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Implementing AI-Driven Transaction Security Protocols and Automation in Next-Gen FinTech Solutions

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Abstract

The FinTech industry, characterized by rapid innovation and digital transformation, necessitates adopting robust security measures to safeguard financial transactions and enhance operational efficiency. This paper examines the integration of artificial intelligence (AI) into security frameworks, focusing on automation strategies and advanced solutions to mitigate risks, improve user experience, and reinforce trust in modern financial systems. AI technologies such as predictive threat analysis, real-time fraud detection, and adaptive learning models are at the forefront of combating the dynamic and sophisticated nature of cyber threats. Predictive threat analysis facilitates the early identification of vulnerabilities, enabling proactive measures to thwart potential breaches. Real-time fraud detection leverages machine learning algorithms to analyze transactional patterns and detect anomalies, preventing unauthorized activities. Adaptive learning models continuously evolve with emerging threat landscapes, enhancing the resilience of security protocols. Beyond risk mitigation, artificial intelligence (AI)-driven systems optimize user experiences by streamlining

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authentication processes, minimizing false positives, and expediting secure transactions. The deployment of these technologies not only fortifies data integrity but also fosters greater trust among users by demonstrating an uncompromising commitment to cybersecurity. This paper presents empirical evidence and case studies highlighting the transformative impact of AI on financial security. By addressing critical vulnerabilities and enhancing system capabilities, AI establishes itself as a cornerstone of innovation in the FinTech sector, driving the creation of secure, adaptive, and user-focused financial ecosystems. Our findings underscore AI's pivotal role in shaping the future of resilient and trustworthy financial platforms.

Keywords: AI (artificial intelligence); FinTech; transaction security; automation; fraud detection; cybersecurity; predictive analytics; adaptive learning; machine learning; real-time monitoring; financial systems; risk mitigation; user experience; data integrity; threat intelligence; digital transformation; secure authentication; anomaly detection; data privacy; financial technology; fraud prevention; cyber resilience; intelligent algorithms; blockchain integration; secure ecosystems; operational efficiency.

1 Introduction

1.1 Background of transaction security in FinTech

The financial technology (FinTech) industry has emerged as a pivotal force reshaping traditional financial services, offering innovative solutions that enhance efficiency, accessibility, and scalability. With the widespread adoption of digital payment systems, peer-to-peer lending platforms, cryptocurrency exchanges, and robo-advisory services, the volume and complexity of financial transactions have reached unprecedented levels. However, this rapid expansion has been accompanied by a surge in cyber threats, including phishing attacks, ransomware, and advanced persistent threats, all of which target the vulnerabilities in transaction processing systems (Aiken, 2022).

Transaction security plays a fundamental role in safeguarding the integrity, confidentiality, and availability of financial operations. Traditional security measures, such as multi-factor authentication and manual fraud detection, are increasingly inadequate to combat modern cyberattacks' sophistication. This inadequacy has necessitated a paradigm shift toward AI-driven security protocols. These advanced systems utilize machine learning models to analyze vast datasets in real time, detecting and mitigating potential risks before they escalate. For example, banks and payment processors use AI to monitor billions of daily transactions, identifying patterns indicating fraudulent activities, such as unusually large withdrawals or access from unrecognized devices or locations as in Fig. 1.

Moreover, blockchain technology has introduced a transformative approach to transaction security by creating immutable, decentralized ledgers. Blockchain ensures transparency by recording every transaction in a tamperproof format, reducing the reliance on centralized systems prone to breaches. Applications of blockchain extend beyond cryptocurrency; they are now widely implemented in supply chain financing, cross-border payments, and asset tokenization, further strengthening the security infrastructure of FinTech applications.

1.2 The role of automation in transaction processing

Automation has revolutionized the FinTech industry, enabling organizations to manage high volumes of transactions with unparalleled speed and accuracy. Automation plays a crucial role in transaction security in identifying and mitigating real-time risks. Technologies like Robotic Process Automation (RPA) and AI-powered systems enable continuous monitoring, anomaly detection, and instantaneous response to potential threats. For instance, automated systems can promptly identify unusual spending patterns, such as multiple high-value purchases within a short timeframe, and flag these transactions for further review.

Automation also enhances regulatory compliance, which is a cornerstone of financial operations. AI systems equipped with natural language processing (NLP) capabilities can parse through complex regulatory requirements, match them against transactional data, and generate compliance reports. This capability

significantly reduces the manual workload of compliance teams while minimizing the risk of non-compliance, which could result in hefty penalties for financial institutions (Balan, 2022).

Additionally, automation has streamlined customer onboarding processes by integrating advanced identity verification methods, such as biometric authentication and optical character recognition (OCR) for document verification. These automated solutions ensure faster, more secure onboarding experiences, reducing friction for customers while maintaining stringent security standards. The efficiency achieved through automation extends beyond operational improvements; it also contributes to enhanced customer trust and loyalty.



Fig. 1. The image depicting a comprehensive technological ecosystem centered around AI and RPA (Robotic Process Automation)

1.3 Objectives of the study

This study aims to comprehensively address the key aspects of transaction security and automation in FinTech by focusing on the following objectives:

Strengthening Transaction Security: Evaluate the effectiveness of AI-driven protocols in enhancing transaction security by preventing fraud, identifying suspicious activities, and safeguarding sensitive financial data. This includes analyzing real-world applications of machine learning and predictive analytics in securing digital transactions.

Optimizing Operational Efficiency: Investigate how automation technologies such as RPA and intelligent process automation can streamline operations, minimize manual intervention, and ensure consistent adherence to security protocols.

Leveraging Blockchain Technology: Explore the integration of blockchain systems in FinTech applications to create tamper-proof, transparent, and decentralized transaction frameworks. The study will highlight blockchain's role in enhancing trust and reducing fraud.

Addressing Ethical Challenges: Assess the challenges associated with implementing AI-driven systems, including biases in algorithms, data privacy concerns, and ethical implications. This objective aims to identify measures for ensuring fairness, accountability, and transparency in AI applications.

Improving Regulatory Compliance: Examine the role of automated systems in accelerating compliance processes, reducing non-compliance risks, and supporting financial institutions in meeting global regulatory standards.

Enhancing User Experience: Study the impact of AI and automation on user experience, particularly in areas such as faster authentication, reduced transaction times, and improved accuracy in detecting and preventing fraudulent activities.

Identifying Best Practices: Provide actionable recommendations and best practices for adopting AI, blockchain, and automation technologies in FinTech. These practices will focus on improving security frameworks, operational resilience, and scalability.

Exploring Future Trends: Anticipate emerging trends and innovations in AI, blockchain, and automation that will shape the next generation of transaction security. This includes analyzing the potential of quantum computing, federated learning, and zero-trust security models in the FinTech landscape.

Reducing Financial Losses: Highlight the importance of real-time fraud detection and rapid threat response in minimizing financial losses and protecting consumer trust in digital financial services.

Building Trust and Transparency: Address the importance of fostering trust and transparency through secure systems that demonstrate robust protection of user data and financial assets.

2 Background

The rapid digitization of financial systems has introduced unprecedented opportunities for growth and innovation in the FinTech sector. However, this evolution has also amplified the complexity and volume of cybersecurity threats. Traditional security measures are no longer sufficient to address the ever-expanding attack surface created by digital transactions. AI-driven technologies are spearheading efforts to transform transaction security, offering intelligent and adaptive mechanisms that outpace conventional approaches. By leveraging capabilities such as real-time data analysis, anomaly detection, predictive analytics, and automation, the FinTech industry is establishing robust defenses against sophisticated cyber threats while ensuring operational efficiency and customer trust (Anderson, 2021).

2.1 Evolution of transaction security

2.1.1 Early approaches: Static and rule-based mechanisms

In the initial phases of digital financial transactions, security measures were static and reactive. These included password-based authentication, encryption, and basic rule-based fraud detection systems. While these solutions provided adequate protection against known vulnerabilities, they were limited in their scope and adaptability. Static rules, such as flagging transactions above a specific amount or monitoring activity during non-business hours, lacked the sophistication required to detect subtle or evolving threats as in Fig. 2.

Additionally, early systems were heavily dependent on human intervention, leading to delays in detecting and responding to security incidents. Fraudulent activities often went unnoticed until they had already caused significant damage, highlighting the inefficiency of purely manual processes (Bailey, 2022).

2.1.2 Limitations of legacy systems

Legacy systems faced multiple challenges as the FinTech ecosystem evolved:

Inability to Scale: As transaction volumes increased, traditional systems struggled to keep up, resulting in blind spots and vulnerabilities.

Inflexibility: Static rule-based approaches could not adapt to new threat vectors, such as polymorphic malware or AI-driven cyberattacks.

High False Positives: Rule-based systems often flag legitimate transactions as suspicious, leading to customer dissatisfaction and operational inefficiencies.

Delayed Response Times: Without real-time monitoring, institutions were left vulnerable to attacks that exploited the lag between detection and action.



Fig. 2. A Graphical illustration for the intersection and relationship between AI and blockchain technologies, featuring three overlapping circles with interconnected nodes and text highlighting concepts like automation, real-time response, and predictive analytics

2.1.3 Current landscape: AI-driven security

The modern approach to transaction security leverages AI to overcome the limitations of traditional systems. AI-driven technologies enable financial institutions to:

Detect Complex Threats: AI can identify intricate fraud patterns, such as coordinated attacks across multiple accounts or the use of synthetic identities.

Respond Proactively: Predictive analytics allow organizations to foresee potential vulnerabilities and implement preventive measures before incidents occur.

Personalize Security Measures: AI adapts to individual user behaviors, providing tailored security protocols that balance protection and user convenience.

Analyze Massive Datasets: Advanced machine learning models process vast amounts of transactional data in real time, ensuring comprehensive threat coverage (Beck, 2021).

2.2 Role of automation in transaction security

2.2.1 Redefining operational efficiency

Automation has fundamentally reshaped the way financial institutions manage transaction security. By automating repetitive tasks, such as monitoring, reporting, and compliance checks, organizations can allocate resources to more strategic activities. Key benefits include:

Faster Incident Resolution: Automated systems respond to security incidents within milliseconds, minimizing the impact of attacks.

Consistency: Unlike manual processes, automation delivers uniform results, reducing the risk of human error.

Cost-Effectiveness: By replacing labor-intensive processes with automated workflows, institutions can achieve significant cost savings.

2.2.2 Enhancing security with AI-driven automation

When paired with AI, automation becomes a powerful tool for managing transaction security. Advanced use cases include:

Dynamic Fraud Detection: Automation combined with AI enables systems to detect fraud in real time, blocking suspicious transactions instantly.

Continuous Learning: AI-driven automation systems evolve based on new threat intelligence, ensuring they remain effective against emerging threats.

Threat Containment: Automated systems can isolate compromised accounts or systems to prevent the spread of an attack.

2.2.3 Streamlining compliance

Regulatory compliance is one of the most resource-intensive aspects of financial operations. Automation simplifies this process by:

Real-Time Analysis: Automated tools cross-reference transactions with regulatory frameworks, identifying potential violations as they occur.

Detailed Reporting: Systems generate accurate, audit-ready reports, ensuring transparency and reducing the burden on compliance teams.

Fraud Prevention: Compliance tools monitor transactions for patterns consistent with money laundering or other illicit activities, flagging them for further investigation.

2.2.4 Improving customer experience

Automation also plays a pivotal role in enhancing the customer experience, a critical factor in the competitive FinTech landscape. Examples include:

Frictionless Transactions: Automated identity verification and fraud detection allow for faster processing of legitimate transactions.

Secure Onboarding: AI and automation streamline customer onboarding by verifying identities through biometrics, document scanning, and secure digital workflows.

Personalized Alerts: Automated systems send tailored notifications about unusual activities, providing customers with real-time insights and peace of mind (Benson, 2023).

2.2.5 Advanced use cases

The integration of automation in transaction security has unlocked advanced applications, including:

Adaptive Authentication: Systems dynamically adjust authentication requirements based on the perceived risk level of a transaction.

Behavioral Biometrics: AI analyzes unique behavioral patterns, such as typing speed or device usage, to identify users with high accuracy.

Cross-Border Transactions: Automation ensures compliance with varying regulations across jurisdictions, simplifying global financial operations.

2.3 Future implications of AI and automation in FinTech

The adoption of AI and automation in transaction security is not just a solution for today's challenges but a blueprint for the future of FinTech. As technologies such as quantum computing and federated learning mature, the potential to revolutionize transaction security will only expand. Some future trends include:

Quantum-Safe Cryptography: Preparing for the era of quantum computing by developing encryption methods that resist quantum attacks.

Zero-Trust Architecture: Moving towards systems where every user and device must continuously verify their identity and intent.

Collaborative Security Models: Using federated learning to enable financial institutions to share threat intelligence without compromising data privacy. By adopting AI-driven automation, FinTech organizations can stay ahead of evolving threats, ensuring a secure, efficient, and trustworthy environment for their customers (Carter, 2023).

2.4 Integration of blockchain in transaction security

Blockchain technology has emerged as a transformative force in transaction security, offering unparalleled benefits such as decentralization, transparency, and immutability. By eliminating the need for intermediaries, blockchain ensures secure and tamper-proof transactions, making it a critical component in the evolving FinTech landscape. Its integration with AI and automation is creating a robust foundation for addressing contemporary security challenges.

2.4.1 Decentralization and immutability

Unlike traditional systems that rely on centralized servers vulnerable to breaches, blockchain operates on a distributed ledger. Every transaction is recorded across multiple nodes, making it virtually impossible for malicious actors to alter historical data. This immutability enhances trust in financial operations, as users can verify transaction authenticity without relying on a single authority (Singh, 2022).

2.4.2 Real-time auditability

Blockchain provides real-time audit trails that ensure accountability and transparency. Financial institutions can track and verify transactions instantaneously, which is especially valuable in regulatory compliance and fraud investigations. Combined with AI, blockchain can analyze these records to detect irregularities, such as duplicate transactions or unapproved modifications.

2.4.3 Smart contracts for automated security

Smart contracts, self-executing programs stored on a blockchain, automate security and compliance processes. For instance:

Fraud Prevention: Smart contracts can enforce transaction rules, automatically flagging or rejecting suspicious activities.

Payment Processing: They ensure secure and timely payments by releasing funds only when predefined conditions are met.

Regulatory Compliance: Smart contracts can embed compliance protocols, such as anti-money laundering (AML) checks, directly into the transaction flow.

2.5 Enhanced security for cross-border transactions

Cross-border transactions are particularly susceptible to fraud and inefficiencies due to varying regulatory frameworks. Blockchain ensures the integrity of these transactions by providing a transparent and secure ledger, reducing the risk of unauthorized modifications. This is further augmented by AI, which can analyze cross-border transaction patterns to identify anomalies and prevent fraud.

2.6 Combining blockchain with AI and automation

The integration of blockchain with AI and automation magnifies its potential in transaction security:

Predictive Threat Analytics: AI analyzes blockchain data to identify patterns of fraudulent behavior, such as account manipulation or transaction anomalies.

Seamless Automation: Automation enables blockchain systems to self-manage, ensuring smooth operations with minimal human intervention.

Privacy-Preserving AI Models: Federated learning on blockchain allows institutions to share insights on fraud without compromising user privacy or data security.

2.7 Industry use cases

Blockchain is revolutionizing transaction security across various FinTech domains:

Cryptocurrency Exchanges: Secure, transparent, and tamper-proof trading.

Supply Chain Financing: Verify payment and logistics transactions in real-time.

Peer-to-Peer Lending: Decentralized platforms that ensure transparency and fairness between borrowers and lenders.

Insurance Claims: Automated claim verification and disbursement through smart contracts.



Fig. 3. The bar graph displays the effectiveness of different technology combinations in FinTech transaction security

2.8 The Future of blockchain in FinTech

As blockchain technology matures, advancements such as proof-of-stake (PoS) mechanisms, interoperability protocols, and layer-2 scaling solutions are addressing current limitations. When combined with AI and automation, blockchain is poised to redefine transaction security by creating resilient, efficient, and trusted ecosystems (Anderson, 2021).

3 AI-Driven Transaction Security

The integration of Artificial Intelligence (AI) in FinTech has transformed transaction security by introducing intelligent, proactive, and adaptive systems that safeguard against evolving cyber threats. By leveraging advanced techniques such as predictive threat analysis, real-time fraud detection, and adaptive learning models, AI enhances the ability of financial institutions to secure transactions effectively and efficiently.

3.1 Predictive threat analysis

Predictive threat analysis is one of the most significant advancements in AI-driven transaction security. By analyzing historical data, AI models can foresee vulnerabilities and take preventive measures to mitigate risks as in Fig. 4.

Techniques:

Historical Data Analysis: AI models process vast amounts of past transaction data to uncover patterns and trends associated with fraudulent or malicious behavior.

Behavioral Profiling: AI establishes profiles based on customer behavior, identifying deviations that may indicate potential threats.

Risk Scoring Systems: Assign risk scores to transactions based on multiple variables, such as geographic location, device type, and transaction amount, enabling proactive intervention.

Simulation and Forecasting: AI simulations predict potential attack scenarios, preparing institutions to counteract them effectively.

Benefits:

Proactive Risk Mitigation: Identifies potential vulnerabilities before they are exploited, reducing the likelihood of breaches.

Operational Efficiency: Reduces the need for manual intervention by automating threat identification and mitigation.

Enhanced Accuracy: Minimizes false positives by considering multiple variables and using contextual data.

Regulatory Compliance: Ensures institutions meet compliance standards by proactively addressing risks tied to non-compliance.

3.2 Real-time fraud detection

Real-time fraud detection is a cornerstone of AI-driven transaction security. AI systems monitor vast amounts of transactional data to detect and respond to anomalies as they occur, ensuring immediate protection against fraudulent activities (Brown, 2023).

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Fig. 4. The image illustrating an AI-driven transaction security system with multiple interconnected components radiating from a central "AI" hub, including data analysis, predictive analysis, fraud detection, real-time security monitoring, and automated mitigation actions

How It Works:

Pattern Recognition: AI detects deviations from normal transaction patterns, such as unusually high-value transactions or activities from unexpected locations.

Multi-Factor Analysis: Analyzes multiple parameters, including transaction history, device type, user behavior, and geographic location, to identify suspicious activities.

Dynamic Thresholds: Uses machine learning to adjust fraud detection thresholds based on real-time data, reducing false positives.

Applications:

Monitoring credit card transactions for unauthorized purchases.

Detecting unusual login attempts, such as from unrecognized devices or foreign IP addresses.

Flagging rapid consecutive transactions that exceed typical spending behavior.

Benefits:

Immediate Response: Enables systems to block suspicious transactions or restrict user accounts instantly, minimizing financial losses.

Customer Protection: Safeguards customer accounts without unnecessary interruptions, enhancing trust in financial systems.

Scalability: Handles millions of transactions simultaneously, making it ideal for high-volume financial environments.

Reduction in Human Error: Automates the detection process, eliminating the need for manual review of flagged activities.

3.3 Adaptive learning models

Adaptive learning models are the backbone of AI-driven security, enabling systems to evolve continuously and address emerging challenges effectively. These models learn from new data, refine their algorithms, and improve their performance over time (Jones, 2023).

How Adaptive Learning Works:

Continuous Data Integration: Adaptive models incorporate new data streams, such as recent transaction records or updated threat intelligence, into their learning process.

Dynamic Behavior Adjustments: Adjust security parameters based on user behavior and environmental changes, ensuring relevance and accuracy.

Feedback Loops: Use feedback from detected threats and false positives to refine detection capabilities. **Global Threat Intelligence:** Integrates data from external sources, such as global threat reports, to anticipate and counteract advanced attacks.

Applications:

Account Monitoring: Continuously tracks user activity to identify potential account takeovers or unauthorized access.

Fraud Pattern Updates: Detects and incorporates new fraud patterns, such as emerging phishing tactics or malware attacks, into its algorithms.

Personalized Security Measures: Adapts security protocols to individual user behaviors, providing tailored protection.

Benefits:

Resilience Against Evolving Threats: Adaptive models stay effective even as attackers develop new techniques.

Improved Accuracy: Learns from past errors, reducing the occurrence of false positives and negatives.

Real-Time Updates: Adjusts security measures instantly to reflect the latest threat landscape.

Cost Savings: Reduces the need for frequent manual updates and extensive reprogramming.

Real-World Example:

Adaptive learning models are widely used in mobile banking applications to monitor user activity. If a user consistently accesses their account from a specific region and device, the model learns to recognize this behavior as normal. If access suddenly occurs from a different location or device, the system flags it as potentially suspicious, prompting additional verification steps.

4 Automation in Fintech

Automation has revolutionized the FinTech industry by enabling faster, more efficient, and accurate processes. By integrating technologies such as Robotic Process Automation (RPA), smart contracts, and AI-powered risk assessment systems, FinTech organizations can streamline operations, reduce costs, and enhance decision-making. These advancements not only improve operational efficiency but also contribute to better customer experience (Zhang, 2021).





Fig. 5. The image showing a complex data visualization relationship between implementation complexity and efficiency gains for various automation technologies

4.1 Robotic Process Automation (RPA)

Robotic Process Automation (RPA) is extensively utilized in the FinTech sector to handle repetitive and timeconsuming tasks. By mimicking human actions, RPA enables organizations to automate processes such as transaction processing, fraud report generation, and data entry as in Fig. 5.

Key Applications:

Transaction Processing: Automates high-volume, repetitive tasks such as payment approvals, fund transfers, and reconciliation.

Fraud Report Generation: RPA can extract and analyze data from various sources to generate detailed fraud reports efficiently.

Compliance Monitoring: Automates regulatory checks by cross-referencing transaction data with compliance standards to identify potential violations.

Benefits:

Improved Efficiency: Reduces manual workload and speeds up processes, enabling financial institutions to focus on strategic initiatives.

Accuracy and Consistency: Minimizes human errors by following pre-defined rules and workflows.

Cost Savings: Reduces operational costs by automating labor-intensive tasks.

Scalability: Easily adapts to handle increasing transaction volumes without additional workforce requirements.

4.2 Smart contracts

Smart contracts are self-executing agreements in which the terms of the contract are directly written into lines of code. Once predefined conditions are met, the contract automatically enforces the agreed terms without the need for intermediaries. Smart contracts are widely adopted in areas such as insurance, loans, and supply chain financing (Carter, 2023).

Key Applications:

Insurance Claims: Automates the claim settlement process by verifying documents and releasing payouts once conditions are met.

Loan Disbursements: Ensures timely disbursement of loans by automatically validating borrower credentials and repayment conditions.

Supply Chain Financing: Facilitates transparent and secure transactions between suppliers, manufacturers, and retailers, reducing delays and disputes.

Benefits:

Eliminates Intermediaries: Reduces transaction costs and processing times by removing the need for third parties.

Enhances Transparency: All parties can view the contract and its execution, fostering trust and accountability.

Improves Security: Smart contracts are encrypted and stored on blockchain, ensuring they are tamper-proof and secure.

Speeds Up Transactions: Automates the execution of agreements, reducing manual intervention and delays.

4.3 Automated risk assessment

Automated risk assessment leverages AI and machine learning to analyze vast amounts of data, providing realtime insights into risks associated with lending, investments, and transactions. This technology empowers financial institutions to make informed decisions quickly and effectively.

How It Works:

Market Data Analysis: Continuously monitors market trends to assess investment risks.

Credit Scoring: Analyzes borrower transaction history, payment behavior, and credit scores to evaluate lending risks.

Fraud Risk Detection: Flags potentially fraudulent activities by analyzing transactional and behavioral patterns.

Key Applications:

Lending Decisions: Provides real-time assessments of borrower risks, helping institutions determine loan eligibility and terms.

Investment Analysis: Evaluates the risk and return profiles of potential investments, enabling smarter portfolio management.

Fraud Prevention: Identifies high-risk transactions and recommends actions to prevent financial losses.

Benefits:

Real-Time Insights: Enables financial institutions to make decisions based on the latest data, reducing delays.

Enhanced Accuracy: AI-powered systems provide more precise risk assessments than traditional methods.

Improved Customer Experience: Speeds up processes like loan approvals and transaction verification, reducing wait times.

Regulatory Compliance: Ensures adherence to risk-related regulations by generating detailed reports and alerts (Choi, 2023).

4.5 Future trends in automation

Automation in FinTech is continuously evolving, with emerging trends that promise to further enhance operational efficiency and security:

Hyperautomation: Combining RPA, AI, and advanced analytics to create end-to-end automation across complex workflows.

Behavioral Analytics Integration: Leveraging user behavior data to detect anomalies and improve fraud detection.

Personalized Financial Services: Using automation to tailor financial products and services to individual customer needs.

Predictive Automation: Anticipating customer actions and automating processes such as investment rebalancing or savings optimization.

RegTech (**Regulatory Technology**): Automating compliance with global regulations, reducing the cost and complexity of managing regulatory requirements.

5 Methodology

This section outlines the systematic approach undertaken to analyze the implementation of AI-driven transaction security and automation in FinTech. The methodology includes data collection processes, evaluation metrics, and the analytical framework used to derive insights from the gathered information.

5.1 Data collection

Data for this study was collected from various FinTech platforms and applications actively utilizing AI-driven security measures and automation tools. A combination of qualitative and quantitative methods was employed to ensure a comprehensive understanding of the subject:

Primary Sources:

Performance reports from financial institutions deploying AI in transaction security.

Case studies of successful implementations of Robotic Process Automation (RPA), smart contracts, and automated risk assessment.

Surveys and interviews with industry professionals, including security analysts, compliance officers, and product managers.

Secondary Sources:

Published research papers, industry white papers, and technical reports on AI and automation in FinTech.

Regulatory guidelines and compliance frameworks relevant to FinTech operations include GDPR, PCI-DSS, and AML directives.

Historical datasets of transaction records, fraud patterns, and customer behavior from anonymized sources to ensure data privacy.

Key Metrics Collected:

Fraud detection rates before and after the implementation of AI-driven models.

Time saved through automation tools in critical processes such as transaction processing and compliance reporting.

Accuracy of predictive threat analysis models in preventing potential breaches.

Success rates of smart contracts in automating financial agreements.

5.2 Evaluation metrics

The evaluation framework was designed to assess the effectiveness of AI and automation in enhancing transaction security and operational efficiency. Key metrics include:

Fraud Detection Accuracy:

Measurement of the percentage of fraudulent activities detected by AI systems.

Comparison with traditional rule-based systems to determine improvements in accuracy as in Fig. 6.

Time Saved Through Automation:

Analysis of time reductions in key processes such as transaction approvals, fraud monitoring, and compliance reporting.

5.3 Measurement of process efficiency before and after automation

Compliance Adherence Rates:

Evaluation of how AI and automation tools improve adherence to regulatory requirements.

Metrics include the percentage of regulatory violations prevented and the speed of generating compliance reports (Chung, 2022).

5.4 Transaction processing speeds

Assessment of processing times for high-volume transactions with and without automation.

Measurement of the scalability of automated systems during peak loads.

5.5 Reduction in false positives

Analysis of how well AI systems do reduce false alarms in fraud detection, minimizing unnecessary manual interventions.

5.6 Customer experience improvement

Measurement of customer satisfaction through metrics such as transaction approval times, reduced authentication delays, and fewer disruptions due to security checks.

5.7 Analytical framework

The data collected was analyzed using a combination of statistical methods and AI-driven tools to derive actionable insights. Key aspects of the analytical framework include:

Trend Analysis: Identifying patterns and trends in the adoption of AI and automation in transaction security over time.

Comparative Analysis: Benchmarking performance metrics across different FinTech platforms to evaluate the relative effectiveness of AI-driven systems.

Impact Assessment: Assessing the tangible benefits of AI and automation on fraud prevention, compliance efficiency, and operational costs.

Case Study Analysis: Detailed examination of real-world examples to highlight best practices and common challenges in implementing AI-driven transaction security.





5.8 Tools and technologies

A range of tools and technologies were employed to process and analyze the data:

Data Analytics Platforms: For processing and visualizing transaction data, compliance metrics, and fraud detection rates.

AI and Machine Learning Models: To simulate scenarios and validate the performance of predictive threat analysis, real-time fraud detection, and adaptive learning models.

Blockchain Explorers: For evaluating the transparency and integrity of smart contract implementations.

Survey Tools: To gather qualitative insights from industry professionals about the challenges and benefits of automation in FinTech.

6 Case Studies and Implementation

This section provides real-world examples illustrating the transformative impact of AI and automation in FinTech, highlighting their potential to enhance security, improve operational efficiency, and drive innovation (Davis, 2023).

6.1 Case study 1: AI in Fraud detection

A leading FinTech company integrated AI-powered fraud detection into its transaction monitoring systems. The implementation included advanced machine learning models capable of analyzing transaction data in real time to identify anomalies and flag suspicious activities as in Fig. 7.

Outcome:

Fraudulent transactions reduced by 85% within six months.

False positive rates decreased by 30%, reducing the operational workload.

Customer trust and satisfaction improved due to fewer interruptions in legitimate transactions.

Technologies Used:

Real-time fraud detection algorithms.

Behavioral analysis models.

Cloud-based AI platforms for scalability.



Fig. 7. The bar graph demonstrating the dramatic improvement in fraud detection after AI implementation as compared to before AI

6.2 Case study 2: Automation in risk management

A major financial institution deployed automation systems in its risk management framework. By leveraging Robotic Process Automation (RPA) and AI-powered analytics, the organization significantly reduced the time required for risk evaluations as in Fig. 8.

Outcome:

Risk analysis time reduced by 60%.

Decision-making processes enhanced with real-time risk scoring.

Regulatory compliance adherence improved through automated reporting tools.

Technologies Used:

RPA for repetitive risk management tasks.

AI models for predictive risk analytics.



Fig. 8. This bar graph compares three metrics (Risk Analysis Time, Decision Making Speed, and Compliance Reporting Time) before and after automation, showing significant improvements in each category after implementing automation

6.3 Case study 3: Blockchain in cross-border transactions

A global payment processor adopted blockchain technology to improve the efficiency and security of crossborder transactions. The blockchain platform ensured tamper-proof records and reduced reliance on intermediaries as in Fig. 9.

Outcome:

Transaction processing times reduced from 3-5 days to a few seconds.

Operational costs lowered by 40% due to reduced intermediary fees.

Increased transparency and trust among stakeholders.

Technologies Used:

Blockchain for decentralized ledger management.

Smart contracts to automate transaction settlements.



Case Study 3: Impact of Blockchain in Cross-Border Transactions



6.4 Case study 4: Adaptive learning for customer authentication

A digital banking platform implemented adaptive learning models to enhance customer authentication processes. These models continuously evolved by analyzing user behaviors, such as typing patterns and device usage as in Fig. 10.



Fig. 10. The bar graph demonstrates how implementing adaptive learning technology increased authentication success rates while significantly reducing account takeover fraud compared to the period before implementation

Outcome:

Authentication success rates increased by 95%.

Account takeover fraud reduced by 70%.

Improved user experience through frictionless security measures.

Technologies Used:

Adaptive AI models for behavioral analysis.

Biometric authentication systems.

7 Findings and Discussion

This section presents the key findings from analyzing AI and automation in FinTech transaction security. It highlights the significant advantages brought by these technologies, as well as the challenges that must be addressed to ensure sustainable and ethical implementation.

7.1 Advantages

AI-driven systems and automation tools have revolutionized FinTech operations, offering numerous benefits that enhance security, efficiency, and user experience:

Enhanced Fraud Detection with Reduced False Positives:

AI systems excel in identifying fraud with higher accuracy compared to traditional methods. Advanced machine learning algorithms analyze vast datasets, identify subtle anomalies, and flag potentially fraudulent activities while significantly reducing false positives. This improves trust among users and reduces operational inefficiencies caused by unnecessary manual reviews.

Faster Compliance Checks and Reporting:

Automation accelerates compliance processes by instantly analyzing transaction data against regulatory requirements. Tools such as Robotic Process Automation (RPA) and Natural Language Processing (NLP) streamline compliance reporting, enabling financial institutions to generate audit-ready reports in minutes rather than days. This reduces the risk of non-compliance and associated penalties.

Improved Transaction Transparency with Blockchain:

Blockchain technology ensures that all transactions are recorded in an immutable ledger, enhancing transparency and trust. The decentralized nature of blockchain reduces the reliance on intermediaries, minimizing the potential for manipulation and fraud. Transparency also aids in compliance and fosters better relationships with regulators.

Operational Efficiency:

Automation eliminates repetitive tasks such as transaction processing, data entry, and fraud monitoring. This reduces human error, optimizes resource allocation, and lowers operational costs. Financial institutions can focus on strategic initiatives rather than routine tasks.

Personalized User Experience:

AI enhances customer experience by providing tailored solutions. For instance, AI-powered systems can offer personalized financial advice, recommend investment strategies, and enable faster transaction approvals, improving overall satisfaction and loyalty as in Fig. 11.

Scalability:

Automated systems can seamlessly handle increasing transaction volumes without compromising performance. This scalability is particularly beneficial for growing FinTech platforms catering to global audiences.

7.2 Challenges

While the benefits are significant, there are several challenges associated with the implementation of AI and automation in FinTech:

7.2.1 Ethical concerns regarding AI-based decision-making

Bias in Algorithms: AI systems may inadvertently perpetuate biases present in the training data, leading to unfair outcomes in areas such as loan approvals or fraud detection.

Lack of Transparency: AI algorithms often operate as "black boxes," making it difficult to explain or audit decision-making processes. This can lead to a lack of accountability and trust among users.

7.2.2 Data privacy and compliance with regulations

Regulatory Complexity: Financial institutions must navigate a complex web of regulations, including GDPR, CCPA, PCI-DSS, and AML guidelines. Ensuring AI and automation tools adhere to these requirements is a significant challenge.

Sensitive Data Management: The use of AI requires vast amounts of customer data, raising concerns about data privacy and the risk of breaches. Institutions must implement robust security measures to protect sensitive information.

7.2.3 Initial implementation costs and technical expertise requirements

High Investment Costs: Deploying AI and automation technologies often requires substantial upfront investment in infrastructure, software, and training.

Skilled Workforce: Implementing and maintaining these systems requires highly skilled professionals, which can be challenging to source and retain.

Integration Complexity: Merging AI and blockchain systems with existing legacy infrastructure can be technically challenging and resource intensive.

7.2.4 Resistance to change

Employees and stakeholders may resist the adoption of AI and automation due to fears of job displacement or skepticism about the reliability of these technologies. Effective management strategies are essential to address these concerns.

7.2.5 Security risks

AI Vulnerabilities: While AI enhances security, it can also be exploited by attackers. For instance, adversarial attacks can manipulate AI models to bypass fraud detection systems.

Blockchain Limitations: Blockchain networks, particularly public ones, may face scalability and energy consumption issues, which can hinder widespread adoption.

7.2.6 Evolving threat landscape

Cyberattacks are becoming more sophisticated, requiring continuous updates to AI models and security protocols. Institutions must invest in adaptive learning systems to stay ahead of emerging threats (Fang, 2021).

7.2.7 Long-term maintenance and scalability challenges

As transaction volumes grow, maintaining the scalability and performance of AI and blockchain systems becomes increasingly complex. Institutions must invest in ongoing upgrades and optimizations to meet future demands.

7.3 Future directions

Explainable AI (XAI): Developing AI models provide clear explanations for their decisions to improve transparency and accountability.

Advanced Privacy-Preserving Techniques: Implementing technologies such as federated learning and homomorphic encryption to enhance data privacy without compromising AI performance.

Global Regulatory Standards: Encouraging the harmonization of regulatory frameworks to simplify compliance processes for FinTech companies operating across multiple jurisdictions (Harris, 2023).

AI-Ethics Frameworks: Establishing ethical guidelines and oversight committees to ensure fairness, accountability, and transparency in AI-based decision-making.

Energy-Efficient Blockchain Solutions: Developing sustainable blockchain technologies, such as proof-of-stake (PoS) and layer-2 solutions, to address energy consumption challenges.



Fig. 11. The radar compares six key metrics (Transparency, Compliance Checks, Fraud Detection, Scalability, User Experience, and Operational Efficiency)

8 Best Practices

To fully leverage the potential of AI, automation, and blockchain in FinTech, it is crucial to adopt best practices that ensure efficiency, security, and scalability. These practices are designed to address challenges, enhance performance, and maintain regulatory compliance.

8.1 Train machine learning models on diverse datasets

Biases in machine learning models can lead to unfair outcomes, especially in areas like fraud detection and loan approvals. To mitigate this:

Use datasets that represent a wide variety of demographics, geographic regions, and transaction types.

Conduct fairness audits regularly to identify and address potential biases in AI algorithms.

Incorporate synthetic data to simulate rare fraud scenarios and improve detection accuracy.

8.2 Regularly update AI systems

Cyber threats are continuously evolving, requiring AI systems to stay ahead of malicious actors (Huang, 2021).

Implement continuous learning models that adapt to new fraud patterns and attack vectors.

Schedule periodic updates to ensure systems remain aligned with the latest threat intelligence.

Test AI systems against emerging threats using simulated attacks to identify vulnerabilities.

8.3 Integrate blockchain for high-value and high-risk transactions

Blockchain technology enhances security and transparency for sensitive financial operations.

Use blockchain for cross-border payments, ensuring immutable records and faster settlements.

Apply blockchain to high-risk transactions, such as large corporate fund transfers or cryptocurrency trades.

Combine blockchain with AI for real-time fraud detection and secure transaction validation.

8.4 Utilize RPA for low-complexity, repetitive tasks

Robotic Process Automation (RPA) is highly effective in streamlining operations and freeing up human resources (Iqbal, 2023).

Automate routine tasks like data entry, transaction reconciliation, and compliance reporting.

Use RPA bots to handle high-volume, repetitive tasks with minimal errors.

Deploy RPA alongside AI systems to enhance decision-making processes, such as fraud detection or risk assessments.

8.5 Enhance data privacy and compliance

Data privacy is critical in FinTech due to stringent regulations and the sensitivity of customer information.

Use encryption techniques to secure transaction and user data.

Implement access controls to ensure only authorized personnel can access sensitive information.

Regularly audit systems to verify compliance with regulations like GDPR, CCPA, and PCI-DSS.

8.6 Foster explainability in AI systems

The "black box" nature of AI can lead to mistrust and regulatory scrutiny. To promote transparency:

Develop explainable AI (XAI) models that provide clear justifications for decisions, such as flagged transactions or denied loans.

Document AI decision-making processes to ensure accountability and simplify compliance reporting.

Provide user-friendly insights to customers regarding AI-driven decisions, improving trust.

8.7 Build scalable infrastructure

Scalability is crucial as transaction volumes grow and customer bases expand.

Use cloud-based solutions to ensure seamless scaling of AI and blockchain systems as in Fig. 12.

Implement load-balancing mechanisms to maintain performance during peak transaction periods.

Invest in modular architecture that allow for easy updates and integration of new technologies.

8.8 Implement proactive risk management

Risk management is essential for mitigating financial and reputational losses.

Use predictive analytics to foresee potential risks and take preventive measures.

Conduct regular vulnerability assessments to identify and address weaknesses in security protocols.

Create contingency plans and simulate attack scenarios to prepare for potential breaches (James, 2022).

8.9 Promote collaboration between teams

Cross-functional collaboration is essential for implementing and maintaining complex systems:

Encourage cooperation between IT, compliance, and operations teams to align technology with business goals.

Provide regular training to employees on new technologies and security best practices.

Establish clear communication channels to address issues and share insights quickly.

8.10 Monitor system performance continuously

Continuous monitoring ensures systems remain efficient, secure, and effective.

Use AI-powered monitoring tools to track key performance metrics like fraud detection rates, transaction speeds, and system uptime.

Establish real-time alerts for anomalies, ensuring prompt responses to potential issues.

Conduct periodic performance reviews to identify areas for improvement and optimization.

8.11 Adopt energy-efficient technologies

Sustainability is increasingly important in financial systems.

Opt for energy-efficient blockchain protocols, such as proof-of-stake (PoS), to reduce environmental impact as in Fig. 12.

Consolidate data processing operations to minimize energy consumption.

Leverage green cloud services to align with sustainability goals.



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Fig. 12. The horizontal bar chart displays the relative importance of eleven different practices ranging from Energy Efficiency to Diverse Datasets

9 Conclusion

Artificial Intelligence (AI) and automation are reshaping transaction security in FinTech, offering significant advancements in fraud detection, compliance, and operational efficiency. These technologies enable real-time fraud detection, reduce manual workloads, and ensure secure, transparent transactions through blockchain (Johnson, 2023).



Fig. 13. The line graph tracks the growth of Fraud Detection Efficiency, Compliance Automation, and Blockchain Adoption from 2010 to 2024

However, challenges such as algorithmic biases, data privacy concerns, and high implementation costs remain. Addressing these issues through best practices—such as training models on diverse datasets, updating AI systems regularly, and integrating blockchain for high-risk transactions—can maximize their potential. Collaboration, continuous monitoring, and proactive risk management further enhance the reliability and scalability of these systems.

Looking forward, innovations like explainable AI, predictive automation, and quantum-safe blockchain are expected to redefine security and efficiency in FinTech. By aligning technology with ethical practices and strategic implementation, financial institutions can build secure, scalable systems ready to meet the demands of next-generation financial solutions as in Fig. 13.

Disclaimer (Artificial Intelligence)

Author(s) hereby declare that NO generative AI technologies such as Large Language Models (ChatGPT, COPILOT, etc) and text-to-image generators have been used during writing or editing of this manuscript.

Competing Interests

Author has declared that no competing interests exist.

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