A Decision Support System for Industry based upon Multivariate Spatial Outlier Detection Techniques

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Abstract

Decision support systems are computer based programs which assists decision makers in effective and efficient decision making. Outliers are the data sets which are highly irrelevant with respect to main data sets. Spatial outliers are the spatial objects with distinct features from their surrounding neighbours. In this paper we have proposed a decision support system for strategic decision makers to establish an industry. We have shown how spatial outlier detection techniques may be used to aid decision makers. Each site is evaluated based upon multiple decision variables. Here, we use two different outlier detection algorithms to detect a site which is most unsuitable to establish an industry. We also perform an experiment on a synthetic dataset of fifteen different sites to locate ambiguous sites. We found that decision making process for site selection may be improved by using spatial outlier detection techniques.

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1. Introduction

Industrial location is an increasingly important factor facing both national and international firms. To establish any industry a manager has to analyse large amount of information regarding available list of sites. Some of the critical factors of industrial location are transportation, labour, raw material, markets, utilities, electricity, climate, tax structure, technology, government policies & etc. The sites which are not suitable are completely discarded from the decision making process and such sites are called as spatial outlier sites. Spatial outlier is an object whose non-spatial attribute value is significantly different from the values of its spatial neighbours. Spatial outlier detection algorithms may be applied to find such ambiguous sites.
2. Related work & background

An approach for discovering outliers using distance metrics was first presented by Knorr et al. They define a point to be a distance outlier if at least a user-defined fraction of the points in the data set are further away than some user-defined minimum distance from that point. In their experiments, they primarily focus on data sets containing only continuous attributes. Related to distance-based methods are methods that cluster data and find outliers as part of the process of clustering [1]. Points that do not cluster well are labelled as outliers. Recently, density-based approaches to outlier detection have been proposed. In this approach, a local outlier factor (LOF) is computed for each point. The LOF of a point is based on the ratios of the local density of the area around the point and the local densities of its neighbours. A Spatial-Temporal approach has been used in medical science to differentiate epidemic risk patterns [2]. Outlier detection approach has also been used for PCB Testing in industry [3]. An application of such techniques has discussed in [4] for outlier detection in high dimensional data. Another application was proposed by [5] for temporal outlier detection in vehicle traffic data. [6] Discuss application of spatial outlier detection for malicious user detection in a cognitive radio cooperative sensing systems. Another application of such techniques has been discussed in [8] where hierarchical clustered outlier detection methods have been used in laser scanner point clouds. Such applications may also be used in crowds. The [9] used one such approach where anomaly detection in crowded scenes is illustrated. Some improved techniques have been discovered for special kind of data. Spatial outlier detection may also be used in security based computing [10]. One such approach is cluster-based congestion outlier detection method on trajectory data [11]. Recently [12] used spatial outlier detection techniques to locate less developed regions in the state of haryana.

3. Problem formulations

Suppose we have a dataset of n sites, denoted as \( D = \{S_1, S_2, S_3, \ldots, S_n\} \). Each site has a set of m attributes, \( A = \{a_1, a_2, \ldots, a_m\} \), where \( a \in A \) represents a decision parameter of the site. Problem is to find a site \( S_a \) which is most ambiguous to establish an industry. \( S_a \), obviously is a site which does not fit into minimum criteria set by the decision maker.

![Fig. 1 An example of spatially distributed sites with non spatial features](image)

Supposing that minimum requirement for choosing a site is denoted by a consolidated numerical threshold, denoted as \( \Phi \), and then \( S_a \) be a site whose measures is below \( \Phi \). Such a site is called as spatial outlier site. In Fig. 1, a set of seven sites are shown. Sites 2, 4, 6 and 7 are normal one, while site 1, 3 and 5 are inconsistent sites where establishing an industry, is extremely difficult or risky. Problem in our case is comprised of locating sites with red dots. These are the sites where establishing an industry is very difficult or not feasible.
3.1 Proposed system

The proposed methodology is discussed step by step in Fig. 2. First step of the proposed system is to collect data regarding various prospective sites to establish an industry. After that data items are categorised into spatial and non spatial attributes. A spatial database of various sites is constructed which contains spatial attributes like longitude, latitude, area, direction and non spatial attributes like availability of electricity, skilled and unskilled manpower, tax structures, raw materials and many more. Finally spatial outlier techniques are applied on this spatial database. We have used two very popular approaches to detect spatial outliers using multiple variables [13]. Other more accurate methods may be applied to get improved results. Finally results so obtained after previous step may be used by decisions makers to decide whether a site is suitable for establishing an industry or not. Result will show a list of spatial outlier sites which will indicate degree of unsuitability of these sites for establishing an industry.

Fig. 2 Methodology for Decision support system using outlier detection

We will use following two important multivariate spatial outlier detection algorithms for our proposed system.

3.2 Mean Algorithm

1. Given the spatial data set \( X = \{x_1, x_2, \ldots, x_n\} \), predefined threshold \( \theta \), attribute function \( f \), and the number \( k \) of nearest neighbors.
2. For each fixed \( j (1 \leq j \leq q) \), standardize the attribute function \( f_j \), i.e., \( f_j(x_i) \leftarrow f_j(x_i) - \mu f_j / \sigma f_j \) for \( i = 1, 2, \ldots, n \).
3. For each spatial point \( x_i \), compute the \( k \) nearest neighbor set \( NN_k(x_i) \).
4. For each spatial point \( x_i \), compute the neighborhood function \( g \) such that \( g_j(x_i) = \text{average of the data set} \{ f_j(x) : x \in NN_k(x_i) \} \), and the comparison function \( h(x_i) = f(x_i) - g(x_i) \).
5. Compute \( d^2(x_i) = (h(x_i) - \mu s)^T \Sigma^{-1} s (h(x_i) - \mu s) \). If \( d^2(x_i) > \theta \), \( x_i \) is a spatial outlier w.r.t. \( A \).

3.3 Median Algorithm

1. Given the spatial data set \( X = \{x_1, x_2, \ldots, x_n\} \), predefined threshold \( \theta \), attribute function \( f \), and the number of nearest neighbors.
2. For each fixed \( j (1 \leq j \leq q) \), standardize the attribute function \( f_j \), i.e., \( f_j(x_i) \leftarrow f_j(x_i) - \mu f_j / \sigma f_j \) for \( i = 1, \ldots, n \).
3. For each spatial point \( x_i \), compute the \( k \)-nearest neighbor set \( NN_k(x_i) \) based on its spatial location.
4. For each spatial point \( x_i \), compute the neighborhood function \( g \) such that \( g_j(x_i) = \text{median of the data set} \{ f_j(x) : x \in NN_k(x_i) \} \), and the comparison function \( h(x_i) = f(x_i) - g(x_i) \).
5. Compute $d^2(x_i) = (h(x_i) - \mu)^T \Sigma^{-1} (h(x_i) - \mu)$. If $d^2(x_i) > \theta$, $x_i$ is a spatial outlier w.r.t. $A$.

4. Experimental results

Suppose we take a simple example to establish an Information Technology industry. Common and critical factors which may affect the successful operation of an IT industry are; availability of skilled labour, communication, transportation, access to latest technologies, availability of projects, electricity, medical facilities and basic needs or amenities.

Consider a sample dataset of fifteen different sites located at different regions. Fig. 3(a) shows the outliers in red color using Mean algorithm. Here, points E, K, L denote the outliers. Fig. 3(b), shows the outliers in red color using Median algorithm. Here E, K denote the spatial outliers. These are the sites where it is risky to establish an IT industry. Following screenshots in Fig. 4 shows the results so obtained on a spatial dataset using two different algorithms. Here we have taken synthetic spatial dataset with three distinct attributes.

![Fig. 3(a) Mean based (b) Medium based spatial outlier detection approach showing two outlier sites](image)

![Fig. 4 a. Mean based approach to detect spatial outlier sites b. Median based approach to detect spatial outlier sites](image)
5. Conclusion

In this paper, we used two spatial outlier detection algorithms using Mahalanobis distance to find most irrelevant sites from a spatial dataset with multiple attributes: one algorithm based on the average of the attribute values from neighborhoods and the other based on median of the attribute values. We have seen how decision making process to establish an IT industry may be improved. It is easy to analyze large numbers of sites which are spatially distributed. Proposed system is suitable in the situations where large numbers of spatial items are to be distinguished based upon their non spatial attributes. Performance of our system may be further improved by choosing more efficient spatial outlier detection methods. Furthermore, such systems may be generalized to use in other real life applications like financial decisions, demographic survey decisions, agricultural decisions and etc.

References

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