

The Use of Artificial Intelligence (AI) to Checkmate Security Challenges in Katsina State (A Case Study of Some Selected Local Government Areas)

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Abstract: Over the past decade, Katsina State has experienced persistent security challenges, including banditry (kidnapping), cattle rustling, and armed robbery, particularly in seven frontline Local Government Areas (LGAs) out of the thirty-four in the state. These recurrent attacks have led to loss of lives and property, as well as mass migration of residents to relatively safer towns. This study investigates the use of Artificial Intelligence (AI) to checkmate security challenges in selected LGAs of Katsina State. The research adopted a descriptive survey design, using structured questionnaires as the primary instrument for data collection. A total of 90 questionnaires were administered to selected respondents, comprising security personnel, community leaders, and residents across Dan-Musa, Batsari, and Safana Local Government Areas (LGAs). Findings reveal a strong consensus among respondents that AI technologies—such as machine learning, intelligent video surveillance, sensor-based monitoring systems, and data analytics—can significantly enhance intelligence gathering, crime detection, and rapid decision-making. The study highlights that AI integration with Closed-Circuit Television (CCTV) systems can improve monitoring of suspicious activities, support search and rescue operations, assist in forensic investigations, and reduce response time to security threats. Respondents also indicated that automation of routine administrative tasks through AI could enable redeployment of security personnel to high-risk areas. The study concludes that effective deployment of AI-based security systems in Katsina State can strengthen crime prevention strategies and improve overall public safety. It recommends government investment in AI infrastructure, capacity building for security agencies, and policy frameworks to ensure ethical and efficient implementation.

Keywords: Artificial Intelligence; Security; Public Safety; Community Leaders; Crime Detection.

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I. INTRODUCTION

Artificial intelligence (AI) technology abstracts the idea of creating machine thinking behaving humanly and logically by using sets of computer algorithms. AI is new and prominent due to the latest improvements in the areas of machine learning, deep learning, and big data science. Over the last decade, AI has contributed to transforming many sectors, resulting in the fifth industrial revolution where with AI technology chunks of data can easily be turned into useful information.

With widespread applications of AI in many areas, it can be used in overcoming criminal activities and tracking of criminals. Digital personal assistants, which are

characteristics of the present operating system of computers and smartphones solely, depend on AI technology to perform some tasks such as online search, recommendations, voice recognition, image recognition, automated personal banks database warehouse (images, fingerprint, insurance details, and vehicle registration) are maintained using AI tools. This serves as an enormous source of reference data for security personnel.

With the increasing number of kidnappers' attack and cyber-arms emerging significantly over the last 10 years, it is apparent that only AI technology can help safeguard against such crimes [1]. In order to conduct routine searches for people, motorcycles, cars, and forests, AI technology has to be coordinated with military personnel

which will result in a new defense mechanism. New aggressive methods for instance the dynamic building of protective perimeters, integrated management, and completely automated responses to the environment, would need the deployment of AI methods and knowledge-intensive technologies [2].

Integration of AI into our familiar world needs urgent attention. Presently, in the US and EU Orbital Insight is practicing machine self-training (AI components) for low-resolution photos from satellites [3]. Since the September 11th, 2001 attack, huge amounts of funds have been spent for national security programs. Nigeria is not left out as federal, state, and local governments dedicated billions of Naira in order to curb the security challenges facing the nation.

A suitable quality system can solve all tasks assigned to it, so it is essential to formulate such a task so that the solution obtained from the machine is not questionable in the interest of man; this process is called AI self-learning (Deep learning). The basic idea is that the purpose of the machine must be determined in the extreme outreach of valuable for the people targets, but it does not have to know in advance what they are. Innovative technologies for example quantum computing, may change how AI's technique to receive information about various issues, and allow it to learn by getting feedback, and possibly even imitate human cognitive contact with the globe. There are numerous routine techniques in security processes that AI can handle with ease such as forecasting of a crime, early identification of threats, and accurate forecast in a diverse timeframe. It provide satisfactory and timely backup with detailed multi-factor scenarios that can vividly improve the fight against crime in Katsina state.

➤ *Problem Statement*

Security remains a fundamental responsibility of government and a prerequisite for economic stability and sustainable development. However, Katsina State has experienced escalating insecurity over the past decade, characterized by banditry, kidnapping, cattle rustling, and armed robbery. Despite multiple federal and state interventions—including military operations such as Operation Sharan Daji, amnesty initiatives, negotiated exchanges, and restrictive measures on markets and fuel sales—the crisis has persisted, particularly in frontline Local Government Areas such as Faskari, Sabuwa, Dandume, Jibia, Batsari, Safana, Danmusa, and Kankara. The activities of armed groups operating around the Rugu Forest and along transnational routes have led to widespread displacement, destruction of property, and loss of lives. The continued rise in attacks, even in the presence of heavy military deployment, highlights the limitations of conventional security strategies.

Given the complexity and organized nature of banditry operations, there is an urgent need for innovative and intelligence-driven approaches to security management. In the digital age, Artificial Intelligence (AI) offers advanced capabilities in data collection, processing, analysis, and real-

time decision-making. Planned criminal activities often leave digital footprints that can be detected through machine learning, predictive analytics, and intelligent surveillance systems. Therefore, integrating AI into the security architecture of Katsina State is justified as a strategic response to complement existing efforts, enhance intelligence gathering, improve rapid response mechanisms, and provide a more proactive and technology-driven solution to curbing insecurity in the state.

➤ *Objectives*

- To examine the level of awareness of security personnel on the use of Artificial Intelligence in combating insecurity in Katsina State.
- To assess the perception and readiness of traditional rulers and community stakeholders toward adopting AI-based security solutions.
- To explore the potential application of machine learning techniques in predicting and preventing planned criminal attacks in selected Local Government Areas.

➤ *Research Questions*

The following research questions are to be answered in this study:

- How is Artificial Intelligence (AI) currently being applied to address security challenges in selected Local Government Areas of Katsina State?
- What specific security threats in the selected LGAs can be effectively detected, monitored, or prevented through the use of AI technologies?
- What factors influence the effectiveness and adoption of AI-based security solutions in addressing security challenges within the selected LGAs of Katsina State?

II. LITERATURE REVIEW

Security challenges such as kidnapping, banditry, terrorism, and communal conflicts have significantly affected Northern Nigeria, particularly Katsina State. The increasing complexity and sophistication of these security threats have exposed limitations in conventional security strategies, including manual surveillance, reactive policing, and intelligence gaps. Scholars argue that modern security challenges require technologically driven solutions capable of predictive analysis, rapid data processing, and real-time response. Artificial Intelligence (AI), defined as computer systems capable of performing tasks that normally require human intelligence such as learning, reasoning, and decision-making, has emerged as a transformative tool in addressing contemporary security concerns [4].

Globally, AI has been integrated into various security architectures, including predictive analytics, automated surveillance, anomaly detection, and cyber threat intelligence. Research by [5] emphasizes that machine learning algorithms are highly effective in identifying patterns within large datasets, enabling early detection of security breaches and suspicious activities. Similarly, [6]

demonstrate how AI-powered computer vision systems can detect abnormal behaviors in real time, thereby enhancing surveillance efficiency in high-risk environments. These technological advancements suggest that AI systems can significantly improve monitoring capabilities in volatile regions.

Predictive policing, a major application of AI in security management, relies on historical crime data to forecast potential hotspots and allocate resources effectively. [7] argue that predictive models enhance operational efficiency by guiding law enforcement agencies toward data-driven decision-making. However, [8] caution that algorithmic bias may reinforce existing inequalities if datasets are not carefully managed. Despite these concerns, predictive analytics remains a promising tool for proactive security management, especially in regions experiencing recurring patterns of banditry and kidnapping such as some Local Government Areas in Katsina State.

AI-driven facial recognition and biometric surveillance systems have also gained prominence in modern security operations. [9] note that facial recognition technologies can assist law enforcement agencies in identifying suspects and tracking criminal networks. In areas with insurgent activities, such systems can support intelligence gathering and enhance rapid response mechanisms. For Katsina State, where mobility of armed groups across communities presents challenges, AI-enabled surveillance infrastructure could strengthen border monitoring and internal community security frameworks.

In addition to physical security applications, AI plays a critical role in cyber security. As security operations become digitized, vulnerabilities in communication systems, databases, and government infrastructure increase. [5] highlight that AI systems can detect anomalies in network traffic, identify malicious patterns, and respond autonomously to cyber threats. This capability is essential for protecting sensitive security information and ensuring coordinated response strategies in high-risk environments.

Within the African context, AI adoption in security remains in its developmental stage but shows considerable potential. [10] discuss the relevance of AI-assisted systems in enhancing border security and monitoring transnational criminal activities in West Africa. Their findings indicate that integrating AI technologies with traditional security structures can improve surveillance efficiency and intelligence coordination. However, infrastructural deficits, limited technical expertise, and funding constraints remain major barriers to effective implementation.

The Nigerian security landscape presents unique socio-economic drivers of insecurity, including unemployment, poverty, and weak institutional capacity. [11] argue that sustainable security solutions must combine technological innovation with socio-economic interventions. AI systems, when integrated into community policing frameworks, could enable early-warning mechanisms, data-driven patrol deployment, and real-time incident reporting. For selected

Local Government Areas in Katsina State, such localized AI deployment may enhance responsiveness and reduce reaction time during security incidents.

Despite its advantages, the deployment of AI in security operations raises ethical and governance concerns. [12] emphasizes the importance of regulatory frameworks to prevent misuse, ensure accountability, and protect citizens' privacy rights. In low-resource settings, [13] further identify infrastructural limitations, poor data quality, and inadequate training as key obstacles to AI adoption. These challenges suggest that while AI presents promising opportunities to checkmate security threats in Katsina State, careful planning, legal oversight, and institutional capacity-building are essential.

In the eighties, a similar wave transformed the world with personal computer technologies, where computational power became very cheap and affordable. In the same way, AI makes prediction cheap and affordable and will lead to immediately automating routine and reproducible works through machines [14]. [15] proposed an approach to artificial intelligence as a tool for combating insecurity in Nigeria. Nigerian government and Katsina state in general can fully adopt and implement a similar approach proposed by [16] to control rising insecurity. [17] proposed a machine-learning based Internet of Things (IoT) smart home system to detect and reduce urban insecurity in Uganda, this research can extended to both rural and urban areas in Katsina state, Nigeria.

Overall, the reviewed literature indicates that Artificial Intelligence has strong potential to transform security management through predictive analytics, surveillance automation, and intelligent threat detection. Although empirical studies focusing specifically on Katsina State remain limited, global and regional evidence supports the feasibility of AI-driven security interventions. This study therefore contributes to existing knowledge by examining how AI technologies can be strategically deployed within selected Local Government Areas of Katsina State to enhance security operations and mitigate emerging threats.

III. METHODOLOGY

A quantitative approach was followed. [18] stresses that quantitative research focuses on gathering numerical data and generalizing it across groups of people or explaining a particular phenomenon. It emphasizes objective measurements and the statistical, mathematical, or numerical analysis of data collected through polls, questionnaires, and surveys or by manipulating pre-existing statistical data using computational techniques. According to [19], a descriptive survey design describes a condition or phenomenon as it exists naturally without manipulations.

➤ *Sample Size*

The sampling technique is a procedure applied to choose among the accessible population, a representative of the same population upon which generalization could be made. In simple terms, it is a method of using a small

portion of the population for investigation, and in the end, generalization is made on the whole population. For the purpose of this research a sampling frame of 90 respondents from Dan-Musa, Batsari and Safana LGAs will be used, 30 respondents from each local government area.

➤ *Instrument for Data Collection*

The techniques employed for data collection is basically the questionnaire, as a primary source of data. It provides an opportunity for respondents to give a frank, anonymous answer. It makes it easier to collect data from respondents and relatively easier to analyze the collected data [20] (Bolarinwa, 2015).

The questionnaire is a structured type because the research considered it to be more appropriate for the study [18] Igwenagu (2016) opined that “Questionnaire is an observational technique which comprises series of items presented to a respondent in a written form, in which the individual is expected to respond, the respondents are given a list of written items which he responds by ticking the one he considers appropriate”.

➤ *Validity of the Research Instrument*

According to [20] validity refers to “the degree to which the instrument measures what it is supposed to be measuring”. In order to make the questionnaire reliable and valid, it will given out to one of the staff of Umaru Musa Yaradua University Katsina for validation.

• *Data Analysis*

Data collected from the respondents are analyze and interpreted using a frequency table.

➤ *Simulation Environment*

The simulation scenario consists of one hexagonal cell with 500m radius. The total bandwidth is 5MHz with 25 resource block per slot of 12 subcarriers spacing.

➤ *Computer Resources*

The proposed machine algorithm will implemented using a system-level simulator [22] (Ikuno et al., 2010) developed in MATLAB (MATLAB R2014a). The Windows 7 64-bit operating system runs on an Intel core i5 machine with a 3.2 GHz core processor and 8GB of Random Access Memory (RAM). The simulation results generated are transferred to an Excel file and are depicted using the Gnutplot graphics utility version 5.1.

➤ *Simulation Setups*

The simulation experiments are conducted to evaluate the performance of the proposed schemes, using the same simulation platform [21] (Ikuno et al., 2010) implemented using object- oriented MATLAB.

➤ *Pseudocode*

The pseudocode for the proposed algorithm is presented as follows:

```
BEGIN AI_Bandit_Detection_Matching_System
```

```
SET ThresholdHigh = 0.80
```

```
SET ThresholdMedium = 0.60
```

```
FOR each Suspect in SuspectDatabase DO
```

```
// STEP 1: Compute Facial Similarity
```

```
FaceScore ← CosineSimilarity(IncomingFaceVector,  
Suspect.FaceVector)
```

```
// STEP 2: Compute Movement Pattern Similarity
```

```
MovementScore ← MovementSimilarity(IncomingMovement  
Vector, Suspect.MovementVector)
```

```
// STEP 3: Compute Location Matching Score
```

```
IF IncomingLocation == Suspect.LastKnownLocation  
THEN
```

```
LocationScore ← 1.0
```

```
ELSE
```

```
LocationScore ← 0.5
```

```
ENDIF
```

```
// STEP 4: Compute Final Weighted Score
```

```
FinalScore ← (0.60 × FaceScore)
```

```
+ (0.25 × MovementScore)
```

```
+ (0.15 × LocationScore)
```

```
// STEP 5: Classify Risk Level
```

```
IF FinalScore ≥ ThresholdHigh THEN
```

```
RiskLevel ← "HIGH RISK"
```

```
AlertStatus ← "TRIGGER ALERT"
```

```
ELSE IF FinalScore ≥ ThresholdMedium THEN
```

```
RiskLevel ← "MEDIUM RISK"
```

```
AlertStatus ← "FLAG FOR MONITORING"
```

```
ELSE
```

```
RiskLevel ← "LOW RISK"
```

```
AlertStatus ← "NO ALERT"
```

```
ENDIF
```

```
// STEP 6: Store Result
```

```
SaveResult(Suspect.ID, FinalScore, RiskLevel)
```

```
ENDFOR
```

```
END AI_Bandit_Detection_Matching_System
```

➤ *Workflow for the Proposed Work*

Figure 1 illustrated the diagram for the workflow for the proposed work. Figure 2 represent the drone with install CCTV monitoring at the Dajin Rugu forest. Figure 3 show the matching area of the suspected bandits and criminal area.

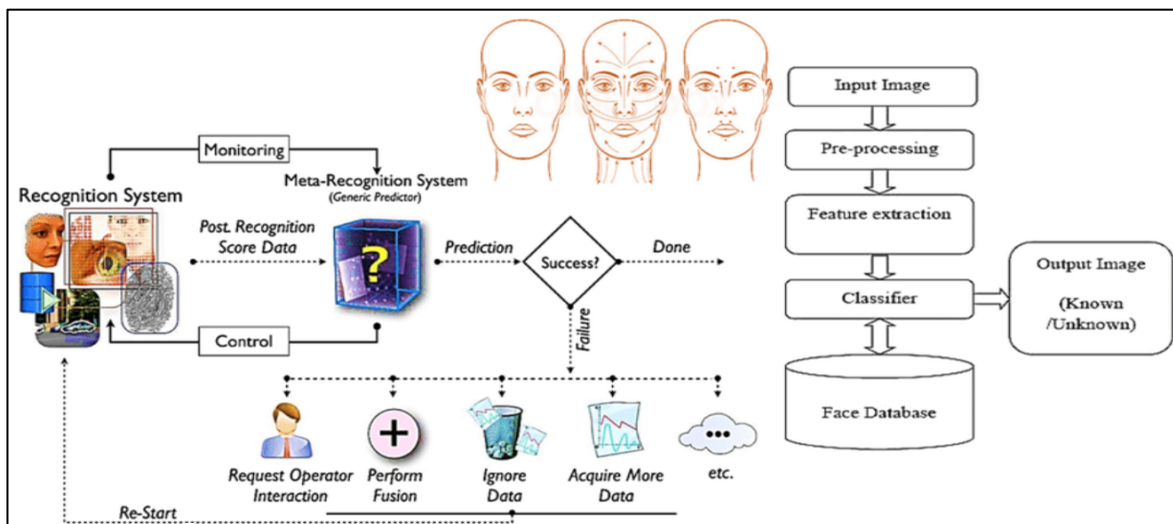


Fig 1 Workflow Diagram



Fig 2 Drone at Dajin Rugu Forest



Fig 3 Suspected Bandits and Criminals Areas

IV. RESULTS AND DISCUSSIONS

The discussion of this results obtained in this research is of twofold. The first segment is the implementation of the pseudocode in section 3.7 and analysis of the results obtained from questionnaire at the Appendix 1

A. AI Bandit Detection System Results Interpretation

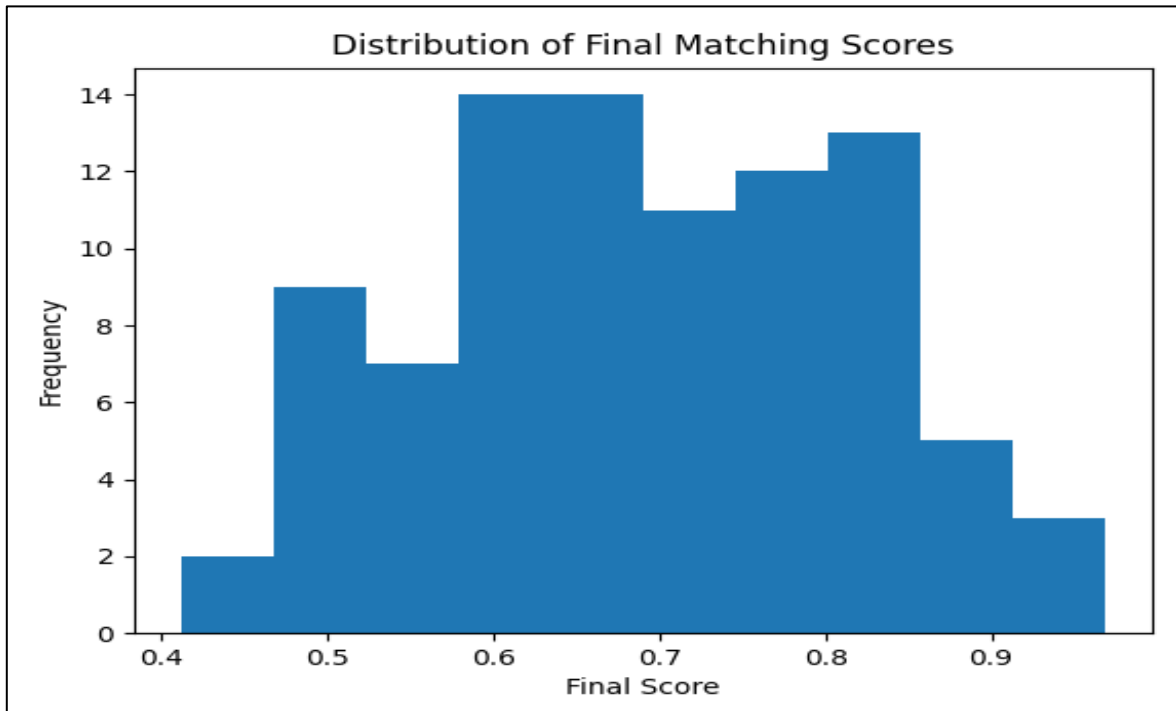


Fig 4 Distribution of Final Matching Scores

Figure 4 shows that most of the final matching scores fall between 0.6 and 0.85, indicating that many suspects are within the medium to high similarity range. This suggests that the system effectively identifies potential matches using combined inputs.

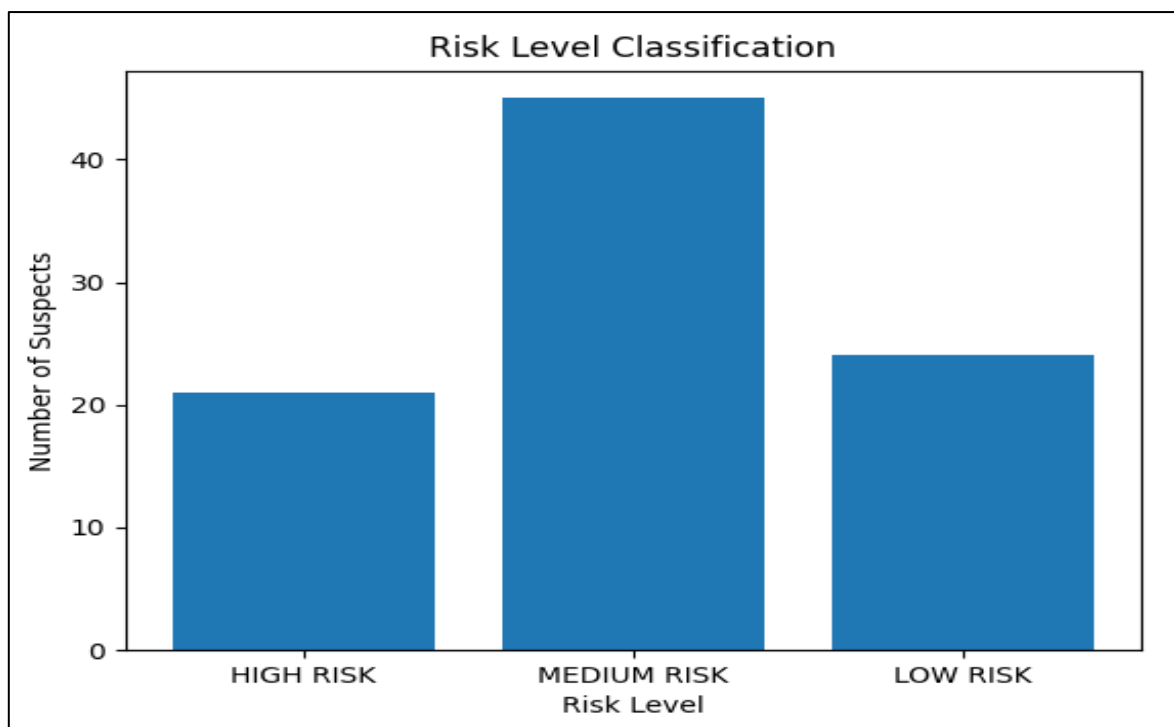


Fig 5 Risk Level Classification

Figure 5 indicates that the majority of suspects fall under the Medium Risk category, followed by High Risk, while fewer are classified as Low Risk. This shows that the

system is sensitive in identifying individuals requiring monitoring.

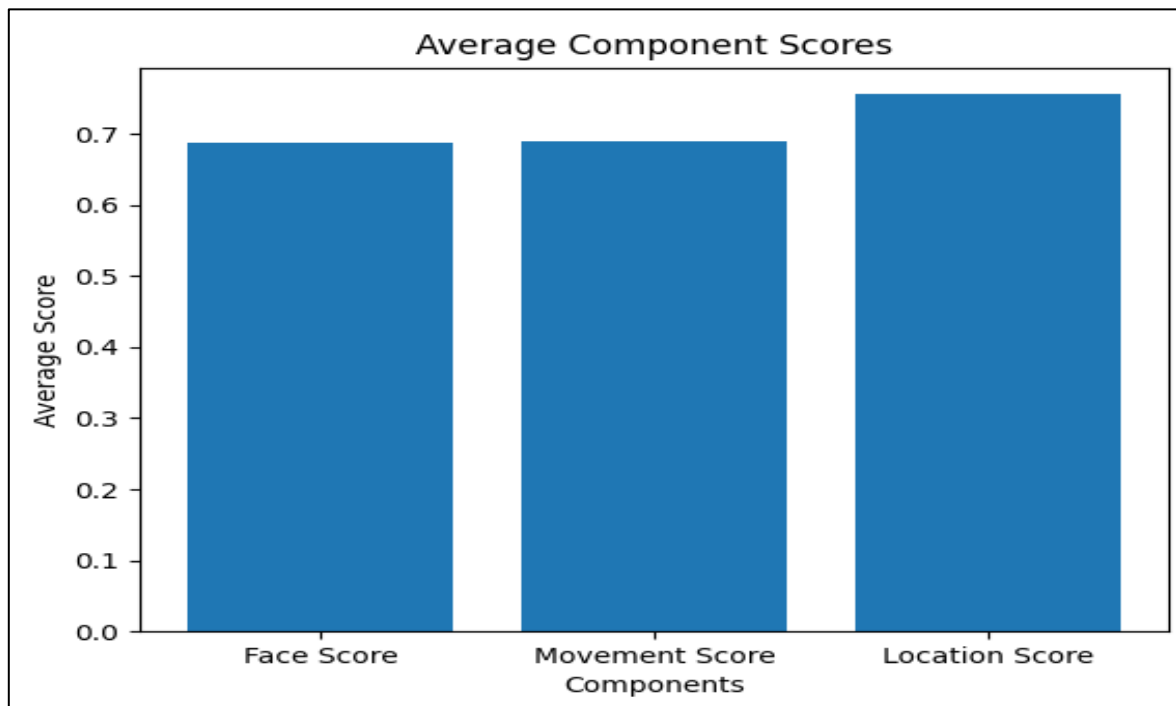


Fig 6 Average Component Scores

Figure 6 reveals that facial recognition has the highest average score contribution, followed by movement pattern, while location has the least contribution. This confirms that facial recognition is the most influential factor in the more.

B. Data Analysis and Presentation from the Questionnaire Results

➤ *Section A: Demographic Information*

Table 1 Gender Distribution

| Gender | Frequency | Percentage (%) |
|--------|-----------|----------------|
| Male | 55 | 61.10 |
| Female | 35 | 38.90 |
| Total | 90 | 100 |

From the Table 1 it shows that majority of the respondents are male (55) representing 61.10% which indicates that they have dominated the sample as a result of their high involvement in educational sector, security sector

as well as traditional rulers. The Female respondents are 35 represent 38.90% indicating low participation in education, security, and traditional rulers.

Table 2 Age Distribution

| Age Group | Frequency | Percentage (%) |
|-----------|-----------|----------------|
| 18–25 | 20 | 22.20 |
| 26–35 | 30 | 33.30 |
| 36–45 | 25 | 27.80 |
| 46+ | 15 | 16.70 |
| Total | 90 | 100 |

Table 2 indicated that majority of the respondents are within 26–35 years presenting 33.30%. This show that the

youth are the leaders in areas such education, security and traditional rulers.

Table 3 Educational Qualification

| Qualification | Frequency | Percentage (%) |
|------------------|-----------|----------------|
| Secondary School | 15 | 16.70 |
| NCE/ND | 25 | 27.80 |

| | | |
|--------------|----|-------|
| HND/BSc | 35 | 38.90 |
| Postgraduate | 15 | 16.70 |
| Total | 90 | 100 |

Table 3 show that 35 of the respondents are HND/BSc holders representing 38.90%, 25 of the respondents hold NCE/ND certificates representing 27.80%, while the

secondary school and postgraduate both have 15 respondents each representing 16.70 respectively.

Table 4 Occupation

| Occupation | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Security Personnel | 20 | 22.20 |
| Civil Servant | 18 | 20.00 |
| Traditional Leader | 10 | 11.10 |
| Business Person | 15 | 16.70 |
| Student | 20 | 22.20 |
| Others | 7 | 7.80 |
| Total | 90 | 100 |

Table 4 illustrated that majority of the respondents are security personnel and students 20 each representing 22.20% respectively. 18 civil servants representing 20.00%,

we have 15 business person in the study representing 16.70%, 10 traditional rulers representing 11.10% while other are 7 representing 7.80%.

Table 5 Distribution of Respondents by Local Government Area (LGA)

| LGA | Frequency | Percentage (%) |
|----------|-----------|----------------|
| Dan Musa | 30 | 33.30 |
| Batsari | 30 | 33.30 |
| Safana | 30 | 33.30 |
| Total | 90 | 100% |

Table 5 shows that the respondents are evenly distributed across the three Local Government Areas, with Dan Musa, Batsari, and Safana each having 30 respondents

representing 33.30% respectively. This indicates equal representation of the study areas, ensuring balanced responses across the selected LGAs.

Table 6 Years of Experience in Security-related Matters

| Years of Experience | Frequency | Percentage (%) |
|---------------------|-----------|----------------|
| 1–5 years | 25 | 27.80 |
| 6–10 years | 20 | 22.20 |
| 11 years and above | 15 | 16.70 |
| Not Applicable | 30 | 33.30 |
| Total | 90 | 100 |

Table 6 indicates that the highest number of respondents 30 (33.30%) reported not applicable in terms of security-related experience, while 25 respondents representing 27.80% have 1–5 years of experience. Additionally, 20 respondents (22.20%) have 6–10 years, and 15 respondents (16.70%) have 11 years and above. This

shows that although many respondents lack direct experience, a considerable number possess relevant security knowledge.

➤ *Section B: Responses to Research Questions*

Table 7 Artificial Intelligence (AI) Technologies are Currently Being Applied in Some Selected LGAs of Katsina State to Support Security Operations.

| Response | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Strongly Agreed | 25 | 27.80 |
| Agreed | 30 | 33.30 |
| Disagreed | 10 | 11.10 |
| Strongly Disagreed | 5 | 5.60 |
| Undecided | 20 | 22.20 |
| Total | 90 | 100 |

Table 7 shows that the majority of respondents agreed (30 representing 33.30%) and strongly agreed (25 representing 27.80%) that AI technologies are currently

being applied in selected LGAs. This indicates a general awareness and acceptance of AI usage in supporting security operations.

Table 8 AI Tools Such as Machine Learning, Intelligent Video Surveillance, and Data Analytics Improve Intelligence Gathering in Addressing Insecurity within the Selected LGAs.

| Response | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Strongly Agreed | 30 | 33.30 |
| Agreed | 35 | 38.90 |
| Disagreed | 8 | 8.90 |
| Strongly Disagreed | 5 | 5.60 |
| Undecided | 12 | 13.30 |
| Total | 90 | 100 |

Table 8 indicates that most respondents agreed (35 representing 38.90%) and strongly agreed (30 representing 33.30%) that AI tools improve intelligence gathering. This

suggests that AI technologies are perceived as effective in enhancing security intelligence in the study areas.

Table 9 AI-Based Surveillance Systems (e.g., Smart Cameras and Drones) can Effectively Detect Crimes Such as Banditry, Kidnapping, and Cattle Rustling in the Selected LGAs.

| Response | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Strongly Agreed | 28 | 31.10 |
| Agreed | 32 | 35.60 |
| Disagreed | 10 | 11.10 |
| Strongly Disagreed | 6 | 6.70 |
| Undecided | 14 | 15.60 |
| Total | 90 | 100 |

Table 9 shows that the majority of respondents agreed (32 representing 35.60%) and strongly agreed (28 representing 31.10%) that AI-based surveillance systems

can effectively detect crimes. This implies strong confidence in AI systems for crime detection within the selected LGAs.

Table 10 Sensor-Based Monitoring Systems and Predictive Analytics can Help Prevent Security Threats before they Occur.

| Response | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Strongly Agreed | 27 | 30.00 |
| Agreed | 33 | 36.70 |
| Disagreed | 9 | 10.00 |
| Strongly Disagreed | 6 | 6.70 |
| Undecided | 15 | 16.70 |
| Total | 90 | 100 |

Table 10 indicates that most respondents agreed (33 representing 36.70%) and strongly agreed (27 representing 30.00%) that sensor-based monitoring and predictive

analytics can prevent security threats. This reflects a positive perception of AI in proactive security management.

Table 11 The Integration of AI into Security Operations Enhances Rapid Decision-Making and Response Time Among Security Agencies

| Response | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Strongly Agreed | 29 | 32.20 |
| Agreed | 34 | 37.80 |
| Disagreed | 8 | 8.90 |
| Strongly Disagreed | 5 | 5.60 |
| Undecided | 14 | 15.60 |
| Total | 90 | 100 |

Table 11 shows that the majority of respondents agreed (34 representing 37.80%) and strongly agreed (29 representing 32.20%) that AI integration enhances rapid

decision-making and response time. This highlights the importance of AI in improving operational efficiency among security agencies.

Table 12 Inadequate Infrastructure (Such as Internet Connectivity, Electricity Supply, and Technical Equipment) Affects the Effectiveness of AI-Driven Security Solutions in the Selected LGAs.

| Response | Frequency | Percentage (%) |
|--------------------|------------------|-----------------------|
| Strongly Agreed | 35 | 38.90 |
| Agreed | 30 | 33.30 |
| Disagreed | 8 | 8.90 |
| Strongly Disagreed | 5 | 5.60 |
| Undecided | 12 | 13.30 |
| Total | 90 | 100 |

Table 12 indicates that most respondents strongly agreed (35 representing 38.90%) and agreed (30 representing 33.30%) that inadequate infrastructure affects

the effectiveness of AI-driven security solutions. This suggests that infrastructural challenges remain a major barrier to AI implementation.

Table 13 Lack of Technical Expertise and Training Among Security Personnel Limits the Successful Adoption of AI Technologies in Katsina State.

| Response | Frequency | Percentage (%) |
|--------------------|------------------|-----------------------|
| Strongly Agreed | 32 | 35.60 |
| Agreed | 33 | 36.70 |
| Disagreed | 10 | 11.10 |
| Strongly Disagreed | 5 | 5.60 |
| Undecided | 10 | 11.10 |
| Total | 90 | 100 |

Table 13 shows that the majority of respondents agreed (33 representing 36.70%) and strongly agreed (32 representing 35.60%) that lack of technical expertise limits

AI adoption. This implies that skill gaps among personnel hinder effective utilization of AI technologies.

Table 14 Community Acceptance and Cooperation Play a Significant Role in the Successful Implementation of AI-Based Security Systems.

| Response | Frequency | Percentage (%) |
|--------------------|------------------|-----------------------|
| Strongly Agreed | 28 | 31.10 |
| Agreed | 36 | 40.00 |
| Disagreed | 8 | 8.90 |
| Strongly Disagreed | 4 | 4.40 |
| Undecided | 14 | 15.60 |
| Total | 90 | 100 |

Table 14 indicates that most respondents agreed (36 representing 40.00%) and strongly agreed (28 representing 31.10%) that community acceptance plays a significant role

in AI implementation. This highlights the importance of public cooperation in security initiatives.

Table 15 Government Support, Funding, and Policy Framework are Essential for the Effective Deployment of AI in Combating Insecurity in the Selected LGAs.

| Response | Frequency | Percentage (%) |
|--------------------|------------------|-----------------------|
| Strongly Agreed | 40 | 44.40 |
| Agreed | 30 | 33.30 |
| Disagreed | 5 | 5.60 |
| Strongly Disagreed | 3 | 3.30 |
| Undecided | 12 | 13.30 |
| Total | 90 | 100 |

Table 15 shows that the majority of respondents strongly agreed (40 representing 44.40%) and agreed (30 representing 33.30%) that government support and funding

are essential for AI deployment. This emphasizes the critical role of government in ensuring successful implementation.

Table 16 I am willing to Support and Adopt AI-Driven Security Solutions as a Sustainable Strategy for Addressing Security Challenges in the Selected LGAs of Katsina State.

| Response | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| Strongly Agreed | 33 | 36.70 |
| Agreed | 32 | 35.60 |
| Disagreed | 8 | 8.90 |
| Strongly Disagreed | 5 | 5.60 |
| Undecided | 12 | 13.30 |
| Total | 90 | 100 |

Table 16 indicates that most respondents strongly agreed (33 representing 36.70%) and agreed (32 representing 35.60%) that they are willing to support and adopt AI-driven security solutions. This suggests a high level of acceptance and readiness among respondents.

RECOMMENDATIONS

- The government should invest in AI-driven security technologies such as intelligent surveillance, drones, and data analytics systems. This will enhance real-time monitoring and improve response to security threats.
- Security personnel should receive regular training on the use of AI tools and technologies. This will ensure effective operation and maximize the benefits of AI in security management.
- AI technologies should be integrated with current security frameworks like CCTV and communication systems. This will improve coordination, intelligence sharing, and decision-making.
- Clear policies and regulations should guide the use of AI in security operations. This will ensure data privacy, accountability, and prevent misuse of technology.
- Communities should be educated and involved in AI-based security initiatives. This will build trust, encourage cooperation, and enhance the effectiveness of security measures.

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➤ Conflict of Interest

The authors declared their in no conflict of interest.

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APPENDIX 1

Department of Computer Science Federal College of Education Katsina Kastina State Nigeria

Dear Respondent

I am researcher from the above department and institution conducting a research with the title “The use of Artificial Intelligence (AI) to checkmate security challenges in Katsina State. (A Case Study of some selected Local Government Areas). I would be glad if you provide me with the information needed in the questionnaire, I promise to treat all information given by you as confidential and would be used only for this research work. Thanks.

A. *Section A*

➤ *Section A: Demographic Information*

Please tick (✓) the appropriate option.

• Gender:

- Male
 Female

• Age Group:

- 18–25 years
 26–35 years
 36–45 years
 46 years and above

• Educational Qualification:

- Secondary School
 NCE/ND
 HND/Bachelor’s Degree
 Postgraduate Degree

• Occupation:

- Security Personnel
 Civil Servant
 Traditional Leader
 Business Person
 Student
 Others (Specify) _____

• Local Government Area (LGA):

- _____

• Years of Experience in Security-related Matters (if applicable):

- 1–5 years
 6–10 years
 11 years and above
 Not Applicable

B. Section B

- Artificial Intelligence (AI) technologies are currently being applied in some selected LGAs of Katsina State to support security operations.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- AI tools such as machine learning, intelligent video surveillance, and data analytics improve intelligence gathering in addressing insecurity within the selected LGAs.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- AI-based surveillance systems (e.g., smart cameras and drones) can effectively detect crimes such as banditry, kidnapping, and cattle rustling in the selected LGAs.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- Sensor-based monitoring systems and predictive analytics can help prevent security threats before they occur.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- The integration of AI into security operations enhances rapid decision-making and response time among security agencies.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- Inadequate infrastructure (such as internet connectivity, electricity supply, and technical equipment) affects the effectiveness of AI-driven security solutions in the selected LGAs.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- Lack of technical expertise and training among security personnel limits the successful adoption of AI technologies in Katsina State.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed

- d) Strongly Disagreed
- e) Undecided

- Community acceptance and cooperation play a significant role in the successful implementation of AI-based security systems.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- Government support, funding, and policy framework are essential for the effective deployment of AI in combating insecurity in the selected LGAs.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided

- I am willing to support and adopt AI-driven security solutions as a sustainable strategy for addressing security challenges in the selected LGAs of Katsina State.

- a) Agreed
- b) Disagreed
- c) Strongly Agreed
- d) Strongly Disagreed
- e) Undecided