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# ENHANCING FRAUD DETECTION USING MACHINE LEARNING BEYOND THE LOGIN EVENT: STRATEGIES AND APPROACHES

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#### Abstract

Account Takeover (ATO) is a prevalent and growing threat in the realm of cybersecurity, often associated with the initial breach of login credentials. However, the threat landscape extends far beyond the initial login phase. Therefore, traditional fraud detection methods focusing solely on login activities are no longer sufficient. This paper explores advanced ML strategies to detect and prevent fraudulent activities that occur after the initial account compromise by analyzing downstream events beyond the login event, thus ensuring a comprehensive security framework.

Keywords Account Takeover, Digital Fraud Detection, Machine Learning, Account Security Beyond Login, Digital Security, Identity Protection, Multi-Factor Authentication (MFA), Threat Intelligence, Data Protection, Continuous Monitoring.

### I. INTRODUCTION

As digital transactions and online services become more pervasive, the threat of Account Takeover (ATO) has grown exponentially. Historically, fraud detection mechanisms have concentrated on thwarting attacks during the login phase. However, attackers often employ sophisticated techniques to exploit accounts long after gaining access, making it essential to expand detection efforts beyond the login event. By implementing ATO beyond login strategies, organizations can enhance their fraud detection capabilities and protect user accounts more effectively.

### II. THE IMPORTANCE OF DOWNSTREAM EVENT MONITORING

While the initial compromise of login credentials is a critical security breach, the actions taken by attacker's post-login often result in the most significant damage. From unauthorized transactions to data exfiltration, these activities exploit vulnerabilities beyond the login event. Downstream monitoring using machine learning offers a proactive approach to detecting fraud and responding in real-time, creating a robust defence system. Strategies to enhance fraud detection by addressing ATO activities post-login include –

### 1. Post-Login Vulnerabilities

After successfully logging in, attackers may engage in various fraudulent activities, including:

- Unauthorized transactions by performing transactions without the account holder's consent.
- **Data exfiltration** by stealing sensitive information stored in the account.
- **Account modification** by changing account settings or linked information to facilitate further fraud
- Service abuse by misusing account privileges for unauthorized purposes.



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### 2. Importance of Continuous Monitoring

Detecting these post-login fraud activities requires continuous monitoring and analysis of user behaviour. By analysing user patterns and detecting anomalies, organizations can respond to threats as they unfold, minimizing the impact on users and systems.

## III. SCENARIOS WHERE AN ACCOUNT TAKEOVER CAN HAPPEN BEYOND THE INITIAL LOGIN

Account Takeover often extends beyond the initial breach of credentials. Attackers exploit various vulnerabilities to maintain control of the account or manipulate its resources. Understanding these scenarios is critical for developing robust countermeasures.

- Weak or Stolen Credentials: Attackers might obtain login credentials through phishing, social engineering, or data breaches. Once they have these credentials, they can access an account without needing to log in again.
- **Session Hijacking**: Attackers can intercept or steal session cookies or tokens, which allow them to take over an authenticated session without needing to know the login credentials. This can happen through cross-site scripting (XSS) attacks or man-in-the-middle attacks.
- Malware or Key loggers: If a user's device is infected with malware or a keylogger, attackers can capture login credentials and use them for account takeovers. After the initial login, the attacker can control the account.
- Forgotten Logins on Public Devices: Users may log into their accounts on public computers and forget to log out. Subsequent users of the same computer may have access to the logged-in account.
- **Session Persistence**: Some web applications may maintain a user's session even after the user logs out. If this is not implemented correctly, it can lead to unauthorized access to the account.
- Multi-Session Vulnerabilities: In some cases, users can open multiple sessions or tabs in a web application. If one of these sessions is compromised, the attacker can gain access to other active sessions without needing to log in again.
- **Password Resets:** If an attacker has access to the user's email or can intercept password reset emails, they can change the account's password, effectively taking it over.
- Account Recovery Weaknesses: Some account recovery processes may have security weaknesses that allow attackers to reset passwords or take over accounts without proper verification.
- Third-Party Integrations: Accounts linked to third-party services or applications may be vulnerable if the third-party service is compromised. An attacker could exploit these integrations to gain access to the user's account.
- **Insider Threats:** An employee or someone with legitimate access may misuse their privileges to take over an account from within the organization.

### IV. STRATEGIES FOR ATO FRAUD DETECTION BEYOND LOGIN

Mitigating ATO risks requires a multi-faceted approach that combines advanced technologies, user behaviour analysis, and robust security protocols. This section outlines effective strategies that organizations can implement to enhance fraud detection capabilities beyond the login phase.

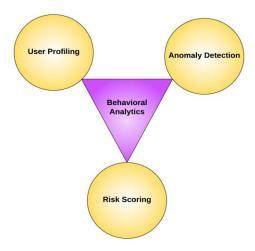
### 1. Behavioural Analytics

Proactive monitoring and anomaly detection are critical in identifying and responding to ATO



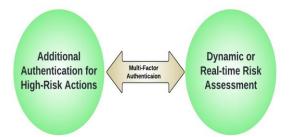
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activities beyond the login phase. Behavioural analytics involves creating profiles of typical user behaviour and identifying deviations from these norms. This method is effective for detecting anomalies that may signify fraud.



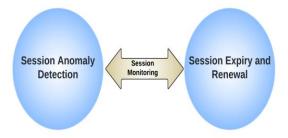
### 2. Multi-Factor Authentication (MFA)

Extending MFA to sensitive post-login actions, such as large transactions or account modifications, adds a crucial layer of security, ensuring that fraudulent activities are mitigated even after account access.



### 3. Session Management and Real-Time Transaction Monitoring

Monitoring session activities allows organizations to detect unauthorized access and mitigate risks associated with session persistence and hijacking. Real-time transaction monitoring further enhances fraud prevention by analysing patterns for anomalies.

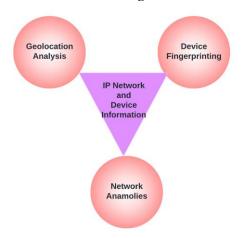




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### 4. IP Network & and Device Information

Analysing the geographical location and device used for account access helps detect inconsistencies, such as access from unfamiliar or high-risk locations.



### V. CASE STUDIES AND EXAMPLES

Implementing ATO detection strategies has yielded tangible results for organizations across industries. The following case studies highlight the effectiveness of ML-driven fraud detection techniques.

### **Case Study 1: Financial Institution**

Sift Science's fraud detection platform has successfully implemented ATO beyond login implementation to enhance their fraud detection capabilities for its customers across various industry verticals. The integration of behavioural analytics and risk-based authentication allowed their fintech client to detect and respond to suspicious activities promptly. Real-time monitoring and anomaly scoring further helped detect and respond to suspicious activities promptly.

### Case Study 2: E-commerce Organisation

Another case study is where Sift platform helped deploy advanced threat detection tools and comprehensive logging on one of their e-commerce customers, significantly enhancing their fraud detection capabilities. The integration of AI and machine learning models enabled the platform to identify and mitigate fraudulent transactions effectively on post-login downstream activities.

### VI. CONCLUSION

While preventing account takeovers at the login phase is crucial, addressing fraudulent activities beyond this point is equally important. Organizations must adopt a comprehensive approach encompassing robust authentication, continuous monitoring, and machine learning-based anomaly detection. By proactively addressing vulnerabilities and implementing these strategies, businesses can significantly strengthen their defences against account takeover threats, safeguarding user data and maintaining trust.



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