

## Disclosure Control of Sensitive Information from OLAP Query Results

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**Abstract**— With the rapid development in technology organizations are able collect and store huge amount of business relevant information. Today the companies wish to convert such huge data into quality information. Organizations are looking forward for decision support system where the data collected from heterogeneous sources can be converted into digitized form and stored in a permanent storage space for business analysis purpose. Data warehouse provides the architecture to serve as a single integrated source of data which is different from operational database and stores current and historical records coalesced from multiple transactional systems. OLAP operation on Data warehouse database allows the client to analyze the data interactively and perform other business intelligence functions. The data in the data warehouse is collected from various sources like insurance companies, healthcare systems also [1] contains sensitive information, OLAP operation on such databases may reveal the information which is private to an individual. In this paper we discussed about the privacy violation of sensitive information by malicious user with valid OLAP queries. We proposed a framework to identify the privacy violation of sensitive information during OLAP operation and modify the query results with suitable privacy preservation technique.

**Keywords**- OLAP, privacy, sensitive information, technology.

### I. INTRODUCTION

The development in software, hardware, and computing and network technology facilitated different organizations to collect and store huge amount of business related information. The organizations realized that the growth of an organization and benefits can only be achieved by sharing the information from other organizations. Today Organizations come forward to store huge information in a single repository and analyze it to improve the service or performance, scientific analysis, research application, customer retention management. The data warehouse is a single integrated source which combines various databases across an entire enterprise. The multidimensional database structure is [3] used to store the data in the data warehouse.

The data in the data warehouse represents the business history of an organization. This historical data is used for analysis that supports business decisions at many levels of abstraction, from strategic planning to performance evaluation of a discrete organizational unit. Data in a data warehouse is organized to support analysis with OLAP rather than to process real-time transactions as in online transaction processing systems (OLTP). Multidimensional cube representation is used to store the data in the data warehouse. A subset of attributes in the database is used to construct a data cube. Few attributes are selected as measure attributes i.e. the attributes whose values are of interest and are aggregated according to the dimensions. Other attributes are selected as dimensions or functional attributes. Figure 1 below depicts 3-D cube model representing sales data having three dimensions i.e. year, product and stores location. Total sales are the measurement of interest. Figure 1 below shows the multidimensional model of data warehouse.

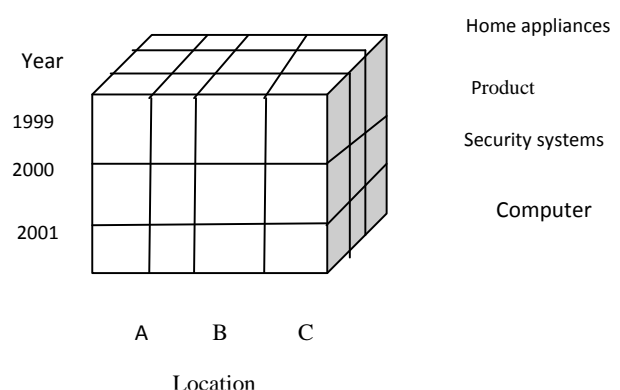


Figure 1. Multidimensional model of data warehouse

OLAP operations are used to perform complex operations on data warehouse data and detect novel patterns and relationships between data items which went unnoticed earlier. OLAP operation also provides a multidimensional

presentation of data warehouse data, creating cubes that organize and summarize the data for efficient and quick analytical operation. OLAP technique facilitates the user to extract and view the organizational database from different aspects. For example a user can submit a request to analyze and show and compare all the companies who sold computer security products in America in the month of January with those for the similar type of products in the month of June. Various OLAP operations are rollup, drill-down, cube, pivot, slice and dice.

Multidimensional view of aggregate data is utilized by OLAP to provide quick access to strategic information for further analysis. The information ranges from basic navigation through slice and dice operation to more complex modeling. OLAP operations have the ability to provide information as and when for effective decision-making.

Generally the output of an OLAP is displayed in the form of a matrix. The rows and column represents the dimensions of the query. The analyst can understand the meaning of organizational data base contained in the data warehouse using multi-dimensional analysis. OLAP operations allow the analyst to traverse through the database through different operations like roll up, drill down, slice, dice and pivot.

## II. PRIVACY ISSUES IN OLAP OPERATION

Over the past few years the significance of data warehouses and OLAP operation for an organization’s decision support system has rapidly grown. At the same time the problem with information privacy violation has been also developed. A data warehouse stores current and historical records consolidated from multiple transactional systems like health care system, banking, and insurance companies also contains sensitive information. The main aim of data warehouse is to provide access to all necessary data without any restriction on data access. But, as portion of the data is very sensitive and private to an individual, OLAP operation may reveal the pattern containing the [6] sensitive information creating security conflict. The revealed sensitive information can be misused without the knowledge of the actual data owner. The misuse of sensitive information may harm the data owner.

One straightforward solution to protect the privacy is to completely hide the sensitive information in the cell of multidimensional database or such data should not be included in the database. But such kind of suppression of information will not provide the valid results for analysis. There is immense need to provide the solution which can give results for analysis purpose but also protect privacy of

an individual. In the proposed framework we investigate the privacy breaches caused by multi-dimensional queries [7] in OLAP (Online Analytic Processing) systems. We propose a novel technique to preserve the privacy of sensitive information during OLAP operation and give valid results for analysis purpose.

## II. PROPOSED FRAMEWORK

The growth in technology has enabled the organizations to collect huge volume of information. Today most of the organizations wish to store the data and also share the data for mutual benefit and growth of organization. Many organizations depend on OLAP results to extract novel patterns which can help them to reduce costs, increase business expansion opportunities.

The proposed framework shown in figure 2 is very useful in a collaborative environment where group of parties want to share the data but restrict themselves from revealing the private information [8] about their clients or customers. This framework helps the parties to share the information but also protect the privacy of sensitive information.

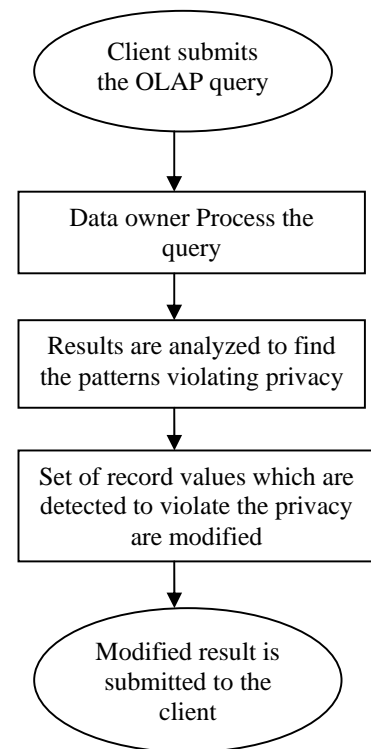


Figure 2. The framework to preserve the privacy during execution of OLAP queries.

### III. EXPERIMENTAL ANALYSIS

In this experimental work a standard health database of patient information contains nearly 1000 records. The different attributes Patient\_id, name, disease, and gender, and age, type of disease, location, bill amount and category.

Patient ID	Name	Gender	Age	Disease	Severity	Location	Bill Amount
981	patient981	Atopic dermatitis	FEMALE	30	NORMAL	Bhubaneswar	10000
982	patient982	Nasal polyps	MALE	35	SEVER	Moradabad	8500
983	patient983	Pilonitis arthris	FEMALE	40	NORMAL	Coraltpur	27000
984	patient984	Ulcerasia	MALE	45	SEVER	Riapur	3500
985	patient985	Glucosme	FEMALE	50	NORMAL	Bhivandi	4000
986	patient986	Asthma	MALE	55	SEVER	Kochi (Cochin)	4500
987	patient987	Throat cancer	FEMALE	5	NORMAL	Jamshedpur	2000
988	patient988	Prostate cancer	MALE	13	SEVER	Bhali	7000
989	patient989	Diabetes	FEMALE	23	NORMAL	Amravati	15000
990	patient990	Conductivites	MALE	33	SEVER	Bilaner	2500
991	patient991	Complicated UTI	FEMALE	48	NORMAL	Warangal	2000
992	patient992	Enteric fever	MALE	53	SEVER	Mir Bhatiyander	3500
993	patient993	Asthma	FEMALE	10	NORMAL	Guntur	4600
994	patient994	Migraine	MALE	15	SEVER	Bhavnagar	8000
995	patient995	Typhoid	FEMALE	20	NORMAL	Durgapur	8500
996	patient996	Altitude sickness	MALE	25	SEVER	Asansol	6500
997	patient997	Leprosy	FEMALE	30	NORMAL	Kohapur	5000
998	patient998	bladder cancer	MALE	35	SEVER	Ajmer	7000
999	patient999	Burns	FEMALE	40	NORMAL	Sahasrapur	4000
1000	patient1000	ENT infection	MALE	45	SEVER	Dehra Dun	8000
1001	2001	rajesh DENGUE	MALE	14	FEVER	WARANGAL	1000
1002	2002	SPRIVAS HYPERSENSITIVITY	MALE	33	NORMAL	WARANGAL	54500
1003	2003	RAMADEVI ENT INFECTION	FEMALE	20	NORMAL	KARIMNAGAR	500
1004	2004	DEVI ZANDIS	FEMALE	21	SEVER	HANAMKONDA	50000

Figure 3. The partial patient information database

The database administrator has the access to all information. But patient\_id and name are not revealed for analysis purpose. The remaining attributes are used to perform various OLAP operations. We considered location as one of the sensitive attribute which can reveal the identity of an individual. The disease and bill amount are considered as the major attribute values to perform analysis. We performed cube operation on the given database. The cube operation gives the summarized information for a given query. From the output result it is found that in some cases there is only record from a particular location.

After getting such result having single record value, an intruder can easily [5] generate other queries and extract other information of that particular person. The proposed framework modifies the location information with a default value and then the modified results are submitted to the client for analysis purpose. Other attributes values like disease, bill amount for particular disease, age with

modified place value can be displayed for analysis purpose. Figure 3 shows the partial database considered for experimental work. Figure 4 below shows the output of cube operation showing the special record with count equal to 1.

CATEGORY	DISEASE	LOCATION	COUNT	SUM(BILLAMT)
			57	12530500
		Jana	6	1350000
		Akola	6	1350000
		Bihar	6	1335000
		Sager	6	1272000
		Belgaum	6	1272000
		Amritsar	7	1484000
		Allahabad	7	1575000
		Mangalore	6	1335000
		Visakhapatnam	7	1557500
	Gout		7	1575000
	Gout	Allahabad	7	1575000
	Hives		7	1557500
	Hives	Visakhapatnam	7	1557500
	Tonic		12	2622000
	Tonic	Jana	6	1350000
	Tonic	Belgaum	6	1272000
	Leprosy		12	2670000
	Leprosy	Bihar	6	1335000
	Leprosy	Mangalore	6	1335000
	Haemorrhage		7	1484000
	Haemorrhage	Amritsar	7	1484000
	Brain cancer		6	1272000
	Brain cancer	Sager	6	1272000
	Tripolardisease		6	1350000
	Tripolardisease	Akola	6	1350000
			20	4751000
		Etawah	6	1500000
		Mysore	7	1750000
		Thirissur	6	1500000
		WARANGAL	1	1000
	DENGUE		1	1000
	DENGUE	WARANGAL	1	1000
	Nausea		6	1500000
	Nausea	Etawah	6	1500000
	Sunburn		7	1750000
	Sunburn	Mysore	7	1750000
	Complicated UTI		6	1500000
	Complicated UTI	Thirissur	6	1500000
		HANAMKONDA	1	50000
	ZANDIS		1	50000
	ZANDIS	HANAMKONDA	1	50000

Figure 4. Showing the output result of cube operation

In figure 4 it is found that the disease DENGUE has only record in the city Warangal. An intruder after getting this result can generate other combination of queries which can identify the identity of actual person.

Hence, after obtaining the OLAP query result, we identify of the combination of output which can violate the privacy of an individual. Fig. 5 below shows portion of the results where two separate records are found with count value 1. because name and patient id is unique it is completely suppressed. But as the place information when linked with other information which is not sensitive may reveal the identity of an individual, so the actual sensitive attribute location value is replaced with some default value related to the same hierarchy of location. Similarly, the age can be modified using additive perturbation. The additive perturbation adds small amount of noise [4] drawn from Gaussian distribution to the values under attribute Age. Finally the modified result is submitted to the client.

Disease	Gender	Age	Type of disease	Location	bill amount
Dengue	Male	14	Severe	ASDF	1000
hypersensitivity	Male	33	normal	ASDF	54500

Figure 5 The two specific records having count equal to 1

Disease	Gender	Age	Type of disease	Location	bill amount
Dengue	Male	16	Severe	INDIA	1000
hypersensitivity	Male	31	normal	INDIA	54500

Figure 6. The modified result to be submitted to the client for analysis

The ability to of an intruder to identify an individual will be difficult, because values under attribute location have been changed to next hierarchy. It will be complex to identify an individual at state level than at city level. Further, privacy is enhanced by modifying the age values. The modified results do not deviate from the desired summarized information and is valid for analysis purpose.

## VI. CONCLUSION

On-Line Analytical Processing operation on data warehouse information enables analysts, knowledge workers to view their own organizational database from different point of view. As the decision support system integrated with data warehouse is utilized by number of users or groups, the possibility of retrieving personal information and re-identifying an individual increases. Hence, inspite of numerous advantages, OLAP operations [9] faces the privacy problem from an opponent who can deduce the private from OLAP query answers. Such type of privacy violation is more dangerous in case of medical, patient history, financial data bases which consists not only the statistical but also confidential information about an individual which should be known only to the privileged persons and relevant organization. Privacy can be protected by query restriction and by modifying the aggregate results

before releasing it for analysis purpose. This paper proposes a framework which analyzes the OLAP query results, and identifies the record attribute values which one can discover the sensitive information stored in the cell and identify and individual's information. This model modifies the summarized information so that one cannot infer the private information stored in the cell.

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