



AI and ML Driven Cyber security: Emerging Technologies and Future Perspectives

¹Dr.M.Kundalakesi, ²Harnisya Siva, ³AbrahamS

¹AssistantProfessor, ^{2,3}Students of BCA, Department of Computer Application, Sri Krishna Arts and Science College,
Coimbatore.

ABSTRACT:

The rapid development of digital technologies and the increasing dependence on connected systems have greatly increased the risk of cybersecurity threats. Organizations and individuals are facing various cyber-attacks such as malware, phishing, ransomware, denial of service attacks, and advanced persistent threats. These attacks are becoming more complex, frequent, and difficult to detect using traditional security methods. Conventional cybersecurity solutions mainly depend on predefined rules and signature-based detection, which are often ineffective against new and unknown threats. To overcome these challenges, Artificial Intelligence (AI) and Machine Learning (ML) have become important technologies in modern cyber security. Artificial Intelligence enables machines to perform tasks that require human intelligence, including learning, reasoning, and decision-making.

Machine Learning allows computer systems to learn from data, identify patterns, and improve their performance without requiring direct programming. In cyber security, AI and ML techniques help analyze large amounts of network traffic, system records, and user behaviour data in real time. These technologies support threat detection, intrusion identification, malware classification, anomaly detection, and prediction of possible cyber-attacks. Different learning methods such as supervised learning, unsupervised learning, and reinforcement learning are used to identify existing threats and discover new attack patterns. AI-based cyber security systems improve detection accuracy, reduce false alarms, and provide faster responses to security incidents.

Machine Learning-based Intrusion Detection Systems (IDS) and Intrusion Prevention Systems (IPS) can identify complex attack behaviours that may not be detected by traditional approaches. AI also helps automate incident response by reducing human dependency and improving response speed. Behaviour analysis using ML further strengthens security by identifying unusual user activities and possible insider threats.



INTRODUCTION:

The increasing use of digital platforms, cloud computing, mobile devices, and interconnected networks has completely changed the way organizations and individuals manage information. Although digital transformation provides better communication, efficiency, and accessibility, it has also introduced several cyber security challenges. Cyber threats such as malware attacks, phishing, ransomware, unauthorized access, and data breaches are continuously increasing in both number and complexity. Traditional security mechanisms based on fixed rules, signatures, and manual analysis are no longer sufficient to protect modern digital environments. Artificial Intelligence (AI) and Machine Learning (ML) have emerged as effective solutions for improving cyber security systems.

AI refers to the ability of machines to perform intelligent activities similar to human thinking, while ML focuses on developing algorithms that learn from previous data and improve their decision-making ability. The application of AI and ML in cyber security allows systems to process large volumes of information generated from networks, applications, and users. These technologies can detect abnormal patterns, identify malicious activities, and predict possible security threats before serious damage occurs. Machine learning models are capable of recognizing known attack patterns as well as identifying previously unknown threats. This makes them highly useful for detecting zero-day attacks and advanced cyber threats that can bypass traditional security tools.

AI-based cyber security solutions are widely used in intrusion detection, malware analysis, fraud detection, vulnerability assessment, and automated incident response. They help security teams by reducing manual workload, improving threat identification, and enabling faster decision-making. Despite these advantages, AI and ML implementation also faces challenges including data availability, privacy concerns, lack of transparency in decision-making, and the possibility of attacks against AI models.

However, continuous improvements in technology are helping overcome these limitations. The integration of AI and ML into cyber security provides a powerful approach for developing intelligent, adaptive, and reliable defence systems. As cyber threats continue to evolve, these technologies will become increasingly important for protecting digital information and infrastructure.





Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

AI AND ML TECHNIQUES USED IN CYBERSECURITY:



Artificial Intelligence and Machine Learning techniques play an important role in improving cyber security by providing automated, intelligent, and adaptive solutions. These methods help organizations detect threats, analyze attacks, and respond effectively.

1. Supervised Learning:

Supervised learning uses labelled datasets where models are trained using examples of normal and malicious activities. These algorithms are commonly used for classification tasks such as malware detection, spam filtering, and intrusion detection.

Common supervised learning algorithms include:

- Decision Tree
- Random Forest
- Support Vector Machine (SVM)
- Naïve Bayes
- Logistic Regression

These models identify patterns from previous attacks and classify new activities as safe or harmful.

2. Unsupervised Learning:

Unsupervised learning works without labelled data and identifies hidden patterns or unusual behaviour. It is mainly used for detecting unknown threats and zero-day attacks.

Techniques include:

- K-Means Clustering
- DBSCAN
- Autoencoders



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

These methods help detect abnormal network traffic, suspicious user behaviour, and unusual system activities.



3. Semi-Supervised Learning:

Semi-supervised learning combines both labelled and unlabelled data. It is useful in situations where collecting labelled cyber security data is difficult. This approach improves detection performance by learning from limited known threats and large amounts of available data.

4. Deep Learning:

Deep learning techniques are advanced ML methods that analyze complex and large-scaled data. They are useful for identifying sophisticated cyber threats.

Common deep learning models include:

- Artificial Neural Networks (ANN)
- Convolutional Neural Networks (CNN)
- Recurrent Neural Networks (RNN)

These models analyze network packets, system logs, and user activity sequences to detect advanced attacks.

5. Reinforcement Learning:

Reinforcement learning allows systems to learn through interaction with their environment. In cybersecurity, it is used for automated decision-making, such as improving firewall rules and selecting appropriate responses during attacks.

6. Natural Language Processing (NLP):

NLP techniques analyze text-based information such as emails, messages, and web content. They are used for detecting phishing emails, social engineering attacks, and malicious communication patterns.

DATASETS AND DATA PREPROCESSING:

Datasets are an essential part of developing Artificial Intelligence and Machine Learning-based cybersecurity systems. High-quality datasets are required for training, testing, and evaluating models used for detecting threats, identifying intrusions, analyzing malware, and



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

preventing cyber fraud. Commonly used cyber security datasets include KDD Cup 99, NSL-KDD, CICIDS2017/2018, UNSW-NB15, DARPA, and Bot-IoT. These datasets contain information related to network traffic, attack behaviour, system activities, and different types



of cyber threats such as denial-of-service attacks, malware, unauthorized access, and probing activities.

Apart from public datasets, organizations also use real-world data sources such as firewall logs, authentication records, email communication data, and endpoint monitoring information to train security models. Data preprocessing is a critical step that improves the accuracy and efficiency of machine learning models. Raw cyber security data often contains missing values, duplicate records, noise, and inconsistent information, which must be processed before analysis.

The main preprocessing steps include:

1. Data Cleaning:

This involves removing duplicate data, correcting errors, and managing missing or incorrect values to improve data quality.

2. Data Transformation:

Raw data is converted into a suitable format for machine learning algorithms. Categorical information such as protocol types and attack labels are transformed into numerical values.

3. Feature Selection and Extraction:

Relevant features are selected from large datasets to reduce complexity and improve model performance. Removing unnecessary attributes helps reduce noise.

4. Data Normalization and Scaling:

Different features may have different value ranges. Normalization ensures that all attributes are balanced and prevents certain features from dominating the learning process.

5. Handling Imbalanced Data:

In cyber security datasets, normal activities usually have more records than attack activities. Techniques such as oversampling, undersampling, and Synthetic Minority Over-sampling Technique (SMOTE) are used to balance the dataset.



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

6. DataSplitting:



The dataset is divided into training, validation, and testing sections to measure the performance and reliability of the developed model. Proper dataset selection and preprocessing are important for creating accurate, efficient, and dependable cyber security solutions.

FACTORS RESPONSIBLE FOR CYBER ATTACKS:



Cyber-attacks occur due to various technical, human, and organizational weaknesses. Understanding these factors helps organizations develop effective security strategies and reduce cyber risks.

1. Rapid Digitalization:

The increasing adoption of internet services, cloud computing, mobile applications, and IoT devices has expanded the attack surface. More connected systems provide attackers with additional opportunities to exploit vulnerabilities.

2. Weak Security Infrastructure:

Outdated software, poor security configurations, and unpatched systems create opportunities for attackers. Weak firewall policies and insufficient monitoring mechanisms increase the possibility of security breaches.

3. Human Error:

Human mistakes are one of the major causes of cyber incidents. Users may unknowingly open phishing emails, use weak passwords, or install harmful applications. Lack of security awareness increases vulnerability.

4. Poor Authentication Practices:



Weak or reused passwords make systems vulnerable to unauthorized access. The absence of multi-factor authentication increases the risk of credential-based attacks.



5. Advanced Attack Techniques:

Cyber attackers continuously develop new methods using malware, ransomware, social engineering, and automated attack tools. These techniques can bypass traditional security controls.

6. Insider Threats:

Employees, contractors, or trusted individuals with system access may accidentally or intentionally compromise security. Insider threats are difficult to detect because they involve authorized users.

7. Lack of Security Awareness:

Organizations without proper security training, policies, and risk management practices are more likely to experience cyber incidents.

8. Financial and Political Motivation:

Many attacks are performed for financial benefits such as data theft, fraud, and ransomware payments. Some attacks are motivated by political goals, espionage, or cyber warfare.

9. Poor Incident Response:

Delayed identification and response to attacks allow attackers to cause more damage. Effective monitoring and recovery plans are necessary to reduce the impact of cyber incidents.

TYPES OF THREAT AGENTS:

A threat agent refers to any individual, group, or organization capable of performing cyber attacks against digital systems and information resources. Threat agents differ based on their skills, resources, and objectives.

1. Cyber Criminals:

Cyber criminals mainly perform attacks for financial benefits. They use methods such as phishing, ransomware, identity theft, and financial fraud to gain illegal access or steal valuable information.



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

2. HackersandScriptKiddies:

Hackers may attack systems for different reasons, including curiosity, recognition, or personal goals. Script kiddies usually use existing hacking tools without advanced technical knowledge but can still create security problems.

3. Insider Threats:

Insiders include employees, contractors, or partners who have authorized access. They may misuse their access intentionally or accidentally expose sensitive information.

4. Nation-State Attackers:

Nation-state groups are supported by governments and conduct highly advanced cyber operations. Their objectives include cyber espionage, political influence, and targeting critical infrastructure.

5. Hacktivists:

Hactivists perform cyber attacks based on social, political, or ideological beliefs. Common activities include website attacks, information leaks, and service disruption.

LIMITATIONS OF TRADITIONAL CYBERSECURITY APPROACHES:



Traditional cyber security methods such as firewalls, antivirus software, access control systems, and signature-based detection have provided protection against many known threats. However, they face limitations when dealing with modern and rapidly changing cyberattacks. One major limitation is the lack of adaptability. Traditional systems depend on predefined rules and require manual updates, making them less effective against new attack techniques and zero-day threats.

Another challenge is the occurrence of false positives and false negatives. Security systems may incorrectly identify normal activities as threats or fail to detect actual attacks. Traditional security solutions also struggle with scalability. The growth of cloud computing, IoT devices, and large-scale networks generates massive amounts of data that are difficult to analyze using conventional methods. Additionally, these approaches depend heavily on human monitoring and manual investigation. Security professionals must analyze alerts and logs, which increases workload and delays response time. Due to these limitations, organizations are



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

moving toward AI and ML-based security systems that provide faster, adaptive, and intelligent threat detection.



ROLE OF ARTIFICIAL INTELLIGENCE IN CYBER SECURITY:

Artificial Intelligence (AI) has become an important technology in modern cyber security due to the increasing complexity and frequency of cyber threats. Traditional security approaches mainly depend on predefined rules and known attack patterns, which makes them less effective against new and advanced threats. AI improves cyber security by enabling systems to analyze large amounts of data, identify hidden patterns, and respond to threats automatically.

One of the major roles of AI in cyber security is threat detection and prevention. AI-based systems continuously monitor network traffic, user activities, and system behaviour to identify suspicious actions. These systems can detect unauthorized access attempts, malware activities, and unusual behaviour patterns before they cause serious damage. AI also plays a significant role in detecting malware and phishing attacks. By analyzing files, emails, links, and user communication patterns, AI can identify malicious content and prevent potential security risks. Natural Language Processing (NLP) techniques help detect phishing messages and social engineering attacks by examining language patterns and suspicious information. Another important application of AI is improving Intrusion Detection Systems (IDS) and Intrusion Prevention Systems (IPS). AI-based IDS and IPS solutions can detect complex attack patterns, reduce false alerts, and improve the accuracy of security monitoring. AI supports automated incident response by performing actions such as blocking harmful activities, isolating affected systems, and generating security alerts. This reduces response time and helps organizations minimize the impact of cyber-attacks.

Furthermore, AI helps in vulnerability management by identifying security weaknesses and prioritizing risks based on their severity. By continuously learning from new data, AI-based systems provide adaptive and proactive security protection. Overall, Artificial Intelligence strengthens cyber security by improving detection speed, accuracy, automation, and decision making. It works together with human expertise to create stronger and more reliable security defence mechanisms.

MACHINE LEARNING TECHNIQUES FOR THREAT DETECTION:

Machine Learning techniques are widely used in cyber security to identify threats by learning patterns from large amounts of data. These techniques help systems detect abnormal behaviour, classify attacks, and predict possible security risks.





1. Supervised Learning Techniques:

Supervised learning models are trained using labelled datasets containing examples of both normal and malicious activities. These models classify new data based on previously learned patterns.

Common algorithms include:

Decision Tree:

Used for classification and decision-making by creating a tree-like structure based on security-related features.

Random Forest:

Combines multiple decision trees to improve detection accuracy and reduce errors.

Support Vector Machine (SVM):

Used to classify network activities and identify malicious behaviour.

Logistic Regression:

Applied for predicting whether an activity belongs to a normal or attack category. These algorithms are commonly used in malware detection, spam filtering, and intrusion detection.

2. Unsupervised Learning Techniques:

Unsupervised learning identifies hidden patterns without requiring labelled attack data. It is useful for detecting unknown threats and zero-day attacks.

Common methods include:

K-Means Clustering:

Groups similar data points together and identifies unusual activities.

DBSCAN:



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

Detects abnormal patterns and separates suspicious activities from normal behaviour.



Autoencoders:

Used to identify anomalies by learning normal system behaviour and detecting deviations.

3. Semi-Supervised Learning:

Semi-supervised learning uses a combination of labelled and unlabelled data. It improves threat detection when only limited examples of cyber attacks are available. This technique is useful in large-scale networks where collecting complete labelled datasets is difficult.

4. Deep Learning Techniques:

Deep learning models analyze complex and high-dimensional cybersecurity data.

Artificial Neural Networks (ANN):

Used for identifying patterns in security-related information.

Convolutional Neural Networks (CNN):

Effective for analyzing complex structures such as network traffic and malware features.

Recurrent Neural Networks (RNN):

Used for analysing sequential data such as user activities and system logs. Deep learning methods are highly effective in detecting advanced persistent threats and sophisticated malware attacks.

5. Reinforcement Learning:

Reinforcement learning enables security systems to learn through continuous interaction with the environment. It helps develop automatic response strategies such as improving firewall rules, blocking attacks, and selecting suitable defence actions.

6. Natural Language Processing (NLP):

NLP techniques are used to analyze text-based cyber threats such as phishing emails, malicious messages, and social engineering attempts. By understanding language patterns, NLP helps



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

identify suspicious communication.

CONCLUSION:

The continuous growth of cyber threats has created the need for intelligent and adaptive security solutions. Traditional cybersecurity methods are no longer sufficient to handle modern attacks due to their limited ability to detect unknown and evolving threats. Artificial Intelligence and Machine Learning provide advanced approaches for improving cybersecurity by enabling automated threat detection, real-time monitoring, intelligent analysis, and faster response. AI and ML techniques can identify suspicious activities, detect malware, prevent intrusions, and support effective incident management. Although challenges such as data privacy, model reliability, and adversarial attacks remain, continuous advancements are improving the effectiveness of AI-based security systems. The integration of Artificial Intelligence and Machine Learning with traditional security methods will play a major role in developing stronger, smarter, and more resilient cyber defence systems in the future.

REFERENCE:

1. **Sommer&Paxson(2010)**
Discusses the challenges of applying machine learning to network intrusion detection systems.
2. **Buczak&Guven(2016)**
Provides a detailed survey of machine learning and data mining techniques in cybersecurity.
3. **Patcha&Park(2007)**
Explains various anomaly detection techniques used for identifying cyber-attacks.
4. **Lavalle et al.(2009)**
Analyses limitations of the KDD Cup 99 dataset used for intrusion detection research.
5. **Shone et al. (2018)**
Proposes deep learning models for accurate network intrusion detection.
6. **Kim et al. (2014)**
Introduces a hybrid intrusion detection approach combining misuse and anomaly detection.
7. **Lippmann et al.(2000)**
Evaluates intrusion detection systems using the DARPA dataset.
8. **Moustafa&Slay (2015)**
Presents the UNSW-NB15 dataset for modern intrusion detection experiments.
9. **Sharfuddin et al.(2018)**
Introduces CICIDS 2017 dataset reflecting real-world network attacks.
10. **Goodfellow et al.(2014)**
Explains adversarial attacks that can fool machine learning models.
11. **Scikit-learn**
Provides widely used machine learning algorithms for cybersecurity research.
12. **TensorFlow**



Indo Asian Journal of Management and Entrepreneurship (IAJOME)

|| ISSN: 2319-2992, Impact Factor 5.007 ||

|| Peer-Reviewed | Open Access | International Journal ||

|| Volume: 17 | Issue: 07 | July 2026 | www.iajome.com ||

Supports deep learning models for threat detection and analysis.

13. MITRE ATT&CK



Offers a knowledge base of adversary tactics and techniques.

14. Cisco Security

Applies AI to detect and respond to enterprise cyber threats.

15. IBM Security

Uses machine learning for advanced threat intelligence and analytics.

16. Symantec

Employs AI-based systems for malware and endpoint protection.

17. FireEye

Focuses on detecting advanced persistent threats using AI.

18. ENISA

Publishes reports on AI applications in European cyber security.

19. NIST

Provides standards and guidelines for secure AI systems.

20. IEEE Xplore

A digital library for research on AI and ML in cyber security.