



**Table 1:** Hash Format of string computation with L=5, M=2

word	Hashed bit string
A	00101
B	01001
C	00011
D	00110
E	10010

**Table 2:** Spatial points with text

P	Wp
P1	a, b
P2	b, d
P3	D
P4	a, e
P5	c, e
P6	c, d, e
P7	b, e
P8	c, d

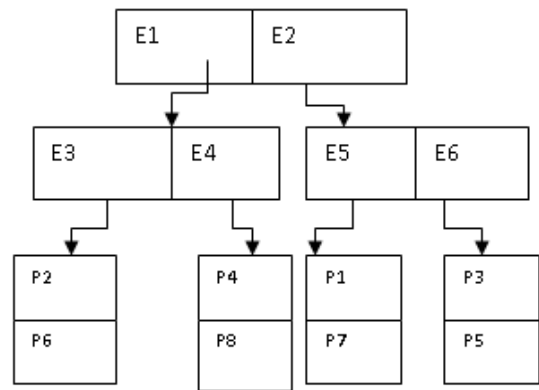
a word / string of length L = 5 and repeated bit M = 2 as shown in Table I. For example: In the above table we have to assume l=5,m=2 means if we have h(a) of a word, third and fifth bit set to 1 from left. In the hashing function we will OR the bit string of all the bit. Suppose I want to calculate signature of a,b so 01101. For finding the query word wq from W the SC perform the membership test in that test it will check whether all 1's of W appear at same location. If not it is sure that wq is not in the W. But sometimes false hit may occurs like assume that we want to check 'c' is member of the a,b using the set of signature h(c)=00011 and h(a,b)=01101 in the h(c) the fourth bit is 1, and h(a,b) fourth bit is 0 so the 'c' is not the member of a,b.

False hit example: Consider the membership test of SC in which 'c' will be test in ( b,d) in h(b,d) the fourth bit is 0 and h(c) fourth bit is 0 and SC report that 'c' is member of (b,d) and that is false hit.

**B. IR2-tree**

IR2-tree is based on the R-tree in fig.1. Each leaf, non-leaf entry is E which is summary of the text object. In the fig.1 we will illustrate the example based on data set of fig.1 and hash value which is represented in the Table 1. The string value 01111 in Fig 1 is the leaf entry, which is P2. The signature of the wp2 = b, d which is the document of the P2.

The string 11111 is the non-leaf entry E3 is the signature of wp2+ wp6 means the signature of the non-leaf entry is the combination of the signature of leaf node. Normally R tree, best first search algorithm is the better option of the NN nearest neighbor search.[1] For IR2-tree we have to fire the query point 'q' with associated text wq. The IR2-tree generate the ascending order of the distance of MBR to 'q' (MBR is the leaf entry). Pruning the entry whose signature which is absent the any one word of Wq. so in the Fig.1 for the verification the algorithm read all node of the tree and fetch the entry of p1,p4,p6 for the word c,d because the Wq is c,d and the final answer is p6 while p2,p4 are the false hit. So in the IR2-tree avoid the false hit which was occurred in signature file.



**Figure 1:** Signature of the entry

**Table 3:** Example of an inverted Index

word	Inverted list
a	p1,p4
b	p1,p2,p7
c	p5,p6,p8
d	p2,p3,p6,p8
e	p4,p5,p6,p7

- 1) Review study of Spatial invert index: Spatial invert index is the best method for accessing the keyword based retrieval .in the following list we will see the how to arrange the inverted index of points and the associated text of that point[5]. According to above list we have to create the list of inverted index which is having query word and associated point which having the same word [1]. One more point is that the list of the word is sorted order with regards point ID. So at the time of the query processing merge step will be performed on list. For example suppose we want find the point which is having words c, d because of that we will compute insertion of the inverted list. In the NN algorithm NN processing is with the IR2-tree. In that the points are retrieved from the index .Specifically NN query q with keyword set Wq the query method of I-index first determine the set of pq of all the points that will contains all the query word and then do —pq— randomly for finding the distance of pq from q.
- 2) Overview of Dbxplorer: The above paper is related to Keyword-Based Search over Relational Databases [2]. In day today internet is very user friendly for accessing the data. In this paper they have to give us the powerful question language. It will find the keyword from the server and retrieve the related web pages for the user.
- 3) Query processing on geographic data: In the geographic search the search engine allow the user to fire the query or find the result based geographic region [7]. It's also called local search, it is also useful for the extracting the knowledge of any location. It is also useful in the GIS. For the geographical search engine we need association of text as well as spatial data.

**C. Basic technology for the geographic search:**

- 1) Geo coding: In the geo coding technique three steps are necessary that are geo extraction, geo matching, geo propagation. Geo extraction: All the elements from a page which indicate query location. That is city name, contact number distance and generate the footprint. For the second step that is geo matching that foot print of same page will

be considered and in the third step that is geo propagation increase the quality scope of the geocoding by analyzing the link structure and the web pages topology [9]. And from that site map they will generate tree result.

2) Geographic query processing: Each query is having text term and query footprint means geographic ranking regarding user request. Thus in the above technique geographic ranking assign the score to each document footprint. Thus our overall ranking function in the form of

$$f(D, q) = g(fd, fq) \sum pr(D) \sum f(D, q) \quad (1)$$

In the above expression  $f(D,q)$ =Term based ranking,  $pr(D)$ =Global ranking,  $g(fd,fq)$ =geographic score And the expression will be calculated from footprint.

#### D. Concept of Bloom filter

Bloom filter [8] is one of the data structures that are useful for the membership queries to a set. The bloom filter needs very less space. Bloom filter avoids the false hit. It is normally used in the network. It is also used in the distributed database. As per above section in the signature file also use the bloom filter for the membership testing. Also it used in the password data structure and spell checking. Let F be a function of  $D = 0, 1, 2, n-1$  to  $R = \perp, 1, 2, 2r$ . According to above expression  $f(x) = \perp$  for all  $x \perp$ .  $\perp$  this symbol is used to represent 0.

#### E. View of Spatial Keyword Index

Above topic name suggest the retrieving concept that is one will retrieve the geographic location as well as associated text with the query. For the spatial keyword retrieval one need to first of all collective answer of the spatial query that full fill the user requirement for that one have to assume the database of spatial multidimensional object and after that one will find the set of keyword[6].

In that case  $q = (\lambda, \psi)$  where  $\lambda$  = location and  $\psi$  = keywords. This type of query is called spatial group keyword. In the above paper IR-tree and approximation algorithms are used.

### 3. Mathematical Model

Let S is a system of relevant Top-k neighbor search with Keywords which can be represented as,

$$S = I, P, O$$

Where,

I = Set of all inputs given to the system represented as,

$I = w_1, w_2, \dots, w_n$  where  $w_1, w_2, \dots, w_n$  are input keywords

P = Set of all possible processes required to get expected

Output it includes

1. Searching iteratively
2. Pruning
3. Word Partitioning

O = Set of all the required output can be represented as,

$O = O_1, O_2, \dots$  where  $O_1, O_2, \dots$  are all output Documents which has required keywords.

This method gives output in non-polynomial time. So, problem becomes NP-Complete.

### 4. Implementation Details

The goal of the implementation is to make an all-around evaluation on important aspects of the geo-textual indices and compare their performance. All indices and algorithms were implemented in Java running the Windows 8 OS. Machine is equipped with Intel(R) Core(TM)i3 @ 1.70GHz, 8GB RAM, and 1TB SATA disk. For ensuring a comparable evaluation results, the same server is used for conducting experiments on the same dataset. The Java Virtual Machine Heap is set to 2GB.

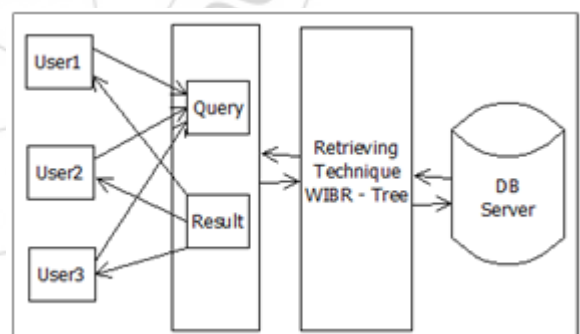
#### A. Datasets

For this study we are considering Hotel dataset which is prepared containing Hotel's details of location and types of Services they are providing. We concentrate on Hotels in Pune region.

#### B. Working Details

The architecture of proposed system is shown in Fig.2. It works in following steps:

1. Merging and distance browsing: In the spatial retrieving the basic problem is the bottleneck so need to avoid it. But in the I-index is having the simple way to recover it. In the I-index we have to preserve co-ordinates of the point in one group in the inverted list and that co-ordinates of the list are used to generate the R-tree. Now discuss how to perform keyword based NN. In this technique NN queries are processed with I index. For answering the query first of all we will access all the points which is having all the query keywords in  $W_q$ . It is very useful if we find the p very early in all the relevant inverted list. In that case we can access the list of element which are having less distance with q. so the p will be discovered all points of the list.



**Figure 2:** Proposed System

By using that we can count the number of copies of same point that will be relevant data. Consider an example if we want to NN search whose query point q and associated text is (c,d). for that search we have to use the list of word(c) and (d).from list

- 1) And now the new access order is depend on the distance of the given q. If we use the kNN then it will reported k nearest neighbor point and finish. Distance browsing is simple in the R-trees because R-tree uses best first algorithm which will give the exactly point with ascending order of the distance to q and R-trees are also the global access of the tree. For example at each step taking the point with the next point and return it. This algorithm is

normally work in the condition when the  $W_q$  small. But if the  $W_q$  is large then the out performance of sequential algorithm will be merged.

- 2) Weighted Independent Binary Representation (WIBR) tree: WIBR-tree [10] is a variant of IR-tree. It aims at partitioning objects into multiple groups such that each group shares as few keywords as possible. To achieve this goal, the objects in  $D$  is partitioned first into two groups using the most frequent word  $w_1$ : one group whose objects contain  $w_1$  and the other group whose objects do not. Then it partition each of these two groups by the next frequent word  $w_2$ . This process is repeated iteratively until each partition contains a certain number of objects. After partitioning, each group of objects becomes the leaf node of the WIBR-tree. Afterwards the tree is constructed following the structure of the IR-tree. When used for processing Boolean queries, the WIBR-tree [10] uses the inverted bitmap to replace the inverted file, which is denoted as the WIBR-tree, where a bitmap position corresponds to the relative position of an entry in its WIBR-tree node. The length of a bitmap is equal to the fan out of a node. Like the IR tree, the WIBR-tree can handle all the three types of query - Boolean kNN query, the top-k kNN query, and the Boolean range query.

### C. Algorithm

#### Searching Algorithm

The search algorithm traverse the tree from the root in a way similar to B-tree. In the following we denote the rectangle part of an index entry  $E$  by  $EI$ , and the tuple-identifier or child pointer part by  $EP$ . Algorithm Search: Given an WIBR-tree whose root node is  $T$ , find all index records whose rectangles overlap a search rectangle  $S$ .

- $S1$  [Search subtrees] If  $T$  is not a leaf, check each entry  $E$  to determine whether  $EI$  overlaps  $S$  For all overlapping entries, invoke Search on the tree whose root node is pointed to by  $EP$ .
- $S2$  [Search Leaf node ] If  $T$  is a leaf, check all entries  $E$  to determine whether  $EI$  overlaps  $S$  If so,  $E$  is a qualifying record.

### 5. Conclusion

This paper provides solution for problem of spatial keyword search and explained the performance limitations of current approaches. This system proposed a solution which is dramatically faster than current approaches and is based on a WIBR- tree. In particular we used the WIBR-Tree and showed how it is better than prior approaches. An efficient incremental algorithm was presented that uses the WIBR – Tree to answer spatial keyword queries. Here we are concentrating on application of searching of hotels at particular geographic area and which satisfy user requirement.

### 6. Acknowledgement

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