



Analyzing the Food Quality Using K-Means Clustering

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Abstract: Clustering plays an important role in a life problem, since we live in a world full of data where we encounter a large amount of information. One of the vital means in dealing with these data is to classify or group them into a set of categories or clusters. In this paper we apply the k-Means clustering algorithm in analyzing the food quality.

Keywords: Clustering, Fuzzy clustering, k-Means clustering.

I. INTRODUCTION

Good nutrition is an important part of leading a healthy lifestyle. Healthy food also increases a person's chances of living longer. It improves one's mood and enhances his mental status. When a person's body is in stress, protein is often broken down into amino acids which aid the body to deal with stress. Protein-rich foods can help to enhance the protein level in the body hence it ensures that one is in good moods. Fuzzy clustering plays an important role in various streams and even in our day to day life like data mining, text mining, grouping the products, surveys, graph theory and etc.

K-Means clustering is used to analyze the food quality. The term "k-means" was first used by James MacQueen in 1967. Cluster analysis was originated in anthropology by Driver and Kroeber in 1932 and introduced to psychology by Zubin in 1938 and [Robert Tryon](#) in 1939. This paper is organized as follows: section II contains the concept of clustering, fuzzy clustering and k-Means clustering, section III deals with application of k-Means clustering, section IV has conclusion.

II. The Concept Of Clustering, Fuzzy Clustering And K-Means Clustering

A. Clustering

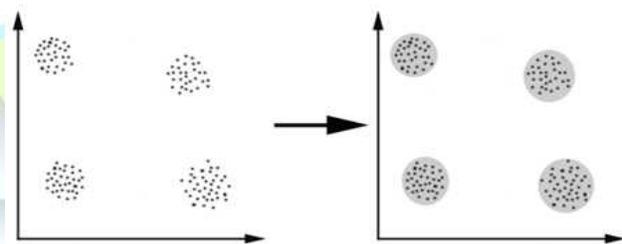
Clustering is the classification of objects into various groups, or more precisely, the partitioning of data set into clusters, so that in each cluster shares some common features, according to some defined distance measure. Clustering finds application in many fields. It is used to discover relevance knowledge in data.

B. Fuzzy Clustering

Fuzzy clustering is a form of partitioning in which every data point can belong to more than one cluster.

It is a classification of objects into various groups or more precisely, the partitioning of data sets into clusters, so that the data in each cluster share some common features, often proximity according to some defined measures.

A cluster is usually represented as grouping of similar data points around a center known as *centroid* or it may be defined as prototype data instance nearest to the centroid. A cluster can be represented with or without a well-defined boundary such that those clusters with well-defined boundaries are called *crisp clusters* whereas those without a well-defined boundary are called *fuzzy clusters*.



Example of clustering in diagrammatical representation

C. k-Means Clustering

k-Means is one of the simple algorithm that solve the well-known clustering problem. Its procedure follows a very simple and easy way to classify a given data set through a certain number of clusters (assume k clusters) fixed apriori.



III. Application of K-Means Clustering

A. Food products grouping

Let us apply the k-Means clustering algorithm in analyzing the food quality using k-Means clustering and obtain four clusters. Here I have chosen seven food products which are not same but with some similarities to each other in the protein and fat content [15].

S.No	Nuts & Seeds	Protein content, P	Fat content, F
1	Almonds	13	38
2	Sesame seeds, Dry	9	24
3	Coconut, Shredded, Sweetened	1	20
4	Brazil nuts	10	47
5	Cashews, Unsalted	12	32
6	Pecans, Raw, Halves	5	35
7	Sunflower seeds	11	26

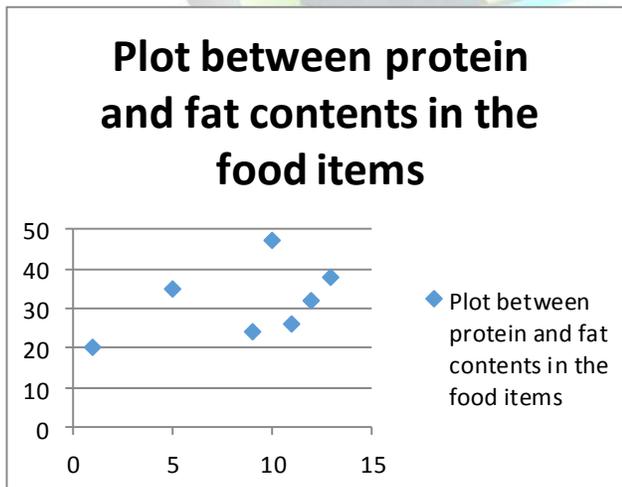


Table 1

Let us plot these points so that we can have better understanding of the problem. Also, we can select the three points which are farthest apart.

We observe that point 1 is close to point 5. So, both can be taken as one cluster. The resulting cluster is called C15 cluster. The value of P for C15 centroid is $(13 + 12)/2 = 12.5$ and F for C15 centroid is $(38 + 32)/2 = 35$.

Upon closer observation, the point 2 can be merged with the C7 cluster. The resulting cluster is called C27 cluster.

The values of P for C27 centroid is $(9 + 11)/2 = 10$ and F for C27 centroid is $(24 + 26)/2 = 25$.

The point 3 is close to point 6. They can be merged into C36 cluster.

The values of P for C36 centroid is $(1 + 5)/2 = 3$ and F for C36 centroid is $(20 + 35)/2 = 27.5$.

The point 4 is not close to any other point. So, it is assigned to C4 with the value of P for C4 centroid as 10 and F for C4 centroid is 47.

Finally, four clusters with three centroids have been obtained.

Cluster number	Protein content, P	Fat content, F
C15	12.5	35
C27	10	25
C36	3	27.5
C4	10	47

Table 3

In the above example it was quite easy to estimate the distance between the points. In cases in which it is more difficult to estimate the distance, one has to use euclidean metric to measure the distance between two points to assign a point to a cluster.

IV. CONCLUSION

Here we have applied k-Means clustering technique in grouping the food products which are available in the market according to their similarities and dissimilarities based on protein and fat content in those food products, so that it helps us to decide which is suitable for our health and what not. And this method can be even applied on grouping the students who're using social networking sites and how the impact of these networking sites on the student's academic performances.

Conflict of interests: The author declared no conflict of interests.



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BIOGRAPHY



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