

Using Artificial Intelligence in Law Enforcement and Policing to Improve Public Health and Safety

Patricia Haley¹ & Darrell Norman Burrell²

¹Capitol Technology University, Laurel, MD, United States

²Capital Technology University and Associate Ethics Fellow, Marymount University, United States

Correspondence: Patricia Haley, Capitol Technology University, United States. E-mail: patriciaaemail@gmail.com
dnburrell@marymount.edu

Received: February 2, 2025

Accepted: February 12, 2025

Online Published: February 13, 2025

Abstract

The integration of artificial intelligence (AI) policing tools and geo-profiling into contemporary law enforcement strategies has revolutionized analysis of concerning behavior, offering unprecedented precision in the identification of psychological risk factors and predictive crime analysis. AI's sophisticated pattern recognition capabilities, powered by machine learning algorithms, enable the dissection of vast datasets to uncover complex behavioral trends, latent correlations, and risk indicators often imperceptible to human cognition. This analytical depth enhances law enforcement's ability to identify links between disparate criminal activities, forecast potential threats, and shift from reactive to proactive crime prevention. Complementing AI's prowess, geo-profiling employs spatial analysis rooted in criminology, psychology, and geographic information systems (GIS) to elucidate crime patterns, identify hotspots, and predict offender anchor points. The synergy between these technologies augments investigative efficiency and mitigates cognitive biases inherent in traditional profiling through data-driven objectivity. Moreover, the implications of AI and geo-profiling extend beyond criminal justice, significantly impacting public health and safety. By enhancing crime detection and enabling early intervention, these technologies contribute to reducing violence-related injuries, mitigating psychological trauma, and fostering resilient communities. Police organizations can leverage AI-driven insights to deploy targeted interventions addressing the root causes of violence, such as socio-economic disparities and mental health challenges. This conceptual study explores the transformative potential of AI and geo-profiling in crime prevention, emphasizing their role in advancing public safety, promoting health equity, and informing data-driven policies. Ultimately, these innovations represent a paradigm shift in law enforcement and public health, fostering integrated approaches to address the multifaceted challenges of modern crime and its societal impacts.

Keywords: artificial intelligence, geo-profiling, law enforcement, criminal justice, public health, public safety, criminal profiling, police investigations

1. Introduction

Geo-profiling, also known as spatial profiling, is an advanced investigative technique utilized in criminal profiling and law enforcement to analyze the geographical distribution of criminal activities. This methodology is predicated on the premise that offenders, particularly serial perpetrators, tend to commit crimes near their residences or familiar territories (Glass & Herbig, 2021). By identifying patterns in the spatial distribution of crime scenes, law enforcement agencies can infer probable offender anchor points, such as their homes, workplaces, or frequented locales. One of the foremost advantages of geo-profiling is its capacity to expedite suspect identification. Law enforcement can efficiently narrow down potential suspects involved in similar offenses by meticulously analyzing crime patterns, thereby enhancing the focus and efficacy of investigative efforts (Butkovic et al., 2019; Ashby & Craglia, 2007; Casey & Burrell, 2010).

Integrating geo-mapping and geo-plotting technologies further augments the precision of crime analysis. These tools provide dynamic visual representations of crime data, enabling investigators to identify spatial correlations, temporal trends, and potential crime hotspots more clearly (Haley & Burrell, 2024). For instance, by overlaying crime incidents on geographic information systems (GIS), law enforcement can detect clusters of activity that may signify serial offending patterns, thus facilitating the allocation of resources to high-risk areas. This geospatial intelligence improves the timeliness and accuracy of incident reporting and enriches the granularity of information available to investigators, including specific crime locations, timeframes, and environmental contexts (Haley &

Burrell, 2024).

Beyond its applications in law enforcement, geo-profiling holds significant implications for public health and safety. By elucidating crime patterns and identifying high-risk zones, geo-profiling informs the deployment of community-based interventions aimed at violence prevention and public health promotion. For example, public health officials can leverage geo-profiling data to allocate mental health services, social support programs, and crisis intervention resources to communities disproportionately affected by crime. This proactive approach mitigates the immediate risks associated with criminal activities and addresses the broader social determinants of health contributing to community vulnerability.

2. The Transformative Role of Artificial Intelligence in Criminal Behavior

Artificial intelligence (AI) has revolutionized the domain of criminal behavior, particularly in identifying psychological risk factors, offering unprecedented depth and precision in behavioral analysis. At the heart of AI's transformative capabilities is its sophisticated pattern recognition function, where machine learning algorithms dissect expansive datasets to uncover intricate behavioral trends, risk indicators, and latent correlations often imperceptible to human cognition (Cekic, 2024). For example, AI can identify nuanced links between disparate criminal activities, such as correlating specific modus operandi with geographic crime hotspots or temporal crime trends, thereby providing law enforcement with insights that were previously inconceivable (Cekic, 2024).

The analytical prowess of AI is further amplified by its formidable data processing capacity. Unlike traditional methods constrained by human limitations, AI can rapidly and meticulously analyze voluminous data encompassing criminal histories, psychological behaviors, and socio-environmental factors. Consider a scenario where AI processes thousands of case files to identify recurring themes in offenders' childhood environments, such as exposure to violence or neglect, that may contribute to recidivism. This endeavor would be prohibitively time-intensive for human analysts alone (Cekic, 2024).

Moreover, AI mitigates the pervasive subjective bias inherent in conventional profiling techniques. By relying on empirical data and objective algorithms, AI reduces the influence of cognitive biases and personal intuitions that may skew human judgment. This data-driven objectivity enhances the reliability and accuracy of psychological risk assessments (Cekic, 2024). For instance, AI can analyze linguistic patterns in written communications or social media activity to detect subtle indicators of aggression, deceit, or psychological distress that might remain obscured in traditional evaluations. Additionally, AI's ability to integrate diverse psychological theories, ranging from behavioral reinforcement models to social cognitive frameworks, enriches its analytical depth, providing a comprehensive lens to explore the motivations underlying criminal behavior (Cekic, 2024).

AI's predictive capabilities further distinguish its role in modern criminal analysis. AI can forecast potential criminal activities and emergent threats with remarkable foresight by synthesizing data from disparate sources, including digital footprints, surveillance feeds, and socio-demographic databases. Imagine an AI system continuously monitoring online forums for radicalization patterns, flagging individuals whose behavioral trajectories suggest an escalating risk of violent extremism. This real-time analysis enhances situational awareness and empowers law enforcement to implement preemptive interventions, shifting the paradigm from reactive to proactive crime prevention (Cekic, 2024).

3. Implications for Public Health and Safety

The integration of AI and geo-profiling into criminal justice systems transcends traditional law enforcement, bearing profound implications for public health and safety. Crime, particularly violent crime, exerts a significant toll on community well-being, contributing to psychological trauma, chronic stress, and adverse health outcomes. By enhancing the precision of crime detection and the efficiency of preventative interventions, AI and geo-profiling technologies play a pivotal role in reducing violence-related injuries, alleviating community fear, and fostering safer, healthier environments.

Public health practitioners can harness insights from AI-driven geo-profiling to inform the strategic deployment of resources, such as mental health services, trauma support programs, and community resilience initiatives. For instance, identifying neighborhoods with high rates of violent crime enables targeted public health interventions aimed at addressing the root causes of violence, such as poverty, social disintegration, and lack of access to healthcare. Furthermore, AI's predictive analytics can support early warning systems for public health crises, such as spikes in substance abuse, domestic violence, or gang-related activities, facilitating timely responses that mitigate harm and promote community well-being.

The convergence of AI and geo-profiling technologies represents a paradigm shift in criminal justice and public health. These innovations enhance the efficacy of crime prevention and investigation and contribute to a broader

vision of public safety that prioritizes health, equity, and community resilience. By fostering data-driven, ethically grounded approaches to crime analysis, we can build safer, healthier societies better equipped to address the multifaceted challenges of modern crime and its impacts on public health.

Integrating AI-based systems into law enforcement investigations has revolutionized the analytical landscape, offering specialized tools to enhance operational efficiency, police analysis, law enforcement investigations, and strategic decision-making (Fernandez-Basso et al., 2024). These sophisticated technologies streamline data collection, facilitate comprehensive analysis, and generate actionable insights, thus augmenting the investigative capabilities of law enforcement agencies (LEAs).

3.1 Automatic Crawling Tools

Automatic crawling tools are at the forefront of data acquisition, engineered to autonomously harvest vast datasets from diverse digital ecosystems, including social media platforms and the dark web. This automation obviates the need for labor-intensive manual data collection, enabling LEAs to swiftly and efficiently capture relevant intelligence. For instance, during investigations of online radicalization, these tools can systematically extract posts, comments, and user interactions that may signify extremist activities, providing a critical foundation for subsequent analyses (Fernandez-Basso et al., 2024).

3.2 Natural Language Processing (NLP) Tools

Complementing data collection, NLP technologies are employed to preprocess and structure unstructured textual data, rendering it amenable to sophisticated analysis. NLP facilitates extracting meaningful information from natural language content, such as identifying keywords, sentiment trends, and contextual themes within criminal communications. For example, in fraud investigations, NLP can detect deceptive language patterns across emails and financial documents, uncovering fraudulent schemes that might otherwise remain hidden (Fernandez-Basso et al., 2024).

3.3 Knowledge Repository (KR)

The KR is a dynamic, centralized repository for processed knowledge continually updated to reflect new data and evolving investigative insights. This repository enriches analytical processes by providing historical context and cross-referencing capabilities, allowing investigators to trace patterns over time. For instance, correlating data from past cybercrime cases with current incidents can reveal recurring threat actors or tactics, enhancing predictive capabilities (Fernandez-Basso et al., 2024).

3.4 Knowledge Discovery (KD) Tools

The system integrates KD algorithms, such as association rule mining, to uncover latent patterns and relationships within collected data. These tools identify complex interconnections that might not be immediately apparent, guiding investigators toward more informed conclusions. In narcotics trafficking investigations, KD tools can reveal distribution networks and transactional linkages between suspects, supporting targeted interdiction efforts (Fernandez-Basso et al., 2024).

3.5 Data Visualization Interfaces

Immersive human-machine Interfaces (HMIs) transform analytical outputs into intuitive visual representations, facilitating the interpretation of complex data. These interfaces enhance situational awareness, enabling LEAs to discern trends, anomalies, and critical insights. For example, visual heat maps depicting crime hotspots can assist in resource allocation decisions and optimizing patrol strategies in urban environments (Fernandez-Basso et al., 2024).

3.6 Early Warning/Early Action (EW/EA) Mechanism

The EW/EA mechanism actively addresses emerging threats by identifying "weak signals" indicative of nascent criminal activities. By detecting subtle anomalies and precursors to organized crime, this system empowers LEAs to implement preventive measures before threats fully materialize. For instance, early detection of unusual financial transactions might preempt money laundering operations linked to terrorist financing (Fernandez-Basso et al., 2024).

3.7 Scenario-Based Use Cases

To contextualize the system's functionality, scenario-based use cases, such as those involving firearms trafficking, are incorporated. These practical examples demonstrate how the system's tools can be applied in real-world investigations, providing LEAs with operational blueprints for leveraging AI capabilities effectively. Such scenarios illustrate technical applications and highlight strategic considerations in complex investigative

environments (Fernandez-Basso et al., 2024).

By equipping law enforcement with these advanced tools, AI-based systems streamline data collection, processing, and analysis and generate profound, actionable insights. This technological synergy enhances investigative operations' precision, efficiency, and overall efficacy, fortifying LEAs' capacity to combat evolving criminal threats (Fernandez-Basso et al., 2024).

4. Problem Statement

Crime remains an omnipresent and evolving phenomenon that transcends socio-economic boundaries, afflicting both affluent and impoverished nations with equal severity. It encompasses a broad spectrum of offenses that undermine individual safety, community cohesion, and societal stability. These criminal acts range from minor infractions such as threats, harassment, and petty theft to more egregious violations, including domestic violence, illegal possession of firearms or narcotics, cybercrimes, and heinous offenses like sexual assault, homicide, and human trafficking. The Global Organized Crime Index 2023 paints a stark picture of the global crime landscape, revealing that an alarming 83% of the world's population lives in environments characterized by high levels of criminality (The Global Initiative, 2024). This statistic underscores modern criminal enterprises' escalating complexity and transnational reach, often employing sophisticated, technology-driven methodologies that outpace traditional law enforcement capabilities.

As crime evolves, adopting more intricate and globalized dimensions, the need for innovation within law enforcement agencies and crime-fighting organizations becomes increasingly urgent. Traditional investigative paradigms, while foundational, are frequently inadequate for addressing the dynamic, borderless nature of contemporary criminal activities. This research study investigates the transformative potential of integrating artificial intelligence (AI) and geo-profiling into law enforcement strategies. Geo-profiling, a spatial analysis technique grounded in criminology, psychology, and geographic information systems (GIS), provides a nuanced framework for analyzing crime patterns, elucidating victimology, and identifying emerging trends among criminal actors. The conceptual premise of this study posits that the synergy between AI and geo-profiling can significantly enhance the precision and efficiency of crime detection, investigation, and prevention efforts. By processing vast datasets to uncover latent patterns and contextualizing these insights within specific geographical locales, AI-driven methodologies can facilitate the prediction of criminal behaviors and the identification of crime hotspots, leaving past understanding of techniques of hot spot identification that led to undermining police legitimacy and upturning law-abiding community members living or functioning in the identified hot spot (Braga & Weisburd, 2012). Precision identification of offenders in hot spot locations have shown positive results in overall crime reduction as demonstrated a 42% reduction in all violent crimes and 50 % reduction in violent felonies (Groff, 2015).

Ultimately, this research seeks to contribute to the evolving discourse on crime prevention, offering empirical insights into how technological advancements can fortify global efforts against crime while bolstering public health and safety infrastructures (Haley & Burrell, 2024).

5. Aim of the Inquiry

The primary aim of this inquiry is to critically examine the efficacy of integrating artificial intelligence (AI) with law enforcement methodologies as advanced tools in combating contemporary crime. This research explores how AI's robust data-processing capabilities, when coupled with the spatial analytical prowess of geo-profiling, can enhance the precision of crime detection, the efficiency of investigative processes, and the effectiveness of preventative strategies. Specifically, the study aims to elucidate how AI-driven algorithms can identify latent criminal patterns that may elude traditional investigative techniques. Furthermore, the inquiry will investigate how geo-profiling and other AI tools can contextualize crime patterns, trends, and evidence within distinct geographical and socio-environmental contexts simultaneously considering the aspects of artificial intelligence applied to predictive models may not necessarily correlate with a reduction in violence if the predictive model is presented from biased data (Fernandes & Zekic, 2023).

By examining this technological synergy, the research aspires to contribute novel insights into optimizing law enforcement practices, particularly addressing organized crime's escalating complexity and transnational dimensions. Additionally, this study explores the implications of AI-driven crime prevention strategies on public health and safety, recognizing that effective crime reduction has profound ripple effects on community well-being and societal resilience (Healy & Burrell, 2024).

6. Significance of the Inquiry

The significance of this inquiry lies in its potential to address a critical gap in contemporary law enforcement

methodologies amid an increasingly complex and globalized crime landscape. With 83% of the world's population residing in environments marked by high criminality (The Global Initiative, 2024), the demand for innovative, data-driven crime-fighting tools has never been more pressing. Traditional investigative approaches often falter when confronted with modern criminal enterprises' sophisticated, borderless nature, necessitating the adoption of technologies that can process and analyze large volumes of diverse data efficiently and accurately. This research is pivotal as it explores how the convergence of AI and geo-profiling can revolutionize crime prevention strategies, offering law enforcement agencies enhanced capabilities for proactive interventions, strategic resource allocation, and identifying emergent criminal trends.

Moreover, the implications of this inquiry extend beyond the realm of criminal justice to encompass broader public health and safety outcomes. Crime is not merely a legal issue; it is a public health crisis that affects mental health, community stability, and overall quality of life. High crime rates are associated with increased stress, trauma, and health disparities within affected communities. By enhancing the precision and timeliness of crime prevention efforts, AI and geo-profiling technologies can reduce violence-related injuries, mitigate the psychological toll of crime on individuals and communities, and foster safer, more resilient environments. Ultimately, this inquiry seeks to fortify the global discourse on crime prevention, providing empirical foundations that could inform policy development, improve public safety outcomes, and strengthen the resilience of justice and public health systems against evolving threats (Haley & Burrell, 2024).

7. Method

This conceptual paper explores innovative solutions to contemporary law enforcement challenges through the strategic application of geo-profiling augmented by artificial intelligence (AI). Drawing from emerging and established scholarly literature, the paper seeks to bridge theoretical paradigms with practical implications, offering a nuanced understanding of how AI-driven geo-profiling can transform investigative methodologies. As a distinguished genre within the academic literature, conceptual papers have experienced a surge in scholarly prominence, serving as critical platforms for synthesizing disparate strands of knowledge, fostering interdisciplinary integration, and advancing theoretical discourse (Jaakkola, 2020).

In contrast to quantitative studies, which are grounded in empirical data collection and statistical analysis, conceptual papers operate within a framework that prioritizes intellectual synthesis over numerical validation. They meticulously curate and analyze existing research, drawing upon theoretical constructs, historical insights, and fragmented academic dialogues to elucidate complex phenomena (Jaakkola, 2020). This methodological approach allows for the interrogation of abstract concepts, the proposition of novel theoretical models, and the exploration of multifaceted issues that may not be readily quantifiable.

Within this scholarly tradition, the present paper harnesses a rich tapestry of interdisciplinary literature to dissect the evolving landscape of law enforcement, particularly in the context of rising global crime rates and the escalating sophistication of criminal networks. By examining the convergence of AI technologies with geospatial analytical techniques, the paper aims to illuminate pathways for enhancing crime detection, predictive policing, and strategic resource allocation. Through this conceptual lens, the study aspires to contribute meaningfully to the academic discourse on law enforcement innovation, offering theoretical insights that could inform future empirical investigations and policy development (Jaakkola, 2020).

8. Research Question

The following research question guided the inquiry: *"How can artificial intelligence (AI) enhance law enforcement's capabilities for crime detection, investigation, and prevention, and its broader implications for public health and safety?"* This question is designed to explore both the operational efficacy of these technologies in combating crime and their potential to mitigate public health risks associated with criminal activities.

9. Key Search Terms

To capture relevant literature, a comprehensive set of key search terms was developed based on the core concepts of the study. These terms include:

- *Artificial Intelligence (AI)*
- *Geo-profiling*
- *Crime Detection*
- *Predictive Policing*
- *Crime Prevention*

- *Public Safety*
- *Organized Crime*
- *Law Enforcement Technologies*
- *Criminal Justice Innovation*
- *Spatial Analysis in Crime*

10. Boolean Search Strategy

To refine and expand search results, Boolean operators (AND, OR, NOT) were employed strategically:

- ("artificial intelligence" OR "AI technologies") AND ("geo-profiling" OR "spatial analysis") AND ("crime detection" OR "crime prevention")
- ("predictive policing" AND "law enforcement") OR ("AI in criminal justice" AND "public safety")
- ("organized crime" AND "data-driven policing") NOT ("non-violent crimes")

These combinations were designed to capture literature that addresses both the technological and criminological aspects of AI and geo-profiling.

11. Criminal Justice Databases

A targeted search was conducted across reputable academic and criminal justice databases to ensure the inclusion of high-quality, peer-reviewed sources. These databases include:

- *ProQuest Criminal Justice Database*
- *National Criminal Justice Reference Service (NCJRS)*
- *Criminal Justice Abstracts with Full Text*
- *PsycINFO (for criminological psychology studies)*
- *Scopus*
- *Web of Science*
- *IEEE Xplore (for AI and technology-focused studies)*
- *Google Scholar*
- *ResearchGate*
- *Academia.edu*

12. Article Inclusion Strategy

The inclusion criteria were meticulously defined to ensure the relevance and quality of the literature reviewed:

12.1 Publication Date

Articles published between 2010 and 2025 capture recent advancements in AI and geo-profiling technologies in policing and law enforcement.

12.2 Language

Only articles published in English were considered.

12.3 Peer-Reviewed Sources

Preference was given to peer-reviewed journal articles, conference papers, and government reports.

12.4 Content Relevance

Studies on applying AI and geo-profiling in law enforcement, crime detection, crime prevention, public safety, and public health were included.

12.5 Screening Process

The screening process involved two stages:

12.5.1 Title and Abstract Screening

Initial screening to identify studies that meet the inclusion criteria based on titles and abstracts.

12.5.2 Full-Text Review

Selected articles were then subjected to a full-text review to confirm their relevance to the research question.

12.6 Data Extraction and Synthesis

Data was extracted using a structured template capturing key elements such as study objectives, methodologies, key findings, technological applications, and implications for law enforcement and public health. A narrative synthesis approach was employed to integrate findings from diverse studies, identify recurring themes, and highlight gaps in the existing literature.

13. What is Geo-Profiling

Geo-profiling represents a sophisticated form of predictive analytics that harnesses the power of geographic information to delineate areas with a heightened probability of criminal activity occurrence, particularly in identifying patterns linked to specific crimes such as sexual assaults (Suguna et al., 2022; Willmott et al., 2021; Glass & Herbig, 2021). This advanced analytical method transcends traditional investigative techniques by providing law enforcement with spatial insights that pinpoint "hotspots," which are geographical zones where crimes are statistically more likely to transpire. For example, in cases of sexual violence, geo-profiling aids in isolating neighborhoods with recurrent incidents, thereby enabling law enforcement agencies to strategically concentrate their resources, optimize patrol routes, and expedite the apprehension of offenders (Suguna et al., 2022; Willmott et al., 2021; Glass & Herbig, 2021). Empirical evidence underscores the efficacy of geo-profiling in augmenting case closure rates for rape and sexual assault cases, thus reinforcing its value as a pivotal tool in modern criminology (Haley & Burrell, 2024).

Central to geo-profiling is the application of spatial analysis techniques, including Geographic Information Systems (GIS), crime mapping, and geographic profiling algorithms, which collectively enable the synthesis and interpretation of complex spatial data (Suguna et al., 2022; Willmott et al., 2021; Glass & Herbig, 2021). GIS technology facilitates the systematic collection, storage, manipulation, and visualization of geographically referenced data, identifying crime-prone areas and deploying proactive preventive measures (Haley & Burrell, 2024). Crime mapping, as an extension of GIS, deciphers temporal and spatial crime trends, such as peak crime hours or crime type concentrations, thereby furnishing actionable intelligence for tactical policing. Furthermore, geographic profiling leverages these crime patterns to hypothesize the probable residential locations of offenders based on the spatial distribution of related criminal events, thus narrowing suspect pools with remarkable precision (Haley & Burrell, 2024).

Recent technological advancements have significantly enhanced the capabilities of geo-profiling. Artificial intelligence (AI) has emerged as a transformative force capable of processing voluminous datasets to uncover latent patterns imperceptible to human analysts (Haley & Burrell, 2024). AI-driven algorithms excel in detecting nuanced crime data correlations, such as spatial-temporal clustering of offenses, and generating predictive models that forecast potential crime hotspots (Özkul, 2021; McDaniel & Pease, 2021; Rowe & Muir, 2021). Beyond pattern recognition, AI can infer behavioral profiles of suspects, drawing from geographic movement data to predict future actions and complementing AI, biometric technologies, which encompasses fingerprint analysis, iris recognition, and voice biometrics to enhance suspect identification accuracy. When integrated with geo-profiling data, biometrics construct a multidimensional view of suspect activities and geolocations, facilitating both real-time tracking and historical movement analysis (Özkul, 2021; McDaniel & Pease, 2021; Rowe & Muir, 2021).

Geographic Information Systems (GIS) remain indispensable in crime prevention, offering a dynamic interface to correlate environmental variables with criminal activity. For instance, GIS can elucidate the interplay between crime rates and socio-economic indicators, urban infrastructure, or public transportation networks, thereby uncovering underlying criminogenic factors (Haley & Burrell, 2024). Predictive modeling within GIS frameworks anticipates emerging crime trends, enabling preemptive resource allocation. Parallel to GIS, big data analytics integrate diverse data streams, from police records and judicial databases to social media footprints, creating a holistic crime landscape analysis. This synthesis facilitates anticipating criminal behavior patterns and identifying at-risk locales (Özkul, 2021; McDaniel & Pease, 2021; Rowe & Muir, 2021).

Predictive analytics further augments crime prevention strategies through machine learning algorithms that extrapolate from historical crime data to identify potential criminal activity hotspots (Haley & Burrell, 2024). This predictive acumen directs law enforcement efforts with surgical precision, optimizing resource deployment to areas of highest need. Additionally, advanced wireless and satellite technologies have revolutionized suspect monitoring capabilities. GPS-enabled tracking devices, for instance, allow for continuous surveillance of suspect movements, while satellite imagery provides macro-level monitoring capabilities across vast terrains.

Collectively, integrating geo-profiling with cutting-edge technologies such as AI, biometrics, GIS, big data analytics, and satellite surveillance heralds a new era in law enforcement efficacy. This multidisciplinary convergence enhances the precision of suspect identification and apprehension and fortifies crime prevention

frameworks, ultimately fostering safer communities (Özkul, 2021; McDaniel & Pease, 2021; Rowe & Muir, 2021).

14. The Integration of Theoretical Frameworks with AI Tools

The synthesis of theoretical frameworks with artificial intelligence (AI) tools in criminal analysis unfolds through multifaceted mechanisms that significantly elevate the comprehension of criminal behavior and underlying motivations (Cekic, 2024). Foremost, applying learning theories, notably behaviorism, enables AI to dissect the intricate ways past behaviors predicate future actions. By processing voluminous datasets encompassing the histories of previous offenders, AI discerns entrenched behavioral patterns akin to identifying recurring motifs in an expansive narrative, which serve as predictive indicators of potential criminal conduct (Cekic, 2024).

Moreover, incorporating criminological theories, such as the general theory of crime, empowers AI to evaluate how constructs like impulsivity, self-control, and the presence of opportunity coalesce to catalyze criminal activities. This theoretical infusion enriches algorithmic models, transforming raw data into nuanced psychological profiles and robust risk assessments that mirror the complexity of human decision-making (Cekic, 2024). Equally pivotal is utilizing motivational theories, which AI leverages to unravel the psychological catalysts behind criminal actions. Through meticulous behavioral data analysis, AI isolates specific triggers, such as financial stressors or interpersonal conflicts, that amplify the propensity for criminality, thereby facilitating the design of precision-targeted intervention strategies (Cekic, 2024).

In addition, AI's prowess in systematic data analysis stands as a cornerstone of its efficacy. By amalgamating disparate data sources, including court transcripts, police records, and digital footprints, AI fosters a cohesive analytical framework that bridges theoretical paradigms with empirical realities. This integration fortifies the validity of psychological analysis and enhances its operational reliability (Cekic, 2024). Developing enhanced predictive models further exemplifies the synergy between theory and technology. Guided by established criminological knowledge, AI constructs sophisticated models capable of simulating complex behavioral scenarios, thereby refining the precision of criminal behavior forecasts with a degree of granularity previously unattainable (Cekic, 2024).

Crucially, embedding theoretical constructs within AI algorithms facilitates objective measurement of psychological traits and risk factors, mitigating the intrusion of subjective bias and bolstering the scientific integrity of profiling methodologies (Cekic, 2024). This objectivity is complemented by a dynamic feedback loop wherein AI-generated insights serve to validate, challenge, or refine existing criminological theories. Such iterative interplay fosters an evolving, evidence-based understanding of criminal behavior, continuously enhancing both theoretical constructs and practical applications (Cekic, 2024).

The confluence of theoretical frameworks with AI technologies in criminal profiling catalyzes a transformative shift in the field. This integration deepens analytical rigor and predictive precision and equips law enforcement and psychological professionals with empirically grounded tools to devise more effective crime prevention and intervention strategies (Cekic, 2024).

15. Advantages of AI in Criminal Investigation

Integrating artificial intelligence (AI) into criminal investigation offers many transformative benefits that significantly enhance investigative processes' precision, efficiency, and objectivity (Cekic, 2024). AI's capacity for enhanced accuracy stems from its ability to process and analyze expansive datasets, uncovering complex patterns and correlations that often elude human cognition. For instance, AI can identify nuanced behavioral trends across thousands of case files, leading to more precise psychological profiles and a sophisticated understanding of criminal behavior that transcends traditional methodologies (Cekic, 2024) beyond assumptions based on past behaviors of a person of concern or a checklist of behaviors, known as profiling. More closely, criminal investigations seek atypical behaviors or contextually inappropriate behaviors for a specific person, isolating concerning behaviors known as pre-incident behaviors (Schweitz, 2021).

In parallel, AI mitigates the pervasive subjective bias inherent in conventional profiling techniques. Whereas human analysts may unconsciously allow personal experiences and cognitive biases to influence judgments, AI employs objective algorithms and data-driven models that minimize the intrusion of such errors. This objectivity fosters a more impartial and scientifically grounded approach to criminal profiling (Cekic, 2024). Furthermore, AI's capability for real-time data processing revolutionizes law enforcement's responsiveness to emerging threats. For example, AI can instantaneously analyze live surveillance feeds or social media activity, enhancing situational awareness and enabling rapid, informed interventions during critical incidents (Cekic, 2024).

The domain of predictive modeling is notably enriched through AI, as its algorithms leverage historical data and identified patterns to forecast potential criminal behaviors. This empowers law enforcement to allocate resources,

proactively preventing crimes before they materialize (Cekic, 2024). Additionally, AI excels at detecting subtle, often imperceptible patterns within data. Advanced techniques such as deep learning reveal hidden connections in criminal activities, offering profound insights into offenses' psychological underpinnings and motives (Cekic, 2024).

Another critical advantage of AI is its scalability. It enables the analysis of voluminous information streams from diverse sources, including social media platforms, financial transactions, and surveillance footage. This capacity ensures that AI can seamlessly integrate across various contexts within criminal justice systems (Cekic, 2024). Moreover, AI facilitates the integration of multidisciplinary theories from criminology, psychology, and sociology, fostering a holistic analysis of criminal behavior and enriching the interpretive depth of profiling efforts (Cekic, 2024).

From an economic perspective, AI enhances cost efficiency by automating labor-intensive data analysis processes. This automation reduces the temporal and resource expenditures traditionally associated with extensive investigations, optimizing criminal justice agencies' operational efficacy (Cekic, 2024). AI also augments investigative procedures by analyzing patterns in financial transactions, social networks, and communication records, thereby unearthing suspicious activities that might remain undetected (Cekic, 2024).

AI's integration with emotional intelligence algorithms proves invaluable in mental health assessment. It aids in evaluating witness testimonies, discerning micro-expressions, and identifying potential deceit, thereby enhancing the veracity of criminal investigations (Cekic, 2024). Additionally, AI's compatibility with other advanced technologies, such as blockchain, ensures secure data storage and sharing, thereby fortifying the integrity and transparency of the criminal justice process (Cekic, 2024).

In conclusion, the deployment of AI in criminal analysis markedly advances the field by elevating analytical accuracy, operational efficiency, and methodological objectivity. These enhancements provide law enforcement and judicial systems with formidable tools to understand, preempt, and mitigate criminal activities, ultimately contributing to a more effective and equitable justice system (Cekic, 2024).

16. Innovations in AI and Policing

The Automated Forensic Examiner (AFE) represents a conceptual advancement in digital forensic science. It is envisioned as an intelligent system designed to revolutionize investigative methodologies by automating data analysis, artifact identification, and evidence correlation through artificial intelligence (Al Fahdi et al., 2013). The AFE's architecture is predicated on integrating cutting-edge technologies that collectively enhance the efficiency and precision of forensic investigations.

Foremost among its capabilities is the automation of evidence processing, wherein the AFE employs advanced computational techniques to manage and analyze voluminous digital datasets. This automation facilitates the rapid extraction of pertinent information, enabling forensic examiners to discern critical artifacts with heightened accuracy while systematically filtering extraneous data. For instance, in a complex cybercrime investigation involving terabytes of email correspondence, the AFE can swiftly isolate communications containing suspicious keywords or anomalous patterns, significantly expediting the investigative timeline (Al Fahdi et al., 2013).

Integral to the AFE's functionality is incorporating artificial intelligence (AI), which augments its analytical prowess through sophisticated algorithms and machine learning models. One such AI technique is the application of Self-Organizing Maps (SOMs), which adeptly manage large datasets by clustering related events based on inherent data relationships. This clustering mechanism not only streamlines the organization of forensic data but also aids in constructing coherent evidence trails, thereby elucidating the sequence and interconnection of digital events within a case (Al Fahdi et al., 2013).

Another pivotal AI-driven feature is the deployment of concerning behavior in algorithms within the AFE framework. By integrating behavioral analysis techniques, the system can contextualize artifacts within the broader spectrum of criminal activity, offering insights into the modus operandi and psychological profiles of potential suspects. For example, when analyzing digital footprints left by a cyberstalker, the AFE can correlate repetitive behavioral patterns with known profiling models, thereby refining suspect identification (Al Fahdi et al., 2013).

To further optimize investigative efficacy, the AFE incorporates technical competency measures, which assess the complexity of digital forensic cases based on the scope and depth of required analyses. This evaluative mechanism serves as a decision-support tool for investigators, guiding them in prioritizing investigative tasks and allocating resources effectively. In scenarios involving multifaceted data sources, such as encrypted communications across diverse platforms, these measures assist in determining the most technically challenging aspects of the investigation (Al Fahdi et al., 2013).

The AI-based law enforcement systems delineated by Fernandez-Basso et al. (2024) signify a paradigm shift in crime investigation methodologies. They offer an array of innovative functionalities tailored to augment the operational capabilities of law enforcement agencies (LEAs). This sophisticated system is underpinned by a multifaceted framework integrating advanced technologies to streamline data acquisition, enhance analytical precision, and foster informed decision-making.

A cornerstone of this system is its data collection and processing capability, which leverages state-of-the-art automatic data crawling mechanisms coupled with natural language processing (NLP) algorithms. This dual approach facilitates the rapid aggregation and synthesis of voluminous datasets from disparate sources, including social media platforms, open web repositories, and the clandestine domains of the dark web. For instance, in tracking illicit activities, the system can seamlessly identify coded language patterns and emerging threats within encrypted communications, expediting the investigative process (Fernandez-Basso et al., 2024).

Central to the system's analytical prowess is the Knowledge Repository (KR), a dynamic database that archives processed intelligence, enabling LEAs to draw upon a reservoir of historical and contextual data. This repository is an intellectual nexus, enriching current investigations with insights derived from past cases, analogous crime trends, and interlinked data nodes. For example, investigators can cross-reference contemporary cybercrime incidents with archived data to detect recurring modus operandi or suspect profiles, enhancing investigative continuity and depth (Fernandez-Basso et al., 2024).

Further amplifying its analytical capabilities are the Knowledge Discovery (KD) tools, which employ sophisticated algorithms to excavate latent patterns and correlations within the amassed data. These tools excel in unveiling obscured relationships, such as connections between disparate criminal networks or the evolution of illicit activities across geographical boundaries. The resultant insights are instrumental in pinpointing crime hotspots and anticipating potential escalations, thus empowering LEAs with predictive foresight (Fernandez-Basso et al., 2024).

To translate complex data into actionable intelligence, the system integrates immersive Human-Machine Interfaces (HMIs) that facilitate intuitive visualization of analytical outcomes. These interfaces enhance situational awareness, offering law enforcement personnel real-time, interactive dashboards that depict crime heat maps, network linkages, and temporal trends. Such visual tools are pivotal in crisis scenarios, where rapid comprehension and swift decision-making are paramount (Fernandez-Basso et al., 2024).

A distinctive feature of the system is its Early Warning/Early Action (EW/EA) methodology, which specializes in detecting "weak signals" indicative of nascent organized crime threats. By identifying subtle anomalies and emerging patterns, the system provides preemptive alerts, enabling LEAs to initiate proactive measures and mitigate risks before they escalate into significant security concerns. For example, slight shifts in darknet marketplace activities can trigger alerts about potential trafficking operations in their formative stages (Fernandez-Basso et al., 2024).

The incorporation of semantic technologies further refines the system's analytical accuracy. The system enhances data relevance through semantic filtering and representation models, ensuring that investigators are presented with the most critical and contextually pertinent information. This semantic layer aids in distilling vast datasets into coherent narratives highlighting key investigative leads (Fernandez-Basso et al., 2024).

Lastly, the system's architecture is designed with interoperability in mind, facilitating seamless integration with other data-driven law enforcement initiatives. This collaborative framework enables synergistic partnerships with existing projects employing data mining and AI technologies, fostering a cohesive ecosystem for crime prevention and investigative excellence (Fernandez-Basso et al., 2024).

In essence, this AI-based system represents a comprehensive, technologically advanced toolset that augments law enforcement agencies' analytical capabilities and transforms the strategic landscape of modern crime investigations.

17. Rawls' Theory of Justice

John Rawls' Theory of Justice, articulated through the principles of "justice as fairness," provides a foundational framework for evaluating the ethical implications of integrating artificial intelligence (AI) in policing and geoprofiling law enforcement. Rawls's Theory of Justice posits that societal institutions must operate under principles that ensure equal basic liberties and equitable opportunities, particularly benefiting the least advantaged (Said & Nurhayati, 2021). Applied to AI-driven crime prevention, this theory mandates that algorithmic models should not reinforce systemic inequalities by disproportionately targeting marginalized communities. For instance, predictive policing algorithms that rely heavily on historical crime data risk perpetuating biases against communities historically subjected to over-policing. Through the lens of Rawlsian justice, the ethical deployment of AI requires designing systems that mitigate such disparities, ensuring that technological advancements in law enforcement

contribute to a fairer, more just society by promoting equal protection under the law and safeguarding civil liberties.

18. Distributive Justice Theory

Distributive Justice Theory emphasizes the equitable allocation of resources, opportunities, and benefits within a society (Wolfe et al., 2018; Charman & Williams, 2022). In the context of AI and geo-profiling in crime prevention, distributive justice necessitates implementing these technologies in ways that do not exacerbate social inequities. For example, if crime prediction models direct disproportionate surveillance towards economically disadvantaged neighborhoods, this could result in over-policing and stigmatization, undermining community trust and exacerbating existing social divides. Ethical implementation, guided by distributive justice, would require the equitable distribution of crime prevention resources, ensuring that all communities benefit from advancements in law enforcement technologies without bearing undue burdens of surveillance or criminalization.

19. Procedural Justice Theory

Procedural Justice Theory focuses on the fairness of the processes through which decisions are made rather than solely on the outcomes of those decisions (Tyler et al., 2015; Nagin & Telep, 2020). This theory is particularly relevant in deploying AI algorithms in law enforcement, where automated systems increasingly influence decisions about surveillance, resource allocation, and suspect identification. Procedural justice demands transparency, accountability, and the inclusion of affected communities in decision-making processes. For example, law enforcement agencies employing predictive policing tools should engage with the communities most impacted by these technologies, ensuring their voices are heard in discussions about data collection practices, algorithmic design, and policy implementation. This participatory approach enhances the legitimacy of law enforcement efforts and fosters public trust, essential for effective crime prevention and community safety.

20. The Capability Approach

The Capability Approach shifts the focus of justice from distributing resources to enhancing individuals' capabilities to achieve well-being (Ayling & Grabosky, 2006; Worrall & Kjaerulf, 2018). In AI-driven crime prevention, this framework highlights the importance of ensuring that technological interventions do not infringe upon individuals' freedoms or capabilities. For example, extensive surveillance enabled by AI and geo-profiling could restrict personal freedoms, such as the freedom of movement and privacy, particularly in communities subjected to heightened scrutiny. Ethical decision-making, guided by the Capability Approach, would involve evaluating the broader societal impacts of these technologies, ensuring that they enhance rather than hinder individuals' abilities to lead secure, autonomous, and fulfilling lives.

21. Accountability and Transparency in AI Ethics (FAT Framework)

The Fairness, Accountability, and Transparency (FAT) Framework has emerged as a key model for addressing ethical concerns in deploying AI technologies (Agrawal, 2024; Matulionyte & Hanif, 2021). This framework emphasizes the need for fairness in algorithmic decision-making, accountability mechanisms to address potential harms, and transparency to ensure that the functioning of AI systems is understandable to all stakeholders (Agrawal, 2024; Matulionyte & Hanif, 2021). Stakeholders include law enforcement, community organizations, and policymakers to ensure that biometric technology solutions are implemented responsibly and effectively (Tyler, 2004; Johnson et al., 2022). In law enforcement, the FAT Framework can guide the ethical implementation of predictive policing tools by ensuring that algorithms are regularly audited for biases, that decision-making processes are transparent to the public, and that there are clear avenues for redress in cases of harm or discrimination. For example, if an AI system disproportionately predicts higher crime rates in minority neighborhoods, the FAT Framework would require an investigation into the data and algorithms used, adjustments to correct any biases, and public disclosure of the findings to maintain trust and accountability. The investigation may include, and not limited to any number of criteria that exacerbate societal harm for example, a need for a unified international legal guidance to address significant gap in the accountability and regulation of AI system (Mulyana, 2023), improvements to regulatory oversight and single reliance on private sector AI corporations and a greater understanding of misuse in creating mis- and disinformation or manipulating emotions, especially in online spaces like social media (Pauwels, 2020).

22. Conclusions

AI-powered algorithms, particularly those leveraging machine learning and natural language processing (NLP), have emerged as transformative tools in the realm of criminal investigations, offering unprecedented capabilities in analyzing vast datasets comprising victim statements, case reports, and forensic evidence (Özkul, 2021; McDaniel & Pease, 2021; Rowe & Muir, 2021). These advanced analytical tools are adept at identifying intricate patterns related to victim characteristics, offender interactions, and trauma responses, thus facilitating the creation

of comprehensive and dynamic victim profiles (Haley & Burrell, 2024). The ability of AI to process unstructured data, such as emotional narratives and qualitative accounts, enhances our understanding of the psychological and emotional impacts of crime on victims. This deeper insight enables the development of more empathetic, targeted support systems that address not just the legal but also the mental health needs of survivors, contributing significantly to public health by mitigating the long-term effects of trauma on individuals and communities.

Geo-profiling, a cornerstone of modern investigative techniques, gains new dimensions of efficacy when integrated with AI-driven geographic information systems (GIS). These systems analyze the spatial distribution of criminal activities, uncovering hidden connections and patterns that traditional investigative methods might overlook (Özkul, 2021; McDaniel & Pease, 2021; Rowe & Muir, 2021). AI enhances geo-profiling by pinpointing potential offender anchor points, mapping travel routes, and identifying crime hot spots with remarkable precision. This capacity not only aids law enforcement in narrowing search areas and optimizing resource allocation but also serves as a critical tool for public health and safety planning. By predicting high-risk areas, public health officials can deploy preventative measures, mental health resources, and community outreach programs to mitigate the broader societal impacts of crime, such as fear, stress-related health conditions, and community disintegration.

For law enforcement agencies to fully harness the potential of AI tools in improving case closure rates, it is imperative to adopt best practices for successful implementation. Central to these practices is data-driven decision-making, which employs crime data and geographic information to create predictive models that can identify potential suspects and high-risk locations with greater accuracy (Haley & Burrell, 2024). This approach streamlines investigative processes, allowing for quicker suspect identification and more efficient deployment of resources. Moreover, integrating current, comprehensive data sources, such as crime databases and GIS, significantly enhances the accuracy of geo-profiling outcomes. Utilizing the latest technologies ensures that law enforcement remains adaptive to evolving criminal tactics, thereby maintaining the efficacy of crime prevention and investigation strategies, for example, AI systems can be used for both legitimate purposes, such as in the entertainment and education industries, as well as for malicious activities, such as creating false evidence for criminal activities or spreading media manipulation and disinformation (Farid & Schindler, 2020; Labuz, 2024).

However, deploying geo-profiling technologies necessitates carefully considering ethical and legal implications, particularly concerning privacy and civil liberties (Fernandez-Basso et al., 2024). Aggregating and analyzing extensive personal data raises profound privacy concerns, as these processes often occur without explicit consent, potentially infringing on individuals' rights to digital privacy. For example, mining mobile location data to identify crime hotspots could inadvertently expose sensitive information unrelated to criminal activity, compromising innocent individuals' privacy. Additionally, there is an inherent risk of algorithmic bias, where predictive models may disproportionately target certain demographics, leading to over-policing in marginalized communities and exacerbating existing social inequalities (Haley & Burrell, 2024).

To address these challenges, AI systems must prioritize transparency and actively mitigate discriminatory outcomes through robust governance frameworks (Fernandez-Basso et al., 2024). Key strategies include ensuring human oversight, where AI is an analytical assistant rather than an autonomous decision-maker. This approach guarantees that all AI-generated insights are critically evaluated and contextualized within ethical and legal frameworks. Additionally, algorithmic explainability is vital; AI tools must be designed to be interpretable by non-technical law enforcement personnel, fostering accountability and reducing the risk of bias in decision-making.

The dynamic Knowledge Repositories (KR) management further supports ethical AI deployment. Regular updates incorporating diverse, representative data sources help counteract algorithmic bias, promoting equitable analytical outcomes. Moreover, systematic ethical audits and bias monitoring are essential to ensure that AI tools align with principles of fairness, justice, and non-discrimination (Fernandez-Basso et al., 2024). Emphasizing the detection of "weak signals" represents a proactive approach to crime prevention, focusing on emerging threats rather than historical crime patterns, thus reducing reliance on potentially biased data.

In the broader public health and safety context, the ethical deployment of AI and geo-profiling technologies can significantly contribute to societal well-being. By enhancing crime detection and prevention capabilities, these technologies improve immediate law enforcement outcomes and support the development of safer, healthier communities. Effective crime prevention reduces the prevalence of violence-related injuries, alleviates community stress, and fosters environments where public health initiatives can thrive. Ultimately, the integration of AI in geo-profiling embodies a dual promise: advancing the frontiers of criminal justice while safeguarding the ethical principles that underpin a just and equitable society (Fernandez-Basso et al., 2024).

23. Recommendations for Future Research

Investigating crime as a complex societal phenomenon necessitates a multidisciplinary lens, integrating insights

from criminal justice, legal jurisprudence, and public health to construct a holistic understanding of its underlying dynamics. This study's innovative application of artificial intelligence (AI) and geo-profiling underscores the transformative potential of technology in crime analysis. However, to cultivate a more comprehensive framework, future research must intertwine these quantitative methodologies with robust qualitative approaches that capture the nuanced, lived experiences of affected individuals and communities. The following recommendations delineate strategic research avenues to deepen our comprehension of crime and enhance interdisciplinary interventions.

24. Grounded Theory Approach

Future research should consider employing grounded theory to delve into crime survivors' lived experiences and perceptions, particularly focusing on the barriers they face in accessing support services and pursuing justice. Grounded theory's inductive methodology allows for the emergence of contextually rich, data-driven insights that illuminate the psychological, sociocultural, and systemic factors influencing survivors' journeys. For instance, examining narratives from diverse survivor populations could reveal patterns in how legal frameworks, community support structures, and cultural stigmas intersect to shape their experiences, thus informing more responsive policy and practice reforms.

25. Qualitative Focus Groups

Conducting qualitative focus groups with key stakeholders, including survivors, victim advocates, law enforcement officials, and mental health professionals, can foster in-depth dialogue on the multifaceted impacts of crime at both individual and community levels. These focus groups can serve as dynamic platforms for exploring divergent perspectives on critical issues such as trauma recovery, public perceptions of safety, systemic biases, and the efficacy of current intervention strategies. For example, discussions with law enforcement and survivor advocacy groups may uncover communication and resource allocation gaps, leading to more integrated and survivor-centered approaches to crime prevention and response.

Future scholarly endeavors can significantly enrich the existing knowledge on crime and public safety by embracing qualitative research methodologies such as grounded theory, focus groups, qualitative case studies, and phenomenological analyses. These approaches will complement the technological advancements presented in this study, facilitating a more comprehensive, empathetic, and context-sensitive response to the profound societal ramifications of crime. Such interdisciplinary integration is essential for crafting evidence-based strategies that address the symptoms of crime and tackle its root causes within diverse sociocultural landscapes.

References

- Agrawal, G. (2024). Accountability, trust, and transparency in AI systems from the perspective of public policy: Elevating ethical standards. In *AI healthcare applications and security, ethical, and legal considerations* (pp. 148–162). IGI Global.
- Ahn, J., Yoon, J., & Choi, Y. (2023). Victim selection in Korean sexual crimes: A latent class analysis. *Psychology, Crime & Law*, 29(1), 1–20.
- Al Fahdi, M., Clarke, N. L., & Furnell, S. M. (2013). Towards an automated forensic examiner (AFE) based upon criminal profiling & artificial intelligence. *11th Australian Digital Forensics Conference*, 2–4 December 2013, Edith Cowan University, Perth, Western Australia. <https://doi.org/10.4225/75/57b3be61fb866>
- Ashby, D., & Craglia, M. (2007). Profiling places: Geodemographics and GIS. In *Handbook of criminal investigation* (pp. 517–546).
- Ayling, J., & Grabosky, P. (2006). Policing by command: Enhancing law enforcement capacity through coercion. *Law & Policy*, 28(4), 420–443.
- Braga, A. A., & Weisburd, D. L. (2019). *Police innovation: Contrasting perspectives*. Cambridge University Press. <https://doi.org/10.1017/9781108278423>
- Butkovic, A., Mrdovic, S., Uludag, S., & Tanovic, A. (2019). Geographic profiling for serial cybercrime investigation. *Digital Investigation*, 28, 176–182.
- Casey, D., & Burrell, P. (2010). LASSO: Linkage analysis of serious sexual offences: A decision support system for crime analysts and investigators. In *Artificial intelligence applications and innovations: 6th IFIP WG 12.5 International Conference, AIAI 2010, Larnaca, Cyprus, October 6-7, 2010. Proceedings* (pp. 70–77). Springer Berlin Heidelberg.
- Cekic, E. (2024). The impact of artificial intelligence tools on criminal psychological profiling. *International Journal of Academic Research in Psychology*, 11(1), 1–13.

- Charman, S., & Williams, E. (2022). Accessing justice: The impact of discretion, 'deservedness' and distributive justice on the equitable allocation of policing resources. *Criminology & Criminal Justice*, 22(3), 404–422.
- Farid, H., & Schindler, D. (2020). Deep fakes: Implications for media and society. *International Journal of Digital Media*, 22(4), 45–61.
- Fernandes, P., & Zekic, K. (2023). Evaluating the effectiveness of predictive policing algorithms in the real world. *Crime Science*, 12(1), 1–12. <https://doi.org/10.1186/s40163-023-00127-x>
- Fernandez-Basso, C., Gutiérrez-Batista, K., Gómez-Romero, J., Ruiz, M. D., & Martin-Bautista, M. J. (2024). An AI knowledge-based system for police assistance in crime investigation. *Expert Systems*, e13524. <https://doi.org/10.1111/exsy.13524>
- Glass, D., & Herbig, F. (2021). Mapping matters: Geoprofiling application in South African serial rape investigation. *Crime, Law and Social Change*, 75, 349–371.
- Haley, P., & Burrell, D. N. (2024). Leveraging geo-profiling to address rape as a public health and criminal epidemic in the United States. *Revista Academiei Fortelor Terestre*, 29(3).
- Jaakkola, E. (2020). Designing conceptual articles: Four approaches. *AMS Review*, 10(1–2), 18–26.
- Łabuz, M., & Nehring, C. (2024). On the way to deep fake democracy? Deep fakes in election campaigns in 2023. *European Political Science*, 23, 454–473. <https://doi.org/10.1057/s41304-024-00482-9>
- Matulionyte, R., & Hanif, A. (2021, October). A call for more explainable AI in law enforcement. In *2021 IEEE 25th International Enterprise Distributed Object Computing Workshop (EDOCW)* (pp. 75–80). IEEE.
- McDaniel, J., & Pease, K. (Eds.). (2021). *Predictive policing and artificial intelligence*. Routledge.
- Mulyana, P. (2023). Industry self-regulation and the lack of binding legal force for AI systems. *Journal of Technology Policy*, 45(3), 75–88.
- Nagin, D. S., & Telep, C. W. (2020). Procedural justice and legal compliance: A revisionist perspective. *Criminology & Public Policy*, 19(3), 761–786.
- Özkul, D. (2021). The algorithmic fix: Location intelligence, placemaking, and predictable futures. *Convergence*, 27(3), 594–608.
- Pauwels, E. (2020). Artificial intelligence and data capture technologies in violence and conflict prevention: Opportunities and challenges for the international community. *Global Center on Cooperative Security*.
- Said, M. Y., & Nurhayati, Y. (2021). A review on Rawls' theory of justice. *International Journal of Law, Environment, and Natural Resources*, 1(1), 29–36.
- Schweit, K. (2021). *Stop the killing: How to end the mass shooting crisis*. Rowman & Littlefield.
- Suguna, M. R., Jeyavathana, R. B., & Kanimozhi, K. V. (2022, May). Comparative analysis of crime predictions using machine learning algorithms with geospatial features. In *AIP Conference Proceedings* (Vol. 2393, No. 1). AIP Publishing.
- The Global Initiative. (2024). *The Global Organized Crime Index 2023*. <https://globalinitiative.net/analysis/ocindex-2023/>
- Tyler, T. R., Goff, P. A., & MacCoun, R. J. (2015). The impact of psychological science on policing in the United States: Procedural justice, legitimacy, and effective law enforcement. *Psychological Science in the Public Interest*, 16(3), 75–109.
- Willmott, D., Hunt, D., & Mojtahedi, D. (2021). Criminal geography and geographical profiling within police investigations: A brief introduction. *Internet Journal of Criminology*.
- Wolfe, S. E., Rojek, J., Manjarrez, V. M., Jr., & Rojek, A. (2018). Why does organizational justice matter? Uncertainty management among law enforcement officers. *Journal of Criminal Justice*, 54, 20–29.
- Worrall, R., & Kjaerulf, F. (2018). Building collaborative capability between law enforcement and civil society leaders to prevent urban violence. *International Journal of Public Health*, 63, 969–976.

Copyrights

Copyright for this article is retained by the author(s), with first publication rights granted to the journal.

This is an open-access article distributed under the terms and conditions of the Creative Commons Attribution license (<http://creativecommons.org/licenses/by/4.0/>).