

# Searching The Best Keyword Using KRR and KNN

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**Abstract**—It is common that the database is associated with keywords based on the particular thing or the situation which indicates the business of the particular thing or the future purpose. In this best keyword cover search has an interesting problem known as Closest Keyword Search which is used to query an object which is known as keyword. It is used to cover the set of query keywords and it has the minimum inter distance. In earlier years, the rising availability and need of key text rating in reference error check rate for the advanced good decision making. This motivates us to search a generic version of nearest Key text search called Best Key text search Cover which considers inter-references path as well as the keyword rating of references. The base algorithm is excited by the methods of nearest Key text search which is based on exactly joined references from various questions key text to create user key text covers. When the sequence of query words is high, the activation of the baseline algorithm stops automatically as a result of heavy user key text covers generated. To fight that backlog, the work proposes a more scalable algorithm called key text nearest next expansion (keyword-NNE). Compared to the base algorithm, keyword-NNE algorithm significantly reduces the number of candidate keyword covers generated. The deep analysis and extensive experiments on actual data group have identified the superiority of our keyword-NNE algorithm.

A. **Keywords**— Spatial database, point of interests, keywords, keyword rating, keyword cover

## II. INTRODUCTION

DRIVEN by cell computing, placed-based services and wide availability of extensive digital maps and satellite imagery (e.g., Google Maps and Microsoft Virtual Earth services), the spatial keywords search problem has attracted much attention recently.

In a spatial database, each tuple represents a spatial object references which is associated with keyword(s) to indicate the data such as its businesses/services/features. Given a set of questions key text, a special task of spatial keywords search is to identify spatial object(s) which are joined with keywords relevant to a set of query key text, and have excited spatial relationships (e.g., close to each other and/or close to a query location). This error has specified value in various applications because users' needs are many expressed as many keywords. For example, a tourist who plans to visit a city may have particular shopping, dining and lodging and boarding needs. It is desirable that all these requirements can be executed without big space going. Due to the remarkable value in

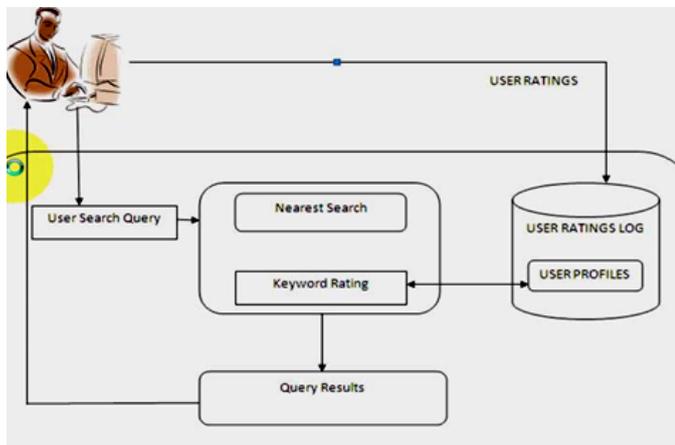
practice, several variants of spatial keyword search error rate have been studied. The works aim to find a number of individual objects, references each of which is exit to a query location and the associated keywords (or called document) are very relevant to a set of query keywords. This paper searches a generic version of CK query, called Best Keyword Cover (BKC) query, which considers interobjects distance as well as keyword rating. It is motivated by the concentration of high availability and priority of keyword rating in decision making. Millions of businesses/services/features surrounding the world have been rated by clients through online business review sites such as Yelp, Citysearch, ZAGAT and Dianping, etc. For example, a restaurant is rated 65 out of 100 (ZAGAT.com) and a hotel is rated 3.9 out of 5 (hotels.com). According to a survey in 2013 (dimensionalresearch.com), an overwhelming 90 percent claimed that purchasing decisions are influenced by online business review/rating. Due to the combination of keyword rating, the answer of BKC query can be very complex from that of CK query. Shows an example. Suppose the query keywords are Hotel, Restaurant and Bar. CK query returns since it considers the distance between the returned objects only. BKC query returns From the keyword text ratings of object are considered in addition to the inter-objects distance. Compared to CK query, BKC query supports more robust object error check rate and thus underpins the better decision making. To overcome this critical backlog, we developed much measurable keyword nearest neighbor expansion (keyword-NNE) algorithm which applies a different strategy. Keyword-NNE selects one query keyword as principal query keyword. The objects associated with the principal query keyword are principal objects references. For each principal object, the local best solution is computed. Among them the highest evaluation correction is the solution of BKC query. Given a principal object, it can be identified by simply retrieving a few nearby and highly rated objects in each non principal query keyword.

## III. LITERATURE SURVEY

Now a days, the enormous use of search engines has made it realistic to write spatial queries in a brand new way. Conventionally, queries focus on objects' geometric properties only, such as whether a point is in a rectangle, or how near two step are from together. We have seen some modern applications that call for the potential to select objects

based on both of their geometric coordinates and their associated texts. For example, It would be fairly useful if a search engine can be used to find the nearest restaurant that offers —steak, spaghetti, and brandy all at the same time. Note that this is not the globally closest restaurant (which would have been returned by a traditional nearest neighbour query), but the nearest restaurant among only those providing all the demanded foods and drinks. 1.2.1 Problem Definition. There are easy ways to support queries that joined spatial and text features. For example, for the above query, we could first fetch all the hotels whose card contain the group of keyvalues {steak, spaghetti, brandy}, and then from the retrieved restaurants, find the nearest one. Equity, one could also do it reversely by targeting first the spatial conditions – browse all the restaurants in ascending order of their distances to the query point until encountering one whose menu has all the keywords. The major backlogs of these straight approaches is that they will fail to provide real time answers on difficult inputs. A typical example is that the real nearest neighbour lies quite away from the searching poin, while all the closer neighbours are missing at least one of the query keywords.

#### IV. ARCHITECTURE



##### Module 1: Customer Registration:

In this module, the user will have to give the details first. Once the user gives the details then he/she can get the application. For registration user have to enter the basic information about himself. User also have to set the username and password. This all registration information is get stored into database. The IMEI number is automatically get stored into database once user do the registration.

##### Module 2: Customer Login:

In this module, after the registration customer can login through mentioned username and password.

##### Module 3: Hotel Registration:

In this module, Admin register the hotel with its famous dish. Hotel owner have to do the registration then only the hotel get search through application. Also hotel owner have to add the menu which is available in the resturent so that client can get the resturent through through. Only registered hotels will be displayed in the application. These hotel’s location will be seen in the map with distance. Each hotel owner will have the separate login id and secret key for access of data.

##### Module 3.1: Hotel Login / Admin:

In this model once Hotel Owner login into application then he can insert the menu or update the menu.

##### Module 4: Searching Keyword:

In this module, the user wants to give the keyword to find for menus present available in restaurant which will nearer from its position. Whenever user will enter keyword (menu name) it will match data with the hotel database server and find the nearest restaurant with the available entered menu by customer. For closest restaurant we are using IR2tree & compression. The IR2-Tree is a joined association of an R-Tree and signature files. In particular, each node of an IR2-Tree contains both spatial and keyword information; the former in the form of a minimum bounding area and the latter in the form of a signature. An IR2-Tree facilitates both top-k spatial queries and top-k spatial keyword queries as we explain below. More formally, an IR2-Tree R is a height-balanced tree data structure, where each leaf node has entries of the form (Obj Ptr, A, S). Object references PTR and A are defined as in the R-Tree while S is the signature of the object referred by Object references PTR. A non-leaf node has entries of the form (Node PTR, A, S). Node PTR and A are defined as in the R-Tree while S is the signature of the node. The signature of a node is the overlapping (OR-ing) of all the signatures of its entries. Thus a signature of a node is equivalent to a signature for all the documents in its sub tree.

##### Module 5: Map view / Searching Location:

In this module, all the label names of restaurant will appear in the list which came from database and find the position in map (google play service library) is required for showing position of restaurant in map and which will be easier to user to get the nearer restaurant from its current position.

##### Module 6: Distance Search:

In this module, customer can find the distance from source to destination. So that it can be easier to find the distance and reached the destination. It will give the distance of the hotel from the current location.

#### Future Scope

In future, we can use this system in different search engine application which will help the user to find the closest object references in faster way by searching keyword. It can useful in location based apps which will needed to find the closest paths for source to destination. Also it provides the fast reactive for the keyword which will expalins the input keyword related details.

#### V. CONCLUSION

Compared to the many relevant CK query, BKC query provides an additional dimension to support more sensible decision making. The introduced base algorithm is inspired by the methods for processing CK query. The base algorithm generates a large number of candidate keyword covers which leads to dramatic performance drop when many query keywords are given. The proposed keyword-NE algorithm applies a different processing strategy, i.e., searching local best solution for each object references in a certain query keyword. As a next sequences of resultant features, the number of candidate keyword covers generated is

significantly reduced. The analysis reveals that the number of user keyword text covers which need to be further processed in keyword-NNE algorithm is optimal and processing each keyword user cover typically generates much less new user keyword text covers in keyword-NNE algorithm than in the baseline algorithm.

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