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## Efficient Keyword Search on Large Scale Graph in a Distributed System

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**ABSTRACT:** Graph Keyword Search has derived interest of number of research scientists, since we can represent graph models in both forms - structured and unstructured database. Keyword Search on the other hand can extract valuable information for users without knowledge of fundamental schema and query language. However, there are number of applications where graph can be much large than the available memory. If user tries to search for a location then first admin should enter a particular domain in the large data set. A web scale graph contains billions of vertices. The State of art approach employs centralized algorithms to process graph Keyword searches which are infeasible for such large graph due to limited computational algorithms to process and storage space of Centralized Server. To address this problem, we have worked and investigated keyword search for Web-scale graphs deployed in a distributed environment. Firstly, we have used a naive search algorithm to answer the query efficiently. Later, we encoded the shortest-path distance from a vertex to any given keyword in the graph. Performing these algorithms, we can find query answers by exploring fewer paths, so that the time and communication costs can be reduced.

**KEYWORDS:** Keyword Search, Distributed Graph, Scale Graph, and graph models.

### I. INTRODUCTION

Graph Keyword Search has derived interest of number of research scientists, since we can represent graph models in both forms - structured and unstructured database. Keyword Search on the other hand can extract valuable information for users without knowledge of fundamental schema and query language. However, there are number of applications where graph can be much large than the available memory. If user tries to search for a location, then first admin should enter a particular domain in the large data set. A web scale graph contains billions of vertices. The State of art approach employs centralized algorithms to process graph Keyword searches which are infeasible for such large graph due to limited computational algorithms to process and storage space of Centralized Server. To address this problem, we have worked and investigated keyword search for Web-scale graphs deployed in a distributed environment. Firstly, we have used a naive search algorithm to answer the query efficiently. Later, we encoded the shortest-path distance from a vertex to any given keyword in the graph. Performing these algorithms, we can find query answers by exploring fewer paths, so that the time and communication costs can be reduced.

### II. LITERATURE SURVEY

B. Ding, J. X. Yu, S. Wang, L. Qin, X. Zhang, and X. Lin found centralized keyword search over graphs, that is, given a set of keywords, we required to return top-k subtrees, subgraphs or h-cliques that contain these keywords. [1]

G. Bhalotia, A. Hulgeri, C. Nakhe, S. Chakrabarti, and S. Sudarshan found top-k subtrees by using backward search in BANKS-I. [3]

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M. R. Garey and D. S. Johnson found the exact top-k subtrees is an instance of the group Steiner tree problem, which is NP-hard. [5]

H. Wang and C. C. Aggarwal calculates a reachability index which is used to answer if two given nodes are reachable in a directed graph. In an answer tree, every leaf node is reachable from the root. However, given leaf nodes, we do not know which vertex of the graph is the root in advance. Therefore, we cannot simply use a reachability index to answer the keyword query. [6]

J. B. Rocha-Junior and K. Nrvag studied top-k spatial keyword queries on road networks, but with different query aims from this paper. The reason that the query in this paper returns the k best objects ranked in terms of both spatial distance to the query location and textual relevance to the query keywords, whereas our studied query returns k answer trees with the smallest total distances. [7]

Y. Yuan, G. Wang, L. Chen, and H. Wang aimed at finding top-k sub trees from a large uncertain graph. The filtering-and verification framework is employed to answer the query efficiently. There are some works to study the variants of graph keyword search. The problem definition and solutions in this paper focus on the probability computation of sub trees over uncertain graphs in a centralized manner, whereas those in this paper concentrate on the parallel and distributed computation of sub trees over deterministic graphs. [8]

M. Qiao, L. Qin, H. Cheng, J. X. Yu, and W. Tian studied the top-k nearest keyword (k-NK) query over a graph. A k-NK query searches for k nearest answer nodes, each of which contains all the query keywords. In contrast, the query in this paper looks for subtrees, in which leaf nodes jointly contain all the query keywords. [9]

Q. Lu, X. Y. Jeffrey, and C. Lijun utilized the power of database and backward search to find answer trees. [20]

### III. PROPOSED SYSTEM

In the system, there are three parameters such as the user, provider and admin. User first registers with all details and login the user searches the location with Domain of keyword, keyword and Location in which he is interested to search. Search location is stored on the graph dataset or normal data set. Admin store the location along with the location name, location keyword and domain of location. When user searches for a location, first he needs the access permission of particular user to search the location. Provider sends the encryption key to user to search the location in application and user Search the location with route.

Explanation:-

User/ Client: User searches the location after valid login and Registration and get the location and later, he finds the route from current location and Search location.

Provider: Provider sends the location and generates a randomly encryption key and checks for user details.

Admin: Admin adds the location with longitude and latitudes.

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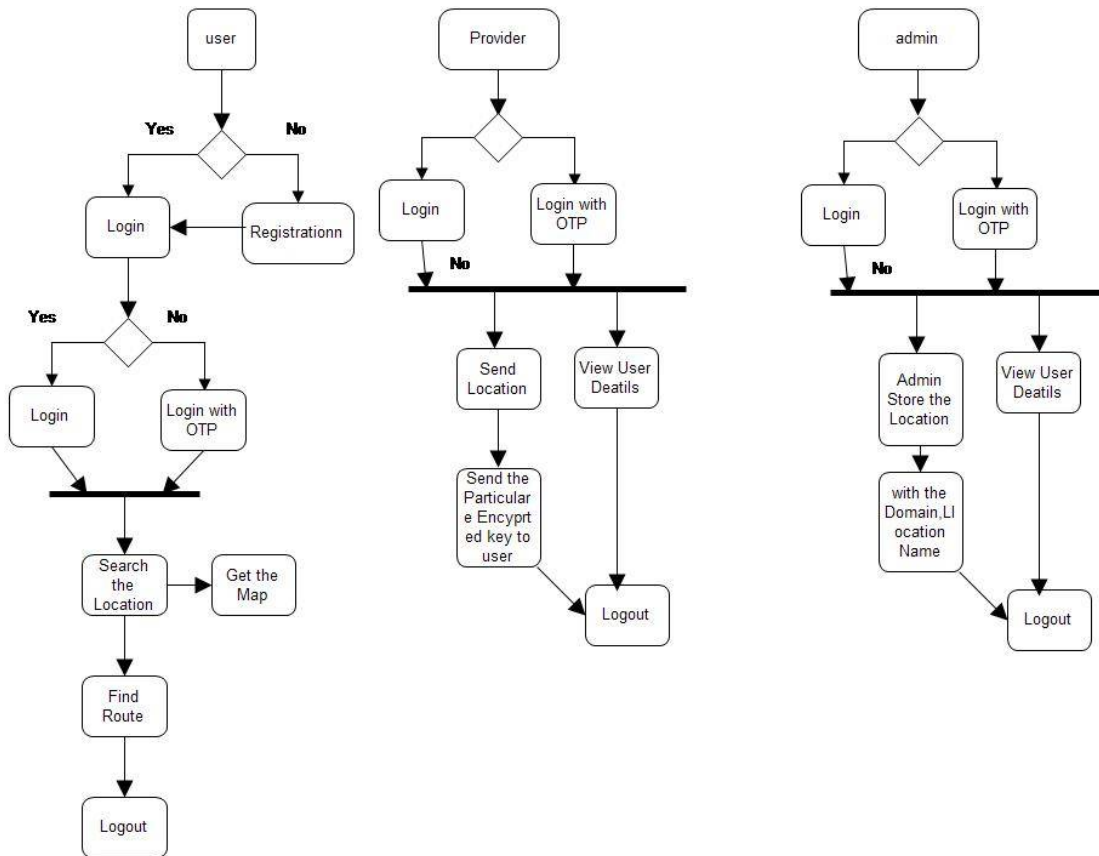


Fig. 1. Block Diagram of Efficient Keyword Search On Large Scale Graph In A Distributed System

## IV. MATHEMATICAL MODEL

Input  $n_i$  Belong  $V(G)$  , maximum Path desitances  $d_c$ , NI index

Output if there is a path from  $n_i$  to  $n_j$  with distance no greater than  $d_c$  return true; otherwise return false;

Let  $S$ , be a system such that,

$S = s, Subi, Pubi, Si, E$

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Where,

S- Proposed System

s- Initial state at

Subi = the Registered Subscriber with attributes Sid,pwd,Li

where

Sid = subscriberid

Li = Locationid

P ubi = theRegisteredSubscriberwithattributesP id, pwd, Li

where

Li = Locationid

EEvent

N N ode

SSubscription

U BUpperBoundSidIdentif ieransubscription

X – Input of the system

– L(Location), S(Searching)andRtreenode

Y Outputof SystemT opK Subscription matchin gand show in Map

## V. CONCLUSION

Applications on big graph usually suffer from extremely high time and space complexities, and thus require solutions of the distributed and parallel computing. In this paper, we have studied the problem of the keyword search over a distributed graph, which returns top-k answer trees. To efficiently find query answers, we give a distributed backward search algorithm and synchronize it over the cluster to terminate the search as early as possible. However, this naive solution incurs massive time and communication costs, due to the flooding nature of the search strategy. To solve this issue, we design a signature-based search mechanism, which explores the approach of spreading summary information about the vertex of hosted keywords in its neighbourhood, and then uses this information for forwarding the distributed backward search. Therefore, the signature-based search algorithm determines query answers by traversing only few paths of the graph, so that it costs very low time and network overhead. In addition, we re-organize the distributed graph to optimize the processes of the Decay and Diffusion Rules. Experiments show that our approach can improve the search efficiency by many times faster than the naive algorithm.

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