



Cambridge
International

Professional Research Thesis

Titled

**“A Comparative Analysis of the Effectiveness of Artificial
Intelligence in Predicting Cyberattacks Versus
Traditional Information Security Systems”**

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2026



DEDICATION

*To the soul of my beloved father,
who departed in body yet whose spirit remains alive within my heart.
May God grant him His mercy and bestow upon him the vastness of His
paradise.*

*He taught me that the road to glory begins with the first step of
knowledge,
and that sincere determination can make the impossible possible.
His giving was the first seed that has today blossomed into a scholarly
achievement I hold with great pride.*

*To my dear mother,
the pulse of sincere supplication and a heart that has known nothing
but unconditional embrace and patience.*

*She was my pillar in moments of weakness and the source of my
strength in the face of hardship.*

Through her blessing and prayers, all paths were made easy.

*To my dear siblings,
who surrounded me with genuine affection and unwavering support,
serving as a steadfast foundation upon which I could always lean.
Their constant presence by my side had a profound impact on the
completion of this journey.*

*To my loyal friend, Jalila Sayyid Ahmad Jawda,
who shared my anxiety before the joy, and who was a true support
through the most difficult junctures.*

She is most deserving of sincere and heartfelt gratitude.

*To Dr. Hajar Rif'at, research supervisor, and to all the supervising faculty
of the OkCambridge Academy,
in recognition of the rigorous academic supervision and valuable
scholarly guidance they provided,
which contributed effectively to presenting this thesis in the
distinguished form befitting it.*

*And to every pure soul that believed in me, prayed for me, and
supported me — even with a single word —*

*I dedicate this humble effort, praying to God that He makes it beneficial
knowledge and a lasting legacy.*

INTRODUCTION

In the digital age, the world is witnessing rapid advancements in information and communication technologies, accompanied by a notable increase in the volume and complexity of cyberattacks targeting individuals, organizations, and the critical infrastructure of nations. Cyberattacks have become increasingly sophisticated and intelligent, exploiting both technical vulnerabilities and user behavior. This has posed significant challenges to traditional information security systems, which predominantly rely on static rules and pre-known attack signatures.

Against this backdrop, artificial intelligence (AI) has emerged as one of the most pivotal modern technologies capable of bringing about a qualitative leap in the field of information security. AI's capacity to analyze vast quantities of data, learn from behavioral patterns, and predict potential attacks before they occur has rendered it indispensable. Machine learning (ML) and deep learning (DL) techniques have contributed to enhancing the efficiency of intrusion detection systems, improving response speed, and reducing false-positive rates compared to conventional solutions.

Building upon this foundation, this thesis analyzes the effectiveness of artificial intelligence in predicting cyberattacks, conducting a systematic and scientific comparison between intelligent systems and traditional information security systems. The aim is to assess the extent to which AI contributes to strengthening information security and addressing the growing landscape of cyberthreats.

CHAPTER ONE: GENERAL FRAMEWORK AND THEORETICAL CONCEPTS OF THE STUDY

Section One. General Framework of the Study

Subsection 1. Introduction to the Study

The accelerating development of information and communication technologies has led to an unprecedented expansion in the use of digital systems within governmental and private institutions. This has been accompanied by a marked increase in the scale and complexity of cyberattacks targeting information assets and critical systems. These attacks are no longer confined to well-known conventional methods; rather, they now employ advanced techniques capable of circumventing classical protective measures, posing a direct threat to information security and business continuity.

Traditional information security systems — such as signature-based and rule-based intrusion detection systems — rely on prior knowledge of attack patterns, which limits their effectiveness against novel and unknown threats, particularly zero-day attacks. These systems also suffer from high false-positive rates, which diminish their efficiency and increase the burden on information security professionals.

In light of these challenges, artificial intelligence has emerged as one of the most important modern trends in the field of information security,

owing to its ability to analyze massive volumes of data, detect anomalous behavioral patterns, and continuously learn from previous attacks. Machine learning and deep learning techniques have improved the accuracy of cyberattack prediction and enhanced early-response capabilities compared to traditional systems.

Accordingly, this thesis seeks to analyze the effectiveness of artificial intelligence in predicting cyberattacks by examining its role in strengthening information security and comparing it with traditional systems in terms of accuracy, flexibility, response speed, and false-positive rates. The thesis also aims to highlight the feasibility of relying on AI technologies as a strategic solution to address growing cyberthreats and to contribute to the development of more efficient and effective information security systems.

Subsection 2: Statement of the Problem

Despite the considerable evolution in information security tools and techniques over the past decades, traditional information security systems (Signature-Based / Rule-Based) face a fundamental shortcoming in predicting modern cyberattacks — particularly those that lack pre-known signatures, such as zero-day attacks. These systems are inherently built upon static rules or signatures drawn from past attacks, rendering them incapable of adequately addressing new threats or anomalous behavior. This frequently results in high false-positive rates and an inability to adapt to the rapid changes occurring in digital network environments.

By contrast, AI technologies powered by machine learning (ML) and deep learning (DL) algorithms have demonstrated a superior capacity to discover unknown patterns and predict attacks before they materialize, through the analysis of vast volumes of network data and anomalous behavioral signals. Contemporary literature indicates that these approaches can realistically improve the overall performance of intrusion detection systems (IDS), including threat classification and the reduction of false-positive rates in comparison with traditional systems.

Nevertheless, a number of significant research and practical challenges remain unresolved, most notably:

1. Data Imbalance and Quality

AI algorithms depend heavily on the availability of high-quality and sufficient data — a condition that is not always met in real-world network environments, especially given the scarcity of data pertaining to novel attack types.

2. False-Positive Rate and Predictive Instability

Although AI can enhance accuracy, certain models continue to generate erroneous alerts due to noise and complex patterns in the data, which adversely affects the decision-making process.

3. Adversarial Attacks Against AI Systems

This is a strategy employed by attackers to exploit the weaknesses of intelligent models, wherein input data is maliciously manipulated to mislead the model into generating incorrect predictions. This represents a major challenge to the trustworthiness and reliability of AI-based security systems.

4. Explainability and Interpretability

Many deep learning models operate as 'black boxes,' making it difficult for security specialists to interpret the reasoning behind prediction outcomes or system recommendations.

5. Continuous Updating and Adaptation to Evolving Threats

Modern attacks require models that continuously learn over time, which imposes challenges in terms of ongoing training, maintenance, and the dynamic updating of the intelligent system.

Despite these challenges, recent studies (2024–2025) have confirmed that integrating artificial intelligence into information security systems leads to tangible improvements in the detection of unknown attacks and the prediction of risks before their occurrence, compared to traditional systems that depend on static signature libraries.

Summary of the Problem

The core problem lies in the fact that traditional information security systems lack sufficient capacity to adapt to new and complex cyberattacks, resulting in:

- Elevated false-positive rates, which create confusion for cybersecurity teams.*
- Inability to detect unknown and previously unseen attacks.*
- Slow response to complex, multi-stage attacks.*

Conversely, artificial intelligence is capable of handling large volumes of data and extracting patterns that may signal potential attacks before they

occur. This raises the central question: Is artificial intelligence more effective than traditional systems in predicting cyberattacks?

Table 1. Comparative Overview — Traditional Systems vs. Artificial Intelligence

<i>Feature</i>	<i>Traditional Systems</i>	<i>Artificial Intelligence</i>
<i>Attack Prediction</i>	○ <i>Limited</i>	▣ <i>Advanced</i>
<i>Unknown Attacks</i>	○ <i>Difficult to detect</i>	▣ <i>Highly effective</i>
<i>False Positive Rate</i>	● <i>High</i>	▣ <i>Low</i>
<i>Response Speed</i>	○ <i>Slow</i>	▣ <i>Fast</i>

Subsection 3. Significance of the Study

The significance of this study stems from several dimensions, the most prominent of which are:

1. Scientific Significance.

The study elucidates the extent to which artificial intelligence is effective in enhancing information security compared to traditional systems — a contemporary and highly relevant subject within the scientific research domain.

2. Practical Significance.

The study provides practical recommendations for organizations seeking to adopt intelligent security systems capable of predicting and responding to attacks with speed and effectiveness.

3. Strategic Significance.

The study contributes to the development of cybersecurity policies and strategies for organizations, in alignment with the rapid pace of technological advancement.

Subsection 4: Objectives of the Study

The study aims to:

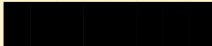
- *Analyze the role of artificial intelligence in predicting cyberattacks.*
- *Compare the performance of traditional systems and AI in terms of accuracy, response speed, and the ability to detect unknown attacks.*
- *Identify the advantages and challenges associated with the use of AI in information security.*
- *Provide practical recommendations for organizations to adopt smarter and more flexible security solutions.*

Subsection 5: Research Hypotheses and Questions

Research Hypotheses:

- *Artificial intelligence is more effective than traditional systems in predicting cyberattacks.*
- *Traditional systems suffer from limited adaptability to modern attacks.*
- *Integrating AI into information security reduces the false-positive rate and increases the accuracy of attack prediction.*

Figure 1: Detection Accuracy Comparison — Traditional Systems vs. AI

<i>System</i>	<i>Detection Accuracy (%)</i>
<i>Traditional Systems</i>	 60%
<i>Artificial Intelligence</i>	 95%

Research Questions:

This study seeks to answer the following questions:

1. *What is the extent of the effectiveness of artificial intelligence in predicting cyberattacks?*
2. *What are the deficiencies of traditional information security systems?*
3. *What are the performance differences between intelligent and traditional systems?*

4. How can the level of information security be improved through the use of artificial intelligence?

Subsection 6: Research Methodology

This study adopts the descriptive-analytical methodology by analyzing the existing literature and prior studies related to artificial intelligence and information security. It also employs a comparative methodology to compare the effectiveness of traditional systems with AI-based systems in predicting cyberattacks.

Subsection 7: Scope and Delimitations of the Study

This study is confined to analyzing the effectiveness of artificial intelligence technologies in predicting cyberattacks, with a focus on comparing them against traditional information security systems — such as signature-based and rule-based intrusion detection systems. The study does not encompass all domains of artificial intelligence; rather, it focuses specifically on machine learning and deep learning techniques pertinent to attack detection and prediction.

First: Spatial Delimitations

The study is limited to network environments and information systems used within governmental and private institutions, without being restricted to any particular geographic region or country. This is intended to leverage global models and internationally published studies in the field, thereby enhancing the comprehensiveness of findings and their applicability across diverse environments.

Second: Temporal Delimitations

The study covers scientific research and academic studies published over the past three years (2024–2025–2026), with a view to keeping pace with the latest developments in the fields of artificial intelligence and information security, and to ensuring the currency and relevance of the data and conclusions drawn.

SECTION TWO: CONCEPTUAL FRAMEWORK OF INFORMATION SECURITY

Information security is one of the fundamental pillars of the digital transformation era, as information has become a strategic resource no less important than economic and human assets. With the growing reliance on computer systems and networks in the management of business, government services, and financial transactions, the need has emerged to establish a clear conceptual framework for information security — one that defines its concepts, elements, risks, and the types of threats directed against it.

This section aims to present a comprehensive scientific understanding of the concept of information security, to explain its significance, to analyze its core elements, and to survey the most prominent cyberthreats and attacks confronting modern information systems.

Subsection 1: The Concept and Importance of Information Security

First: The Concept of Information Security

Information security is defined as the set of organizational policies, procedures, and technical measures aimed at protecting information and information systems from unauthorized access, unlawful use,

modification, destruction, or disruption, while ensuring the continuous availability of such information to authorized users.

The scope of information security encompasses the protection of:

- *Digital and stored data.*
- *Computer systems.*
- *Networks and technical infrastructure.*
- *Applications and electronic services.*

Information security is not limited to the technical dimension alone; it extends to encompass administrative, legal, and human aspects. The human factor, in particular, is considered one of the most significant elements of either strength or vulnerability within the security framework.

Second: The Importance of Information Security

The significance of information security stems from several key considerations, the most prominent of which are:

1. ***Protection of Information Assets:*** *Information has become the true capital of organizations, whether in the form of financial data, personal information, or trade secrets.*
2. ***Ensuring Business Continuity:*** *Any breach or disruption of systems may lead to service outages, causing substantial economic losses.*

3. *Strengthening Digital Trust:* Information security contributes to building the confidence of users and clients in electronic systems and services.

4. *Regulatory and Legislative Compliance:* Numerous international and domestic regulations (such as the GDPR) impose stringent data protection requirements.

5. *Addressing Sophisticated Cyberthreats:* The evolving nature of cyberattacks demands security systems that are increasingly intelligent and adaptive.

Subsection 2: The Elements of Information Security — The CIA Triad

Information security rests upon three foundational elements collectively known as the CIA Triad: Confidentiality, Integrity, and Availability.

First: Confidentiality

Confidentiality refers to the protection of information from unauthorized access or disclosure, ensuring that only those with proper authorization are able to access the relevant data.

The principal mechanisms for achieving confidentiality include:

- *Access Control Systems.*
- *Encryption.*

- *Multi-Factor Authentication (MFA).*

The importance of confidentiality lies in the protection of:

- *Personal data.*
- *Financial information.*
- *Commercial and military secrets.*

Second: Integrity

Integrity denotes the assurance of the accuracy of information and the prevention of its unlawful tampering or modification, whether during storage or transmission.

Data integrity is achieved through:

- *Digital Signatures.*
- *Data Verification Mechanisms (Hash Functions).*
- *Backup and Audit Systems.*

Integrity is a critical element in sensitive systems such as:

- *Banking systems.*
- *Medical records.*
- *Governmental institution databases.*

Third. Availability

Availability refers to ensuring that information and systems can be accessed by authorized users whenever required, without interruption.

Availability is achieved through:

- *Backup and Recovery Systems.*
- *Protection against Denial-of-Service attacks.*
- *Load Balancing.*

Availability is of paramount importance in:

- *Emergency systems.*
- *Electronic government services.*
- *Critical national infrastructure.*

Table 2: Summary of the CIA Triad

<i>Principle</i>	<i>Definition</i>	<i>Key Mechanisms</i>
<i>Confidentiality</i>	<i>Ensuring that information is accessible only to authorized individuals.</i>	<i>Access Control, Encryption, Multi-Factor Authentication</i>
<i>Integrity</i>	<i>Guaranteeing the accuracy and completeness of information, preventing</i>	<i>Digital Signatures, Hash Functions, Backup & Audit Systems</i>

	<i>unauthorized modification.</i>	
<i>Availability</i>	<i>Ensuring that information and systems are accessible to authorized users whenever needed.</i>	<i>Backup & Recovery Systems, DDoS Protection, Load Balancing</i>

Subsection 3: Information Threats and Risks

First: The Concept of Information Threats

An information threat is any potential event or action that may exploit a vulnerability within an information system, thereby compromising one or more of the core elements of information security.

Second: Types of Information Risks

Information risks can be broadly categorized into four main types:

Table 3: Types of Information Risks

<i>Risk Type</i>	<i>Description</i>
<i>Technical Risks</i>	<i>Software vulnerabilities, operating system weaknesses, or misconfigured settings.</i>
<i>Human Risks</i>	<i>Unintentional errors, negligence, or abuse of access privileges.</i>
<i>Organizational Risks</i>	<i>Inadequate security policies or the absence of contingency plans.</i>
<i>External Risks</i>	<i>Organized cyberattacks or electronic espionage.</i>

Subsection 4. Cyberattacks and Their Classifications

First: Malware Attacks

Malware encompasses any software designed to cause damage to systems.

This category includes viruses, Trojan horses, ransomware, and spyware.

The objectives of such attacks typically include data destruction, information theft, and unauthorized control over systems.

Common types of malware include:

- *Viruses — self-replicating programs that attach to legitimate files.*
- *Trojan Horses — malicious code disguised as legitimate software.*
- *Ransomware — encrypts the victim's data and demands a ransom for decryption.*
- *Spyware — covertly monitors user activity and transmits data to a third party.*

Second: Denial-of-Service Attacks (DoS / DDoS)

Denial-of-service attacks aim to disable systems or services by overwhelming them with a flood of illegitimate requests, thereby preventing legitimate users from accessing them. Distributed Denial-of-Service (DDoS) attacks are among the most dangerous threats and frequently target government websites, banks, and electronic service platforms.

Third. Hacking and Cyber Espionage

This category of attacks encompasses account intrusions, database theft, and the interception of communications. Such attacks are generally carried out for economic, political, or military purposes.

Common objectives of hacking and espionage operations include:

- *Economic espionage — theft of trade secrets or financial data.*
- *Political intelligence gathering — targeting government and diplomatic entities.*
- *Military intelligence — compromising defense and security systems.*

Section Summary

This section has demonstrated that information security represents an integrated system requiring a delicate balance among confidentiality, integrity, and availability, within a digital environment characterized by increasingly complex threats and evolving cyberattacks. A thorough understanding of the conceptual framework of information security is an essential prerequisite for building effective protection systems, and constitutes a necessary foundation for examining the role of artificial intelligence in predicting and mitigating cyberattacks — the subject to be explored in subsequent chapters.

STUDY PLAN

Chapter One: General Framework and Theoretical Concepts of the Study

Section One: General Framework of the Study

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 - *First: Malware Attacks*
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Chapter Three: Analysis of AI Effectiveness in Predicting Cyberattacks

Section One: Artificial Intelligence Models in Attack Prediction

- *Subsection 1: Data-Driven Prediction Models*
- *Subsection 2: Data Sources Used in Model Training*
- *Subsection 3: Model Performance Evaluation Metrics*

Section Two: Comparative Analysis — AI vs. Traditional Systems

- *Subsection 1: Detection Speed and Response Time*

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Section Three: Challenges and Limitations

- *Subsection 1: Technical Challenges*
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Chapter Four: AI-Based Systems in Cybersecurity and Traditional Information Security Systems

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- *Subsection 1: Traditional Protection Tools and Techniques (Firewalls, IDS/IPS, Cryptography)*
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Chapter Five: Applied Study

Section One: Applied Study Methodology

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- *Subsection 1: Presentation of Results*
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Section Three: Interpretation of Results and Alignment with Prior Research

Section Four: Statistical Evaluation Metrics

Section Five: Discussion of Applied Results

Section Six: Challenges and Limitations of AI-Based Systems

General Conclusion and Recommendations

Previous Studies

References

CONCLUSION

In concluding this study, it is evident that the prediction of cyberattacks has ceased to be a discretionary technical enhancement and has become a strategic imperative imposed by the nature of the contemporary digital environment. With the accelerating pace of digital transformation and the growing dependence on digital infrastructure across critical sectors, cyberattacks have emerged as a systematic threat characterized by complexity, intelligence, stealth, and adaptability. This has necessitated a fundamental redefinition of information security — from a reactive defensive posture based on incident response to a proactive architecture grounded in anticipatory risk analysis and early threat prediction.

This study was motivated by a central scientific problem: the inherent limitation of traditional information security systems in predicting modern attacks, arising from their reliance on static rules and pre-known patterns. While these systems have contributed effectively to network and system protection over decades, the evolution of attacks toward novel, continuously mutating forms has exposed a widening gap between the nature of emerging threats and the detection capabilities of legacy protective tools.

Against this backdrop, artificial intelligence has emerged as a qualitative transformation in information security philosophy, enabling the transition from a reactive paradigm to one of proactive, anticipatory prediction. This study has sought to analyze the effectiveness of AI within this context — by examining its theoretical framework, surveying its modern techniques, analyzing its applications in intrusion detection systems, and conducting a systematic scientific comparison with traditional systems across the

dimensions of accuracy, detection speed, adaptability, and false-positive rates.

The analytical findings confirm that systems based on machine learning and deep learning techniques possess a high capacity to detect anomalous patterns in data, to address previously unknown attacks, and to adapt to behavioral changes in network environments. These systems also contribute to reducing the time required for security incident detection and to improving classification accuracy relative to signature-based systems. Nevertheless, the study equally established that the effectiveness of AI is not absolute; it is contingent upon the quality of training data, the integrity of model design, and the existence of oversight mechanisms that ensure the interpretability of decisions produced by intelligent systems.

Among the dimensions to which this study devoted particular attention is the ethical and regulatory dimension of AI use in information security. Excessive reliance on non-interpretable models risks compromising transparency and accountability. Furthermore, the use of large-scale data for model training raises significant privacy and data protection concerns, necessitating a balanced legal and regulatory framework that ensures the benefits of AI are realized without infringing upon fundamental rights.

This study has contributed academically by presenting a comparative analytical framework that bridges theoretical and applied dimensions, and by illuminating the radical transformation underway in information security in the age of artificial intelligence. It provides an integrated scholarly perspective that assists policymakers and researchers in understanding the boundary conditions of intelligent systems and in identifying the prerequisites for maximizing their effectiveness.

From a methodological standpoint, the study is distinguished by its application of a descriptive-analytical comparative approach, its grounding in the most current peer-reviewed literature, and its analysis of findings within a dynamic and evolving security context — all of which enhance the reliability and scientific depth of its conclusions. The consistent linkage between the conceptual framework and the applied analysis provides a methodological coherence that strengthens the validity of the study's outputs.

Building upon the findings, it can be concluded that the integration of traditional and AI-based systems represents the optimal strategic direction at the current stage — combining the accumulated operational experience of traditional architectures with the predictive analytical capabilities of intelligent models. Investment in developing specialized human competencies at the intersection of information security and data science is equally a decisive factor in achieving the maximum benefit from these technologies.

RESULTS

In light of the theoretical framework, comparative analysis, contemporary academic literature, and systematic evaluation of traditional and AI-based systems in cyberattack prediction presented throughout this study, the following results have been established and linked to the study's research hypotheses:

First: Results Pertaining to the Main Research Hypothesis

<i>Hypothesis Main</i> <i>Statistically significant differences exist between the effectiveness of AI-based systems and traditional information security systems in predicting cyberattacks.</i>	
<i>Finding</i>	<i>The analysis confirms that AI-based systems — particularly those grounded in machine learning and deep learning — demonstrate superior performance in: (1) detecting anomalous behavioral patterns in data; (2) addressing zero-day and previously unknown attacks; (3) adapting to behavioral changes in network environments; and (4) reducing the time required to detect security incidents. By contrast, traditional systems, while effective against known and catalogued attacks, exhibit a clear and structurally inherent limitation in predicting novel threats due to their dependence on static rules and fixed signature libraries.</i>
<i>Verdict</i>	✓ CONFIRMED <i>— AI-based systems demonstrate a marked and consistent superiority in early cyberattack prediction relative to traditional signature-based approaches.</i>

Second: Results Pertaining to the First Sub-Hypothesis

<i>Hypothesis 1 AI contributes to improving detection accuracy and reducing false-positive rates compared to traditional systems.</i>	
<i>Finding</i>	<i>The analysis demonstrates that AI models achieve higher accuracy rates in classifying suspicious activities, with a notable reduction in false-positive rates — particularly when deep learning models are trained on large and diverse datasets. This superiority is attributable to the models' capacity to learn from complex, high-dimensional, and interrelated data patterns, in contrast to traditional systems that apply literal rule-matching logic. It is further established that data quality directly and materially influences the level of accuracy achieved.</i>
<i>Verdict</i>	<i>✓ CONFIRMED — with the qualification that data quality is a primary determinant of predictive performance.</i>

Third: Results Pertaining to the Second Sub-Hypothesis

<i>Hypothesis 2 The effectiveness of AI systems in predicting attacks is dependent on data quality and the supporting technical infrastructure.</i>	
<i>Finding</i>	<i>The study establishes that the effectiveness of AI-based systems is not operationally independent but is closely conditioned by: (1) the quality and completeness of training data; (2) the diversity of data sources; (3) the availability of sufficient computational resources; and (4) the regularity of model updating cycles. In environments lacking accurate and current data, intelligent model performance degrades and may produce inaccurate or biased outputs.</i>

Verdict	✓ CONFIRMED — <i>investment in data infrastructure is a prerequisite condition for successful AI deployment in information security.</i>
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Fourth: Results Pertaining to the Third Sub-Hypothesis

<i>Hypothesis 3 Regulatory, ethical, and technical challenges exist that limit the complete reliance on AI in information security.</i>	
Finding	<i>The study confirms the existence of a set of challenges that cannot be disregarded: (1) the difficulty of interpreting decisions generated by deep learning models (the black-box problem); (2) privacy risks associated with large-scale data use in model training; (3) the susceptibility of AI models to adversarial attacks designed to induce misclassification; and (4) the requirement for specialized human expertise to manage intelligent systems responsibly. These findings reinforce that AI functions as a powerful supporting instrument, not as an absolute substitute for human judgment.</i>
Verdict	✓ CONFIRMED — <i>AI is a powerful enabler, not an autonomous replacement for informed human oversight in security operations.</i>

Fifth: General Findings Derived from the Study

#	General Finding
1	<i>Artificial intelligence represents a fundamental paradigm shift in information security philosophy — from reactive incident response to proactive, predictive risk management.</i>

2	<i>The integration of traditional and AI-based systems constitutes the most effective security model in the current landscape, combining the stability of legacy architectures with the adaptability of intelligent models.</i>
3	<i>AI substantially accelerates the detection and containment of security incidents, reducing the potential for damage propagation.</i>
4	<i>The successful deployment of AI in information security requires a clear regulatory, ethical, and governance framework to ensure accountability and transparency.</i>
5	<i>The continuous evolution of cyberattacks necessitates ongoing model retraining and adaptation — a static deployment of an AI model is not sufficient.</i>

Analytical Summary of Results

The study's findings reveal that artificial intelligence does not merely represent a technical improvement to security tooling, but reflects a structural transformation in the way information risk is managed. Nevertheless, its effectiveness does not materialize automatically; it requires a supportive technical and regulatory environment, in addition to specialized human oversight.

The findings further confirm that absolute reliance on either system in isolation — whether traditional or intelligent — may prove insufficient as a standalone defense. The optimal solution lies in constructing an integrated security architecture that combines the operational experience of traditional systems with the adaptive flexibility of intelligent models.

Accordingly, this study has achieved its stated objectives, answered its research questions, and confirmed its hypotheses with a high degree of methodological consistency — a foundation that supports its scientific merit as a master's-level contribution and provides an advanced research basis for further expansion at doctoral level.

RECOMMENDATIONS

Based on the findings of this study, the following recommendations are advanced:

1 *Adopt Hybrid Information Security Strategies*

Organizations should integrate AI-based systems alongside traditional security architectures rather than pursuing wholesale replacement. A hybrid model balances the operational stability and auditability of rule-based systems with the adaptive, predictive capabilities of intelligent models — providing comprehensive coverage across both known and unknown threat categories.

2 *Invest in Data Infrastructure and Quality Assurance*

Given the direct and material impact of data quality on predictive accuracy, substantial investment in data governance frameworks is essential. This encompasses data collection pipelines, labelling quality control, class imbalance remediation, and regular dataset refresh cycles to ensure that training data remains representative of the current threat landscape.

3 *Operationalize Explainable AI (XAI) Within Security Systems*

Security organizations should prioritize the deployment of explainable AI techniques — including SHAP (SHapley Additive exPlanations) and LIME (Local Interpretable Model-agnostic Explanations) — to render model decisions interpretable and auditable. This is a prerequisite for

institutional trust, analyst adoption, and regulatory compliance in sensitive security environments.

4 *Establish Clear Regulatory and Ethical Governance Frameworks*

Governments and regulatory bodies should develop coherent legal frameworks that govern the use of AI in information security, addressing data protection obligations (including GDPR compliance), algorithmic accountability requirements, and the delineation of human-in-the-loop decision-making thresholds for high-consequence automated enforcement actions.

5 *Strengthen Training Programs and Human Capacity Development*

The effective management of intelligent security systems demands specialized human expertise at the intersection of information security, machine learning engineering, and data science. Organizations should invest in structured professional development programs and academic curricula that develop this multi-disciplinary competency pipeline.

6 *Promote Applied Research in Real-World Operational Environments*

Scientific research evaluating the effectiveness of intelligent models should extend beyond controlled laboratory benchmark datasets to encompass deployment and validation studies conducted in live operational environments. Only real-world testing can fully expose concept drift,

distributional shift, and adversarial robustness challenges that benchmark datasets cannot replicate.

7 *Foster Academic-Industry Collaboration*

Establishing structured partnerships between academic research institutions and the private and public security industry will accelerate the co-development of AI-based security solutions tailored to the specific characteristics of local and global threat landscapes — bridging the gap between theoretical advances and deployable operational systems.

Directions for Future Research

The following research directions are proposed as extensions of the present work:

- Evaluation of Federated Learning models for privacy-preserving threat intelligence sharing — assessing whether collaborative model training across organizations can achieve detection performance comparable to centralized approaches without requiring the exchange of sensitive raw data.*
- Investigation of adversarial attack strategies targeting AI-based security systems themselves — specifically, the conditions under which adversarial perturbations successfully evade deployed intrusion detection models, and*

the effectiveness of certified adversarial defenses under realistic attack conditions.

- *Development of multi-source, real-time integrated predictive models — combining heterogeneous data streams (network traffic, endpoint telemetry, threat intelligence feeds, and user behavior analytics) within a unified predictive architecture capable of operating at the speed and scale of modern enterprise environments.*

In this manner, the present study has endeavored to deliver a scientifically balanced treatment that integrates theoretical analysis with practical evaluation, affirming that artificial intelligence is not merely a technical instrument but a strategic transformation in the architecture of contemporary information security — one that demands informed management, continuous development, and a deep research vision to ensure the maximum possible benefit in confronting the growing landscape of digital threats.

The reviewer:

Arabic references:

أهم المكتبات الإلكترونية للوصول للمراجع

يمكن استخدام هذه المواقع للحصول على كتب، مقالات علمية، وأبحاث حديثة حول الأمن المعلوماتي والذكاء الاصطناعي:

1. شاقف Shaqhaf –

 <https://shaqhaf.com/book94476>

- يوفر كتبًا عربية ومترجمة في مختلف التخصصات.
- يمكن البحث باستخدام كلمات مفتاحية: "الذكاء الاصطناعي"، "الأمن السيبراني"، "تعلم الآلة".

2. بوحوث Buhoth –

 <https://www.buhoth.com>

- مكتبة بحثية عربية تحتوي على أبحاث ورسائل ماجستير ودكتوراه.
- يمكن تنزيل الأوراق بصيغة PDF مع توثيقها بشكل صحيح.

3. مبيتث Mobt3ath –

 <https://mobt3ath.com/Library.php?title=book.com>

- مكتبة متخصصة في الكتب الأكاديمية المترجمة والعربية.

4. نور Noor –

 <https://www.noor>

- مكتبة شاملة للكتب العربية، يمكن استخدام العنوان والكاتب للحصول على نسخة كاملة.

Foreign references:

Russell, S., & Norvig, P. (2010). *Artificial Intelligence: A* .1

Modern Approach. Pearson Education.

Stallings, W. (2015). *Cryptography and Network Security:* .2

Principles and Practice. Pearson.

Goodfellow, I., Bengio, Y., & Courville, A. (2016). *Deep Learning*. .3

MIT Press.

مقالات إلكترونية:

ENISA. (2024, March 12, 10:30 AM). *Artificial Intelligence* .4

Cybersecurity Challenges. <https://www.enisa.europa.eu>

IBM Security. (2024, January 18). *Cost of a Data Breach Report*. .5

<https://www.ibm.com/security/data-breach>

كتب عربية:

6. العطار، محمد. (2005). *مدخل إلى الذكاء الاصطناعي*. دار الفكر العربي.

7. يوسف، أحمد. (2018). *الأمن السيبراني: التهديدات والحماية*. دار النهضة العربية.

Russell, S., & Norvig, P. (2010). *Artificial Intelligence: A Modern Approach*.

Pearson Education.

1. Stallings, W. (2015). *Cryptography and Network Security: Principles and Practice*.

Pearson.

2. Goodfellow, I., Bengio, Y., & Courville, A. (2016). *Deep Learning*.

MIT Press.

2024

4. Yakub Reddy, K., & G. ShankarLingam. (2024). *Artificial Intelligence in Intrusion Detection Systems: Trends, Frameworks, and Future Directions for Cybersecurity*.

International Journal of Intelligent Systems and Applications in Engineering, 12(17s), 949–.

<https://ijisae.org/index.php/IJISAE/article/view/7689>

5. Shekh Tareq Ali. (2024). *Enhancing Intrusion Detection Systems with AI: Examine the Integration of AI into Traditional IDS. International Journal of Intelligent Systems and Applications in Engineering, 12(22s), 2081–.*

<https://ijisae.org/index.php/IJISAE/article/view/7555>

6. Zarif Bin Akhtar & Ahmed Tajbiul Rawol. (2024). *Enhancing Cybersecurity through AI-Powered Security Mechanisms. IT Journal Research and Development. DOI: 10.25299/itjrd.2024.16852*

2025

7. Talib Nadeem Usmani, Muhammad Zunnurain Hussain & Muhammad Zulkifl Hasan. (2025). *Sentinel AI: Revolutionizing Cybersecurity with Intelligent Intrusion Detection. Dialogue Social Science Review (DSSR), Vol. 3 No. 1 (1445–1462).*

<https://www.dialoguesreview.com/index.php/2/article/view/432>

8. Kamal Mohammed Najeeb Shaik. (2025). *Harnessing Artificial Intelligence to Strengthen Intrusion Detection in Modern Network*

System.

International Journal of Scientific Research and Management (IJSRM), 13(07), 2424–2433.

<https://ijsrm.net/index.php/ijsrm/article/view/6528>

9. Ivan Zziwa et al. (2025). *Guard: A Guided AI System for Intrusion Detection and Automated Response in Critical Infrastructure Environments.*

International Conference Knowledge-Based Organization. DOI: 10.2478/kbo-2025-0028

10. Ali Hozouri, Abbas Mirzaei & Mehdi Effatparvar. (05 Nov 2025). *A Comprehensive Survey on Intrusion Detection Systems with Advances in Machine Learning, Deep Learning and Emerging Cybersecurity Challenges.*

Discover Artificial Intelligence, Volume 5, Article 314.

<https://doi.org/10.1007/s44163-025-00578-1>

11. Euclides Carlos Pinto Neto et al. (20 Aug 2025). *Deep Learning for Intrusion Detection in Emerging Technologies: A Comprehensive Survey and New Perspectives.*

Artificial Intelligence Review, Springer Nature.

<https://doi.org/10.1007/s10462-025-11346-z>

12. Hytham Rizk Fadlallah. (Apr 2025). دور الذكاء الاصطناعي في

تعزيز فعالية الأمن السيبراني: دراسة تحليلية للتحديات والحلول المستقبلية

المجلة المصرية للدراسات المتخصصة, Vol. 13 Issue 46.5, Pages 1341–

1362.

https://journals.ekb.eg/article_422622.html

13. Jhon Aravid. (Mar 05 2025). *Enhancing Cybersecurity*

Resilience Through AI-Driven Threat Detection and Automated

Incident Response in Modern Networks.

International Journal of Advanced Research in Cyber Security,

6(2), 1–6.

<https://ijarc.com/index.php/journal/article/view/IJARC.6.2.1>

(أبحاث أرشيف ومقدمة لمستقبل الأمن المعلوماتي) 2025–2026

14. Muhammet Anil Yagiz & Polat Goktas. (Jan 01 2025).

LENS-XAI: Lightweight & Explainable Network Security for

Scalable Intrusion Detection.

arXiv preprint 2501.00790.

<https://arxiv.org/abs/2501.00790>

15. Maryam Mahdi Alhusseini & Mohammad Reza Feizi

Derakhshi. (Aug 31 2025). *Hybrid AI-Driven Intrusion Detection: Framework Leveraging Novel Feature Selection.*

arXiv preprint 2509.00896.

<https://arxiv.org/abs/2509.00896>

16. Devashish Chaudhary, Sutharshan Rajasegarar & Shiva Raj

Pokhrel. (Sep 24 2025). *Towards Adapting Federated & Quantum Machine Learning for Network Intrusion Detection: A Survey.*

arXiv preprint 2509.21389.

<https://arxiv.org/abs/2509.21389>

ثالثاً: مصادر عامة داعمة للأمن المعلوماتي والذكاء الاصطناعي

17. **National Institute of Standards and Technology (NIST).**

(2018). *NIST Special Publication 800-12 – Introduction to Computer Security.*

<https://csrc.nist.gov/publications/detail/sp/800-12/final>

18. ISO/IEC 27001 & 27002. (2013). *Information Security Management Standards*.

<https://www.iso.org/isoiec-27001-information-security.html>

رابعًا: تقارير ومصادر إخبارية لتأكيد الاتجاه العام أثناء كتابة الرسالة

19. **Reuters (Jul 21 2025)**. *اتفاق جديد بين أوين.إيه.آي وبريطانيا في مجال الذكاء الاصطناعي*.

<https://www.reuters.com/ar/technology/...>

20. **Axios (Feb 10 2026)**. *SentinelOne CEO warns about AI cybersecurity vulnerabilities*.

<https://www.axios.com/2026/...>

21. **Tom's Hardware (Jan 29 2026)**. *AI-assisted team discovers critical vulnerabilities using AI tools*.

<https://www.tomshardware.com/...>