# Advanced Privilege Escalation Prevention System in Windows

#### 01. Introduction

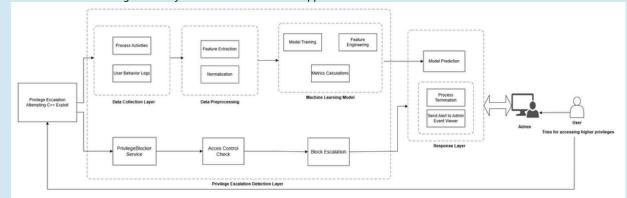
In the evolving landscape of cybersecurity threats, privilege escalation remains one of the most critical attack vectors. Once attackers gain initial access to a system, they often attempt to elevate their privileges to gain control over sensitive resources, execute administrative tasks, or maintain persistence. This project introduces "PrivilegeBlocker", a lightweight yet effective Windows-based detection and prevention system designed to monitor, detect, and stop privilege escalation attempts in real time. Unlike traditional antivirus or endpoint protection software, PrivilegeBlocker focuses specifically on process-level monitoring, token analysis, and security log auditing to identify anomalous behavior. To reinforce the system's intelligence, a machine learning model was externally developed to simulate detection capabilities, training on synthetic features that characterize malicious versus normal process behavior. This hybrid approach of system-level monitoring and intelligent detection helps demonstrate a proactive defense mechanism against privilege abuse.

# 02. Objective

Simulate a privilege escalation attack in a windows environment. Design and implement a prevention system (PrivilegeBlocker) that monitors and blocks such attacks. Include a machine learning model to analyze behavior and improve detection accuracy.

### 03. Methods

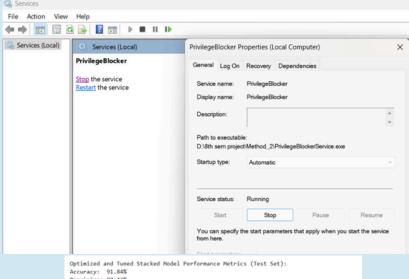
- Attack Simulation (Privilege Escalation)
- Developed a custom C++ exploit that attempts to gain administrative privileges from a standard user context.
- The exploit simulates real-world token manipulation techniques often used in privilege escalation attacks.
- o Consequences of the simulated attack are demonstrated to highlight system vulnerabilities.
- Detection Mechanism
  - A Windows Service (PrivilegeBlocker) written in C# runs in the background.
- It monitors:
- New Process Creation
- Parent-child process chains
- User Tokens & Privilege levels
- Suspicious or unauthorized privilege changes are flagged in real-time.
- Prevention System
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- Uses event log correlation and process inspection to identify privilege abuse patterns.
- Sends alerts or notifications when a potential escalation is blocked.
- Machine Learning Integration (External)
- Built a stacked ensemble ML model (XGBoost + Random Forest + SVM) to simulate detection accuracy.
- Trained on synthetically generated behavioral features of normal vs. escalated processes.
- Used SMOTE for class balance and Optuna for hyperparameter tuning.
- Achieved high accuracy and ROC-AUC scores to support model effectiveness.

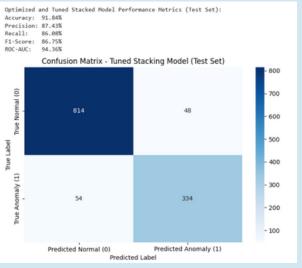


# 04. Results/Findings

- Successfully simulated privilege escalation using custom C++ code.
- Deployed PrivilegeBlocker a Windows Service that detects and blocks unauthorized privilege elevation attempts.
- Designed and trained a machine learning model using Random Forest, XGBoost & SVM to predict malicious privilege escalation behavior with high accuracy of 91.42%
- System logs suspicious behavior and alerts user/admin via Event Viewer.
- Demonstrated effectiveness in real Windows environment (not VM)







#### 05. Conclusion

- This project successfully developed a Windows-based Privilege Escalation
  Detection and Prevention System that addresses one of the most critical
  threats in cybersecurity: unauthorized privilege elevation. By simulating
  real-world attacks through custom C++ code and integrating a lightweight,
  proactive service (PrivilegeBlocker), the system effectively monitored and
  intercepted suspicious activity in real time.
- The inclusion of an externally trained Machine Learning model enhanced detection accuracy and supported behavior-based analysis of system processes. This dual-mode detection—combining rule-based monitoring and ML-based prediction—ensured higher reliability in identifying and blocking attacks.
- Overall, the solution demonstrated practical feasibility, strong performance, and adaptability across Windows environments without relying on virtual machines or external admin tools like Mimikatz. It provides a valuable contribution toward building self-defending endpoints and strengthens awareness of OS-level security mechanisms.

# 06. References

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