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Intrusion Detection for Grid and Cloud Computing

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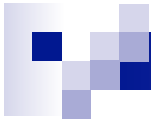
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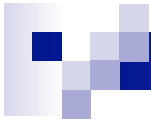
Introduction

- Providing security in a **distributed system** requires more than user authentication with passwords or digital certificates and confidentiality in data transmission. The Grid and Cloud Computing Intrusion Detection System integrates **knowledge** and **behavior** analysis to detect intrusions.
- Because of their distributed nature, grid and cloud computing environments are easy targets for intruders looking for possible vulnerabilities to exploit.
- To combat attackers, intrusion-detection systems can offer additional security measures.



Introduction

- IDS (intrusion-detection systems) must monitor each node and, when an **attack occurs**, **alert other nodes** in the environment.
- This kind of communication requires **compatibility between heterogeneous hosts**, **various communication mechanisms**, and **permission control over system maintenance and updates**—typical features in grid and cloud environments.
- Cloud middleware usually provides these features, so we propose an IDS service offered at the **middleware layer**.

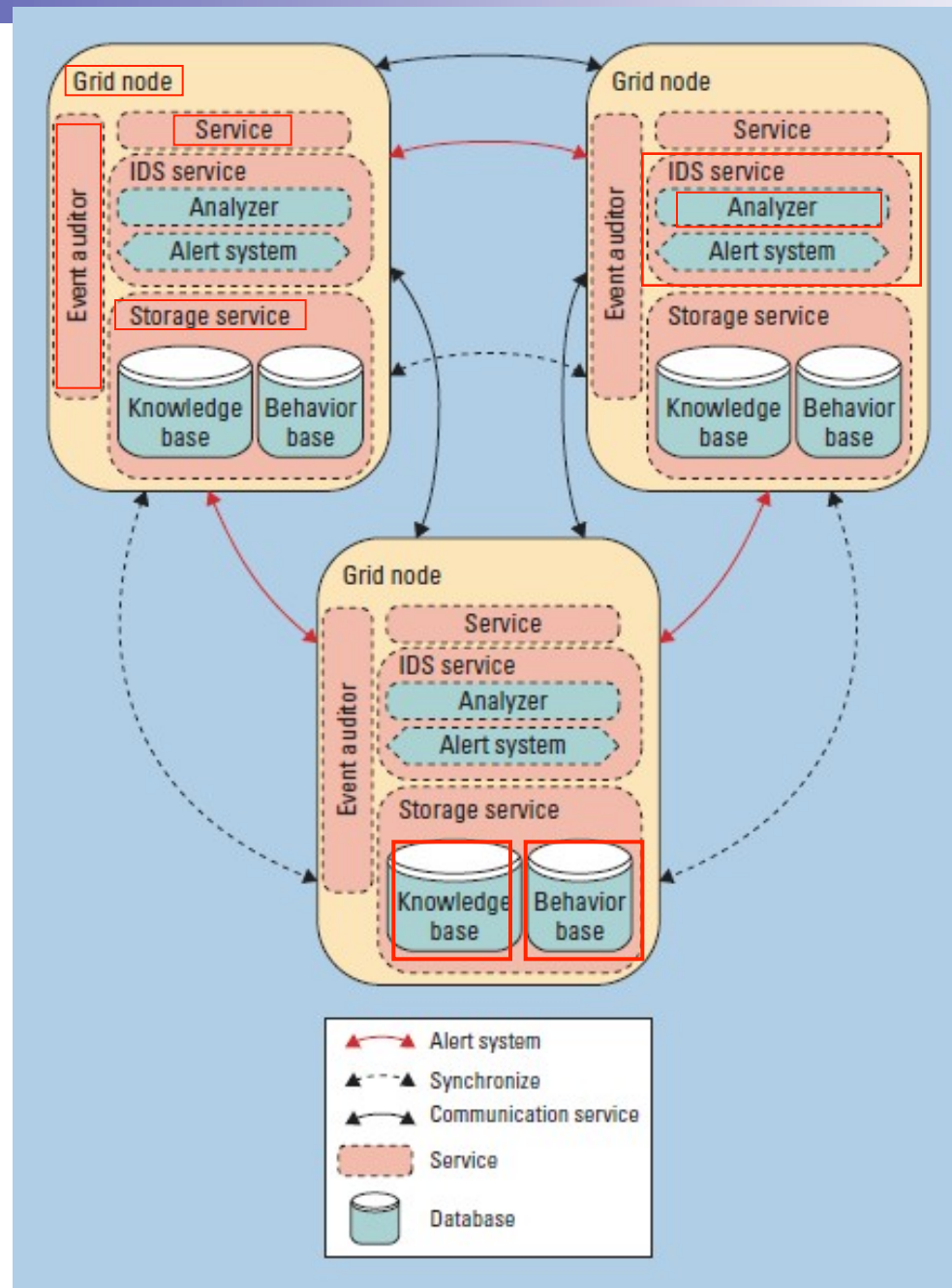


Introduction

- An attack against a cloud computing system can be **silent**, because cloud-specific attacks don't necessarily leave traces in a **node's operating system**.
- In this way, traditional IDSs can't appropriately identify suspicious activities in a grid and cloud environment.
- We propose the Grid and Cloud Computing Intrusion Detection System (GCCIDS), which has an **audit system** designed to cover attacks.

Figure 1

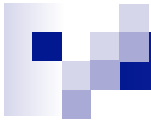
- The architecture of grid and cloud computing intrusion detection. Each node identifies local events that could represent security **violations** and sends an **alert to the other nodes**.



Out Proposed Service

- Figure 1 depicts the **sharing** of information between the IDS service and the other elements participating in the architecture: the **node**, **service**, **event auditor**, and **storage service**.
 - Node : **resources**, which are accessed homogeneously through the middleware.
 - Service : provides its **functionality** in the environment through the middleware, which facilitates communication.
 - Event Auditor : is the key piece in the system. It **captures data** from various sources, such as the log system, service, and node messages.
 - Storage Service : holds the data that the IDS service must analyze. It's important for **all nodes to have access to the same data**.

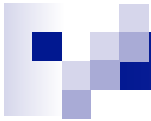




IDS Service

- The IDS service increases a cloud's security level by applying two methods of intrusion detection.
- The **behavior-based** method dictates how to **compare recent user actions** to the usual behavior.
- The **knowledge-based** method **detects known trails left by attacks** or certain sequences of actions from a user who might represent an attack.

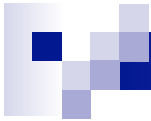




IDS Service - Analyzer

- The analyzer uses a profile **history database** to determine the distance between a **typical user behavior** and the **suspect behavior** and **communicates this to the IDS service**.
- With these responses, the IDS calculates the probability that the **action represents an attack** and **alerts the other nodes** if the probability is sufficiently **high**.





Behavior Analysis

- Numerous methods exist for behavior-based intrusion detection, such as **data mining**, **artificial neural networks**, and **artificial immunological systems**.
- We use a feed-forward **artificial neural network**, because this type of network can **quickly process information**, has **self-learning capabilities**, and can **tolerate small behavior deviations**. These features help overcome some IDS limitations.





Behavior Analysis

- Using this method, we need to recognize **expected behavior** (legitimate use) or a **severe behavior deviation**.
- For a given intrusion sample set, the **network learns** to identify the intrusions using its **retropropagation algorithm**.
- However, we focus on identifying user behavioral patterns and **deviations** from such patterns.
- With this strategy, we can cover a wider range of **unknown attacks**.

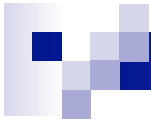




Knowledge Analysis

- Knowledge-based intrusion detection is the most often applied technique in the field because it results in a **low false-alarm rate and high positive rates**, although it **can't detect unknown attack** patterns.
- Using an **expert system**, we can describe a **malicious behavior with a rule**. One advantage of using this kind of intrusion detection is that we can **add new rules without modifying existing ones**.
- In contrast, behavior-based analysis is performed on learned behavior that **can't be modified without losing the previous learning**.

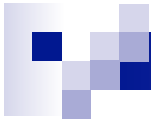




Increasing Attack Coverage

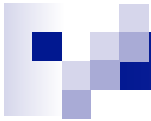
- The two intrusion detection techniques are distinct.
- The knowledge-based intrusion detection is characterized by a **high hit rate of known attacks**, but it's **deficient** in detecting **new attacks**. We therefore complemented it with the behavior based technique.
- The **volume of data** in a cloud computing environment can be **high**, so administrators don't observe each user's actions—they **observe only alerts from the IDS**.





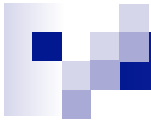
Results

- We developed a prototype to evaluate the proposed architecture using **Grid-M**, a middleware of our research group developed at the Federal University of Santa Catarina.
- We prepared three types of simulation data to test.
 - First, we created data representing **legitimate action** by executing a set of known services simulating a regular behavior.
 - Then, we created data representing **behavior anomalies**.
 - Finally, we created data representing **policy violation**.



Evaluating the Event Auditor

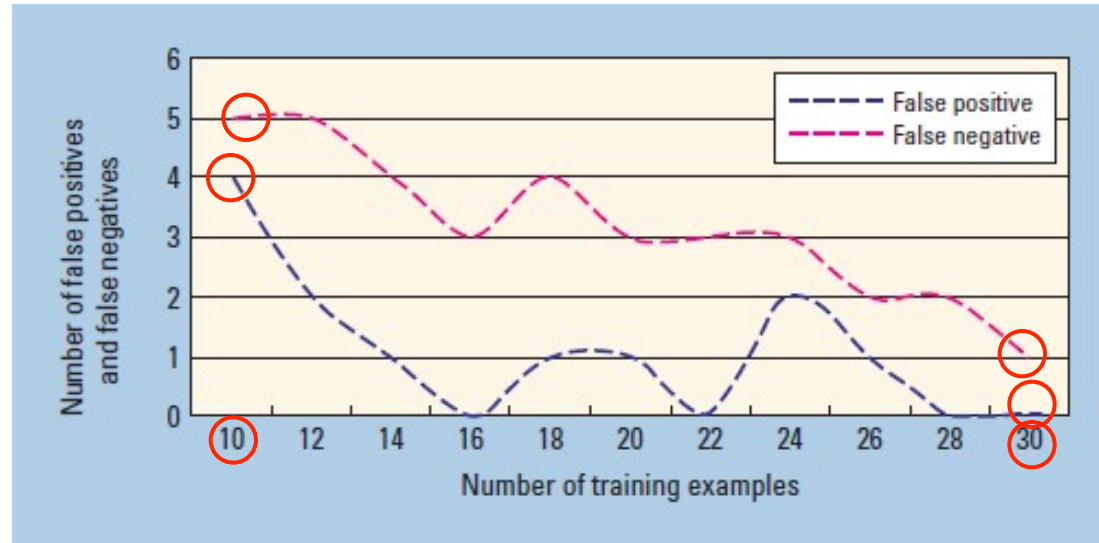
- The event auditor **captures all requests** received by a node and the corresponding responses, which is fundamental for behavior analysis.
- In the experiments with the behavior-based IDS, we considered using audit data from both a **log and a communication system**.
- Unfortunately, data from a log system has a **limited set of values** with **little variation**.



Evaluating the Event Auditor

- This made it **difficult to find attack patterns**, so we opted to explore **communication** elements to evaluate this technique.
- We evaluated the behavior-based technique using **artificial intelligence** enabled by a feedforward **neural network**.
- In the simulation environment, we monitored **five intruders and five legitimate users**.

Evaluating the Event Auditor



- We initiated the neural-network training with a data set representing 10 days of usage simulation.
- Using this data resulted in a **high number of false negatives** and a **high level of uncertainty**.
- Increasing the sample period for the learning phase improved the results.