

Digital Dreams: Privacy Preserving And Cyber-Security Access Control Mechanism For Sensitive Information Using Hypercube

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ABSTRACT

Data privacy refers to the concern of how to preserve the information in data cells. Privacy preserving deals with hiding the sensitive information without sacrifice the usability of data. Every organization today have become well aware of the privacy intrusions of their sensitive data and not want an unauthorized user to access all the information stored in data cube. The major area of concern is that non-sensitive data even may deliver sensitive information, including personal information, facts or patterns. Several techniques of privacy preserving data mining have been proposed in literature. In this paper, we have building a multi-viewing cube of data for analyzing, and applying various access control mechanism for privacy preserving of factual data, and to secure against the unauthorized access to sensitive information

KEYWORDS: - Data-mining, Privacy preserving, Sensitive data, Data-cube.

INTRODUCTION

Data mining is the process used to analyze large amount of data and gather useful information from them. It extracts the information from large heterogeneous databases in many different dimensions and finally summarizes it into categories and relations of data [1]. The primary goal of privacy preserving is to hide the sensitive data before it gets published. Organizations collect and analyze data to improve their services. Access Control Mechanisms is used to ensure that only authorized information is available to users. However, sensitive information, even after the removal of identifying attributes, is still susceptible to linking attacks by the unauthorized users. The privacy is achieved at the cost of accuracy and imprecision is introduced in the authorized information under an access control policy.

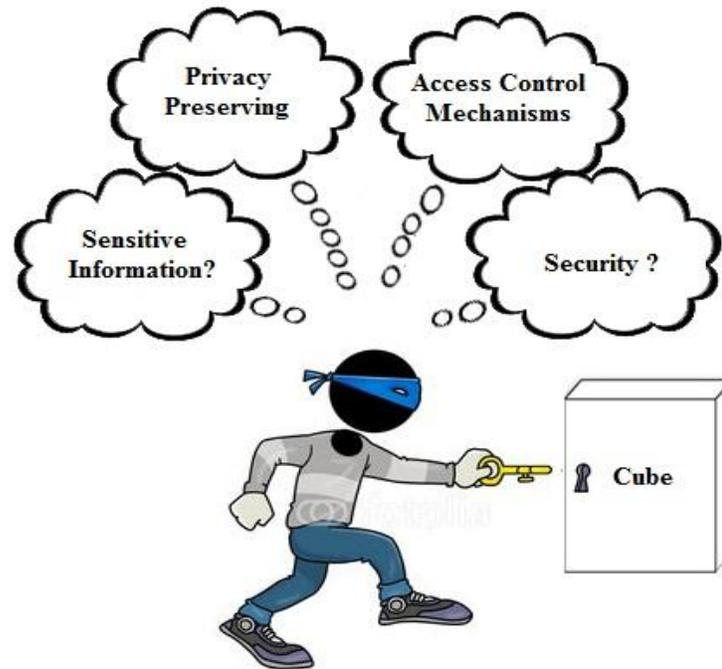


Figure 1: - Privacy Preservations and leaks of Sensitive data.

OBJECTIVES OF THIS RESEARCH STUDY

This proposed study has been focused on building a multidimensional cube and securing privacy preserving mechanism of sensitive data against unauthorized access control. Specifically, the following research objectives have been defined for the proposed study:

1. To perform demographic analysis of a dataset and identifying various dimensions, levels of dimension and building a multi-dimensional cube based on basic elements required for the structure of multidimensional cube.
2. To avail sensitive information and provide protection and preservation to sensitive information against unauthorized access.

In this paper, we discuss different approaches and techniques in the field of Privacy Preserving Data Mining. The paper is organized as. First, we give the basic concept of data mining and privacy. Second, we describe Privacy Preserving data mining with its framework. Then, define different techniques of privacy preservation. And finally, we describe conclusions.

PRIVACY PRESERVING DATA MINING

Privacy preserving has originated as an important concern with reference to the success of the data mining. Privacy preserving data mining deals with protecting the privacy of individual data or sensitive knowledge without sacrificing the usefulness of the data [2].

LITERATURE SURVEY

The review of literature based on various research articles and papers upon multi-dimensional modeling under privacy preserving mechanisms. The researcher arranges all the based studies into chronological order for granularity level refinements of research area

PRIVACY PRESERVING DATA MINING TECHNIQUES

In this, we focus on the different techniques which are developed like cryptographic techniques, data perturbation, blocking based, etc.

(a) Cryptographic Technique- *Cryptography* is a technique which sensitive data can be encrypted available in the databases. It is better technique to preserve the data. Cryptographic technique introduced because it provides security and safety of sensitive attributes. The cryptographic technique includes various encryption and decryption algorithms in order to secure the data so that no individual or no computer program can recognize that data [3]. The encryption is mostly usage in this technique and its result is more secure when data sent to the other party. But, this technique is no longer reliable when there are large numbers of parties sharing the data [4]. Also, these algorithms are unusable where millions of data are generated frequently. It can break the individual's privacy and can lead to attacks.

(b) Data Perturbation - Data Perturbation is a technique for modifying data using random process. This technique sensitive data is changed using by adding, subtracting or any other mathematical formula [5] [6]. Data perturbation technique can handle different-different data types: character type, Boolean type and integer type. In discrete data it is required to preprocess the original data set [5]. The preprocessing of data is classified into attribute coding and obtaining sets coded data set. The method of average data is used here. Discrete formula introduced [7] is: $A(\max) - A(\min)/n = \text{length}$. A is continuous attribute, n is number of discrete, and length is the length of the discrete interval. The technique does not reconstruct the original data values, it only reconstructs the distribution

(c)Blocking based technique -In blocking based technique, a sensitive classification rule which is used for hiding sensitive information from others. In this technique, there are two steps which are used for preserving privacy. First is to identify transactions of sensitive rule and second is to replace the known data values to the unknown data values [8] [9]. In this technique, there is scanning of database and identifying the transactions sensitive rule. And then for each transaction, this algorithm replaces the sensitive data with unknown data values. In this hide the actual value, they replace '1' by '0' or '0' by '1' or with any unknown values in a specific transaction [8]. The replacement of these values does not depend on any specific rule. The objective of this technique is to preserve the sensitive data from unauthorized access. The algorithm replaces unknown values in the place of attribute for every transaction which supports that sensitive rule. These steps will continue till all the sensitive attributes are hidden by the unknown values.

IMPLEMENTATION

This phase describes the implementation of proposed work, how the work has been done and various problems occur during the implementation phase. In this proposed study, we build a cube and secure against the unauthorized access to sensitive information. The diagram defines the complete overview of the study which is going to be accomplished.

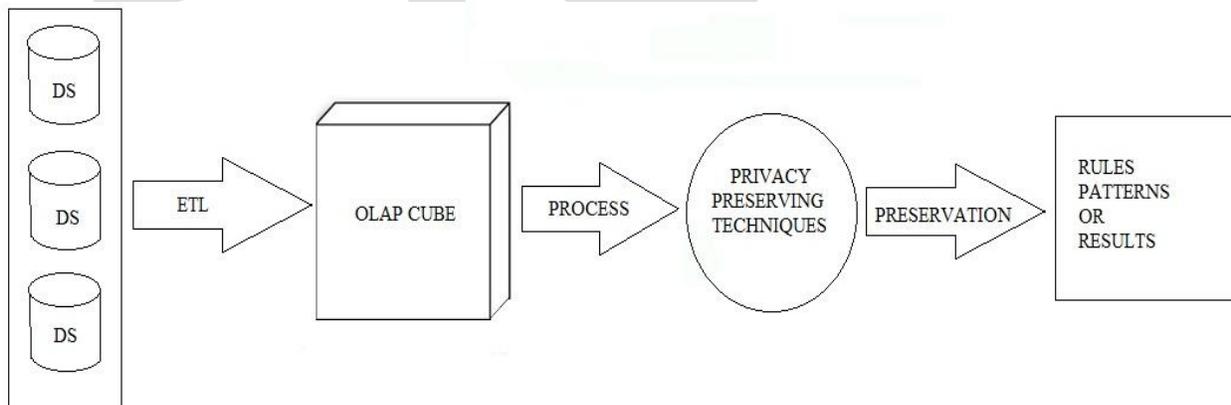


Fig. 2: -Context diagram of privacy preserving data mining

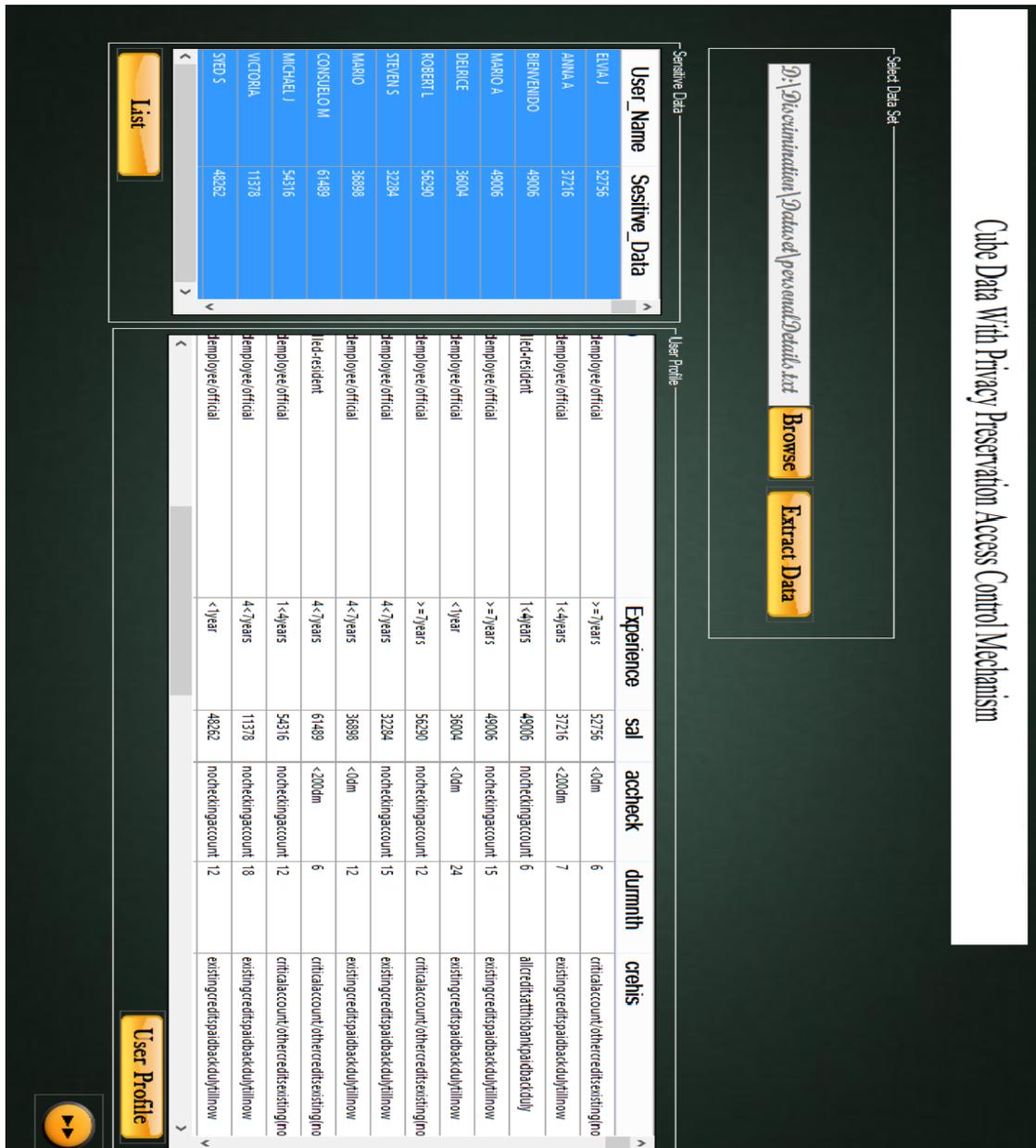
In this framework for privacy preserving Data Mining is shown. Data from different data are collected and are preprocessed using ETL tools. This transformed and clean data from Level 1 is stored in the data cube. Data in data cube is used for mining. In level 2 privacy preservation

techniques are used to protect data from unauthorized access. Sensitive data of an individual can be prevented from being misused. Basically, this paper is focused on applying various privacy preserving mechanisms to protect sensitive data from the unauthorized access control because a single data item in a data cube is not likely to be accessed alone, but a number of data are often aggregated to give summarized information and the trends of the database [10].

APPLYING ANONYMIZATION TECHNIQUE

Anonymization technique use suppression and generalization of records to satisfy privacy requirements with minimal distortion of micro data. The privacy is achieved at the cost of accuracy and imprecision is introduced into the authorized information under an access control policy. The concept of imprecision is bound for permission to define a threshold on the amount of imprecision that can be tolerated. The anonymity technique is used by an access control mechanism to ensure both security and privacy of the sensitive information. Data access control is based on the notion of privileges- the authorization to perform a particular operation. Privileges are required to gain access to information in the data warehouse. However, it is well known that data access control is insufficient in controlling information leakage because information is not released directly by manipulating legitimate queries about aggregated information.

TESTING AND RESULTS



Screenshot 1: - Obtaining Sensitive information from dataset.

In this point of practical implementation, identified the sensitive information from the whole dataset of employees of an organisation.

Here for the privacy preserving point of view, anonymization is applied on the basis of the range value of the threshold to the perspective view of salary details of the employees of the organization.

Cube Data With Privacy Preservation Access Control Mechanism

Load Dataset

Experience	sal	accheck	dur
>=7years	52756	<0m	6
1<4years	32716	<200m	7
1<4years	49006	nocheckingaco...	6
>=7years	49006	nocheckingaco...	15
<1year	38004	<0m	24
>=7years	56290	nocheckingaco...	12
4<7years	32294	nocheckingaco...	15
4<7years	39898	<0m	12
4<7years	61409	<200m	6
1<4years	54316	nocheckingaco...	12
4<7years	11378	nocheckingaco...	18
<1year	48262	nocheckingaco...	12
1<4years	49585	<0m	18
<1year	51853	nocheckingaco...	9
>=7years	50362	nocheckingaco...	7
1<4years	18143	nocheckingaco...	14
4<7years	59818	nocheckingaco...	5
1<4years	49006	<0m	11
4<7years	29956	nocheckingaco...	15

Anonymized Data

Experience	sal	accheck	dumth
>=7years	> 50K	<0m	6
1<4years	<= 50K	<200m	7
1<4years	<= 50K	nocheckingaco...	6
>=7years	> 50K	nocheckingaco...	15
<1year	<= 50K	<0m	24
>=7years	<= 50K	nocheckingaco...	12
4<7years	<= 50K	nocheckingaco...	15
4<7years	> 50K	<0m	12
4<7years	> 50K	<200m	6
1<4years	<= 50K	nocheckingaco...	12
4<7years	<= 50K	nocheckingaco...	18
<1year	<= 50K	nocheckingaco...	12
1<4years	<= 50K	<0m	18
<1year	> 50K	nocheckingaco...	9
>=7years	> 50K	nocheckingaco...	7
1<4years	<= 50K	nocheckingaco...	14
4<7years	<= 50K	nocheckingaco...	5
1<4years	> 50K	<0m	11
4<7years	<= 50K	nocheckingaco...	15

Load Dataset

Anonymized to Salary

Screenshot 2: - Applying anonymization based on threshold range.

Cube Data With Privacy Preservation Access Control Mechanism

The screenshot displays a software interface with a dark background and a white title bar. The title bar contains the text "Cube Data With Privacy Preservation Access Control Mechanism". Below the title bar, there are four data tables, each with a header "Anonymized Data". Each table has four columns: "Gender", "Education", "Foreigner_or", and "Salary". The data rows in the tables are as follows:

Gender	Education	Foreigner_or	Salary
male	11th	Yes	> 50K
male	bachelors	No	> 50K
male	masters	No	> 50K
female	bachelors	No	> 50K
female	hs-grad	No	> 50K
male	bachelors	No	> 50K
male	some-coll...	No	> 50K
female	bachelors	No	> 50K
male	hs-grad	No	> 50K
female	some-coll...	No	> 50K

Gender	Education	Foreigner_or	Salary
male	hs-grad	No	<= 50K
male	preschool	No	<= 50K
female	some-coll...	Yes	<= 50K
female	hs-grad	No	<= 50K
male	9th	No	<= 50K
female	hs-grad	No	<= 50K
male	bachelors	Yes	<= 50K
female	some-coll...	No	<= 50K
male	bachelors	No	<= 50K
male	1st-4th	No	<= 50K

late_N Age	Gender	Education	Salary
J	Old	11th	> 50K
A	Old	bachelors	> 50K
	Young	masters	> 50K
JEL...	Old	bachelors	> 50K
ER L	Young	hs-grad	> 50K
	Old	bachelors	> 50K
W	Old	some-coll...	> 50K

Gender	Education	Foreigner_or	Salary
male	11th	Yes	> 50K
male	bachelors	No	> 50K
male	masters	No	> 50K
female	bachelors	No	> 50K
female	hs-grad	No	> 50K
male	bachelors	No	> 50K
male	some-coll...	No	> 50K
female	bachelors	No	> 50K
male	hs-grad	No	> 50K
female	some-coll...	No	> 50K

At the bottom of the interface, there are four yellow buttons with black text:

- Anonymizing Data in Min Sal
- Anonymizing Data in Max Sal
- Country Min Data Anonymized
- Country Max Data Anonymize

Screenshot 3: - Multiview perspective view of sensitive information.

Finally, multi-perspective view of sensitive information by anonymizing data from different levels as min-salary, max-salary and use the min and max section from the perspective view of residential address.

CONCLUSION

This present study provides a better solution to protecting sensitive data in OLAP data cubes from unauthorized accesses. In this study a privacy-preserving access control framework for cube data has been proposed. This framework is a combination of privacy protection mechanisms and access control. The access control mechanism allows only authorized query predicates on sensitive data. The privacy preserving module anonymizes the data to meet privacy requirements and imprecision constraints on predicates set by the access control mechanism. The result is secure that both privacy preserving and unauthorized accesses are eliminated.

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