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# Application of K-means Clustering Technique for Analysis of Students Academic Performance in School Education

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**Abstract:** *The critical issue to the academic community is the ability to monitor the progress of students' academic performance. The status or image of any institute is specifically depends upon of students' academic performance. This paper introduced a model to categorized students of a class on the basis of their academic performances. A data of 40 student of class 11th is used for the experimental work. K-means clustering algorithm is used for analyzing students' result data. The results are obtains in the form of clusters through which institution can easily trace their average or week students and make some more efforts on their studies.*

**Keywords:** *Academic performance, Average percentage, Overall performance score, K-means, Cluster.*

## I. INTRODUCTION

Average percentage making of all subjects is commonly used as an indicator of academic performance of any student in many schools. Many universities set a minimum average percentage that should be maintained in order to take admission in the degree program. In many universities, the minimum percentage set for the student is 60 to take admission for any graduate program. Average percentage of 70 and above is considered an indicator of excellent academic performance. Therefore, Average percentage still remains the most common factor used by the academic planners to evaluate progression in an academic environment [1]. Many factors could act as barriers to students attaining and maintaining a high, Average percentage that reflects their overall academic performance during their tenure in academic institutions. These factors could be targeted by the faculty members in developing strategies to improve student learning and improve their academic performance by way of monitoring the progression of their performance. Therefore, performance evaluation is one of the bases to monitor the progression of student academic performance in school. Base on this critical issue, grouping of students into different categories according to their subject making or average percentage has become a complicated task.

Some of data mining techniques, such as clustering algorithm, the discovery of key characteristics from the students' performance is possible and possibly use those characteristics for future prediction. Some promising results from applying k-means clustering algorithm have been achieved. There have been some promising results from applying k-means clustering algorithm with the Euclidean distance measure, where the distance is computed by finding the square of the distance between each scores, summing the squares and finding the square root of the sum [2].

Cluster analysis could be divided into hierarchical clustering and non-hierarchical clustering techniques. Examples of hierarchical techniques are single linkage, complete linkage, average linkage, median, and Ward. Non-hierarchical techniques include k-means, adaptive k-means and fuzzy clustering. To determine which algorithm is good is a function of the type of data available and the particular purpose of analysis. In more objective way, the stability of clusters can be investigated in simulation studies [3]. The problem of selecting the "best" algorithm/parameter setting is a difficult one. A good clustering algorithm ideally should produce groups with distinct non-overlapping boundaries, although a perfect separation can't typically be achieved in practice. Figure of merit measures (indices) such as the silhouette width or the homogeneity index [4] can be used to evaluate the quality of separation obtained using a Clustering algorithm. The concept of stability of a clustering algorithm was considered in [5]. The idea behind this validation approach is that an algorithm should be rewarded for consistency. In this paper, we implemented traditional K-means clustering algorithm [6] was chosen to be used in the analysis of the students' scores.

In this paper we use k-means clustering algorithm as a simple and efficient tool for the analysis of students' academic performance is school. The aim of this research is to produce a model for academic institutions to find their best, average and poor student with respect to their academic performances. This system helps in decision making for academic institutions to make some or more efforts towards students learning.

## II. METHODOLOGY

K-Means clustering algorithm developed [7] three decades ago is one of the K-means clustering aims to partition  $n$  observations into  $k$  clusters in which each observation belongs to the cluster with the most popular clustering algorithm used in variety of domains. A prior knowledge of number of clusters are must for K-means clustering algorithm. K-Means is defined over continuous data[8]. K-Means algorithm calculates its centers iteratively[9]. Let  $D = \{d_i | 1, \dots, n\}$  be a data set having  $K$ -clusters,  $C = \{c_k | 1, \dots, k\}$  be a set of  $K$  centers and  $S_k = \{d_i | d_i \text{ is the o of cluster } k\}$  is member of cluster  $k$  be the set of samples that belong to the  $k^{\text{th}}$  cluster. K-Means algorithm minimizes the following function which is defined as a cost function

$$\text{Cost} = \sum_{i=1}^n \text{dist}(d_i, c_k) \quad (1)$$

where  $\text{dist}(d_i, c_k)$  measures the Euclidean distance between a pattern  $d_i$  and its cluster center  $c_k$ .

nearest mean. K-means starts with a single cluster with its center as the mean of the data. This cluster is split to two and the means of the new clusters are trained iteratively. These clusters again split and the process continues until the specified number of clusters is obtained. If the specified number of clusters is not a power of two, then the nearest power of two above the number specified is chosen. Then the least important clusters are removed and the remaining clusters are again iteratively trained to get the final clusters [10].

Generalised Pseudocode of Traditional k-means is as follows [11][12]:

- 1) Step 1: Accept the number of clusters to group data into and the dataset to cluster as input values
- 2) Step 2: Initialize the first  $K$  clusters
  - a) Take first  $k$  instances or
  - b) Take Random sampling of  $k$  elements
- 3) Step 3: Calculate the arithmetic means of each cluster formed in the dataset.
- 4) Step 4: K-means assigns each record in the dataset to only one of the initial clusters
  - a) Each record is assigned to the nearest cluster using a measure of distance (e.g Euclidean distance).
- 5) Step 5: K-means re-assigns each record in the dataset to the most similar cluster and re-calculates the arithmetic mean of all the clusters in the dataset.

In this research we use the application of K-means technique to find cluster and then calculate average percentage of each cluster for the analysis of overall performance of students in a cluster. The performance index which is presented in table I is used to predict the performance students in cluster. Better results are obtained using changes in no of clusters in  $k$  means method.

TABLE I  
PERFORMANCE INDEX

| Average Percentage range | Performance of Student |
|--------------------------|------------------------|
| Above 70                 | Excellent              |
| 60-70                    | Very Good              |
| 50-60                    | Good                   |
| 40-50                    | Fair                   |
| Bellow 40                | Poor                   |

## III. EXPERIMENTAL WORK AND RESULT ANALYSIS

For analysing the academic performance of students, data of 40 student of class 11th (Maths stream) of Vedic Convent School, Bilaspur (CG), is taken presented in Table II. The data contain 7 attributes where first column have name of all 40 student, next five columns contain makes of students in Maths, Physics, Chemistry, Hindi and English subject respectively and the last 7th column contain the total average percentage of each student. K-means clustering technique is then implemented in this data for the grouping of different student with respect to their performances in last academic year.

TABLE III  
11<sup>TH</sup> CLASS DATA OF 40 STUDENT

| S. No | Student Name | Marks Obtained |         |           |       |         | Average Percentage |
|-------|--------------|----------------|---------|-----------|-------|---------|--------------------|
|       |              | Maths          | Physics | Chemistry | Hindi | English |                    |
| 1     | S1           | 44             | 56      | 52        | 58    | 60      | 54                 |
| 2     | S2           | 80             | 88      | 90        | 84    | 92      | 86.8               |
| 3     | S3           | 60             | 79      | 79        | 79    | 77      | 74.8               |
| 4     | S4           | 59             | 66      | 63        | 68    | 64      | 64                 |
| 5     | S5           | 40             | 50      | 49        | 50    | 51      | 48                 |
| 6     | S6           | 70             | 69      | 66        | 63    | 60      | 65.6               |
| 7     | S7           | 69             | 77      | 73        | 77    | 78      | 74.8               |
| 8     | S8           | 62             | 77      | 80        | 79    | 77      | 75                 |
| 9     | S9           | 64             | 77      | 78        | 80    | 76      | 75                 |
| 10    | S10          | 86             | 88      | 84        | 83    | 80      | 84.2               |
| 11    | S11          | 50             | 57      | 60        | 64    | 63      | 58.8               |
| 12    | S12          | 60             | 75      | 70        | 77    | 88      | 74                 |
| 13    | S13          | 60             | 74      | 69        | 70    | 75      | 69.6               |
| 14    | S14          | 57             | 69      | 67        | 69    | 70      | 66.4               |
| 15    | S15          | 69             | 77      | 77        | 76    | 79      | 75.6               |
| 16    | S16          | 57             | 59      | 60        | 65    | 67      | 61.6               |
| 17    | S17          | 61             | 76      | 78        | 80    | 76      | 74.2               |
| 18    | S18          | 51             | 59      | 54        | 64    | 66      | 58.8               |
| 19    | S19          | 65             | 66      | 69        | 70    | 72      | 68.4               |
| 20    | S20          | 67             | 70      | 61        | 64    | 65      | 65.4               |
| 21    | S21          | 54             | 60      | 62        | 66    | 69      | 62.2               |
| 22    | S22          | 57             | 70      | 72        | 74    | 68      | 68.2               |
| 23    | S23          | 55             | 64      | 68        | 69    | 68      | 64.8               |
| 24    | S24          | 69             | 89      | 88        | 84    | 76      | 81.2               |
| 25    | S25          | 54             | 67      | 68        | 70    | 65      | 64.8               |
| 26    | S26          | 65             | 77      | 79        | 83    | 86      | 78                 |
| 27    | S27          | 71             | 89      | 87        | 91    | 86      | 84.8               |
| 28    | S28          | 60             | 71      | 77        | 74    | 79      | 72.2               |
| 29    | S29          | 40             | 67      | 61        | 54    | 55      | 55.4               |
| 30    | S30          | 53             | 60      | 65        | 61    | 62      | 60.2               |
| 31    | S31          | 47             | 56      | 57        | 60    | 67      | 57.4               |
| 32    | S32          | 61             | 67      | 69        | 75    | 74      | 69.2               |
| 33    | S33          | 62             | 69      | 71        | 75    | 67      | 68.8               |
| 34    | S34          | 78             | 79      | 84        | 82    | 88      | 82.2               |
| 35    | S35          | 45             | 56      | 59        | 61    | 56      | 55.4               |
| 36    | S36          | 60             | 76      | 78        | 75    | 77      | 73.2               |
| 37    | S37          | 40             | 66      | 56        | 58    | 61      | 56.2               |
| 38    | S38          | 61             | 66      | 69        | 70    | 70      | 67.2               |
| 39    | S39          | 44             | 40      | 60        | 56    | 56      | 51.2               |
| 40    | S40          | 70             | 68      | 71        | 78    | 77      | 72.8               |

The experimental work is done with two cases on the basis of no of cluster decided initially.

Case 1: In this case the no of cluster is set as 3 initially. The no of cluster with silhouette value after implementation of K means is shown in figure 1. Table III presents the cluster size and overall performance of students in any cluster. The size of cluster 1 ,2 and 3

is 16, 14 and 10 respectively. The overall performance of cluster 1, cluster 2 and cluster 3 is 77.42, 66.15 and 55.54. The graphs are generated in figures 3, where the overall performance is plotted against the cluster size. Through the results it is clearly seen that 16 out of 40 students are performing “Excellent” in their studies. As same 14 out of 40 students are “Very Good” in their studies. But 10 out of 40 students are just “Fair”. But this cluster configuration is not appropriate because most objects have low value.

TABLE IIIII  
OBTAINED RESULTS FOR CLUSTERS

| Cluster# | Cluster size | Overall Performance |
|----------|--------------|---------------------|
| 1        | 16           | 77.42               |
| 2        | 14           | 66.15               |
| 3        | 10           | 55.54               |

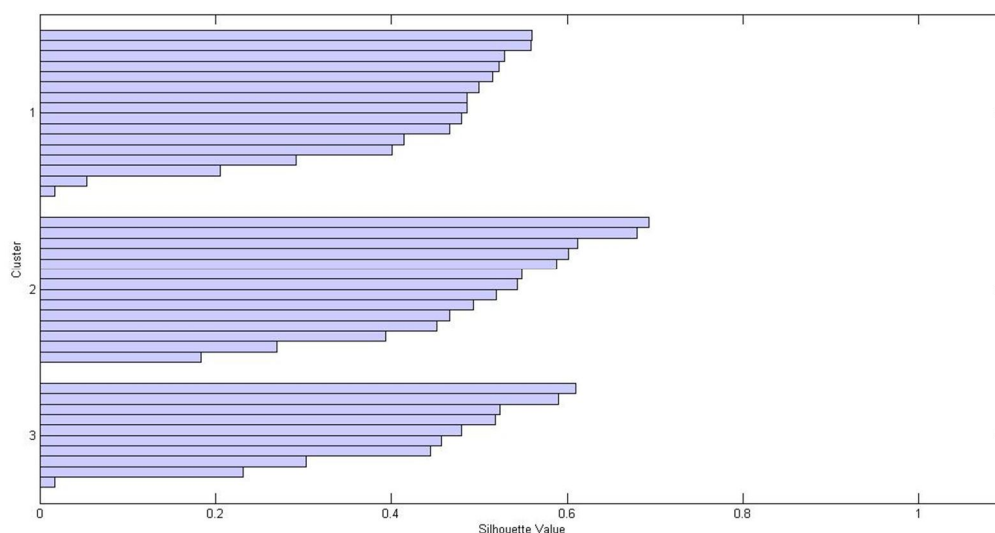


Fig. 1 Different silhouette value for each object of cluster

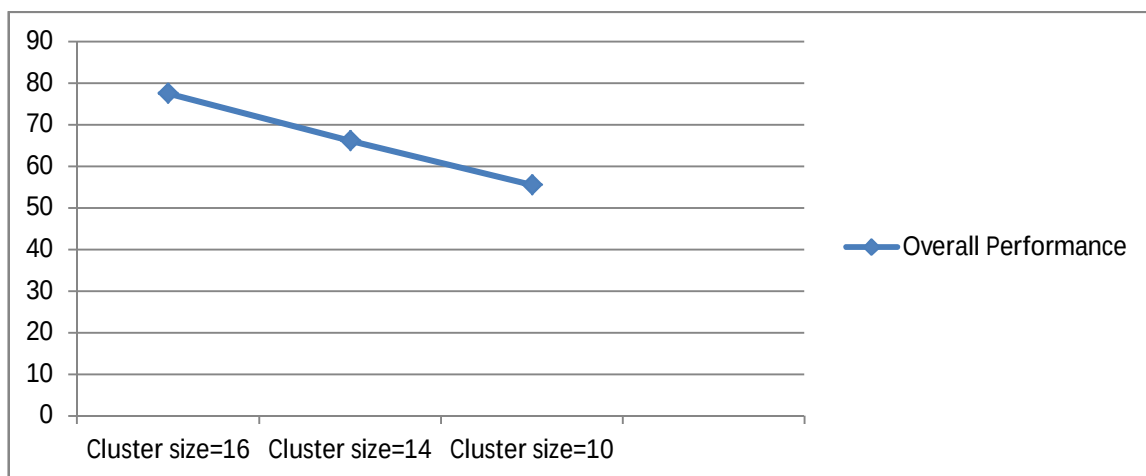


Fig. 2 Overall performance of each cluster

Case 2: In this case the no of cluster is set as 4 initially. The no of cluster with silhouette value after implementation of K means is shown in figure 3. Table IV presents the cluster size and overall performance of students in any cluster. The size of cluster 1, 2, 3 and 4 is 10, 5, 14 and 11 respectively. The overall performance of cluster 1, cluster 2, cluster 3 and cluster 4 are 55.54, 83.84, 66.15 and 74.5. This cluster configuration is highly appropriate because maximum objects have higher value with comparison to what we get in case 1. The graphs are generated in figures 4, where the overall performance is plotted against the cluster size. Through the



results it is clearly seen that 16 out of 40 students are performing “Excellent” in their studies. As same 14 out of 40 students are “Very Good” in their studies. But 10 out of 40 students are just “Fair”.

TABLE IV  
OBTAINED RESULTS FOR CLUSTERS

| Cluster# | Cluster size | Overall Performance |
|----------|--------------|---------------------|
| 1        | 10           | 55.54               |
| 2        | 5            | 83.84               |
| 3        | 14           | 66.15               |
| 4        | 11           | 74.5                |

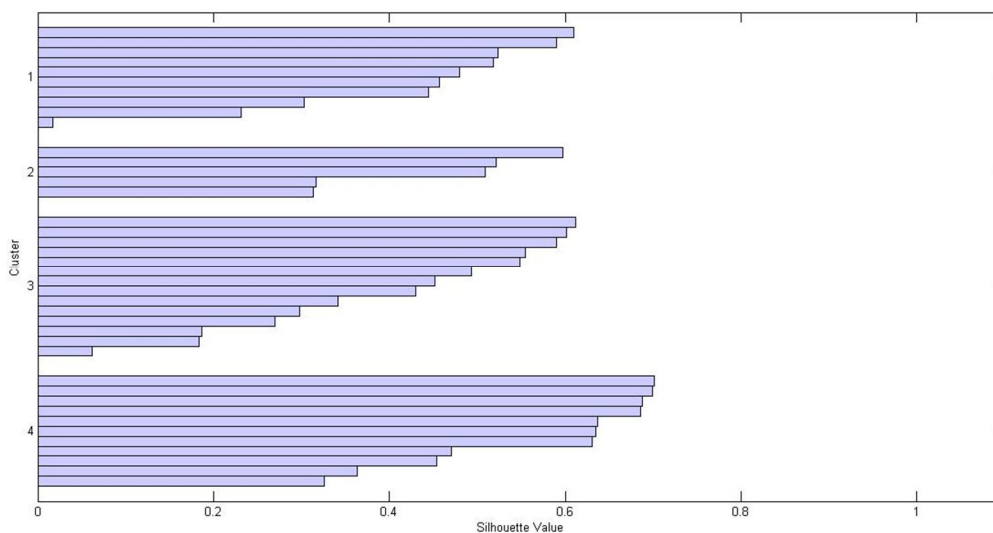


Fig. 3 Different silhouette value for each object of cluster

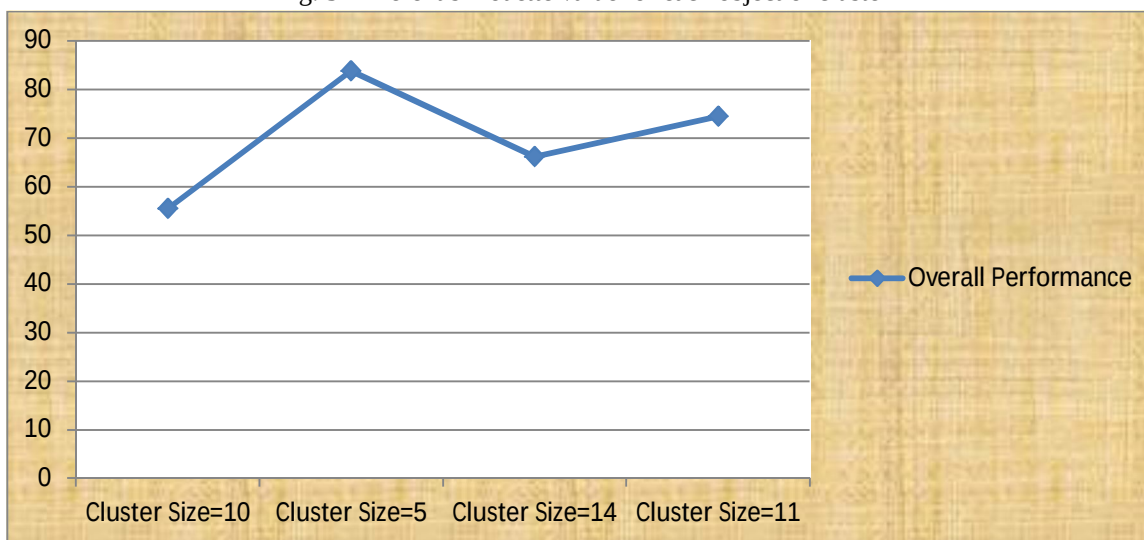


Fig. 4 Overall performance of each cluster

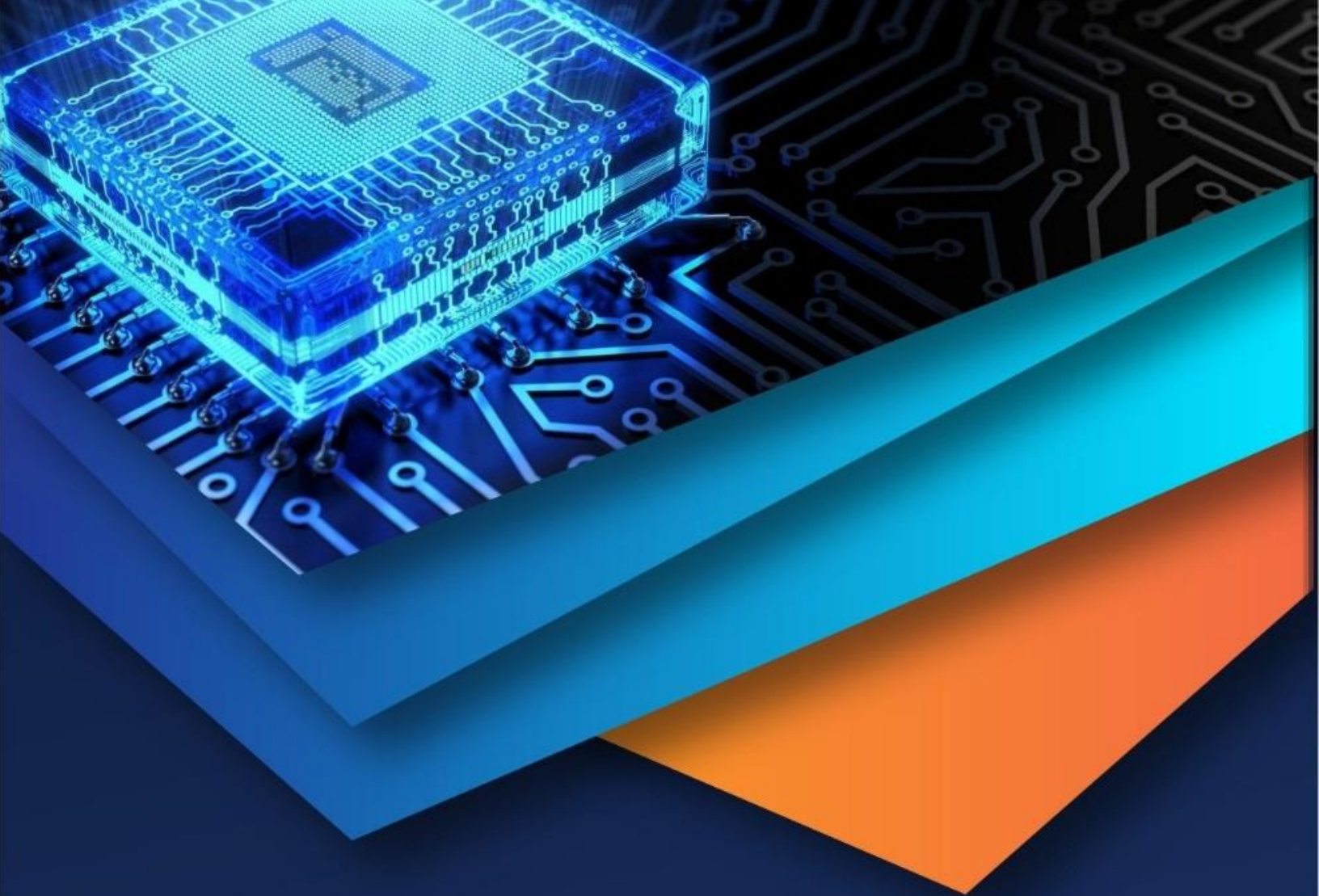
It is clearly seen from Table III that cluster 4 have 11 objects or student whose performances are “Fair” that indicates that those students are not up to the mark as other students. So through this result school can traces those students and can do some extra work in field of student learning. This effort may be in the type of extra classes by subject expert teachers or using so effective turning techniques.

#### IV.DISCUSSION AND CONCLUSION

Analysis of student's academic performance is the major issue for the status and reputation of any academic institution. This analysis helps in decision making for academic institutions to improve their education techniques. A common approach is to use average percentage to evaluate the academic performance of any student. In this research we introduced a model to evaluate students' academic performance. We use the application of K- means that creating no of clusters and the find the performance of students of every cluster using performance index. A cluster with 'Fair' or 'Poor' performances indicates the academic institution to make effective changes to their education techniques for those students. In this research K-means method is applied with initialization of no of cluster is 3 and 4. As results shown above the process is appropriate when cluster is 4. The performance of cluster 1 is 'Fair' that indicates that institution should take care of those students.

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