

Resource and Knowledge Discovery in Global Information Systems: A Multiple Layered Database Approach

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Abstract

With huge amounts of information connected to the global information network (Internet), efficient and effective discovery of resource and knowledge from the “global information base” has become an imminent research issue, especially with the advent of the Information Highway. In this article, a multiple layered database (MLDB) approach is proposed to handle the resource and knowledge discovery in global information base. A multiple layered database is a database formed by generalization and transformation of the information, layer-by-layer, starting from the original information base (treated as layer-0, the primitive layer). Information retrieval, data mining, and data analysis techniques can be used to extract and transform information from a lower layer database to a higher one. Layer-1 and higher layers of an MLDB can be modeled by an extended-relational or object-oriented model, constructed automatically, and updated incrementally. Information at all the layers except the primitive one can be stored, managed and retrieved by the available database technology; resources can be found by controlled search through different layers of the database; and knowledge discovery can be performed efficiently in such a multiple layered database.

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1 Introduction

With the rapid expansion of information base and user community in the Internet, efficient and effective discovery and use of the resources in the global information network has become an important issue in the research into global information systems.

Although researches and developments of database systems have been flourishing for many years, with different kinds of database systems successfully developed and delivered to the market, a global information system stores much huger amount of information in a much more complicated and unstructured manner than any currently available database systems. Thus, the effective organization, discovery and use of the rich resources in the global information network poses great challenges to the database and information system researchers.

The first major challenge of a global information system is the **diversity of information** in the global information base. The current information network stores hundreds of tera-bytes of information including documents, softwares, images, sounds, commercial data, library catalog, user directory data, weather, geography, other scientific data, and many other types of information. Since users have the full freedom to link whatever information they believe useful to the global information network (why should not?), the global information base is huge, heterogeneous, in multimedia form, mostly unstructured, dynamic, incomplete or even inconsistent, which creates tremendous difficulty in systematic management and retrieval than the structured, well-organized data in most commercial database systems.

The second challenge is the **diversity of user community**. The Internet currently connects over 2 million workstations [4], and the user community is still expanding rapidly. Users may have quite different backgrounds, interests, and purposes of usage. Also, most users may not have a good knowledge about the structure of the information system, may not be aware of the heavy cost of a particular search (e.g., a click may bring megabytes of data over half of the globe), and may easily get lost by groping in the “darkness” of the network, or be bored by taking many hops and waiting impatiently for a piece of information.

The third challenge is the **volume of information** to be searched and transmitted. The huge amount of unstructured data makes it unrealistic for any database systems to store and manage and for any queries to find *all* or even *most* of the answers by searching through the global network. The click-triggered massive data transmission over the network is not only costly and unbearable even for the broad bandwidth of the communication network, but also too wasteful or undesirable to many users. Search effectiveness (e.g., hit ratio) and performance (e.g., response time) will be bottlenecks for the successful applications of the global information system.

There have been many interesting studies on information indexing and searching in the global information base with many global information system servers developed, including Archie [11], Veronica, WAIS [18], etc. Although these tools provide indexing and document delivery services, they aim at a very specific service like FTP or gopher [24]. Attempts have also been made to discover resources in the World Wide Web [3, 27]. Spider-based indexing techniques, like the WWW Worm [23], RBSE database [9], Lycos [22] and others, create a substantial value to the web users but generate an increasing Internet backbone traffic. They not only flood the network and overload the servers but also lose the structure and the context of the documents gathered. These wandering software agents on the World Wide Web have already created controversies [21]. Other indexing solutions, like ALIWEB [20] or Harvest [5], behave well on the network but still struggle with the difficulty to isolate information with relevant context. Essence [17, 5], which uses a “semantic” indexing, is one of the most comprehensive indexing systems known up to now. However, it still cannot solve most of the problems posed for systematic discovery of resources and knowledge in the global information base.

In this article, a different approach, called a **Multiple Layered DataBase (MLDB)** approach is proposed to facilitate information discovery in global information systems. A **multiple layered database (MLDB)** is a database composed of several layers of information, with the lowest layer (i.e., *layer-0*) corresponding to the primitive information stored in the global information base and the higher ones (i.e., *layer-1* and above) storing generalized information extracted from the lower layers.

The proposal is based on the previous studies on *multiple layered databases* [26, 16] and *data mining* [25, 14] and the following observations.

With the development of data analysis, transformation and generalization techniques, it is possible to generalize and transform the diverse, primitive information in the network into reasonably structured, classified, descriptive and higher-level information. Such information can be stored into a massive, distributed but structured database which serves as the layer-1 database in the MLDB. By transforming an unstructured global information base into a relatively structured “global database”, most of the database technologies developed before can be applied to manage and retrieve information at this layer.

However, the layer-1 database is usually still too large and too widely distributed for efficient browsing, retrieval, and information discovery. Further generalization should be performed on this layer at each node to form higher layer(s) which can be then merged with the corresponding layered database of other nodes at some backbone site in the network. The merged database can be replicated and propagated to other remote sites for further integration [6]. This integrated,

higher-layer database may serve a diverse user community as a high-level, global information base for resource discovery, information browsing, statistical studies, etc.

The multiple layered database architecture transforms a huge, unstructured, global information base into progressively smaller, better structured, and less remote databases to which the well-developed database technology and the emerging data mining techniques may apply. By doing so, the power and advantages of current database systems can be naturally extended to global information systems, which may represent a promising direction.

In this article, we propose a multiple layered database model for global information systems and study how to construct and maintain such an MLDB and how to perform effective information discovery using MLDBs. The remaining of the paper is organized as follows. In Section 2, a model for global MLDB is introduced. Methods for construction and maintenance of different layers of the global MLDB are proposed in Section 3. Resource and knowledge discovery using the global MLDB is investigated in Section 4. A discussion of the benefits of the MLDB model and the related issues are presented in Section 5. Finally, the study is summarized in Section 6.

2 A Multiple Layered Database Model for Global Information Systems

Although it is difficult to construct a data model for the primitive (i.e., layer-0) global information base, advanced data models can be applied in the construction of better structured, higher-layered databases. To facilitate our discussion, we assume that the nonprimitive layered database (i.e., layer-1 and above) is constructed based on an **extended-relational model** with capabilities to store and handle complex data types, including set- or list- valued data, structured data, hypertext, multimedia data, etc. MLDBs can also be constructed similarly using other data models, including object-oriented and extended entity-relationship models.

Definition 2.1 *A global multiple layered database (MLDB) consists of 3 major components: $\langle S, H, D \rangle$, defined as follows.*

1. *S: a database schema, which contains the meta-information about the layered database structures;*
2. *H: a set of concept hierarchies; and*
3. *D: a set of (generalized) database relations at the nonprimitive layers of the MLDB and files in the primitive global information base. □*

The first component, a **database schema**, outlines the overall database structure of the global MLDB. It stores general information such as structures, types, ranges, and data statistics about the relations at different layers, their relationships, and their associated attributes. Moreover, it describes which higher-layer relation is generalized from which lower-layer relation(s) (i.e., a route map) and how the generalization is performed (i.e., generalization paths). Therefore, it presents a route map for data and meta-data (i.e., schema) browsing and for assistance of resource discovery.

The second component, a **set of concept hierarchies**, provides a set of predefined concept hierarchies which assist the system to generalize lower layer information to high layer ones and map queries to appropriate concept layers for processing.

The third component consists of the whole **global information base** at the primitive information level (i.e., layer-0) and the **generalized database relations** at the nonprimitive layers.

We first examine the **database schema**. Because of the diversity of information stored in the global information base, it is difficult to create relational database structures for the primitive layer information base. However, it is possible to create relational structures to store reasonably structured information generalized from primitive layer information. For example, based on the accessing patterns and accessing frequency of the global information base, layer-1 can be organized into dozens of database relations, such as *document*, *person*, *organization*, *software*, *map*, *library_catalog*, *commercial data*, *geographic_data*, *scientific_data*, *game*, etc. The relationships among these relations can also be constructed either explicitly by creating relationship relations as in an entity-relationship model, such as *person-organization*, or implicitly (and more desirably) by adding the linkages in the tuples of each (entity) relation during the formation of layer-1, such as *adding URL¹ pointers pointing to the corresponding authors (“persons”) in the tuples of the relation “document” when possible*.

To simplify our discussion, we assume in this paper that the layer-1 database contains only two relations, **document** and **person**. Other relations can be constructed and generalized similarly.

Example 2.1 Let the database schema of layer-1 contain two relations, **document** and **person**, as follows (with the attribute type specification omitted).

1. *document(file_addr, authors, title, publication, publication_date, abstract, language, table_of_contents, category_description, key_words, index, URL_links, multimedia_attached, num_pages, form, first_page, size_doc, time_stamp, access_frequency, ...)*.
2. *person(last_name, first_name, home_page_addr, position, picture_attach, phone, e-mail, office_address, education, research_interests, publications, size_of_home_page, time_stamp, access_frequency, ...)*.

¹Uniform Resource Locator. Reference is available by anonymous FTP from ftp.w3.org as /pub/www/doc/url-spec.txt

Take the *document* relation as an example. Each tuple in the relation is an abstraction of one *document* at layer-0 in the global information base. The first attribute, *file_addr*, registers its file name and its “URL” network address. There are several attributes which register the information directly associated with the file, such as *size_doc* (size of the document file), *time_stamp* (the last updating time), etc. There are also attributes related to the formatting information. For example, the attribute *form* may indicate the format of a file: .ps, .dvi, .tex, .troff, .html, text, compressed, uuencoded, etc. One special attribute, *access_frequency*, registers how frequently the entry is being accessed. Other attributes register the major semantic information related to the document, such as *authors*, *title*, *publication*, *publication_date*, *abstract*, *language*, *table_of_contents*, *category_description*, *key_words*, *index*, *URL_links*, *multimedia_attached*, *num_pages*, *first_page*, etc. □

Layer-1 is a detailed abstraction (or descriptor) of the layer-0 information. It should be substantially smaller than the primitive layer global information base but still rich enough to preserve most of the interesting pieces of general information for a diverse community of users to browse and query. Layer-1 is the lowest layer of information manageable by database systems. However, it is usually still too large and too widely distributed for efficient storage, management and search in the global network. Further compression and generalization can be performed to generate higher layered databases.

Example 2.2 Construction of an MLDB on top of the layer-1 global database.

The two layer-1 relations presented in Example 2.1 can be further generalized into layer-2 database which may contain two relations, *doc_brief* and *person_brief*, with the following schema,

1. *doc_brief*(*file_addr*, *authors*, *title*, *publication*, *publication_date*, *abstract*, *language*, *category_description*, *key_words*, *major_index*, *URL_links*, *num_pages*, *form*, *size_doc*, *access_frequency*).
2. *person_brief* (*last_name*, *first_name*, *publications*, *affiliation*, *e-mail*, *research_interests*, *size_home_page*, *access_frequency*).

Further generalization can be performed on layer-2 relations in several directions. One possible direction is to partition the *doc_brief* file into different files according to different classification schemes, such as category description (e.g., *cs_document*), access frequency (e.g., *hot_list_document*), countries, publications, etc., or their combinations. Choice of partitions can be determined by studying the referencing statistics. Another direction is to further generalize some attributes in the relation and merge identical tuples to obtain a “summary” relation (e.g., *doc_summary*) with data distribution statistics associated [14]. The third direction is to join two or more relations. For

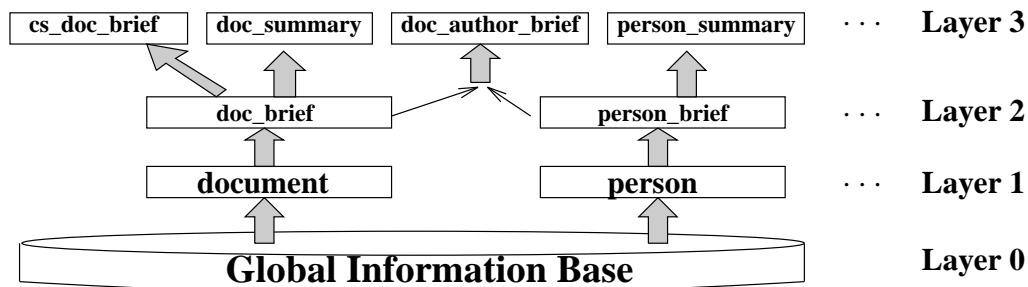


Figure 1: A conceptual route map of the global information base

example, *doc_author_brief* can be produced by generalization on the join of *document* and *person*. Moreover, different schemes can be combined to produce even higher layered databases.

A few layer-3 relations formed by the above approaches are presented below.

1. *cs_doc*(*file_addr*, *authors*, *title*, *publication*, *publication_date*, *abstract*, *language*, *category_description*, *key_words*, *URL_links*, *num_pages*, *form*, *size_doc*).
2. *doc_summary*(*affiliation*, *field*, *publication_year*, *count*, *first_author_list*, *file_addr_list*).
3. *doc_author_brief*(*file_addr*, *authors*, *affiliation*, *title*, *publication*, *pub_date*, *category_description*, *key_words*, *num_pages*, *URL_links*, *form*, *size_doc*).
4. *person_summary* (*affiliation*, *research_interest*, *year*, *num_publications*, *count*).

In general, the overall global MLDB structure should be constructed based on the study of frequent accessing patterns. It is also plausible to construct higher layered databases for a special interested community of users (e.g., ACM/SIGMOD, IEEE/CS) on top of a common layer of the global database. This customized local higher layer acts as cache which may drastically reduce the overall network traffic [7, 2].

One possible schema of a global MLDB (containing only two layer-1 relations) is presented in Fig. 1. □

3 Construction and maintenance of MLDBs

A philosophy behind the construction of MLDB is information abstraction, which assumes that most users may not like to read the details of large pieces of information (such as complete documents) but may like to scan the general description of the information. Usually, the higher level of abstraction, the better structure the information may have. Thus, the sacrifice of the detailed level of information may lead to a better structured information base for manipulation and retrieval.

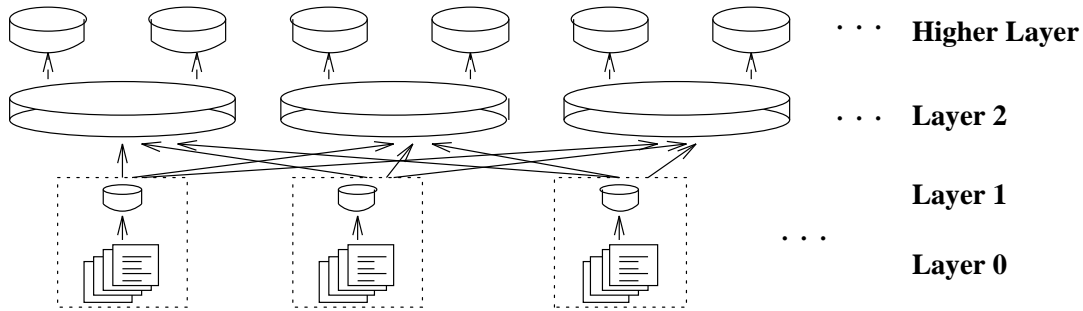


Figure 2: The general architecture of a global MLDB and its construction.

Fig. 2 presents the general architecture of a multiple layered global information base, where the existing global information base forms layer-0, and the abstraction of layer-0 forms layer-1. Further generalization of layer-1 and their integration from different sites form layer-2, which can be replicated and propagated to each backbone site, and then be further generalized to form higher layers.

3.1 Construction of layer-1: From global information base to structured database

The goal for the construction of layer-1 database is to transform and/or generalize the unstructured data of the primitive layer at each site into relatively structured data, manageable and retrievable by the database technology. Three steps are necessary for the realization of this goal: (1) **standardization of the layer-1 schema**, (2) **development of a set of softwares which automatically perform the layer-1 construction**, and (3) **layer construction and database maintenance at each site**.

Obviously, it is neither realistic nor desirable to enforce standards on the format or contents of the primitive layer information in the global information network. However, it is desirable to construct a rich, shared and standardized layer-1 schema because such a schema may lead to the construction of a structured information base and facilitate information management and sharing in the global information network.

A standard layer-1 schema can be worked out by studying the accessing history of a diverse community of users and predicting the future accessing patterns by experts. Such a schema can be constructed incrementally in the process of increasingly popular use of the information network, or standardized by some experts or a standardization committee. However, it is expected that such a schema may need some infrequent, incremental modifications, and the layer-1 database should be modified accordingly in an automatic and incremental way by applying some upgrading software(s).

To serve a diverse community of users for flexible information retrieval, the layer-1 schema should

be rich enough to cover most popular needs and detail enough to reduce excessive searches into the layer-0 information base. Because of the diversity of information in a global information base, there often exist cases that data have complex structures or cannot match the specified schema. Mechanisms should be developed to handle such diversity. We examine a few such cases.

1. **Attribute values with complex structures or in multimedia forms.**

Some attributes may contain set- or list- valued data or nested structures. For example, the attribute “*authors*” contains a list of authors each having a structure: *name*, *affiliation*, *e-mail*, etc. Some attributes may be of a long text form (e.g., *abstract*, *first_page*) or in a multimedia form (e.g., the *picture* of a *person* could be a *photo* or a segment of *video*). Such nested structures, sets, lists, hypertext or multimedia data can be defined by extended data types, as in many extended-relational or object-oriented database systems [10, 19].

2. **Missing attribute values.**

Since different users may have different conventions to connect the information to the network, it often happens that some attribute values may be missing. For example, *publication*, *publication_date*, and *table_of_contents* may not exist in a particular document.

Although missing values can be handled straightforwardly by introducing a null value, such as *value_unavailable*, efforts should be paid to dig up the values implicitly stored in the global information base. For example, softwares can be developed to extract title, authors and their affiliations from a .tex file or a .ps file. Index can be constructed automatically and/or selectively by searching through a document file for frequently occurring terms. Keywords may need to be extracted, if not exist, from the frequently appearing technical terms in the text. The document publication information, if not exist in the *document*, may need to be extracted from the corresponding authors’ publication record, or other relevant information. A person’s research interest, may be extractable from his/her .plan file or from the researcher’s frequently used keywords in his/her publications. The bi-directional URL linkages between a document and its authors’ home pages can be created, if not already exist, by searching through the corresponding files.

3. **Inconsistent or variant attribute values.**

Some attributes of an entry may consist of multiple, potentially inconsistent values due to multiple entries of information. For example, a person may have several working addresses. Also, some attributes may contain a set of variant values. For example, one document may have several variant forms (such as .tex, .dvi, or .ps forms) on the network.

The layer-1 schema should be flexible enough to allow variations. Multiple variations or their pointers should be stored in the corresponding attributes, with flags identifying their distinctions. Also, warning messages should be produced (e.g., in some .log file) to tell local information managers about the potential inconsistency of the data.

4. System- or network- dependent values.

Some information, such as the network address of a document, may be subjected to change due to the network reconstruction. It may also be necessary to create a system-generated object identifier, such as *doc_id*, to identify the object, independent of its URL address.

Since the layer-1 construction is a major effort, a set of softwares should be developed to automate the construction process. (Notice that some existing global information index construction softwares, like the Harvest Gatherer [5], have contributed to such practice and could be further developed to meet our needs). Unless certain information network construction standard is enforced, a software may not always be able to determine certain properties of a file, e.g., a file category (document vs. library catalog), which may need some minimal human assistance.

The layer-1 construction softwares, after being developed and tested, should be released to the information system manager in a regional or local network, which acts as a “*local software robot*” for automated layer-1 construction. Customization may need to be performed on some softwares, such as handling multilingual information, etc., before they can be successfully applied to their local information bases to generate consistent layer-1 local databases. Softwares for upgrading database structures and information transformers should also be released timely to local information system managers to keep their local layer-1 database upgraded and consistent with others.

3.2 Generalization: Formation of higher layers in MLDB

Since local layer-1 databases are all connected via Internet, they collectively form a globally distributed, huge layer-1 database. Although information retrieval can be performed directly on such a global database, performance will be poor if global-wide searches have to be initiated frequently.

Higher layered databases can be constructed on top of the layer-1 database by generalization techniques. Generalization will reduce the size of the global database, make it less distributed (by replicating the smaller, higher-layer databases at, for example, network backbone sites or local server sites), while still preserving the general descriptions of the layer-1 data.

Clearly, successful generalization becomes a key to the construction of higher layered databases. Following our previous studies on attribute-oriented induction for knowledge discovery in relational

databases [14, 15], an attribute-oriented generalization method has been proposed for the construction of multiple layered databases [16]. According to this method, data in a lower layer relation are generalized, attribute by attribute, into appropriate higher layer concepts. Different lower level concepts may be generalized into the same concepts at a higher level and be merged together, which reduces the size of the database.

We examine in detail the generalization techniques for the construction of higher layered databases.

3.2.1 Concept generalization

Nonnumeric data (such as keywords, index, etc.) is the most popularly encountered type of data in the global information base. Generalization on nonnumerical values should rely on the concept hierarchies which represent necessary background knowledge that directs generalization. Using a concept hierarchy, primitive data can be expressed in terms of generalized concepts in a higher layer.

Concept hierarchies should be provided explicitly by domain experts or stored implicitly in the database. For the global MLDB, a set of relatively stable and standard concept hierarchies should be provided as a common reference by all the local databases in their formation of higher layered databases and in their browsing and retrieval of information using different levels of concepts.

The concept hierarchies for keywords and indices can be obtained by referencing a standard concept hierarchy catalog which specifies the partial order of the terms frequently used in the global information base.

A portion of the concept hierarchy for *keywords* is illustrated in Fig. 3, and the specification of such a hierarchy and alias is in Fig. 4. Notice that a **contains**-list specifies a concept and its immediate subconcepts; and an **alias**-list specifies a list of synonyms (aliases) of a concept, which avoids the use of complex lattices in the “hierarchy” specification. The introduction of alias-lists allows flexible queries and helps dealing with documents using different terminologies and languages. Also, the dashed lines between concepts in Fig. 3 represent the possibility to have other layers of concepts in between.

Such a concept hierarchy can be constructed as follows.

1. Collect the frequently used words, technical terms and search keys for classification.
2. Build up a skeleton classification hierarchy based on the technical term specification standards in each field, such as *ACM Computing Review: Classification System for Computing Reviews*, etc. Notice that most of such classification standards are on-line documents.

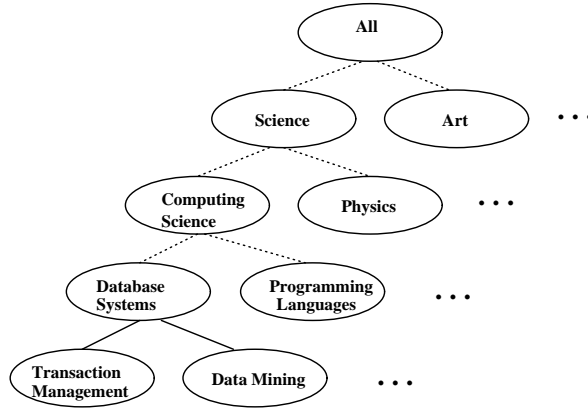


Figure 3: A possible concept hierarchy for *keywords*

All	<u>contains:</u>	Science, Art, ...
Science	<u>contains:</u>	Computing Science, Physics, Mathematics, ...
Computing Science	<u>contains:</u>	database systems, Programming Languages, ...
Computing Science	<u>alias:</u>	Information Science, Computer Science, Computer technologies, ...
Database Systems	<u>contains:</u>	Data mining, transaction management, query processing, ...
Database Systems	<u>alias:</u>	Database technologies, Data management, ...
Data mining	<u>alias:</u>	Knowledge discovery, data dredging, data archaeology, ...
Transaction management	<u>contains:</u>	concurrency control, recovery, ...
...		

Figure 4: Specification of hierarchies and aliases.

3. Consult on-line dictionaries (such as Webster dictionary and thesaurus, etc.) for automatically attaching the remaining words to appropriate places in the hierarchy (which may need some human interaction).
4. Consult experts in the fields to make sure that the hierarchy is reasonably complete and correct.
5. Incrementally update such a hierarchy, when necessary, due to the introduction of new terminologies.

The constructed concept hierarchies should be replicated to each server, together with the higher layered databases, for information browsing and resource discovery.

Generalization on numerical attributes can be performed in a more automatic way by the examination of data distribution characteristics [1, 13, 8]. In many cases, it may not require any predefined concept hierarchies. For example, the size of document can be clustered into several

groups, such as $\{below\ 10Kb,\ 10Kb-100Kb,\ 100Kb-1Mb,\ 1Mb-10Mb,\ over\ 10Mb\}$, according to a relatively uniform data distribution criteria or using some statistical clustering analysis tools. Appropriate names can be assigned to the generalized numerical ranges, such as $\{tiny-size,\ small-size,\ middle-size,\ large-size,\ huge-size\}$ by users or experts to convey more semantic meaning.

With the availability of concept hierarchies, we examine two kinds of generalizations, *data generalization* and *relation generalization*.

3.2.2 Attribute-oriented data generalization

Attribute-oriented data generalization generalizes data in each attribute to certain high concept layer(s). For a long data item, such as multimedia data, long text data, etc., or a group of data containing a set or list of elements, generalization needs to be performed to compress data in a meaningful way. Also, for terms with nonstandard usages, generalization may imply to standardize it for common usage.

Techniques to be applied for data generalization usually depends on the types of data to be generalized.

1. Multimedia data:

Some attribute at layer-1 or layer-0 may contain complex graphics, images, voice, music, and other forms of audio/video information. Such multimedia data are typically quite large, stored as sequences of bytes with variable lengths, and segments of data are linked together for easy reference. Because of the size of multimedia data, it is preferable to either remove it or generalize it to less bulky, more descriptive forms.

Generalization on multimedia data can be performed by recognition and extraction of the essential features and/or general patterns of such data. There are several ways to extract the essential features or general patterns from segments of multimedia data. For an image, the size, color, number of objects, and the shape of the objects in the image can be extracted by some pattern recognition algorithms or aggregation/approximation techniques. For a segment of music, its melody can be summarized based on the approximate patterns that repeatedly occur in the segment, and its style can be summarized based on its tone, tempo, major musical instruments played, etc. In general, it is a challenging task to generalize multimedia data to extract interesting knowledge implicitly stored there. The successful storage and manipulation of generalized multimedia data at high database layers will depend on the research progress on knowledge mining from multimedia data.

2. Long text data:

A long text data could be an article, a book, an e-mail, or other kinds of documents. Depending on the type of the long text, it could be generalized in several ways. For an article, the title, authors, publications, abstract, table of contents, the subject and index terms frequently occurring in the article, etc. may serve as generalization results. Thus the specification of the schema for a document itself represents the usual outline of its generalization. Similarly, e-mails, treated as another information type, may be outlined with a schema containing senders, recipients, sending time, receiving time, subject, keywords, index, etc. for automatic generalization.

3. Structure-valued data:

Complex structure-valued data includes set-valued and list-valued data and data with nested structures. Taking the set-valued attribute as an example, it can be generalized in several ways:

- (a) If a set contains a large number of elements each associated with some weight factors (such as occurrence frequency), the lower weighted elements may be removed from the set in generalization. For example, the indices of an article can be generalized by removing the indices with low occurrences.
- (b) Generalize each value in a set into its corresponding higher level concepts and compress them when possible, and with a *count* associated with the preserved element. For example, the *hobby* of a person is a set-valued attribute which contains a set of values, such as $\{tennis, hockey, chess, violin, nintendo\}$, which can be generalized into a set of high level concepts, such as $\{sports(3), music(1), video_games(1)\}$.
- (c) Derivation of the general behavior of a set, such as the number of elements in the set, the types or value ranges in the set, the weighted average for numerical data, etc.

Notice that generalization can be performed by applying different generalization operators to explore alternative generalization paths. In this case, the result of generalization is a heterogeneous set.

3.2.3 Attribute-oriented relation generalization

Data generalization refers generalizing data within an attribute in a relational tuple, such as merging generalized data within a set-valued data item; whereas relation generalization refers to generalizing

a relation, which often involves merging generalized, identical tuples in a relation.

By removing nongeneralizable attributes (such as long text data, etc.) and generalizing data in other attributes into a small set of values, some different tuples may become identical at the generalized concept level and can be merged into one. A special attribute, *count*, should be associated with each generalized tuple to register how many original tuples have been generalized into the current one. This process reduces the size of the relation to be stored in a generalized database but retains the general description of the data of the original database at a high concept level. Such a summarized view of data may facilitate high-level information browsing, statistical study, and data mining.

With data and relation generalization techniques available, the next important question is how to selectively perform appropriate generalizations to form useful layers of databases. In principle, there could be a large number of combinations of possible generalizations by selecting different sets of attributes to generalize and selecting the levels for the attributes to reach in the generalization. However, in practice, a few layers containing most frequently referenced attributes and patterns will be sufficient to balance the implementation efficiency and practical usage.

Frequently used attributes and patterns should be determined before generation of new layers of an MLDB by the analysis of the statistics of query history or by receiving instructions from users and experts. It is wise to remove rarely used attributes but retain frequently referenced ones in a higher layer. Similar guidelines apply when generalizing attributes to a more general concept level. For example, for a document, the further generalization of layer-1 *document* to layer-2 *doc_brief* can be performed by removing the less frequently inquired attributes *table_of_contents*, *first_page*, etc., and generalize the *index* into *major_index*.

Notice that a new layer could be formed by performing generalization on one relation or on a join of several relations based on the selected, frequently used attributes and patterns. Generalization [14] is performed by removing a set of less-interested attributes, substituting the concepts in one or a set of attributes by their corresponding higher level concepts, performing aggregation or approximation on certain attributes, etc.

Since most joins of several relations are performed on their key and/or foreign key attributes, whereas generalization may remove or generalize the key or foreign key attributes of a data relation, it is important to distinguish the following two classes of generalizations.

1. **key-preserving generalization**, in which all the key or foreign key values are preserved.
2. **key-altering generalization**, in which some key or foreign key values are generalized, and thus

altered. The generalized keys should be marked explicitly since they usually cannot be used as join keys at generating subsequent layers.

It is crucial to identify altered keys since if the altered keys were used to perform joins of different relations, it may generate incorrect information [16]. Notice that join on generalized attributes, though undesirable in most cases, could be useful if the join is to link the tuples with *approximately* the same attribute values together. For example, for search documents, one may like to consider some closely related but not exactly the same subjects. Such kind of join is called an **approximate join** to be distinguished from the **precise join**.

Usually, only *precise join* is considered in the formation of new layered relations using joins because approximate join may produce huge sized joined relations and may also be misleading in the semantic interpretation at different usages. However, approximate join will still be useful for searching some weakly connected concepts in a resource discovery query.

For layer formation, we have the following regulation.

Regulation 3.1 Join in layer formation in the global MLDB must be a precise join. (i.e., join cannot be performed on the generalized attributes.)

3.2.4 An MLDB construction algorithm

Based on the previous discussion, the construction of an MLDB can be summarized into the following algorithm, which is similar to attribute-oriented generalization in knowledge discovery in databases [14].

Algorithm 3.1 Construction of an MLDB.

Input: A global information base, a set of concept hierarchies, and a set of frequently referenced attributes and frequently used query patterns.

Output: A global multiple layered database (MLDB).

Method. A global MLDB is constructed in the following steps.

1. Determine the multiple layers of the database based on the frequently referenced attributes and frequently used query patterns.

2. Starting with the global information base (layer-0), generalize the relation step-by-step (using the given concept hierarchies and generalized schema) to form multiple layered relations (according to the layers determined in Step 1).
3. Merge identical tuples in each generalized relation and update the *count* of the generalized tuple.
4. Construct a new schema by recording the definitions of all the generalized relations, their relationships and the generalization paths. □

Rationale of Algorithm 3.1.

Step 1 indicates that the layers of an MLDB should be determined based on the frequently referenced attributes and frequently used query patterns. This is reasonable since to ensure the elegance and efficiency of an MLDB, only a small number of layers should be constructed, which should provide maximum benefits to the frequently accessed query patterns. Obviously, the frequently referenced attributes should be preserved in higher layers, and the frequently referenced concept levels should be considered as the candidate concept levels in the construction of higher layers. Steps 2 and 3 are performed in a way similar to the attribute-oriented induction, studied previously [14, 16]. Step 4 constructs a new schema which records a route map and the generalization paths for information browsing and knowledge discovery, which will be discussed in detail in Section 4. □

Example 3.1 An portion of relation *doc_brief* is presented in Table 1.

file_addr	authors	title	publication	pub_date	key_words	...
http://fas.sfu.ca/9/cs/research/projects/HMI-5/documents/papers/han/coop94.ps.gz	J. Han Y. Fu R. Ng	Cooperative Query Answering Using Multiple Layered Databases	Proc. 2nd Int'l Conf. Cooperative Info. Systems	May 1994	data mining, multiple layered database,
...
ftp://ftp.cs.colorado.edu/pub/cs/techreports/schwartz/FTP.Caching-PS	P.B.Danzig R.S.Hall M.F.Schwartz	A Case for Caching File Objects Inside Internet-works	Proc. SIGCOMM	Sept. 1993	caching, ftp,
...
http://sobolev.mit.edu/people/jphill/publications/shap.dvi	J.R.Phillips H.S.J. Zant	Influence of induced magnetic fields on Shapiro steps in Josephson junction arrays	Physical Review B 47	1994	magnetic fields, Josephson array, Shapiro step,
...

Table 1: A portion of *doc_brief* extracted from *document* at layer-1.

By extraction of only the documents related to *computing science*, a layer-3 relation *cs_doc* can be easily obtained. Also, performing attributed-oriented induction on *doc_brief* leads to another layer-3 relation *doc_summary*, a portion of which is shown in Table 2.

affiliation	field	pub_year	count	first_author_list	file_addr_list	...
Simon Fraser Univ.	Database Systems	1994	15	Han, Kameda, Luk,
Univ. of Colorado	Global Network Systems	1993	10	Danzig, Hall,
MIT	Electromagnetic Field	1993	53	Bernstein, Phillips,
...

Table 2: A portion of *doc_summary* extracted from *doc_brief* at layer-2.

Notice that backward pointers can be stored in certain entries, such as *first_author_list* and *file_addr_list*, in the *doc_summary* table, and a click on a first author or a file address will lead the presentation of the detailed corresponding entries stored in layer-2 or layer-1. \square

3.3 Distribution and maintenance of the global MLDB

3.3.1 Replication and distribution of the global MLDB

A global MLDB is constructed by extracting extra-layers from an existing (layer-0) global information base using generalization and transformation techniques. A higher layer database is usually much smaller than the lower layered database. However, since the layer-1 database is resulted from direct, detailed information extraction from the huge global information base, its size is still huge. It is unrealistic to have this layer replicated and distributed to other servers.

A possible implementation is to store each local layer-1 database at each local network server site, but replicate the higher layered databases, such as layer-2 and above, and propagate them to remote backbone and/or ordinary network servers. Load can be further partitioned between backbone and ordinary servers. For example, one may store a complete layer-2 database at the backbone site but only the relatively frequently referenced portions of layer-2 and/or higher layers at the corresponding sites. Also, specifically projected layers (e.g., medical database) can be stored at the closely relevant sites (e.g., hospitals and medical schools).

By doing so, most information browsing and brief query answering can be handled by searching within the local network. Only detailed requests will be forwarded to the backbone servers or further to the remote sites which store the information. Only when the full document is explicitly requested by a user (with the awareness of its size), will the full layer-0 document be sent across the network to the user site. This will substantially reduce the amount of data to be transmitted

across the network and improve the response time.

Moreover, some higher layered database could be defined by users for easy reference. For example, a user may define a new database at a high layer as “*all the documents related to heterogeneous databases published in major conferences or journals since 1990*”. An information manager cannot construct a new database for every user’s definition. Most such definitions will be treated like views, i.e., no physical databases will be created, and queries on such views will be answered by the query modification technique [10, 19]. Only if such a view is shared and frequently referenced, may it be worthwhile to create a new database for it.

Notice that the replication of higher-layered databases at network servers may take nontrivial disk space. However, the rapid progress of computer hardware technology has reduced the cost of disk space dramatically in the last decade. It could be more beneficial to trade disk space, as long as it is reasonable, for reduced network traffic and prompt query answering.

3.3.2 Incremental updating of the global MLDB

The global information base is dynamic, with information added, removed and updated constantly at different sites. It is very costly to reconstruct the whole MLDB database. Incremental updating could be the only reasonable approach to make the information updated and consistent in the global MLDB.

In response to the updates to the original information base, the corresponding layer-1 and higher layers should be updated incrementally. Incremental update can be performed on every update or at every night at the local site and propagate the updates to higher layers.

We only examine the incremental database update at insertion and update. Similar techniques can be easily extended to deletions. When a new file is connected to the network, a new tuple t is obtained by the layer-1 construction algorithm. The new tuple is inserted into a layer-1 relation R_1 . Then t should be generalized to t' according to the route map and be inserted into its corresponding higher layer. Such an insertion will be propagated to higher layers accordingly. However, if the generalized tuple t' is equivalent to an existing tuple in this layer, it needs only to increment the count of the existing tuple, and further propagations to higher layers will be confined to count increment as well. When a tuple in a relation is updated, one can check whether the change may affect any of its high layers. If not, do nothing. Otherwise, the algorithm will be similar to the deletion of an old tuple followed by the insertion of a new one.

4 Resource and Knowledge Discovery in the Global MLDB

4.1 Additional relational operators for information discovery in the global MLDB

Similar to other extended-relational database systems, a global MLDB system treats the requests for information browsing and resource discovery like relational queries. However, since concepts in a global MLDB are generalized at different layers, search conditions in a query may not match exactly the concept level of the currently inquired or available layer of the database. For example, to find papers related to a particular topic, such as “attribute-oriented induction”, a query may put this term as a search key. However, the current layer may only contain terms corresponding to a higher concept level, such as “induction techniques”, or “data mining methods”. In this case, it is unlikely to find in the current layer an exact match with the provided search key, but is likely to find a more general concept that absorbs the search key. On the other hand, a search key in a query may be at a more general concept level than those at the current layer. For example, a search key “sports”, though conceptually covers the term “baseball”, does not match it in the database.

Therefore, key-oriented search in an MLDB leads us to introduce several additional relational operations to extend the semantics of traditional selection and join.

Definition 4.1 In the global MLDB system, four additional relationships: *coverage*, *covered_by*, *synonym*, and *approximation*, are defined as follows.

1. coverage (\supset): A concept A covers another concept B , denoted as $A \supset B$, if A or A 's synonym is an ancestor of B or B 's synonym in the same concept hierarchy.
2. covered_by (\subset): A concept A is covered by another concept B , denoted as $A \subset B$, if A or A 's synonym is a descendant of B or B 's synonym in the same concept hierarchy.
3. synonym (\simeq): A concept A is a synonym of another concept B , denoted as $A \simeq B$, if A and B are in the same alias list in the same concept hierarchy.
4. approximation (\sim): A concept A is an approximate of another concept B , denoted as $A \sim B$, if A or A 's synonym is a sibling of B or B 's synonym in the same concept hierarchy. \square

Based on these relationships, additional selection and join operations can be defined as follows.

Definition 4.2 Let σ be a selection performed on the i -th attribute (column) of relation R using the selection constant c . Four addition selection operations are defined as follows,

1. **coverage-selection**: if the selection predicate is $c \supset \$i$, i.e., the selection operation is $\sigma_{c \supset \$i} R$,
2. **covered_by-selection**: if the selection predicate is $c \subset \$i$, i.e., the selection operation is $\sigma_{c \subset \$i} R$,
3. **synonym-selection**: if the selection predicate is $c \simeq \$i$, i.e., the selection operation is $\sigma_{c \simeq \$i} R$,
and
4. **approximation-selection**: if the selection predicate is $c \sim \$i$, i.e., the selection operation is $\sigma_{c \sim \$i} R$. □

Similarly, one can define four corresponding join operations in the global MLDB systems by replacing the query constant c in the selection predicate of the definition with the j -th column of a relation S .

4.2 A query language for information discovery in the global MLDB

With the construction of the global MLDB, a query language, **NetQL**, can be defined for resource and knowledge discovery using a syntax similar to the relational language SQL [10, 19]. Four newly introduced operators have their correspondent language primitives in NetQL, as shown in Table 3.

NetQL primitive	operation	Name of the operation
covers	\supset	coverage
coveredby	\subset	covered_by
like	\simeq	synonym
closeto	\sim	approximation

Table 3: New NetQL primitives for additional relational operations.

```

{ select | list | describe } { attributes_name_list | * }
from relation_list
[ related-to name_list ]
[ in location_list ]
where where_clause

```

Table 4: The top level syntax of NetQL.

The top-level NetQL query syntax is presented in Table 4. At the position for the keyword **select** in SQL, an alternative keyword **list** can be used when the search is to browse the summaries at a high

layer; **describe** can be used when the search is to discover and describe the general characteristics of the data; whereas **select** remains to be a keyword, indicating to find more detailed information. Two optional phrases, “**related-to** *name_list*” and “**in** *location_list*”, are introduced in NetQL for quickly locating the related subject fields and/or geographical regions (e.g., Canada, Europe, etc.). They are semantically equivalent to some phrases in the where-clause, such as “*keyword coveredby field_names*” and/or “*location coveredby geo_areas*”, etc. But their inclusion not only makes the query more readable but also helps system locate the corresponding layer-3 relation if there exists one. The where-clause is similar to that in SQL except that several new operators may be used.

Moreover, Mosaic-like [28] graphics user interfaces can be developed on top of NetQL, making browsing and searching “instruction-free”, i.e., mainly relying on field-filling and button-clicking operations. The interactive interface will facilitate users to browse the information base, quickly locate the resource, or progressively deepen the search or browse to lower information layers [12, 28].

The concrete NetQL examples will be presented and analyzed in the next two subsections.

4.3 Resource discovery in the global MLDB

Most search engines currently available on the Internet are keyword-driven, and the answers presented are a list of URL anchors related to the keywords. The global MLDB allows us to apprehend and solve the resource discovery issues in two different ways: (1) presenting a list of pointers to documents corresponding to the request, (2) allowing the user to progressively and interactively browse detailed information leading to a targeted set of documents.

The resource discovery led by direct addressing uses the relations in a high layer, and possibly those in the lower layers as well, when necessary, to find a list of URL addresses of the documents or objects corresponding to the criteria specified in the query. By clicking at an entry in the list, the user has the choice to either first access the detailed descriptors of the document stored in the layer-1 or directly fetch the layer-0 document.

On the other hand, the resource discovery led by progressively detailed information browsing suits the users who do not have a clear mind on what are the exact resources that they need. The system first presents the top-layer high-level view with selected statistics to a relatively vague, or preliminary query, and works interactively with the user to progressively focus the search and deepen the layer. Such a search adopts a top-down approach, takes advantages of the available concept hierarchies and multiple information layers, and allows users to either interactively adding more constraints, such as “located in British Columbia”, “published since 1994”, etc. to refine the query, or to focus at a subset of high-level answers by selecting appropriate tuples in the answer

list to go down to a lower layer for more detailed information. Finally, by clicking the entries in the last selected list, the detailed information can be down-loaded from layer-1 or the right documents can be fetched from the layer-0 information base.

Following are examples of queries for resource discovery.

Example 4.1 To find the documents related to data mining and written by Ted Thomas, a simple NetQL query is presented as follows.

```
select      *
from        document
related-to  computing science
where       “Ted Thomas” in authors and one of keywords like “data mining”
```

Notice that “select *” means to print all the important attributes in a relation at a high layer, and “from *document*” does not indicate to find the document relation at layer-0 or layer-1, but indicates to find the top-most layer of the *document* relation which fits the query. Therefore, “*document*” is a clue to the system to find the appropriate relation at a high layer. We adopt this convention since it is system’s responsibility to find the best match and it is unreasonable to ask users to remember all the relation names at different layers. Moreover, “‘*Ted Thomas*’ in *authors*” means that ‘*Ted Thomas*’ is in the set of *authors*; whereas “one of *keywords* like ‘*data mining*’” means that there exists an entry in the set *keywords* which is a synonym of ‘*data mining*’.

To execute this query, the MLDB system uses the phrase “from *document*” and “related-to *computing science*” to locate the top layer relation *cs_doc* (which has the same schema as *doc_brief* but contains only cs-related documents, see Example 3.1) and then use the two search keys in the where-clause to locate a set of URL addresses of the required documents, together with the brief descriptions: *authors*, *title*, *publication*, *publication date*, *keywords*, etc. User may have the choice to either find more detailed layer-1 descriptions or layer-0 documents by clicking at different buttons. □

Example 4.2 The query, *list the documents in North America, and related to “database recovery”*, is presented as follows.

```
list      *
from      document
in        north_america
where     one of keywords close-to “database recovery”
```

Notice that the phrase, “one of *keywords* close-to ‘*database recovery*’”, implies that one of the technical terms is *database systems*, a subfield of “*computing science*” (by consulting the *concept hierarchy*). Thus a smart system will still locate the top-layer as *cs_doc*. Moreover, the phrase “in *north_america*” confines the search to be within *North America* which will be mapped into concrete countries using concept hierarchy. The keyword list indicates that the query is to briefly browse the information, and therefore, it searches *cs_doc* using the where-clause as a constraint. Since close-to implies to find those documents containing a keyword located in the same hierarchy but as siblings (and synonyms) of the concept “*database recovery*” in *cs_doc*. The topics like “*concurrency control*”, “*transaction management*”, will also be in the candidate list. In this case, a relatively large set of answers will be returned. An interactive process to deepen the search will usually be initiated by users after browsing the answer set. □

4.4 Information browsing and knowledge discovery in the global MLDB

A major weakness of the current global information system services is their difficulty at supporting efficient and effective *information browsing* operations. The global MLDB is a meta-data based relational database and this architecture allows us to submit queries about the meta-data. In a global MLDB, the higher layer database stores the abstract or summary data and statistics of the global information base, and information browsing can be easily performed by searching through the higher layered databases.

Requesting and looking over meta-data (the higher layered database or a query-relevant portion of it) itself is one kind of information browsing. Notice that such information browsing may lead to resource discovery as shown in Example 4.2. However, another application of information browsing, which could be of its major purpose, is to visualize the information about the global information base and the artifacts it includes. This does not necessarily mean to find physical pointers on the internet or the documents themselves but it may indicate to find interesting high level information about the global information base.

The following examples show that much more can be done using the MLDB beyond simple “*browsing* → *resource discovery*” search pattern.

Example 4.3 Let the query be: “*based on the document information on the Internet, list the universities in Europe which are productive in 1990s on database-related research*”.

Suppose that the inquired information is covered by the information stored in the document summary relation and the user is aware of the schema of that relation. If the concept level for the

constants in the query is not higher than those stored in the *doc_summary* relation, it is easy to work out a query on the summary relation as follows.

```
select    affiliation
from      doc_summary
in        Europe
where     affiliation belong_to "university" and field = "database systems"
          and publication_year > 1990 and count > 20
```

In this query, “productive” is measured as those containing more than 20 published papers since 1990 related to database systems on the Internet, which should either be obtained based on the result of an initial browsing of the *doc_summary* table, or justified by some interactive queries. The query can be answered using a simple NetQL query on the generalized (meta)-data relation because the generalized relation *doc_summary* covers the concepts provided in the query. □

However, one can be hardly so lucky to rely fully on the generalized relation. If a query constant contains more detailed information (i.e., a concept at a level lower than those provided in the generalized relation), generalization must be performed on the fly, using a knowledge discovery query.

Example 4.4 Suppose the query is to “*describe the general characteristics in relevance to authors’ affiliations, publications, etc. for those documents which are popular on the Internet and are on “data mining”.*”

Let the concept level of the attribute *field* in the generalized relation *doc_summary* correspond to the keywords like “*database systems*”, “*programming languages*”, etc. (as shown in Example 3.1), which is higher than “*data mining*”. The provided generalized relation, at the concept level like “*database systems*” does not offer much help because a document on “*database systems*” may not be on “*data mining*” but that on “*data mining*” is surely on “*database systems*”.

Therefore, generalization must be performed on a relation containing lower level concepts, such as *doc_brief*, or *cs_doc*, using a knowledge discovery query, characterized by the keyword “describe” as shown below.

```
describe  authors.affiliation, publication, publication_date
from      document
related to Computing Science
where     one of keywords like "data mining" and access_frequency = "high"
```

Notice that `access_frequency` in the `doc_brief` or `cs_doc` contains concrete numbers, such as 150 (since January 1994), and “*high*” is an element in a small concept hierarchy defining the high level concepts for `access_frequency`.

The discovery query will be first executed as a retrieval to collect from `cs_doc` the data which are relevant to “*authors.affiliation, publication, publication_date*” and satisfy the where-clause. Then the attribute-oriented induction is performed on the collected data, which generalizes “*publication*” into groups, such as major AI journals, major database conferences, and so on, and generalizes publication date to year, etc.

The generalized results are collected in a relation table, which can be presented in several alternative forms. One is to present information about individual attributes. For example, it may present a statement like “*30% of such documents are in several universities in North America, including George Mason U., Rutgers U., UCLA, U. of Wisconsin, SFU, . . . , 40% are in several industry research labs, including GTE Labs, IBM Almaden Research Center, AT&T, . . . , and so on*”. It may also present another statement in relevance to publication date, like “*20% were published in 1980s, 30% in 1990-1992, 25% in 1993, etc.*”. Another alternative is to present information in relevance to more than one attribute, such as “*For IBM Almaden Research Center, 70% were published in major database conferences between 1990-1993, etc.*”. This presentation mechanism has been implemented in the DBLearn system in our previous work [14, 16]. □

These examples, though simple, demonstrates that a broad spectrum of knowledge discovery queries can be posed to the global MLDB system to explore some interesting, general or statistical information about the global information base.

5 Discussion

The multiple layered database architecture provides the following advantages for information discovery in global information systems.

1. **Application of database technology:** The MLDB architecture transforms an unstructured global information base into a structured, global database, which makes the database technology (not just storage management and indexing techniques) applicable to resource management, information retrieval, and knowledge discovery in the global information network.
2. **High-level, declarative interfaces and views:** The architecture provides a high-level, declarative query interface on which various kinds of graphics user-interfaces can be constructed for

browsing, retrieval, and discovery of resource and knowledge. Moreover, multiple views can be defined by different users or user communities, cross-resource linkages can be constructed at different layers, and resource search can be initiated flexibly.

3. **Performance enhancement:** The layered architecture makes most searches confined to local or less remote sites on relatively small and structured databases, which will reduce the network bandwidth consumption, substantially enhance the search efficiency, and lead to relatively precise locating of resources and quick response of user's requests.
4. **A global view of database contents:** By preprocessing and generalizing primitive data, a global MLDB system may transform semantically heterogeneous, primitive level information into more homogeneous, high-level data at a high layer. It may provide a global view of the current contents in a database with summary statistics, which will assist users to browse database contents, pose progressively refined queries, and perform knowledge discovery in databases. Users could even be satisfied with the general or abstract data at a high layer and not bother to spend time and network bandwidth for more details.
5. **Intelligent query answering and database browsing:** A user may not always know the exact need when searching in the global information base. In the global MLDB system, a query is treated like an information probe, being mapped to a relatively high concept layer and answered in a hierarchical manner. This will provide with users a high-level view of the database, statistical information relevant to the answer set, and other associative and summary information at different layers.
6. **Information resource management:** Incremental updating can be performed on different layers using efficient algorithms, as discussed in Section 3. With the MLDB architecture, it is relatively easy to manage the global MLDB and make it consistent and up-to-date. For example, it is easy to locate weakly-consistent replicas [6] based on their property similarity at higher layers (rather than searching through the whole global information base!). Based on accessing statistics, one can also decide whether a duplicate should be removed or be preserved for resource redirection.

However, it is also important to note that cost should be paid for the construction of such a global MLDB, as presented below.

1. **Extra disk spaces:** Extra disk spaces are needed to store and replicate multiple layers and concept hierarchies. With the low cost of computer disks and hardwares, this seems not to

be a bottleneck. Division of labor among different nodes and trade-offs between disk space and network bandwidth should also be considered in the construction of local or backbone MLDBs.

2. **DBMS softwares:** A subset of the functionalities of a database system, including storage management, indexing, query processing and recovery, should be considered as essential for construction and maintenance of MLDB and query processing in the global MLDB. Moreover, preliminary data mining facilities are also essential for the success of resource and knowledge discovery in the global MLDB. With the fast expansion of relational DBMS market, such preliminary database facilities (which are suggested to be specially designed, developed and distributed for the development of global information systems) should be affordable financially in the near future for most backbone and distributed servers.
3. **New softwares for layer construction and query processing:** Softwares should be developed for the construction and maintenance of the global MLDB, especially the extraction of different kinds of information from the global information base, and the implementation of query processing with additional relational operations introduced here. Some existing global information indexing and servicing softwares can be improved and adapted to the construction and use of MLDBs.
4. **Reasonable standardization:** Similar to the library catalog standardization, a classification standard for the documents in the global information base may need to be introduced and enforced to help reduce errors and enhance the quality of service in the development of the global MLDB.

6 Conclusions

Different from the existing global information system services, a new approach, called *multiple layered database (MLDB) approach*, has been proposed and investigated for resource and knowledge discovery in global information systems. The approach is to construct progressively a global multiple layered database by generalization and transformation of lower layered data, store and manage multiple layered information by database technology, and perform resource and knowledge discovery by query transformation, query processing and data mining techniques.

The major strength of the MLDB approach is its promotion of a tight integration of database and data mining technologies with resource and knowledge discovery in global information systems.

With the dynamically growing, highly unstructured, globally distributed and huge information base, the application of the mature database technology and promising data mining techniques could be an important direction to enhance the power and performance of global information systems.

Our study shows that the global MLDB can be constructed automatically and updated incrementally by integration of information retrieval, data analysis and data mining techniques, information at all of the nonprimitive layers can be managed by database technology, and resource and knowledge discovery can be performed efficiently and effectively in such a multiple layered database.

Our study presents a general framework of the MLDB approach for the resource and knowledge discovery in the global information system. More studies are needed in the construction and utilization of the global multiple layered databases. We are currently developing softwares and performing experiments for automatic construction of the global MLDB on top of the global information base and for discovery of resource and knowledge in such a MLDB. Modifications and refinements to our initial design are expected, along with the progress of the research and developments. The effectiveness of the approach will be tested in the environment of the global information network, and further investigation and experimentation will be reported in the future.

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