The USA Higher Education Institutes Segmentation Using **Clustering Technique**

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Abstract: The huge quantities of data and information about universities, students, faculty members, personnel, material resources, etc., can in most cases contains valuable information and patterns for higher education. American higher education institutions are usually classified according to indicators such cost including tuition fees, living and educational expenses. Also, the percentage of doctoral students and graduates of each faculty, reflecting the scientific level of the faculty of each university and the availability of facilities such as laboratories, libraries and research centers, can provide significant insights and patterns. In the present research, cluster analysis and K-Means techniques are used to categorize and analyze data related to te mentioned indicators.

Keywords: Data mining, Clustering, K-means, Patter Recognition

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Introduction I.

As one of the most dynamic educational systems, the higher education system is responsible for the most important roles, and because of dealing with important data about students, professors, personnel and material resources, etc. is one of the areas where data mining is most used. The present study aims to achieve useful results using the data from US higher education institutions and data mining techniques.

II. **Literature Review**

Data mining or discovery in the databases refers to non-obvious extraction of potentially useful information from data that has previously been unknown and Data Mining is the process of extracting hidden knowledge from large volumes of raw data. [1, 2]. Pandey et al. proposed the performance of clustering algorithm using heart disease dataset. They evaluated the performance and prediction accuracy of some clustering algorithms. The performance of clusters will be calculated using the mode of classes to clusters evaluation. Finally they proposed Make Density Based Cluster with the prediction accuracy of 85.8086%, as the most versatile algorithm for heart disease diagnosis [3]. Finding such patterns and knowledge in data can be useful in making future decisions such as eco-efficiency [4] and other computational method such as Data Envelopment Analysis (DEA) [5], simulation model [6] and Fuzzy expert system [7]. The stages of the data mining process can be observed in Figure 1.

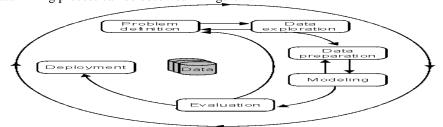


Figure 1 - Data mining process

III. Clustering

Clustering refers to grouping the similar data in a data mass. The basic issue of clustering is the distribution of data to k different groups such that the data of each group being similar and the data of different groups are dissimilar. A good clustering method generates high quality clusters based on the following two criteria:

- high similarity of the internal points of each cluster
- low similarity of the points of different clusters
 - K-means method is one of the most important methods of clustering algorithm

IV. Data Set

The dataset used in this study has been extracted from the www.ics.uci.edu website which are related to higher education institutions located in the United States until the end of 2007. Table 1 represents the fields of this database.

| Field name | Field type | Description | | |
|-----------------------------|-----------------------|--|--|--|
| University name | Nominal | University name along with state name | | |
| State code | Numerical | Each state has a unique code | | |
| Public / Private university | Numerical (1/2) | Public / private university | | |
| admission application | Numerical | Number of admissions applications at the university | | |
| Accepted application | Numerical | Number of applications accepted at university | | |
| New entrance | Numerical | Number of new students | | |
| Elite entrance | Numerical | Number of elite entrance students | | |
| Graduates | Numerical | Number of students on graduation | | |
| Tuition | US currency (dollars) | The tuition fee of each university | | |
| Number of rooms | Numerical | Number of rooms per university | | |
| Tuition | US currency (dollars) | Expensive study fees like dormitory and | | |
| Living expenses | US currency (dollars) | personal expenses | | |
| PhD students | Numerical (percent) | Number of PhD students | | |
| College students | Numerical (percent) | Student rate of each faculty | | |
| Graduates percent | Numerical (percent) | The graduation rate of each university | | |

| Table | 1 | – the | database | fields |
|-------|---|-------|----------|--------|
| Lanc | 1 | -uc | uatavase | ncius |

V. Preprocessing and Data Preparation

The data preparation, is the most important and time consuming stage in the data mining project. As the data are the input into this project, the more accurate the inputs, the more accurate the output. There are several preprocessing methods introduced in the literature. In [8] several data mining clustering algorithms were evaluated to find the most accurate one in heart disease prophecy and a particular preprocessing filtering method are applied to find the superior algorithms. In [9], normalization technique is used as an effective preprocessing method before training a data-mining model. This technique is very efficient especially for a large dataset [10]. Another strong preprocessing method is called principle components analysis (PCA) which reduces the dimension of the dataset. In [11], PCA is used for cell recognition.

The data was coded in order to prepare data, so that they can be easily used in the Clementine12 software. Considering the numerous fields in this database, we selected some of its fields to be used in the data mining process. Table 2 shows the fields used in data mining and how they are encoded.

| Table 2 - The fields used in data mining and coding | | | | |
|---|-----------------------|----------------------------|--------------|--|
| Field name (attribute) | Туре | Coding | Feature Type | |
| | Less than \$ 5,000 | Less than \$ 5000 = 1 | | |
| | 10,000 \$ - 5000 \$ | 10,000 \$ - 5000 \$=2 | | |
| Tuition | 15,000 \$ - 10,000 \$ | 15000 \$ - 1000 \$=3 | Numerical | |
| | 20,000 \$ - 15,000 \$ | 20,000 \$ - 1500 \$=4 | | |
| | More than \$ 20,000 | More than \$ 20,000 = \$ 5 | | |
| | Less than \$ 1,000 | Less than \$ 1000 = 1 | | |
| Education secondary costs | 1,500 \$ - 1,000 \$ | \$ 1500 - \$ 1000=2 | Numerical | |
| Education secondary costs | 2000 \$ - 1500 \$ | \$ 2000 - \$ 1500=3 | | |
| | More than \$ 2,000 | More than 2000 \$ = 4 | | |
| | Less than \$ 100 | Less than \$ 100 = 1 | | |
| Living expenses | 2000 \$ - 100 \$ | \$ 2000 - 100 \$ | Numerical | |
| Living expenses | 2000 \$ - 1500 \$ | \$ 2000 - \$ 1500=3 | Numericai | |
| | More than \$ 2,000 | More than 2000 \$ = 4 | | |
| | Less than 25% | Less than 25% = 1 | | |
| Number of doctoral | %50 - %25 | %50 - % 2 5=2 | NT ' 1 | |
| students | %75 - %50 | %75 - %50 =3 | Numerical | |
| | %100 - %75 | %100 - %75=4 | | |
| Graduates percent | Less than 25% | Less than 25% = 1 | | |
| | %50 - %25 | %50 - %25 %50 - %25=2 | | |
| | %75 - %50 | %75 - %50 =3 | Numerical | |
| | %100 - %75 | %100 - %75=4 | | |

Table 2 - The fields used in data mining and coding

VI. Problem Statement

Cluster analysis refers to a group of multivariate techniques whose primary purpose is to cluster objects (respondents, products, and other entities) so that every object of the group is very similar to the other members of the cluster (based on predetermined criteria).

The observations in the final clusters, should have a high homogeneity (within clusters) and external heterogeneity (between clusters). This analysis is also helpful when the researcher intends to test hypotheses about the nature of data or to develop a new hypothesis in this regard. The number of clusters in this analysis can be predefined or determined based on the maximum variance within clusters (or between the clusters).

The variable concept is a fundamental issue in this technique, however it has completely different methods from the other techniques. The clustering variable is a set of variables that show the features used to compare objects in cluster analysis. There are always three main objectives of "describing empirical groups", "data simplification" and "determining relationships" in this technique along with the inclusion of two assumptions on the Representativeness and Multicollinearity between the independent variables in question.

The K-means clustering technique has been used in this study to classify US higher education institutions.

As can be seen from Table 2, the fields, used in the data mining process, were introduced and after applying the K-means technique, five clusters were obtained, which are shown in Table 3 of the features of each cluster.

| Table 5. Features of each cluster | | | | |
|-----------------------------------|-------------------|------------------|--|--|
| Cluster | Number of records | Percent of total | | |
| First cluster | 259 | 20% | | |
| Second cluster | 139 | 11% | | |
| Third cluster | 429 | 33% | | |
| Fourth cluster | 195 | 14% | | |
| Fifth cluster | 280 | 22% | | |

Table 3. Features of each cluster

Considering the clustering mentioned in Table 3, the majority of universities are in the third cluster. Table 4 provides the information on the third cluster with the highest frequency

Table 4. The third cluster features

| Cluster | Tuition | Education secondary costs | Living expenses | The number of doctoral students | Graduation Rate | Percent of total |
|---------------|----------------------------|---------------------------|-----------------------|---------------------------------------|-----------------|------------------|
| Third cluster | \$ 10,000 to \$ 150,000 | \$ 1500 to \$ 2000 | \$ 2000 to \$ 4000 | 50% to 75% | 50% to 75% | 32% |

Accordingly, most of the US higher education institutions have the features noted in Table 4.

Now, the indicators are individually examined within other clusters which results are presented in Table 5 (tuition, education secondary costs, living expenses, doctoral and graduate students' rates).

| Table 5 – Individual frequency of each indicator | | | | | |
|--|-------------------------|------------------------------|------------------|---------------------------------|-----------------|
| Index | Tuition | Education secondary costs | Living expenses | The number of doctoral students | Graduation Rate |
| Cluster | Second cluster | Fourth cluster | First cluster | Second cluster | Second cluster |
| The range with the most frequency | \$ 5000 to \$ 10,000 | More than \$ 2,000 | Less than \$ 100 | 25% - 50% | 25% - 50% |

 Table 5 – individual frequency of each indicator

VII. Conclusion

The abundance of higher education institutions in the United States and their extraordinary spread throughout the country has led to a variety of unique indicators such as living expenses, training, university tuition, the number of doctoral students and graduation rates. The present study was conducted with the purpose of segmentation of these institutions using cluster analysis and K-means technique, which resulted in five clusters so that highest frequency was belonged to the third cluster. Also, the indices of tuition, the number of doctoral students and the graduation rate in the second cluster have the most frequencies in the individual review of each field; in terms of education secondary expenses, the fourth cluster has the most frequency, and finally, the living expenses index has the most frequency in the first cluster.

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