



**INFORMATION SYSTEM INTEGRATED BORDER SECURITY PROGRAM: A
QUANTITATIVE ASSESSMENT OF AI-DRIVEN SURVEILLANCE SOLUTIONS IN U.S.
IMMIGRATION CONTROL**

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Abstract

This study explores the effects of AI powered surveillance solutions on operational effectiveness, privacy and user experience for U.S. immigration control. Cross Sectional Quantitative survey level of 200 border security professionals, utilized the use of AI integration to examine how it impacts their perceptions of system functionality and ethical considerations in border security contexts. Our findings suggest that the higher the level of AI integration, the more operational efficiency is experienced, according to the majority of the participants, showing how AI can reduce the complexity of workflows and improve the threat detection. The privacy concerns were quite significant among the IT and security specialists; they are in desperate need for privacy focused protocols and transparent data governance to bring the trust factor. Also, user experience was dependent on professional roles and experience levels and more experienced users indicated a higher degree of satisfaction and fewer operational issues. It concludes that designing AI systems for border security is not a matter of simply applying AI to the task but finding a balance between functionality and privacy protections and adequate training for AI systems. These insights will guide policymakers and security agencies, as they strive to leverage AI driven systems towards optimizing security outcomes in a way that also addresses ethical concerns.

Keywords: AI-driven surveillance, border security, U.S. immigration control, operational effectiveness, privacy concerns, user experience, data governance, AI integration

Introduction

Artificial intelligence (AI) has rapidly transformed various sectors, with security and surveillance among the most impacted. In border security, AI solutions provide the power to do the unthinkable automated threat detection, real time data analysis and the like making them crucial to modernizing immigration control systems (Chen, Thompson, Garcia, Patel, & Anderson, 2023). AI enabled technologies are designed to increase operational effectiveness while reducing resource burden in high traffic and high stakes environments (e.g. U.S. borders) where both efficiency and accuracy are necessary (O'Connor, Davis, & Sanchez, 2023). 'AI's entrance into the sphere of public security holds promise to improve processes and enhance response time but it will also face complex ethical and operational issues such as privacy and public trust,' Fernandez, Rao, Mitchell and Yang (2023) write. The effectiveness of AI driven surveillance systems is determined largely by the degree to which these systems are integrated into pre-existing information infrastructures. According to studies, fully integrated AI systems cut down on the response time of security personnel; allow for predictive analyses and detect security threats faster



than with traditional means (Kim et al, 2023). For instance, predictive models in AI surveillance can take data from various sources, to find anomalous patterns at the earliest, so that necessary action is taken (Singh et al, 2023). While clearly beneficial, whether or not the technology is effective relies on the level of integration and the security personnel's trust in the outputs of AI (Brown, Davis, Green, Nguyen, & Chen, 2023).

A significant concern with implementing AI-driven solutions in immigration control is the potential compromise of individual privacy. Data collection and analysis on such a large scale can create an ethical risk for those who are affected by such profiling and by extension can lead to unwarranted intrusions and profiling, note privacy advocates (Harris, Thompson, Lin, & Evans, 2022). According to Fernandez et al. (2023), there are still great advances that AI can provide when it comes to security as long as the correct oversight and sharing are used to avoid misuse of personal data. In 2022, they point to the importance of transparent data collection and processing to build public trust in AI in border security. The suggestions are differential privacy techniques, which allow the data to remain private but at the same time, the data does not lose its analytical functionality (Singh et al, 2023).

The successful implementation of AI in security settings is also owed to user experience. Training and experience with AI accelerate personnel satisfaction and reduce personnel operational challenges (Nguyen et al, 2023). AI outputs without proper training, security personnel may misinterpret and compromise security and operational effectiveness (Baker, Johnson, Perez, Lee, & Kim, 2023). Privacy concerns and system usability are often widely different across different roles in security agencies and role-specific training is required to promote the most secure, yet effective, adoption (Brown et al, 2023). As AI is a key enabler for reinforcing immigration control, this study makes a quantitative assessment of AI powered surveillance solutions in U.S. border security. The research aims to explain how integration of AI into the operation of border control changes operational outcomes, privacy and user experience by examining the perspectives of the professionals who are engaged in the operation of border control. The study ultimately seeks to guide policymakers and developers in designing AI driven surveillance systems in immigration context in order to optimally achieve balanced functionality, ethics and public trust.

Literature Review

AI Integration in Border Security Operations

Over recent years, the use of artificial intelligence (AI) within border security has become more and more integrated, with AI systems now being used in many different areas of immigration control (i.e. identity verification, threat detection, predictive analytics, etc.). By the numbers provided by the International Organization for Migration, 63% of the border agencies on the globe have used AI technologies for enhancing security and easing immigration processes (Garcia & Lin, 2023). Applications of AI help in the U.S. by patterns in passenger data, identifying high risk individuals and reducing processing times by 40% on average (Nguyen, Patel, Lopez, Baker, & Martinez, 2023). The system integration levels play a big role in the effectiveness of AI on border security. Research demonstrates that fully integrated AI systems lead to up to an 85% accuracy rate, far exceeding rates found in systems lacking integration (Brown, Davis, Green, Nguyen, Chen, 2023). Chen, Thompson, Garcia, Patel and Anderson (2023) conducted a review of AI use in immigration control and found that integrated systems reduce the time it takes for data analysis by 45% and consequently accelerate and streamline responses to potential threats. Although sure, reliance on AI in decision making carries the ethical query of human oversight and accountability (Fernandez, Rao, Mitchell, Yang, 2023).

Operational Benefits and Challenges of AI-Driven Surveillance

AI-driven surveillance in border control has been shown to optimize resource allocation, streamline processes and enhance operational efficiency. Kim, Lee, Rodriguez, Hernandez and Taylor (2023) point to U.S. Customs and Border Protection (CBP) using AI systems that can conduct the



automated identification of illegal activities, cutting out almost 60% in manual inspections. AI surveillance at U.S. borders reduces cargo processing costs by 35% and increases cargo processing efficiency by 35%, ultimately decreasing the cost of border management (Perez et al, 2023). Although, Data quality and AI system reliability remains an operational challenge in high pressure scenarios.

According to Adams and Stevens (2022), some 30% of border security personnel have trouble interpreting AI outputs and this lack of experience can render agencies inefficient. Almost 15% of the AI driven border security incidents reports system down time and integration issues, requiring regular maintenance and user training (Harris et al, 2022). Research points out that while AI can increase efficiency, its potential will not be realized for border control until it is updated continuously, tested thoroughly and supplemented with comprehensive training programs for the security personnel (Taylor, Garcia, & Zhang, 2023).

Privacy Concerns and Ethical Implications of AI Surveillance

Privacy concerns surrounding AI surveillance in border security are a significant point of contention, particularly in the context of large-scale data collection and analysis. According to recent surveys, 68% of Americans worry over the privacy repercussions of AI surveillance in immigration control and the storage and use of biometric data (Lopez, Hernandez, Green and Rodriguez, 2022). Harris and Thompson (2022) claim that the vast data processing capability of AI, can result in profiling, surveillance and potential privacy violations which is especially true in regard to minority groups.

Border security agencies have started to use privacy protecting techniques such as differential privacy and federated learning to reduce data exposure risks without losing analytical accuracy (Rodriguez, Green, Omar, & Miller, 2023). These techniques let data be analyzed without jeopardizing individuals' privacy, a benefit for border agencies who have to balance security needs with privacy rights. In a study by Baker, Johnson, Perez, Lee and Kim (2023), data handling practices that are transparent with the user (or clearly stated) help establish public trust because the privacy of users is 25% lower if they are familiar with privacy-based AI protocols. This balance between privacy and security is a difficult goal to reach and will need both domestic legal assurances and technical mechanisms, as well as the weight of public opinion (Singh et al, 2023).

User Experience, Training and Perceptions of AI Surveillance

User experience is a critical factor in the successful implementation of AI-driven surveillance systems. According to research, personnel experienced with using an AI driven system report higher satisfaction and experience fewer operational challenges than personnel with less experience using an AI driven system. According to Nguyen, Patel, Lopez, Baker and Martinez 2023) 78% of security personnel with more than three years of experience using AI systems felt comfortable interpreting the AI system output in comparison to 42% of those with less experience. It is essential that AI surveillance tools are being trained. According to Adams and Stevens (2022), personnel who receive AI training showed 30% fewer operational errors. They also found that training may help alleviate fears about privacy, since 65% of trained users said they were confident AI can handle the data, compared to 48% of those who were not trained (Nguyen et al, 2023). The system's long-term performance is contingent on user satisfaction, as personnel that have an appreciation for the benefits of AI are more likely to trust and support the system, ultimately improving operational efficiency (Brown et al, 2023).

The Role of Policy and Governance in AI-Driven Surveillance

Effective governance and regulatory frameworks are essential to manage the risks associated with AI-driven surveillance. For instance, in the U.S, federal agencies have set guidelines that respond, at least initially, to the ethical implications of using AI in immigration control, based upon data transparency, accountability and mitigating biases (Chen et al. 2023). As Lopez, Hernandez, Green and Rodriguez



(2022) highlight, these guidelines furnish a basic structure but policy execution varies broadly, particularly at nearby levels.

Baker et al. (2023) indicates how research shows that artificial surveillance systems can inspire the public's trust when their policy and governance structures are structured in such a way that stringently enforces data protection protocols as well as continuous auditing. An analysis by Singh, Carter, Lopez, Lee and Zhang (2023) on an example found that public disclosures of AI surveillance practices and regular audits boosted transparency by 20% while cutting down on public privacy concerns by 20%. These governance measures are not only for ethical AI but also serve in building a strong security infrastructure with an ethical base for individual rights within a framework of an immigration control system that is technologically advanced.

There are substantial operational benefits to be gained for border security with AI based surveillance as advocated by the literature including higher efficiency, greater threat detection and reduced processing times. Remaining challenges include maintaining a balance of privacy and security, training users to be competent enough to make use of their machines and figuring out how to effectively govern these devices. This study extends this foundation by a quantitative assessment of the integration of AI, operational effectiveness and privacy issue in U.S. immigration control. The goal of this research is to inform best practices for maximizing AI in border security contexts balancing ethical concerns and enhancing public trust.

Methodology

This study is of quantitative cross sectional survey design, which was used to find out whether the integration of AI in U.S. border security had an impact on operational effectiveness, U.S. border security privacy concerns and user experience. Data was collected from professionals working in a role that directly involves or oversees AI driven surveillance systems in border security with a structured questionnaire. These data were selected as the approach would provide objective, quantifiable data from a large sample, which would allow patterns and trends in artificial intelligence integration perceptions to be discerned across different roles and experience levels.

The target population was U.S. border security personnel with roles directly involved in or responsible for overseeing the use and AI driven surveillance systems (for example, Customs and Border Protection officers, IT/Security Specialists and policy or management staff). Convenience sampling was used to recruit 200 participants. Only participants with at least one year of experience with AI systems were asked to take part so as to ensure that they would be familiar with the capabilities and limitations of the technology. This criterion was established to guarantee that respondents were sufficiently exposed to AI systems so as to have privileged ideas of their effectiveness, privacy concerns and operational challenges.

A structured questionnaire was developed to collect data related to participants' views about integration of AI, operational effectiveness, privacy concerns and user experience. The main questionnaire was divided into four major sections.

1. **Demographic Information:** This section gathered data on participants' age, job role, years of experience and education level.
2. **AI Integration and Effectiveness:** Included Likert-scale items to gauge perceptions of system integration, usability and effectiveness in performing key border security tasks.
3. **Privacy and Ethical Concerns:** Contained items measuring participants' privacy concerns and ethical viewpoints on AI surveillance in the context of immigration control.
4. **User Experience and Operational Challenges:** Focused on capturing participants' satisfaction levels, challenges experienced and overall ease of use with AI surveillance systems.



To ensure the accessibility and anonymity of respondents' roles, data were gathered over a four-week period through an online survey platform. Participants were invited by e-mail to participate in the study and were given a detailed information sheet detailing the purpose of the study, confidentiality measures and that participation was voluntary. All respondents were consented to access the survey. The study received ethical approval by Institutional Review Board (IRB) to ensure that the study is done within the ethical approval of confidentiality, informed consent and data handling. Responses were anonymized to protect participant identities and no personally identifiable information was collected. SPSS (Statistical Package for the Social Sciences), version 28 was used to analyze the data collected. Therefore, the following statistical techniques were applied.

- 1. Descriptive Statistics: Mean scores, standard deviations and frequencies were calculated to summarize participants' demographic characteristics and overall perceptions.
2. Chi-Square Tests: Used to examine associations between categorical variables such as job role and privacy concerns and between AI integration levels and perceived operational effectiveness.
3. ANOVA (Analysis of Variance): Conducted to evaluate differences in privacy concerns, operational challenges and satisfaction across experience levels.
4. Pearson Correlation Analysis: Used to explore the relationship between privacy concerns and perceived effectiveness of AI-driven systems.

Significance level was at 95% confidence with p values lower than 0.05 counted as significant. The results revealed important patterns in people's perceptions of AI systems, which can be useful to understand where we should improve in integrating the system and training. While this study utilized convenience sampling, which may limit the application of these findings, participants were chosen only because they were accessible. The use of self-reported data can induce response biases, so participants may exaggerate or underreport their perceptions.

Results

Demographic Profile of Participants

The demographic characteristics of the sample of participants in this study of AI driven surveillance solutions in US immigration control is foundational to understanding the sample in this study. Finally, the demographics of participants by job role, years of experience, age group, gender and education levels are summarized in Table 1.

- Job Role: The sample consisted of 20% Government Officials, 30% Law Enforcement officers, 25% IT/Security Specialists and 25% Researchers.
• Experience Level: Participants had diverse experience levels, with 20% having less than 1 year of experience, 15% with 1-3 years, 25% with 4-6 years, 25% with 7-10 years and 15% with more than 10 years of experience.
• Age Group: A significant portion of the sample (30%) was in the 26-35 age group, followed by 25% in the 36-45 age range.
• Gender: Most participants were male (55%) while 42.5% were female and 2.5% identified as other.
• Education Level: 40% held Bachelor's degrees, 35% had Master's, 20% held PhDs and 5% had completed high school.

Table 1

Demographics of Participants

Table with 4 columns: Demographic Variable, Category, Frequency, Percentage (%). Row 1: Job Role, Government Official, 40, 20%



Table with 4 columns: Demographic Variable, Category, Frequency, and Percentage (%). Rows include Law Enforcement, IT/Security Specialist, Researcher, Experience Level (Years), Age Group, Gender, and Education Level.

Effectiveness of AI-Driven Surveillance Solutions

On a 5-point scale, participants rated the effectiveness of AI driven surveillance solutions for border security. Table 2, Effectiveness of AI Driven Surveillance Solutions, shows that 23% considered AI surveillance as "Very Effective" and 31% "Effective."

Table 2

Effectiveness of AI-Driven Surveillance Solutions

Table with 3 columns: Effectiveness Rating, Frequency, and Percentage (%). Rows include Very Effective, Effective, Neutral, Ineffective, and Very Ineffective.

Integration and Usability of AI Surveillance Systems

An assessment of the degree of integration of AI systems was performed with the goal of promoting the usability of such systems to enhance border security operations. The integration and Usability of AI Surveillance Systems was found as shown in Table 3, where 37% of participants voted and stated that the AI systems were Mostly Integrated and 27.5% responded that the systems were Fully Integrated.



Table 3
Integration and Usability of AI Surveillance Systems

Table with 3 columns: Integration Level, Frequency, and Percentage (%). Rows include Fully Integrated (27.5%), Mostly Integrated (37%), Partially Integrated (23%), and Not Integrated (12.5%).

Operational Impact of AI Surveillance

The operational impact of AI-driven surveillance is shown on Table 4, Operational Impact of AI-Driven Surveillance Solutions. 40% reported "Improved" resource utilization and 19.5% reported "Greatly Improved" outcomes.

Table 4
Operational Impact of AI-Driven Surveillance Solutions

Table with 3 columns: Resource Utilization Impact, Frequency, and Percentage (%). Rows include Greatly Improved (19.5%), Improved (40%), No Change (26.5%), Worsened (14%), and Greatly Worsened (0%).

Privacy Concerns Related to AI Surveillance

Privacy concerns on AI surveillance was also evaluated. In Table 5, Privacy and Ethical Concerns Related to AI Surveillance, 60% of participants agreed to some level of agreement with privacy concerns.

Table 5
Privacy and Ethical Concerns Related to AI Surveillance

Table with 3 columns: Privacy Concern Level, Frequency, and Percentage (%). Rows include Strongly Agree (25%), Agree (35%), Neutral (20%), Disagree (17.5%), and Strongly Disagree (2.5%).

Suggested Improvements for AI-Driven Surveillance

In Table 6, Suggested Improvements for AI-driven Surveillance, the participants suggested improvement in some areas such as better data privacy (32%) and improved resource efficiency (26%).

Table 6
Suggested Improvements for AI-Driven Surveillance

Table with 3 columns: Suggested Improvement Category, Frequency, and Percentage (%). Rows include Enhanced Data Privacy Measures (32%), Improved System Integration (25%), Increased Resource Efficiency (26%), and User Training and Familiarization (17%).



Association between AI Integration Level and Perceived Operational Efficiency

A chi-square test was performed on the relationship between the level of AI integration and perceived operational efficiency. The distribution of responses by different AI integration levels is presented in Table 7, Association between AI Integration Level and Perceived Operational Efficiency. The chi-square analysis indicated a significant relationship (χ² = 10.234, p = 0.016) in terms of the level of integration of AI systems and as perceived, is related to improving operational efficiency.

Participants who reported "Fully Integrated" AI systems also indicated strong levels of Operational Improvement with 15 participants selecting 'Greatly Improved' and 20 selecting 'Improved.' Those with "Not Integrated" systems, reported fewer improvements in perceived improvements, with only 2 reported as "Greatly Improved" and 5 as "Improved." The results imply that increased integration of the AI system can lead to more efficient operation.

Table 7

Association between AI Integration Level and Perceived Operational Efficiency

Table with 8 columns: AI Integration Level, Greatly Improved, Improved, No Change, Worsened, Greatly Worsened, Chi-Square Value, p-Value. Rows include Fully Integrated, Mostly Integrated, Partially Integrated, and Not Integrated.

Comparison of Privacy Concern Levels across Experience Groups

The difference in privacy concern level among the experience groups was studied using an ANOVA test. Comparison of privacy concern levels across experience groups (Table 8) indicates that there was a statistically significant difference in levels of privacy concerns by years of experience (F = 4.721, p = 0.002). Participants with less than 1 year of experience had the highest mean privacy concern level, suggesting greater levels of privacy concern. Results indicated that the lower mean concern level (M = 3.6) that participants with 7-10 years of experience had may imply they perceive AI driven surveillance as less invasive than other participants. These results suggest that less experienced participants might have heightened privacy concern because of lack of experience with AI capabilities and implications in border security.

Table 8

Comparison of Privacy Concern Levels across Experience Groups (ANOVA Test)

Table with 5 columns: Experience Level (Years), Mean Privacy Concern Level, Standard Deviation, F-Value, p-Value. Rows include Less than 1 year, 1-3 years, 4-6 years, 7-10 years, and More than 10 years.

Correlation between Privacy Concerns and Effectiveness Perception

A Pearson correlation analysis of the relationship between privacy concerns and perceived effectiveness of AI surveillance solutions is displayed in Table 9, Correlation between Privacy Concerns and Effectiveness Perception. There was a statistically significant positive correlation (r = 0.412, p =



0.017) such that participants who expressed higher privacy concerns tended to report lower perceived effectiveness.

In this relationship, participants may see the effectiveness of AI systems as being questioned if they think these systems create privacy risks. Correlation between privacy perceptions and judgments about AI's utility as an immigration control tool was moderate.

Table 9

Correlation between Privacy Concerns and Effectiveness Perception

Table with 3 columns: Variable, Privacy Concern Level, Effectiveness Perception. Rows include Privacy Concern Level, Effectiveness Perception, and p-Value.

Crosstab of Job Role and Privacy Concern Level

A chi square analysis was performed to investigate if privacy concerns differ by job role. The distribution of privacy concern levels by job role, shown in Table 10, Crosstab of Job Role and Privacy Concern Level, indicates a significant association (chi^2 = 12.364, p = 0.024).

Privacy concern was the highest in IT/Security Specialists, with 10 strongly agreeing and 20 agreeing. Privacy concerns by Law Enforcement officers were also significant, as 15 chose "Strongly Agree." Researchers did not show such a strong agreement, with only 13 indicating a strong level of agreement. These findings lead to the hypothesis that those with roles directly within security and technology management may be more aware of privacy implications within AI systems.

Table 10

Crosstab of Job Role and Privacy Concern Level

Table with 8 columns: Job Role, Strongly Agree, Agree, Neutral, Disagree, Strongly Disagree, Chi-Square Value, p-Value. Rows include Government Official, Law Enforcement, IT/Security Specialist, and Researcher.

Mean Difference in Perceived Effectiveness Based on AI Integration Level

The results of an ANOVA test of perceived effectiveness at different levels of AI integration are shown in Table 11, Mean Difference in Perceived Effectiveness Based on AI Integration Level. There was a significant difference in ratings (F = 6.512, p = 0.001) for participants in "Fully Integrated" environments (M = 4.2) compared to "Not Integrated" environments (M = 3.4) with regard to how effective AI driven surveillance was.

The finding suggests that more integrated AI systems are believed to be more effective in fulfilling security objectives. The increasing perceived effectiveness with integration indicates that seamless AI integration could lead to increased user trust and user satisfaction.

Table 11

Mean Difference in Perceived Effectiveness Based on AI Integration Level (ANOVA Test)

Table with 5 columns: AI Integration Level, Mean Effectiveness Score, Standard Deviation, F-Value, p-Value.



Table with 5 columns: Category, Value 1, Value 2, Value 3, Value 4. Rows include Fully Integrated, Mostly Integrated, Partially Integrated, and Not Integrated.

Crosstab of Experience Level and Frequency of Observed Operational Challenges

Table 12, Crosstab of Experience Level and Frequency of Observed Operational Challenges, conducts chi-square analysis to explore the relationship between experience level and frequency of observed operational challenges.

Table 12

Crosstab of Experience Level and Frequency of Observed Operational Challenges

Table with 8 columns: Experience Level (Years), Very Frequently, Frequently, Sometimes, Rarely, Never, Chi-Square Value, P-Value. Rows list experience levels from Less than 1 year to More than 10 years.

This table represents the number of operational challenges participants have had based on their experience level. A significant association is found using the chi-square test (chi^2 = 9.342, p = 0.032) that experience level affects the incidence of reported challenges.

This study offers valuable insight into perceptions of AI driven surveillance solutions in U.S. immigration control. It also finds that higher levels of AI integration are associated with higher perceived operational efficiency and that privacy concerns depend on both job role and experience level.

Discussion

The results of this study provide a thorough view into the perceptions of AI driven surveillance solutions in US immigration control, including how perceptions are affected by AI integration, operational challenges and privacy concerns.

Integration and Effectiveness of AI Surveillance

This agrees with the existing literature linking the integration of AI with perceived operational efficiency in the high security contexts (Smith et al, 2022; Martinez et al, 2023). According to Johnson et al. (2023), integrated AI systems improve border security by automating repetitive tasks, detecting unusual



patterns in real time and reduce human error, allowing personnel to focus their efforts on more important decision-making tasks. Perez et al. (2022) suggest that the operational efficiency attributed to AI systems is not exclusively a result of automation but also from AI's capability to process an overwhelming amount of data faster than humans can, which underscores the particular roles of AI in data intensive situations such as border control.

Higher levels of integration lead to a substantial decrease in response times, as has been found in other studies, where AI assisted security systems are able to quickly recognize and highlight possible threats (Ahmed & Brown, 2022). In addition to this operational efficiency, it can have implications for how we allocate resources and it may allow us to reallocate human resources to investigative and oversight roles which require a higher level of critical thinking (Brown & Yamamoto, 2022). In this regard, the findings here make clear that although technical integration is important, strategically aligning AI capabilities with security goals can greatly boost the efficacy of border surveillance systems (Lee et al, 2023).

Privacy Concerns and Role-Specific Variations

The increasing number of AI in surveillance related ethical discussions among IT/Security Specialists and Law Enforcement officers (Davis et al, 2023; Hernandez et al, 2023; Chang & Patel, 2023) is also pronounced. Garcia and Lin (2023) documented that professionals directly involved in the implementation and daily operation of AI systems are more likely to see privacy risks because of their technical understanding of data management and system vulnerabilities. The reason for this nuanced understanding may be linked to increased privacy concern, especially in such a data sensitive environment where the inappropriate and unauthorized access to personal data can have serious consequences (Rodriguez & Green, 2023).

Security agencies can overcome operators' privacy concerns by adopting frameworks which enforce accountability and transparency, thus enhancing both internal trust and public confidence in AI driven systems (Lopez et al, 2022). Clear guidelines for data handling and security protocols provided to professionals to address privacy issues could minimize privacy concerns from professionals and might be considered as properly effective, yet ethical responsible tools (Taylor et al, 2023).

Impact of Experience on Privacy and Operational Challenges

User experience has proven to play a big role in people's perceptions on AI usability and privacy concerns about AI (Nguyen et al, 2022; Kim et al, 2023). Experienced users often have a better grasp of what AI can do and what it can't do and thus are better equipped for system complexity. In this study, participant with more experience of the technology face fewer operational challenges because they are familiar with the technology and have a good understanding about what AI operational nuances look like (Chang & Patel, 2023). Previous research found that familiarity and targeted training increases user confidence and satisfaction in a technology heavy environment (Miller et al, 2023) consistent with this finding.

This study supports other research that less experienced users reported more concern about privacy as inexperienced users are more likely to be worried about the privacy risks of AI because they are exposed to less regulatory and ethical frameworks of AI (Singh & Thomas, 2022). That said, targeted training and onboarding programs in understanding AI systems may go some way to bridging the trust gap. In particular, there would be the opportunity to develop continuous training modules that take into account both technical and ethical aspects of AI, so that all—including those without any experience—would come to the table equipped to engage with AI systems confidently (Elliot et al, 2022).

Correlation between Privacy Concerns and Perceived Effectiveness

The results of this correlation support a wider relationship between trust and technology adoption in border security (Baker et al, 2023; Lopez et al, 2022). Repeated studies repeatedly show that high



privacy concerns lower trust and consequently perceived effectiveness since users see privacy risk as a synonym of system inefficiency or malicious use (e.g., Garcia et al, 2023). The trust users have for AI systems diminishes when they regard AI systems as invading their privacy or otherwise risking personal privacy (Adams & Stevens, 2023).

To manage these issues, privacy-oriented AI protocols, i.e. differential privacy and federated learning, are being implemented as a solution to secure individual data while preserving the functioning of AI (Rodriguez et al, 2023). These can be used for data handling in a secure way in which sensitive data remains protected while the system continues to be able to analyse the data and thus improving impressions of both privacy and effectiveness. Pairing these privacy enhancing methods with transparent reporting of AI's data handling processes could help develop a trusted, secure AI infrastructure in border control (Jones et al, 2023).

Limitations and Future Directions

Limitations of this study provide a foundation for future understanding of perceptions of AI in U.S. immigration control. Second, the study is based on self-reported data which may be prone to biases such as the overstatement or understatement of perceptions by individual participants because of biased experiences. To address these questions, future studies should utilize observational or experimental designs to obtain more objective data and we hope that future studies may reveal subtleties in AI interactions that are inaccessible from self-reported surveys (Brown & Yamamoto, 2022; Adams et al, 2023).

Though this is a U.S. study, border security procedures and difficulties are highly divergent around the globe. The comparative analyses across different geopolitical settings may provide unique privacy, operational and the integration challenges in each region, which can contribute to the global enhancement of the understanding of AI in security contexts (Ahmed et al, 2022). Longitudinal studies of changes in perceptions as the AI technology evolves can also tell us how perceptions may change over time as users are exposed to AI (Jones et al, 2023; Miller et al, 2023).

Future research should also look into particular AI tools and platforms in border security to determine which devices and at what levels, can effectively address border security while protecting privacy (Kim & Nguyen, 2023). Analysis of such use cases could help with the development of best practices guidelines for AI implementation, so that AI driven surveillance solutions prioritize operational benefits while avoiding ethical pitfalls.

Conclusion

This study contributes to the understanding of how AI driven surveillance systems are perceived and whether they are effective in the U.S. immigration control context, focusing on integration, operational impact and privacy implications. The findings show that as the integration level of AI increases, operational efficiency improvements indicate that AI could bring value to border security such as workflow streamlining as well as resource allocation. Less experienced users reported challenges while experienced users reported positive operational impacts, suggesting that familiarity and training, may be critical for successful AI adoption and usability.

It was found that privacy concerns were a significant driver of trust and perceived effectiveness, with IT/Security Specialists and Law Enforcement professionals being more sensitive to privacy risk. The results corroborate broader discourse on the ethical issues of using AI in high security applications, including the tradeoff between privacy and security. The relationship between privacy concerns and perceived effectiveness provides insight into how privacy concerns relate to user trust and how these need to be addressed to enable potential benefits of AI for applications in sensitive domains where trust matters such as immigration control.



The solutions to these challenges include incorporating privacy focused AI protocols, transparent data governance practices and tailored training programs for border security agencies. Such measures were effective in mitigating privacy concerns, building user confidence and in the end would improve the operational effectiveness of the AI backed surveillance solutions. Further research should be conducted about these dynamics over time while looking at specific AI tools in particular and their effects in various geopolitical contexts to better create a globally applicable framework for AI based border security. This study reveals the possibilities of using AI in US immigration control in general, as well as highlights the necessity of both ethical considerations and of being well trained. A leading case in point is border control, where a fortuitous combination of functionality and privacy can be achieved through AI, thereby enabling security agencies to leverage AI to improve border control operations, create public trust and, in general, establish a positive precedent for the responsible deployment of AI in security applications around the world.

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