

Machine Learning Approaches for Detecting and Mitigating Privilege Escalation Attack

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Abstract: Privilege escalation attacks are a serious threat to cloud computing security. In these attacks, an attacker exploits vulnerabilities in a system to gain elevated privileges, which can then be used to steal data, launch further attacks, or disrupt operations. Because of the recent exponential rise in attack frequency and sophistication, the proliferation of smart things has created significant cybersecurity challenges. Even though the tremendous changes cloud computing has brought to the business world, its centralization makes it challenging to use distributed services like security systems. Valuable data breaches might occur due to the high volume of data that moves between businesses and cloud service suppliers, both accidental and malicious. The malicious insider becomes a crucial threat to the organization since they have more access and opportunity to produce significant damage. Unlike outsiders, insiders possess privileged and proper access to information and resources. In this work, a machine learning-based system for insider threat detection and classification is proposed and developed a systematic approach to identify various anomalous occurrences that may point to anomalies and security problems associated with privilege escalation. By combining many models, ensemble learning enhances machine learning outcomes and enables greater prediction performance. Multiple studies have been presented regarding detecting irregularities and vulnerabilities in network systems to find security flaws or threats involving privilege escalation. But these studies lack the proper identification of the attacks. This study proposes and evaluates ensembles of Machine learning (ML) techniques in this context. This project implements machine learning algorithms for the classification of insider attacks.

Keywords: Artificial Intelligence, Industry, Intents, Insider Attack, Classification, Machine Learning Approaches, Networks, TF-IDF

I. INTRODUCTION

Multiple studies have been presented regarding detecting irregularities and vulnerabilities in network systems to find security flaws or threats involving privilege escalation. But these studies lack the proper identification of the attacks. This study proposes and evaluates ensembles of Machine learning (ML) techniques in this context. They utilized the “CERT Insider Threat Tools” dataset since obtaining genuine business system logs is extremely challenging. Employee computer actions logs are included in the CERT dataset and certain organizational data such as employee’s departments and responsibilities. They built insider-threat detection models to emulate real-world companies using machine learning-based methods. Privilege escalation attacks involve an attacker gaining higher-level access permissions than originally intended, potentially compromising the entire cloud infrastructure. Traditional security measures may not be sufficient to detect and prevent these sophisticated attacks. This research explores the integration of machine learning into cloud security to fortify defences against privilege escalation threats. To detect privilege escalation attempts, our system employs supervised machine learning models trained on historical data and anomaly detection algorithms. These models analyse patterns of user behaviour, system interactions, and access requests to identify deviations from normal activities. By continuously learning and adapting to evolving threat landscapes, the system can identify suspicious activities indicative of privilege escalation attempts.

II. LITERATURE SURVEY

Title	Year	Description	Limitations
P.Oberoi Survey of various security attacks in basedin clouds-based environments.	2017	In this paper we have studied the various security attacks (in general) with reference to the Clouds (as per The Treacherous 12 – Cloud Computing.	Less Applicable.
D.C. Le and A. N. Zincir Heywood, Machine Learning based insider threat modelling and detection.	2019	This paper proposes a new framework in constructing a user- centred machine learning based insider threat detection system on multiple data granularity levels.	More Complex
D. Tripathy, R. Gohil, and T. Halabi ‘Detecting SQL injection attacks in cloudSaaS using machine learning.’	2020	The SQL injection attack is one of the most potential threat to a SaaS application. This may result in loss of sensitive and important data. The purposeof this research paper is to investigate the potential of using machine learning techniques for SQL injection detection onthe application level.	Less Applicable
U.A. Butt. R. Amin. H. Aldabbas. S. Mohan. B. Alouffi. and A. Ahmadian Cloud- based email phishing attack using machine and deep learning algorithm.	2022	Cloud computing refers to the on-demand availability of personal computer system assets, specifically data storage and processing power. Without the client’s input. Emails are commonly used to send and receive data.	Less Secure
Ajmal, S. Ibrar, and R. Amin ‘‘Cloud computing platform: Performance analysis of prominent cryptographic algorithms,’’	2022	In this paper cloud cryptography is a technique that uses encryption algorithms to secure data. The significant advantage of cloud storage is no difficulty to get to, low protection and fixing cost so every association is working with the cloud.	More Complex
MUHAMMAD MEHMOOD1, RASHID AMIN,2MUHANA MAGBOUL ALIMUSLAM 3,JIANG XIE 4 Privilege Escalation Attack Detection and Mitigation in Cloud Using Machine Learning	2023	This paper implements machine learning algorithms for the classification of insider attacks. A customized dataset from multiple filles of the CERT dataset is used.	More complex

III. PURPOSE

Attackers target data sources because they have the most valuable and sensitive information. Every cloud user’s privacy and security are affected if data is lost. Insider threats are harmful operations carried out by people with authorization. In this problem we are providing best solution to avoid attack and detection of attack location. We evaluate the proposed system using real-world datasets and simulated privilege escalation scenarios. Performance metrics such as precision, recall, and false positive rates will be analysed to assess the effectiveness of the machine learning models in detecting and mitigating privilege escalation attacks.

OBJECTIVE OF SYSTEM

- **Detection Enhancement:** Improve the current capabilities of privilege escalation attack detection by leveragingmachine learning algorithms.

- **Automated Mitigation:** Develop and integrate automated mitigation strategies to respond promptly to detected privilege escalation attempts.
- **Feature Engineering:** Explore and optimize relevant features for machine learning model training, including user behaviour, access timestamps, and resource utilization.
- **Comprehensive Data Sources:** Utilize diverse data sources, including logs from authentication systems, access controllists, and system call traces, to provide a holistic view of user activities.
- **Evaluation Metrics:** Evaluate the proposed system's performance using real-world datasets and simulated scenarios.

IV. PROPOSED SYSTEM

The recent exponential rise in attack frequency and sophistication, the proliferation of smart things has created significant cybersecurity challenges. Even though the tremendous changes cloud computing has brought to the business world, its centralization makes it challenging to use distributed services like security systems. Valuable data breaches might occur due to the high volume of data that moves between businesses and cloud service suppliers, both accidental and malicious.

V. ACTIVITY DIAGRAM

Attackers target data sources because they have the most valuable and sensitive information. Every cloud user's privacy and security are affected if data is lost. Insider threats are harmful operations carried out by people with authorization. With the fast growth of networks, many companies and organizations have established their internal networks. The malicious insider becomes a crucial threat to the organization since they have more access and opportunity to produce significant damage. Unlike outsiders, insiders possess privileged and proper access to information and resources. Insider risks may be defined and addressed using criteria including insider indications, detection approaches, and insider kinds. There are two sorts of analysis intervals: real-time, which may identify malicious activity in real-time, and offline anomaly detection, which gathers log data and looks for certain patterns.

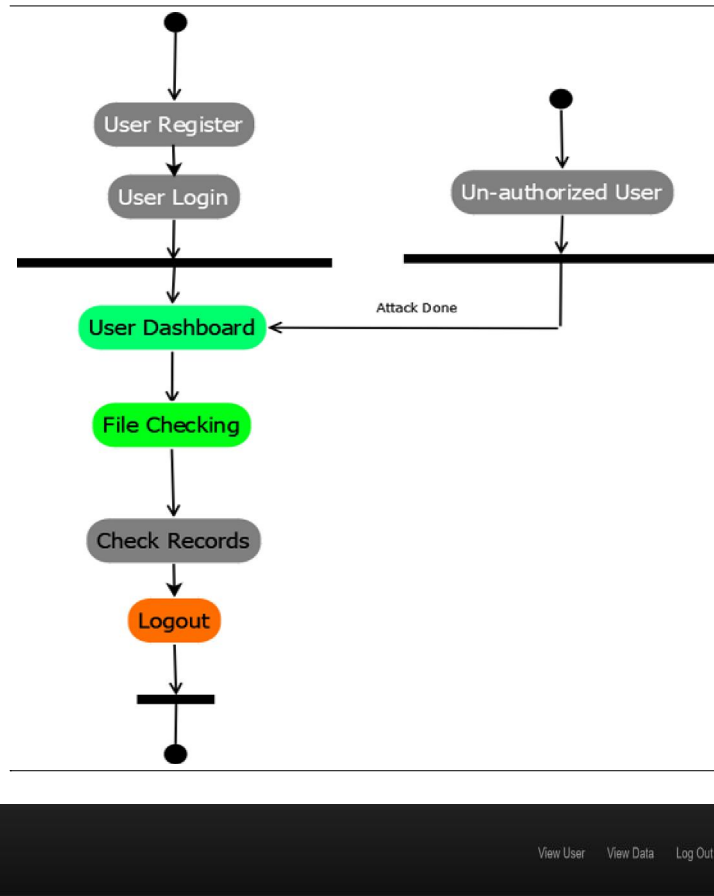
VI. APPLICATIONS

- Anomaly Detection: ML algorithms can analyze user behavior to detect unusual patterns, indicating potential privilege escalation attempts.
- User and Entity Behavior Analytics (UEBA): ML models can identify abnormal activities among users and entities, helping to detect unauthorized access or privilege escalation.
- Banking Application.
- Company Application.

VII. RESULT

Admin Dashboard:

After admin login successfully admin can access his dashboard in below image admin can check register users.



Detection Escalation Attack

Id	Username	Password	Email	Mobile No	Country	State	City
1	Divya	1234	dipaichaudhary15@gmail.com	9874561230	India	MAHARASHTRA	NASHIK
2	Dipali	123	mayurdhondwad225@gmail.com	04474569832	India	MAHARASHTRA	NASHIK

Back

Fig .1 Admin Dashboard

Trained Data:

After admin login admin can train dataset to detect attacks.

User Registration Log Out				
Detection Escalation Attack				
ID	Body Data	Keyword	Label	User NAME
1	So ? pay first lar Then when is da stock comin	?	ham	Dipali
2	So ? pay first lar Then when is da stock comin	pay	ham	Dipali
3	So ? pay first lar Then when is da stock comin	first	ham	Dipali
4	So ? pay first lar Then when is da stock comin	lar	ham	Dipali
5	So ? pay first lar Then when is da stock comin	da	ham	Dipali
6	So ? pay first lar Then when is da stock comin	stock	ham	Dipali
7	So ? pay first lar Then when is da stock comin	comin	ham	Dipali
8	So ? pay first lar Then when is da stock comin	?	ham	Dipali
9	So ? pay first lar Then when is da stock comin	pay	ham	Dipali
10	So ? pay first lar Then when is da stock comin	first	ham	Dipali
11	So ? pay first lar Then when is da stock comin	lar	ham	Dipali
12	So ? pay first lar Then when is da stock comin	da	ham	Dipali
13	So ? pay first lar Then when is da stock comin	stock	ham	Dipali
14	So ? pay first lar Then when is da stock comin	comin	ham	Dipali

Fig.2 Train data

Send Mail

User can send mail if mail contain spam so system will detect spam or ham.

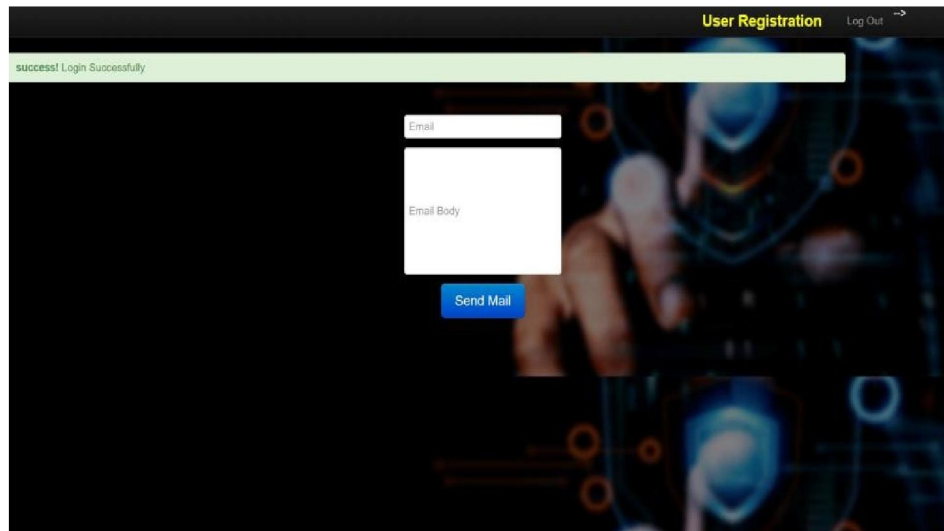


Fig 3. Send mail.

Detection of spam

User Registration Log Out				
Detection EScalation Attack				
ID	Body Data	Keyword	Label	User NAME
1	So ? pay first lar Then when is da stock comin	?	ham	Dipali
2	So ? pay first lar Then when is da stock comin	pay	ham	Dipali
3	So ? pay first lar Then when is da stock comin	first	ham	Dipali
4	So ? pay first lar Then when is da stock comin	lar	ham	Dipali
5	So ? pay first lar Then when is da stock comin	da	ham	Dipali
6	So ? pay first lar Then when is da stock comin	stock	ham	Dipali
7	So ? pay first lar Then when is da stock comin	comin	ham	Dipali
8	So ? pay first lar Then when is da stock comin	?	ham	Dipali
9	So ? pay first lar Then when is da stock comin	pay	ham	Dipali
10	So ? pay first lar Then when is da stock comin	first	ham	Dipali
11	So ? pay first lar Then when is da stock comin	lar	ham	Dipali
12	So ? pay first lar Then when is da stock comin	da	ham	Dipali
13	So ? pay first lar Then when is da stock comin	stock	ham	Dipali
14	So ? pay first lar Then when is da stock comin	comin	ham	Dipali

Fig 4 Spam mail

Graph

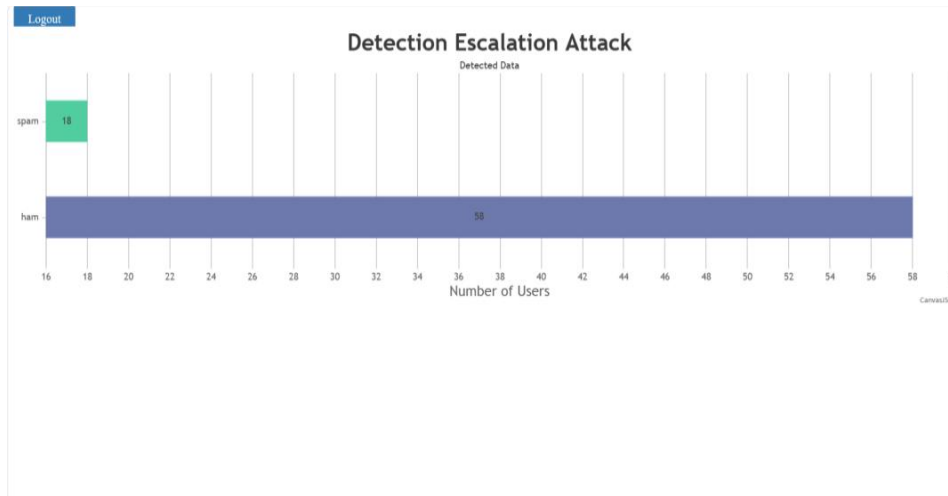


Fig 5 Graph

VIII. CONCLUSION

The malicious insider becomes a crucial threat to the organization since they have more access and opportunity to produce significant damage. Unlike outsiders, insiders possess privileged and proper access to information and resources. This paper proposed machine learning algorithms for detecting and classifying an insider attack. A customized dataset from multiple files of the CERT dataset is used in this work. Using these supervised machine learning algorithms, this paper demonstrated the effective experimental results having higher accuracy in the classification report.

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