

# DATA LEAKAGE DETECTION

## ABSTRACT

A data distributor has given sensitive data to a set of supposedly trusted agents (third parties). Some of the data is leaked and found in an unauthorized place (e.g., on the web or somebody's laptop). The distributor must assess the likelihood that the leaked data came from one or more agents, as opposed to having been independently gathered by other means. We propose data allocation strategies (across the agents) that improve the probability of identifying leakages. These methods do not rely on alterations of the released data (e.g., watermarks). In some cases we can also inject "realistic but fake" data records to further improve our chances of detecting leakage and identifying the guilty party.

## OBJECTIVE

- A data distributor has given sensitive data to a set of supposedly trusted agents (third parties).
- Some of the data is leaked and found in an unauthorized place (e.g., on the web or somebody's laptop).
- The distributor must assess the likelihood that the leaked data came from one or more agents, as opposed to having been independently gathered by other means.
- We propose data allocation strategies (across the agents) that improve the probability of identifying leakages.
- These methods do not rely on alterations of the released data (e.g., watermarks). In some cases we can also inject “realistic but fake” data records to further improve our chances of detecting leakage and identifying the guilty party.
- Our goal is to detect when the distributor's sensitive data has been leaked by agents, and if possible to identify the agent that leaked the data.

## EXISTING SYSTEM

- We consider applications where the original sensitive data cannot be perturbed.  
  
Perturbation is a very useful technique where the data is modified and made “less sensitive” before being handed to agents.
- However, in some cases it is important not to alter the original distributor’s data.
- Traditionally, leakage detection is handled by watermarking, e.g., a unique code is embedded in each distributed copy.
- If that copy is later discovered in the hands of an unauthorized party, the leaker can be identified.
- Watermarks can be very useful in some cases, but again, involve some modification of the original data.
- Furthermore, watermarks can sometimes be destroyed if the data recipient is malicious.

## PROPOSED SYSTEM

- After giving a set of objects to agents, the distributor discovers some of those same objects in an unauthorized place.
- At this point the distributor can assess the likelihood that the leaked data came from one or more agents, as opposed to having been independently gathered by other means.
- If the distributor sees “enough evidence” that an agent leaked data, he may stop doing business with him, or may initiate legal proceedings.
- In this project we develop a model for assessing the “guilt” of agents.
- We also present algorithms for distributing objects to agents, in a way that improves our chances of identifying a leaker.
- Finally, we also consider the option of adding “fake” objects to the distributed set. Such objects do not correspond to real entities but appear.
- If it turns out an agent was given one or more fake objects that were leaked, then the distributor can be more confident that agent was guilty.

### **Hardware Requirements**

- **SYSTEM : Pentium IV 2.4 GHz**
- **HARD DISK : 40 GB**
- **FLOPPY DRIVE : 1.44 MB**
- **MONITOR : 15 VGA colour**
- **MOUSE : Logitech.**
- **RAM : 256 MB**
- **KEYBOARD : 110 keys enhanced.**

### **Software Requirements**

- **Operating system :- Windows XP Professional**
- **Front End :- Microsoft Visual Studio .Net 2005**
- **Coding Language :- C#**
- **Database :- SQL SERVER 2000**

## **MODULE DESCRIPTION:**

### **1) Login / Registration:**

This is a module mainly designed to provide the authority to a user in order to access the other modules of the project. Here a user can have the accessibility authority after the registration.

### **2) DATA TRANSFER:**

This module is mainly designed to transfer data from distributor to agents. The same module can also be used for illegal data transfer from authorized to agents to other agents

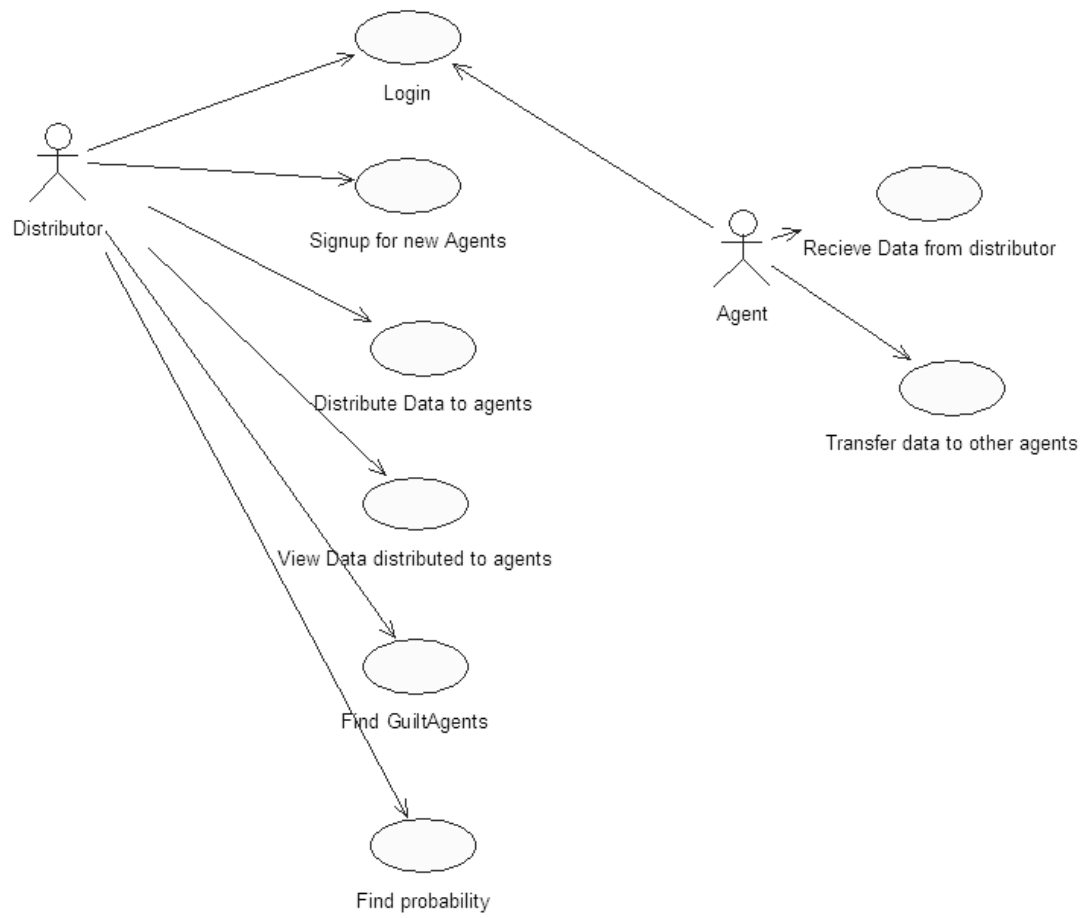
### **3) GUILT MODEL ANALYSIS:**

This module is designed using the agent – guilt model. Here a count value(also called as fake objects) are incremented for any transfer of data occurrence when agent transfers data. Fake objects are stored in database.

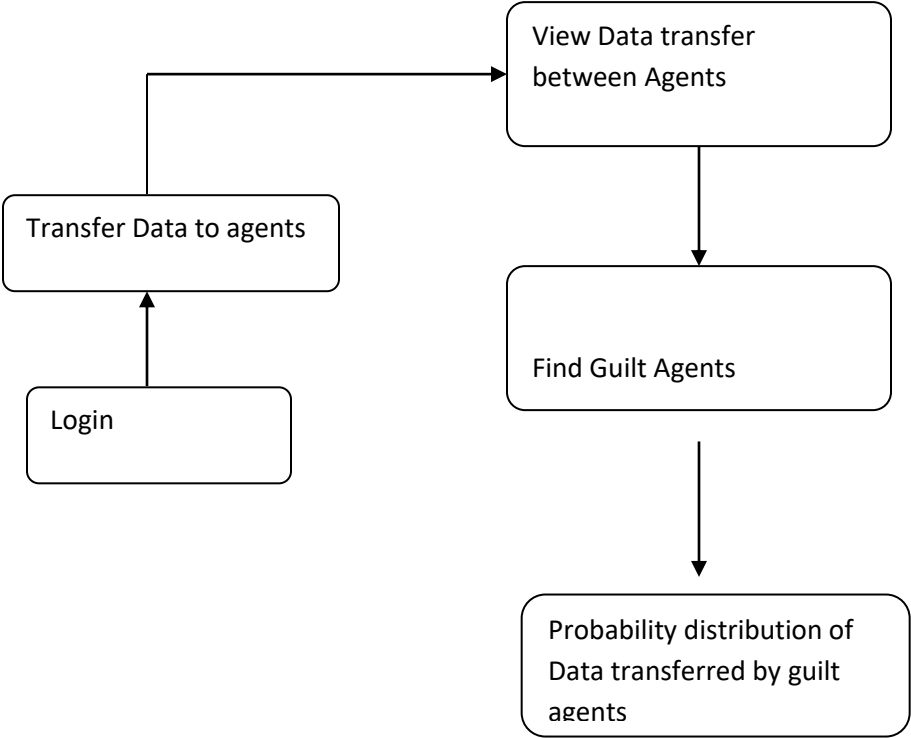
### **4)AGENT-GUILT MODEL:**

This module is mainly designed for determining fake agents. This module uses fake objects (which is stored in database from guilt model module) and determines the guilt agent along with the probability. A graph is used to plot the probability distribution of data which is leaked by fake agents

## USE CASE DIAGRAM:

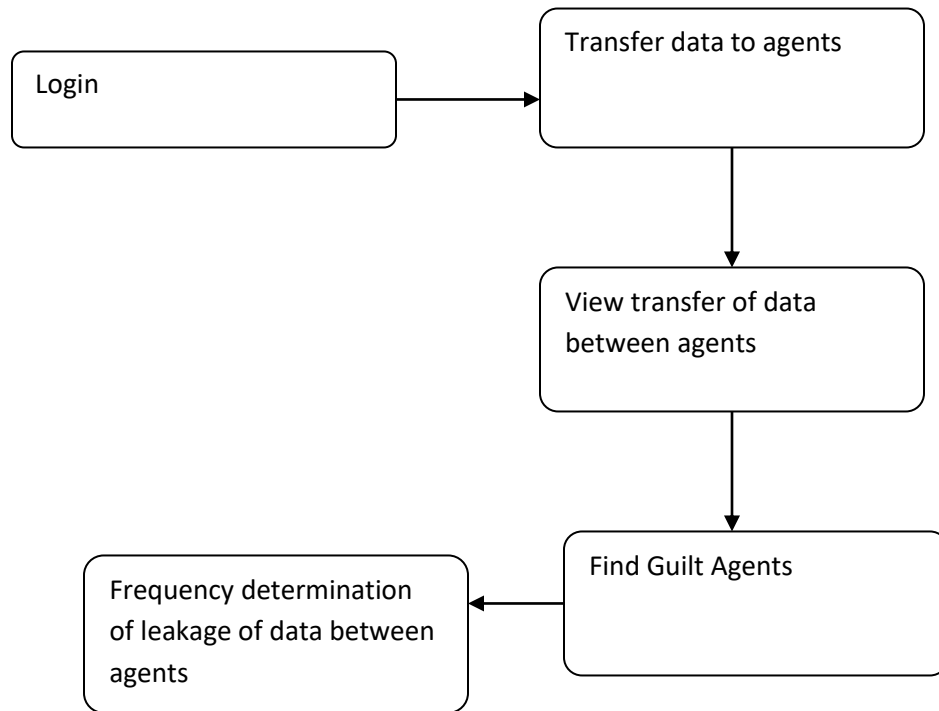


**MODULE DIAGRAM:**

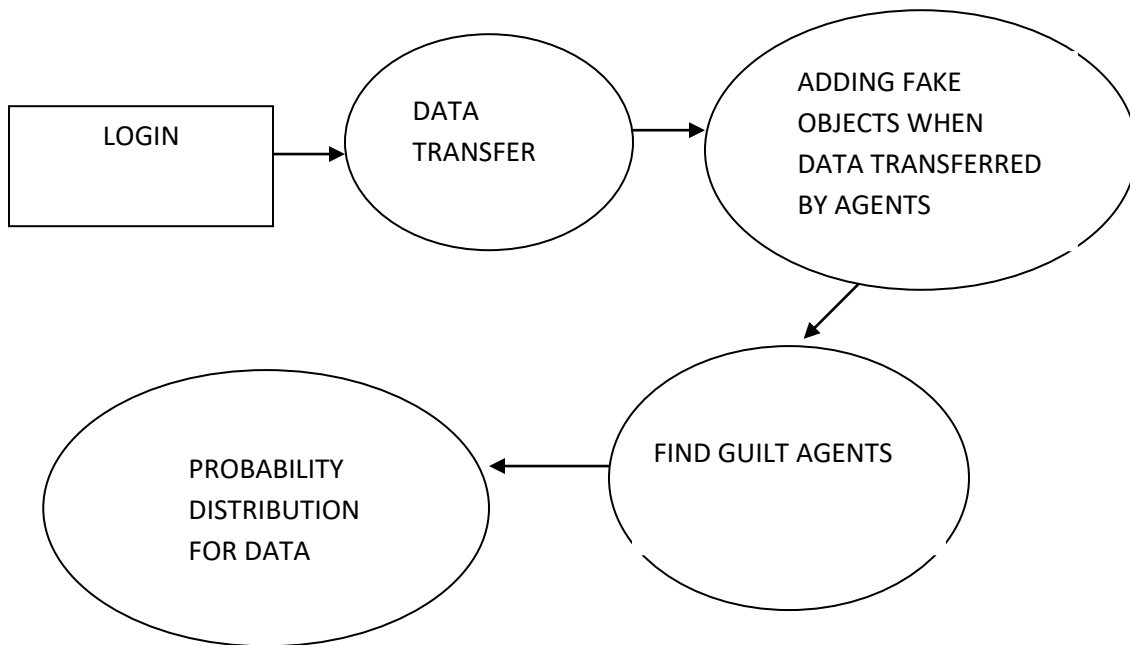




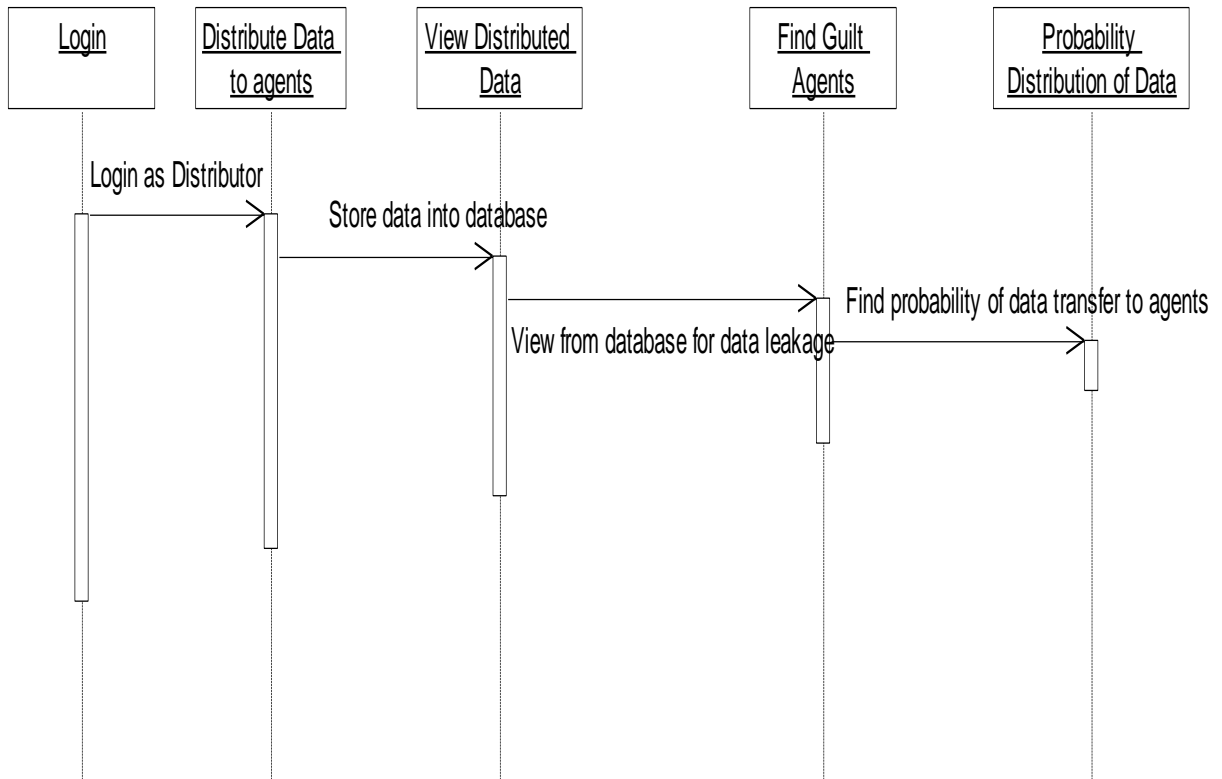
## Object Diagram:



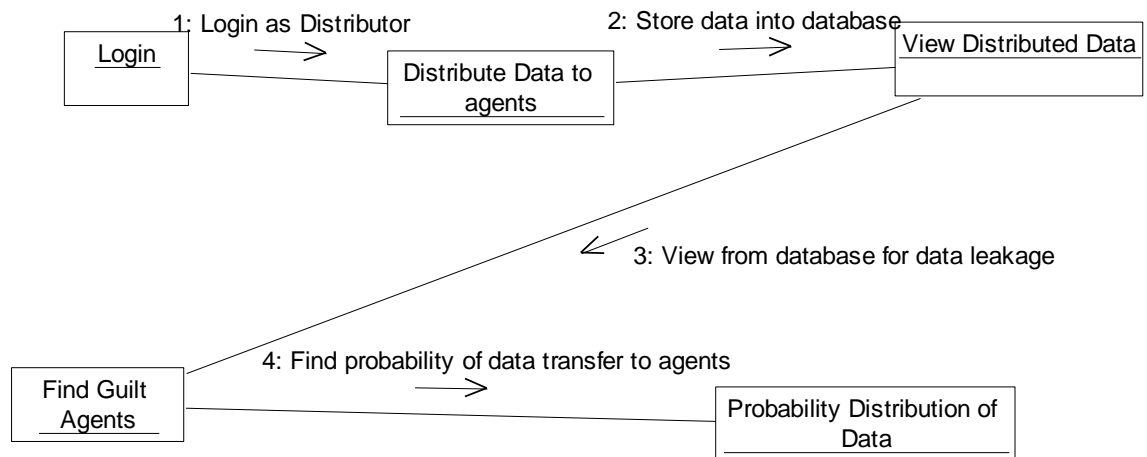
**Project Flow Diagram:**



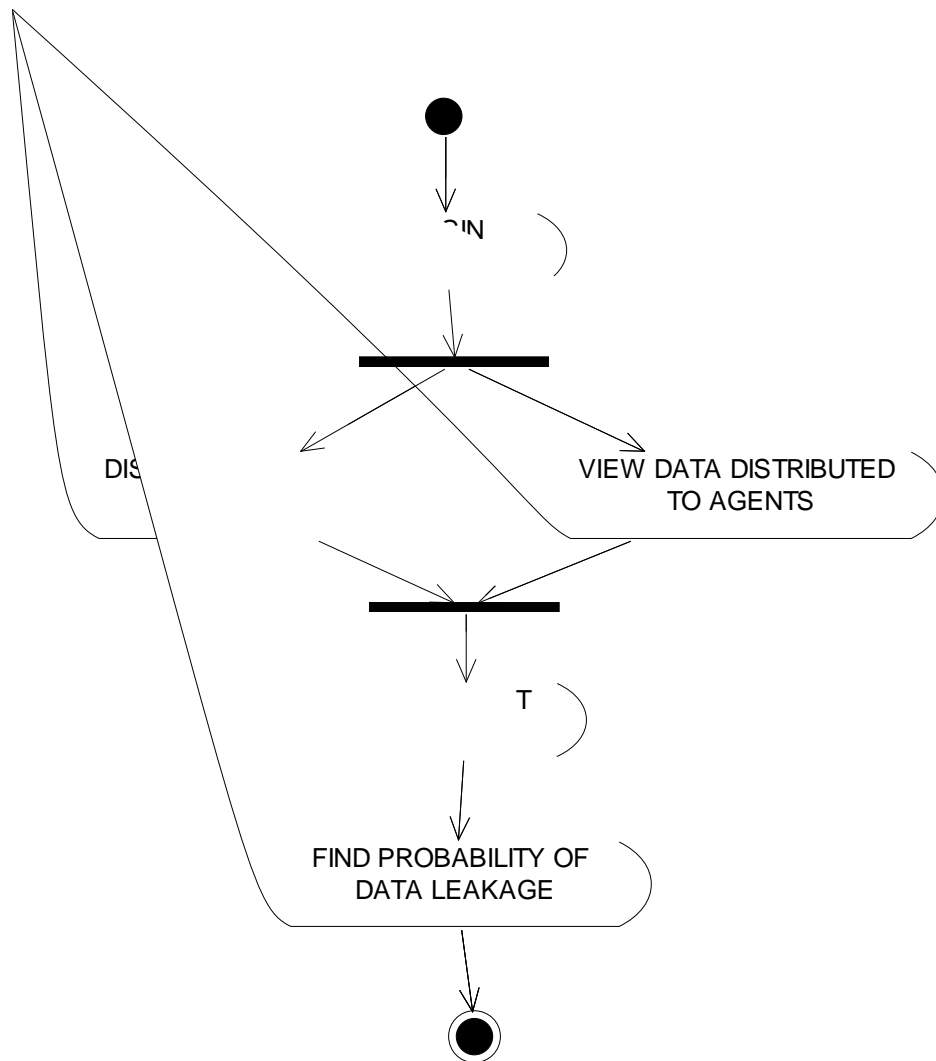
## SEQUENCE DIAGRAM:



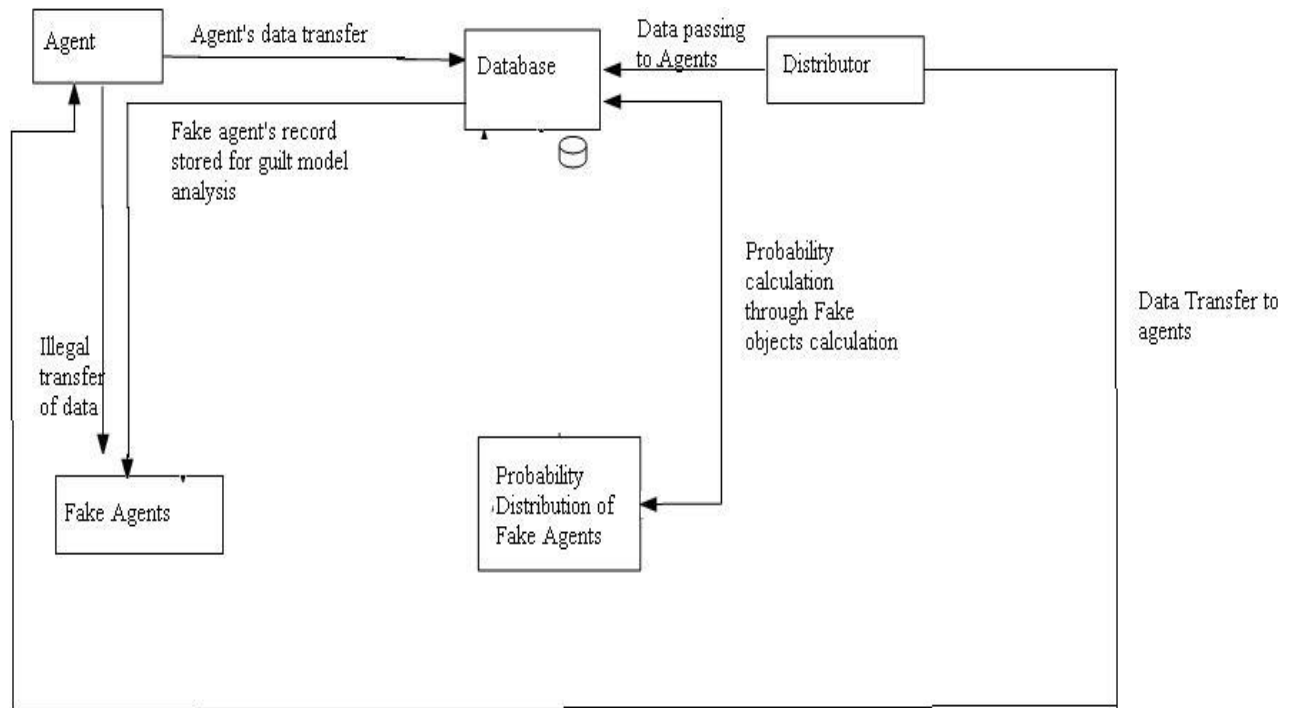
## COLLABORATION DIAGRAM:



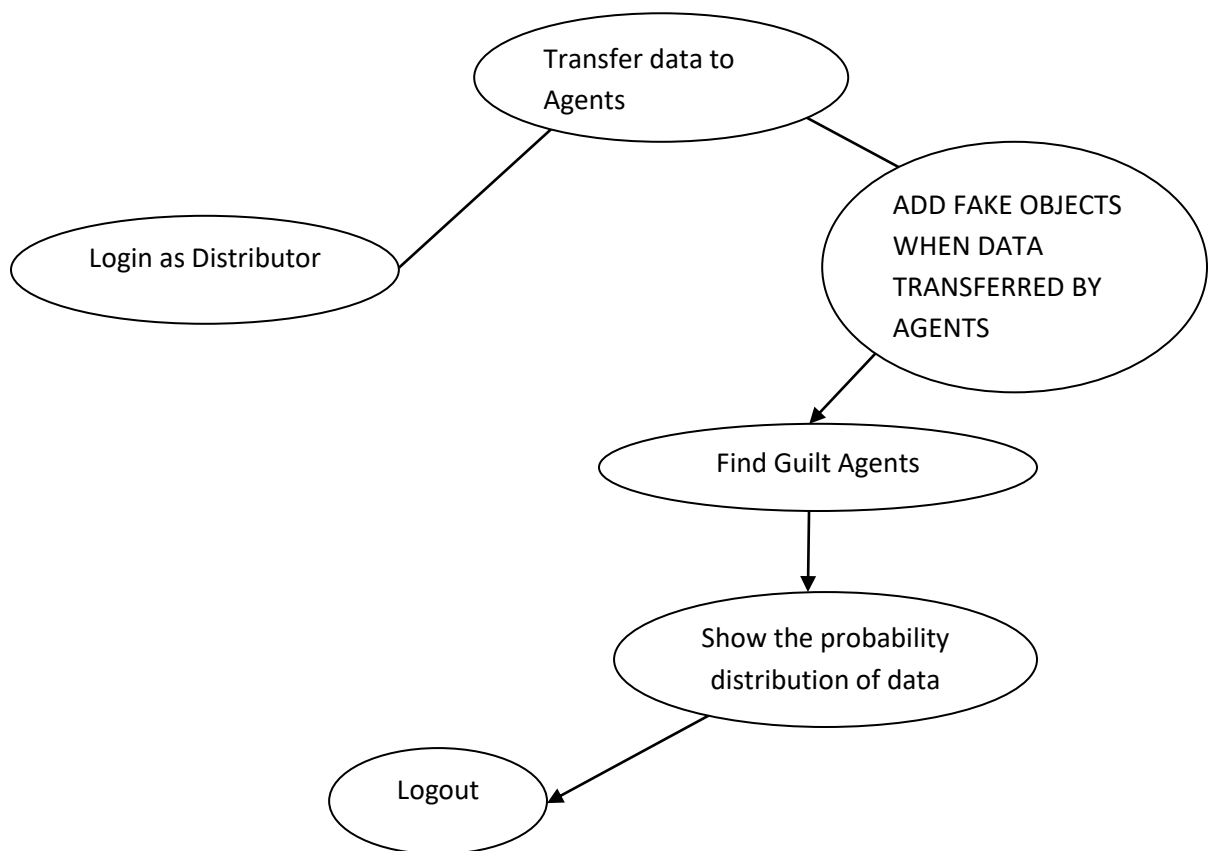
**ACTIVITY DIAGRAM:**



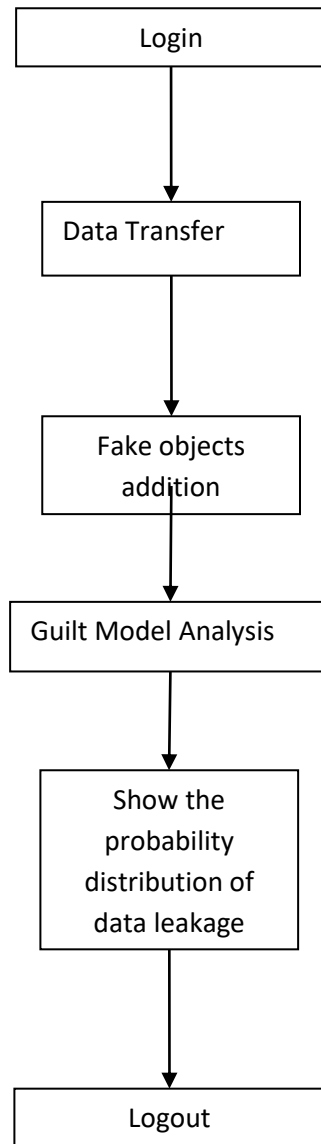
## ARCHITECTURE DIAGRAM:



## E-R DIAGRAM:



## DATA FLOW DIAGRAM:



CONCLUSION

The likelihood that an agent is responsible for a leak is assessed, based on the overlap of his data with the leaked data and the data of other agents, and based on the probability that objects can be “guessed” by other means. The algorithms we have presented implement a variety of data distribution strategies that can improve the distributor’s chances of identifying a leaker. We have shown that distributing objects judiciously can make a significant difference in identifying guilty agents, especially in cases where there is large overlap in the data that agents must receive.

## REFERENCES



- [1] R. Agrawal and J. Kiernan. Watermarking relational databases. In VLDB '02: Proceedings of the 28th international conference on Very Large Data Bases, pages 155–166. VLDB Endowment, 2002.
- [2] P. Bonatti, S. D. C. di Vimercati, and P. Samarati. An algebra for composing access control policies. *ACM Trans. Inf. Syst. Secur.*, 5(1):1–35, 2002.
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- [5] Y. Cui and J. Widom. Lineage tracing for general data warehouse transformations. In *The VLDB Journal*, pages 471–480, 2001.

## KEY TERMS

### Data Leakage:

A data breach is the unintentional release of [secure information](#) to an untrusted environment.

## Data Privacy:

Information privacy, or data privacy is the relationship between collection and dissemination of [data](#), [technology](#), the public [expectation of privacy](#), and the [legal](#) and [political](#) issues surrounding them.

[Privacy](#) concerns exist wherever [personally identifiable information](#) is collected and stored - in digital form or otherwise. Improper or non-existent [disclosure](#) control can be the root cause for privacy issues.

## Fake records:

Records which are false or containing misleading appearance.

## Unobstrusive Techniques:

Unobtrusive technique is a technique of data collection. They describe methodologies which do not involve direct elicitation of data from the research subjects. The unobtrusive approach often seeks unusual data sources.

## WHY USE DATAMINING

Data mining is the process of extracting patterns from [data](#). Data mining is becoming an increasingly important tool to transform the data into information. It is commonly used in a wide range of [profiling practices](#), such as [marketing](#), [surveillance](#), [fraud](#) detection and scientific discovery.

Data mining can be used to uncover patterns in data but is often carried out only on samples of data. The mining process will be ineffective if the samples are not a good representation of the larger body of data. Data mining cannot discover patterns that may be present in the larger body of data if those patterns are not present in the sample being "mined". Inability to find patterns may become a cause for some disputes between customers and service providers. Therefore data mining is not foolproof but may be useful if sufficiently representative data samples are collected. The discovery of a particular pattern in a particular set of data does not necessarily mean that a pattern is found elsewhere in the larger data from which that sample was drawn. An important part of the process is the [verification and validation](#) of patterns on other samples of data