Survey on Different Grid Based Clustering Algorithms

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Abstract: The grid-clustering algorithm is the most important type in the hierarchical clustering algorithm. The grid-based clustering approach considers cells rather than data points. In grid-based clustering all the clustering operations are performed on the segmented data space, rather than the original data objects. Grid-based methods are highly popular compared to the other conventional models due to their computational efficiency but to find optimal grid size is a key feature in grid-based clustering algorithm. There exists some algorithm in that they achieve optimal grid size but in real life data can be dense or sparse.

Keywords: Clustering; grid; GRPDBSCAN; GDILC; GGCA; OPT-GRID(S).

I. INTRODUCTION

Data mining is the extraction of hidden, predictive information patterns from large databases. Data Mining is especially useful now-a-days when there is massive amount of data and identifying the useful portions of it can be a tedious job in itself. Data mining allows us to be proactive about situations rather than retrospective – this means that we can now try and predict the future trends rather than identifying them after they have already taken place.

A cluster is a subset of objects which are “similar”. Clustering is a process of partitioning a set of data (or objects) into a set of meaningful sub-classes, called clusters. It helps users to understand the natural grouping or structure in a data set. Clustering can be considered the most important unsupervised learning technique; so, as every other problem of this kind, it deals with finding a structure in a collection of unlabeled data. A good clustering method will produce high quality clusters in which (1) the intra-class (that is, intra-cluster) similarity is high. (2) The inter-class similarity is low.

There are various applications of clustering such as Economic Science (especially market research), Document classification, Cluster Weblog data to discover groups of similar access patterns, Pattern Recognition, Create thematic maps in GIS by clustering feature spaces, Image Processing, Information Retrieval, Text Mining.
The grid-clustering algorithm is the most important type in the hierarchical clustering algorithm. The grid-based clustering approach considers cells rather than data points. This is because of its nature grid-based clustering algorithms are generally more computationally efficient among all types of clustering algorithms. In fact, most of the grid-clustering algorithms achieve a time complexity of $O(n)$ where $n$ is the number of data objects. It allows all clustering operations to perform in a gridded data space.

Grid-based methods are highly popular compared to the other conventional models due to their computational efficiency. The main variation between grid-based and other clustering methods is as follows. In grid-based clustering all the clustering operations are performed on the segmented data space, rather than the original data objects [1]. Then any topological neighbor search is used to group the points of the closer grids. The grid-based clustering methods face the following challenges. First, it has to determine an appropriate size of the grid structure. If the grid size is too large, two or more clusters may be merged into single one. When the grid size is very small, a cluster may be divided into several sub-clusters. Therefore, finding the suitable size of grid is a challenging issue in grid clustering methods. The second problem is with the data of clusters with variable densities and arbitrary shapes in case of which a global density threshold cannot result the clusters with less densities. This is known as the problem of locality of cluster. The third one is the selection of merging condition to form efficient clusters. Considering these issues, various grid based algorithms have been proposed.

### II. INTRODUCTION OF VARIOUS GRID BASED ALGORITHMS

- **A new shifting grid clustering algorithm [2]**

  The algorithm is based on shifting the grid. The algorithm is a non-parametric type means which does not require users inputting parameters. It divides each dimension of the data space into certain intervals to form a grid structure in the data space, shifting of the whole grid structure is introduced to obtain a more descriptive density profile based on the concept of sliding window. As a result, we are able to improve the accuracy of the results. This algorithm is provides effective and efficient clustering because it clusters data in a way of cell rather than in points. It exhibits tremendous performance to deal with arbitrary shaped cluster. The algorithm does not always suffer from the problem of memory limitation when handling large data set. The algorithm has disadvantage is that it requires the additional computational efforts.

  - **Advantages:**
    
    1. Its computational time is independent to the number of data points.
    2. It shows good performance to deal with arbitrary shaped cluster
    3. It is a non-parametric algorithm.
    4. The algorithm does not always suffer from the problem of memory limitation when handling large data set.

- **Grid-based DBSCAN Algorithm with Referential Parameters [3]**

  The algorithm which is combination of the grid partition technique and multi-density based clustering algorithm. The key issue is to be how to deal with the data changes and how assure the validity of data class’ association rules. The algorithm solved this problem. According to the character of data mutations in dynamic data test, and the association between grid partition technique and multi-density base clustering algorithm: DBSCAN is used. The algorithm chooses the Eps and Minpts parameters automatically and differentiates between noises and discovery clusters of arbitrary shapes including data changes, so they were more objective (independent). The algorithm has improved its efficiency. Finally, the algorithm makes answer more precise and the result of this improved algorithm is robust.
Advantages:

(1) It deals with massive data changes and noise.

(2) It can find clusters of arbitrary shape and remove noise.

(3) It makes the answer more precise and it is more robust.


The idea of GDILC is that the density-isoline figure depicts the distribution of data sample very well. Here, assumes that the entire data samples are normalized. The algorithm starts from the density-isoline figure of data samples to find the densely populated regions, which meets the requirements of the clustering or which are the clusters that we hope to find. In that first, a grid-based method is working to calculate the density of each data sample and chosen a proper density threshold from density-isoline figure. Then, those clusters bounded by the chosen isoline are combined. GDILC can calculate automatically the distance threshold and the density threshold according to the size and distribution of a given data set. So it is non-supervising clustering algorithm because it require no human iteration. The advantage of these algorithm is the high speed and accuracy & mainly removing outlier and finding the clusters of various shapes.

Advantages:

(1) It has the cluster of arbitrary shapes and locate the outliers.

(2) It has linear time complexity.

A general grid-clustering approach [5]

A general grid-clustering approach (GGCA) under a common assumption about hierarchical clustering. The key features of the GGCA include: (1) It is the combination of the divisible and the agglomerative clustering algorithms into a unifying generative framework; (2) the purpose of key input parameters: an optimal grid size for the first time; and (3) the application of a two-phase merging process to combined all data objects. The GGCA is a non-parametric algorithm in which it does not require users to input parameters. With the partitioning index and the depth index, the algorithm solves two critical problems of conventional (predictable) grid-clustering algorithms: (i) grid size and (ii) merging condition.

The characteristics of GGCA are: (1) GGCA consists of a set of local high-density data objects and, therefore, the problem in the existing clustering algorithms is that a global density threshold is unable to identify all clusters in a given density-diverse dataset. Also, the GGCA does not reject those low-density clusters, while many other existing clustering approaches often may not be able to find them. (2) A good clustering method should have the ability to accept noisy data and outliers in the dataset. Using a two-phase merging process enhancements the robustness of the clustering results by the GGCA. Since the boundary data objects of each cluster are often located in the overlapped area of diverse clusters, these leads to incorrect clustering results in the existing clustering approaches. However, these boundary objects cannot be contained in any core grid, and thus do not affect the clustering results, so the risk of incorrect clustering results by a clustering algorithm decreases. GGCA gives excellent performance in dealing with not well-separated and shape-diverse clusters.

Advantages:

(1) It is parameter-free algorithm.

(2) It has the ability to bear noisy data and outlier in the dataset.
• **OPT-GRID(S)**

The grid-based clustering finds the optimal grid-size using the boundaries of the clusters. Here, In dataset D there are n data points and m dimensional space. Initially a single grid which is used to represent all the given n points. This grid are taken as the min. and max. attribute value in each dimension. Then single grid is partitioned into two equal volume grids. So, all the data points are distributed to these two grids. After each round of partitioning of grid it is necessary to check the presence of the new cluster. In the next round of partitioning, the two grids are partitioned into four equal volume grids in another chosen dimension. In this way all the grids are bisected and partitioning processes is continue until the optimal grid structure is generated. The boundary grids are used to find the optimal grid structure. Each cluster is surrounded by the boundary grids. The volume of the cluster decreases as the partitioning process continues and as the same time the number of surrounding boundary grids increases. The problem of outlier is solved with local outlier factor (LOF).

➢ **Advantages:**

1. It is non-parametric algorithm means user does not require to input the parameter.

### III. CONCLUSION

Grid based clustering is good technique to make cluster of data. Iteration by iteration grids is created and according to that clusters will create. This paper gives detail study of the various grid based algorithm like GRPDBSCAN, GDILC, GGCA, OPT-GRID(S) based on the different parameters which gives different arbitrary shapes. To develop an algorithm that can find optimal grid size in any type of dataset in dense or sparse with appropriate accuracy or maintaining the accuracy with less time is a challenging task in a grid clustering.

<table>
<thead>
<tr>
<th>Algorithm Name</th>
<th>Technique Used</th>
<th>Input parameter</th>
<th>Arbitrary shape</th>
<th>Noise</th>
</tr>
</thead>
<tbody>
<tr>
<td>Shifting grid clustering algorithm</td>
<td>Grid + Density</td>
<td>Non-parametric (parameter-free)</td>
<td>Yes</td>
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<td>Eps and Minpts</td>
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<td>Yes</td>
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<td>distance threshold RT and density threshold DT</td>
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<td>Yes</td>
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<tr>
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<td>Hierarchical Grid + Density</td>
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<td>Yes</td>
</tr>
<tr>
<td>OPT-GRID(S)</td>
<td>Grid + Density</td>
<td>Non-parametric (parameter-free)</td>
<td>No</td>
<td>Yes</td>
</tr>
</tbody>
</table>

**Table-1: Comparison Table for grid based algorithms:**

**References**