clustering model in this study to handle them. From a wide perspective, the recommended clustering model—which is a significant advancement over the preceding clustering models—might cluster the decision-makers. It was discovered that the weight vector of the clusters was crucial for the decision-making process, so a novel weight determination method that includes the network centre and the group size was created. By considering cluster size and, the new weight determination model can produce more rational weights. significance into consideration. Additionally, the Neutrosophic were ranked instantly using a centroid-based ranking technique, which can prevent information loss and generate more reliable and accurate findings. As decision-making processes become more complex in the actual world, it is projected that this method will be widely used. The effectiveness of the suggested method was demonstrated in two illustrative tests, and extensive comparisons with other methods were also made to demonstrate the system's superiority. The suggested approach can be expanded in the future to include fuzzy extensions sets

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