



**Institute of Technology Management and Entrepreneurship**

**INVESTIGATION ON PREDICTIVE POLICING ADOPTION,  
INNOVATIVE OFFICER PERFORMANCE AND CRIME  
MITIGATION AMONG ABU DHABI POLICE**

اونيور سيتي تیکنیکل ملیسیا ملاک  
UNIVERSITI TEKNIKAL MALAYSIA MELAKA  
**Hind Rashed Saleh Al Shamsi**

**Doctor of Philosophy**

**2024**

**INVESTIGATION ON PREDICTIVE POLICING ADOPTION, INNOVATIVE  
OFFICER PERFORMANCE AND CRIME MITIGATION AMONG ABU DHABI  
POLICE**

**HIND RASHED SALEH AL SHAMSI**



**Institute of Technology Management and Entrepreneurship**

**UNIVERSITI TEKNIKAL MALAYSIA MELAKA**

**2024**

## DEDICATION

This work is dedicated to the inspiring person in my life... my dear father and my dear mother who always wanted me to have the best, for her love, and the prayers that she made for me.

...To my husband ...

My husband is wonderful, a great companion and so much more in my life. He is always a source of love, courage and strength at every tough instant throughout these years.

...To my great guide...

My dear supervisor, Datuk Assoc.Prof.Dr.Su'aidi Dato' Safei for his visions in artificial intelligence.

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

I would like to appreciate the dedication of my beloved family who educated me and motivate me to learn until this level, and the lecturers and friends who constantly give me support and advice throughout the research. Without their blessing and encouragement, this research is impossible to complete on time

## ABSTRACT

In the face of rapid technological advancements and globalization, Abu Dhabi Police encounters challenges in the adoption of predictive policing. This approach, characterized by technological integration, data privacy concerns, and potential resistance among law enforcement personnel, requires overcoming obstacles such as training requirements and adapting to new methodologies. The global trend of widespread adoption of predictive policing, leveraging artificial intelligence and big data, underscores the urgency to combat crime, enhance surveillance, and keep law enforcement agencies abreast of criminal activities. In Abu Dhabi Police, predictive policing emerges as a potential linchpin in the criminal justice system, aiding investigations and bolstering public safety initiatives. Nonetheless, uncertainties surround the adoption behavior of this technological paradigm, prompting the study to delve into how predictive policing, incorporating key components of artificial intelligence and big data, can effectively mitigate crime through officer training and collaborative learning within the General Command of Abu Dhabi Police. The research draws on theoretical foundations such as activity theory, complexity theory, crime theory, and technology adoption to establish a conceptual framework for analysis. Embracing a quantitative and systematic approach rooted in positivism, the study employs a deductive approach for theoretical inference, employing theory and hypotheses to validate evidence in the field. A survey targeting 2,500 police officers engaged in crime scene management at Abu Dhabi Police, with a sample size of 357 ( $n = 357$ ) using a fairly stratified sampling approach within the criminal security sector, serves as the primary data collection method. The study utilizes a validated survey questionnaire, emphasizing validity and reliability to enhance overall research credibility. Prior to the main data collection, a pilot study was conducted to validate the data collection tool. The findings underscore that the adoption of predictive policing enhances officer performance in innovation, collaborative learning, and overall crime mitigation. Notably, the study reveals a positive correlation between predictive policing and innovative officer performance, with officer innovation performance subsequently positively impacting crime reduction performance. Collaborative learning serves as a significant mediator, enhancing the effect of officer innovation on crime reduction performance within Abu Dhabi Police. The implications of this study extend to government, corporate entities, academia, and society, offering insights into the transformative potential of predictive policing in enhancing law enforcement effectiveness and public safety.

**KAJI SELIDIK TERHADAP PENGGUNAAN KEPOLISAN  
RAMALAN, PRESTASI PEGAWAI INOVATIF DAN PENCEGAHAN  
JENAYAH DALAM KALANGAN POLIS ABU DHABI**

**ABSTRAK**

*Dalam menghadapi kemajuan teknologi dan globalisasi yang pesat, Polis Abu Dhabi menghadapi cabaran dalam mengamalkan kepolisan ramalan. Pendekatan ini, yang dicirikan oleh penyepaduan teknologi, kepentingan kerahsiaan data, dan potensi penolakan dalam kalangan anggota penguatkuasa undang-undang, memerlukan inisiatif mengatasi halangan seperti keperluan latihan dan menyesuaikan diri dengan metodologi baharu. Aliran global penggunaan meluas kepolisan ramalan, memanfaatkan kecerdasan buatan dan data besar, menekankan keperluan mendesak untuk memerangi jenayah, meningkatkan pengawasan dan memastikan agensi penguatkuasaan undang-undang terus berada di barisan hadapan dalam menangani aktiviti jenayah. Kepolisan ramalan dalam polis Abu Dhabi merupakan projek yang direkayasa sebagai sebahagian daripada strategi kecerdasan buatan di Polis Abu Dhabi. Kepolisan ramalan muncul sebagai tulang belakang yang berpotensi dalam sistem keadilan jenayah, membantu siasatan dan mengukuhkan inisiatif keselamatan awam. Memandangkan etika penggunaan paradigma teknologi baharu ini masih dipersoalkan, kajian ini bertujuan untuk menyiasat bagaimana penggunaan kepolisan ramalan (komponen utamanya kecerdasan buatan dan data besar) boleh membantu mitigasi jenayah melalui latihan pegawai dan pembelajaran kolaboratif - Perintah Am Polis Abu Dhabi. Kajian ini berasaskan teori utama, termasuk teori aktiviti, teori kerumitan, teori jenayah, dan penggunaan teknologi bagi mewujudkan rangka kerja konseptual untuk analisis. Melihat kepada teori-teori ini dan menyokong topik-topik kritikal kajian, rangka kerja konsep untuk analisis diperkenalkan. Penyelidikan menggunakan pendekatan kuantitatif dan sistematik dari perspektif positivisme. Pendekatan deduktif terhadap inferens teori juga berasaskan teori dan hipotesis bagi mengesahkan bukti teori yang wujud dalam lapangan terbabit. Tinjauan menyasarkan 2,500 anggota polis yang terlibat dalam pengurusan tempat kejadian jenayah di Polis Abu Dhabi, dengan saiz sampel 357 ( $n = 357$ ) menggunakan pendekatan persampelan yang agak berstrata dalam sektor keselamatan jenayah, berfungsi sebagai kaedah pengumpulan data utama. Kajian ini menggunakan soal selidik tinjauan yang disahkan, menekankan kesahan dan kebolehpercayaan untuk meningkatkan kredibiliti penyelidikan keseluruhan. Sebelum pengumpulan data utama, kajian rintis telah dijalankan untuk mengesahkan alat pengumpulan data. Dapatan kajian menemui kesan positif penggunaan kepolisan ramalan, yang meningkatkan prestasi pegawai dalam inovasi, pembelajaran kolaboratif, dan pengurangan jenayah secara keseluruhan. Kajian ini mendedahkan korelasi positif antara kepolisan ramalan dan prestasi pegawai yang inovatif, dengan prestasi inovasi pegawai. Pembelajaran kolaboratif berfungsi secara khusus dalam memudahkan kesan inovasi pegawai terhadap prestasi mitigasi jenayah dalam Polis Abu Dhabi. Implikasi kajian ini meluas kepada kerajaan, entiti korporat, akademia dan masyarakat, menawarkan pandangan tentang potensi transformatif kepolisan ramalan dalam meningkatkan keberkesanan penguatkuasaan undang-undang dan keselamatan awam.*

## ACKNOWLEDGEMENTS

Alhamdulillah, Praise to Almighty Allah (SWT) who blesses me for the success completion of my thesis. This doctoral research has been challenging with full of twist and turn, and finally reached a satisfying end. Thank you for Allah (SWT) and the prayers from my family and friends. I would never have done this alone without the support from them.

Most of all, I intend to take this chance to express my sincere acknowledgement to my supervisor, Datuk Assoc.Prof.Dr.Su'aidi Dato' Safei from the Institute of Technology Management and Technopreneurship (IPTK), Universiti Teknikal Malaysia Melaka (UTeM) for his exemplary guidance, essential supervision, support and constant encouragement with the completion of this thesis. I will carry along the blessing, guidance and advice that he gave from time to time in the journey of life on which I am about to embark.

Besides that, a special thanks to UTeM and all the staffs involved that provided various facilities and help towards the completion of this thesis. Next, an extremely evince my heartfelt appreciation to my beloved family and friends with their endless support, encouragement and care throughout completion of this thesis. Finally, an honestly heart to show my thankfulness to those who involved either directly or indirectly towards this thesis writing. Hopefully, this thesis would become a reference to the others in the future.

## TABLE OF CONTENTS

	<b>PAGE</b>
<b>DECLARATION</b>	
<b>APPROVAL</b>	
<b>DEDICATION</b>	
<b>ABSTRACT</b>	<b>i</b>
<b>ABSTRAK</b>	Error! Bookmark not defined.
<b>ACKNOWLEDGEMENTS</b>	<b>ii</b>
<b>TABLE OF CONTENTS</b>	<b>ii</b>
<b>LIST OF TABLES</b>	<b>vi</b>
<b>LIST OF FIGURES</b>	<b>viii</b>
<b>LIST OF ABBREVIATIONS</b>	<b>x</b>
<b>LIST OF APPENDICES</b>	<b>xi</b>
<b>LIST OF PUBLICATIONS</b>	<b>xii</b>
<b>CHAPTER</b>	
<b>1. INTRODUCTION</b>	<b>1</b>
1.1. Introduction	1
1.2 Background	1
1.3 Problem statement	5
1.4 Research questions	9
1.5 Research objectives	10
1.6 Significant of the research	11
1.7 Scope of the research	13
1.8 Limitations of the research	13
1.9 Operational definitions	14
1.10 Structure of the thesis	15
1.11 Summary	16
<b>2. LITERATURE REVIEW</b>	<b>17</b>
2.1 Introduction	17
2.2 Review on UAE Crime Index	17
2.3 AI in Crime	21
2.3.1 The Theoretical Basis of the study	23
2.3.2 Ethical Implications	24
2.4 Artificial intelligence (AI) in organizations	26
2.5 Factors for the AI technological shift in the UAE	28
2.6 Utilizing Artificial Intelligence to aid Government Operations	29
2.7 AI for citizen services and happiness	31
2.8 Theoretical framework of the research	34
2.8.1 The activity theory	35
2.8.2 The complexity theory	38
2.8.3 The theory of the crime	40
2.8.4 The technology adoption model	42
2.8.5 Collaborative cognitive load theory	43
2.9 Hypotheses Development	45

2.9.1	The contribution of predictive policing to innovative officer performance	45
2.9.2	Innovative officer performance and crime mitigation	46
2.9.3	The need for specialised training to control predictive policing	47
2.9.4	The ultimate need for collaboration in learning and training	48
2.10	Hypotheses development	48
2.11	Research gap	49
2.12	Research framework	50
2.13	Summary	51
<b>3. RESEARCH METHODOLOGY</b>		<b>52</b>
3.1	Introduction	52
3.2	Location of research	53
3.3	Research philosophy – critical realism	54
3.4	Research design	55
3.4.1	Approaches to theory development – abductive research	56
3.4.2	Methodological choices – quantitative research	57
3.4.3	Research strategy and time horizon – cross-sectional survey	58
3.5	Sources of data	59
3.6	Measurement of variables and instrumentation	59
3.7	Population and sampling techniques	62
3.7.1	Population	62
3.7.2	Sampling techniques	62
3.8	Data collection processes	64
3.9	Research credibility: validity and reliability	64
3.9.1	Pilot study	65
3.10	Data analysis techniques	66
3.10.1	Quantitative analysis	66
3.11	Ethics and limitations	67
3.12	Summary	68
<b>4. RESULT AND DISCUSSION</b>		<b>69</b>
4.1	Introduction	69
4.2	Preliminary data analysis and data screening	71
4.2.1	Response rate, input accuracy analysis and missing data	72
4.2.1.1	Response rate	72
4.2.1.2	Input accuracy	73
4.2.1.3	Unengaged responses and suspicious data	76
4.2.1.4	Missing data analysis and imputation	77
4.2.2	Normality assessment and outliers	79
4.2.2.1	Normality assessment	80
4.2.2.2	Outliers analysis	83
4.2.3	Correlations, multicollinearity and common method variance analysis	85
4.2.3.1	Correlations	85
4.2.3.2	Multicollinearity	86
4.2.3.3	Common method bias	87
4.2.4	Summary of preliminary analysis	89
4.3	Study demographics and profiles of study participants	90



4.3.1	Respondents distribution based on gender	90
4.3.2	Respondents distribution based on age group	91
4.3.3	Respondent distribution across organisational levels	92
4.3.4	Analysis of respondents based on experience	94
4.4	Descriptive statistics on key variables of the study	95
4.5	Exploratory factor analysis (EFA)	98
4.5.1	KMO, Bartlett's test and goodness of fit tests	98
4.5.2	Variance explained analysis and scree plot	101
4.5.3	Factor structure analysis and pattern matrix	104
4.5.4	Summary of EFA analysis	106
4.6	Global and local tests for measurement model	107
4.6.1	Measurement model	107
4.6.2	Factor loadings and model fit analysis (Global tests)	110
4.6.3	Validity and reliability of measurement model (Local Tests)	113
	4.6.3.1 Reliability analysis	113
	4.6.3.2 Validity analysis	115
4.6.4	Summary of global and local tests for measurement model	117
4.7	Hypotheses testing	118
4.7.1	Predictive policing and innovative officer performance	120
4.7.2	Innovative officer performance and crime mitigation performance	121
4.7.3	The moderation effect of technology training on predictive policing and officer innovation	123
4.7.4	The mediatory role of collaborative learning on officer innovation on crime mitigation performance	125
4.8	Additional control analysis and competing models	127
4.8.1	Level in Organisation as a control variable within the model	127
4.8.2	Time spent in organisation (experience) as a control variable within the model	128
4.9	Discussions	129
4.9.1	The role of predictive policing in innovative officer performance	129
4.9.2	The role of innovative officer performance in crime mitigation performance	130
4.9.3	Moderating impact of specialized technology training	131
4.9.4	Mediating impact of collaborative learning	132
4.10	Summary of results and analysis chapter	132
<b>5. CONCLUSION AND RECOMMENDATION</b>		<b>135</b>
5.1	Introduction	135
5.2	Readdressing the research objectives	135
5.3	Contributions of the research	143
	5.3.1 Theoretical contribution	143
	5.3.2 Practical contribution	145
	5.3.3 Contribution for research methodology	146
5.4	Limitations of the research	146
5.5	Research recommendations and future researchs	148
5.6	Conclusion of the research	149

**REFERENCES**  
**APPENDICES**

**152**  
**167**



## LIST OF TABLES

<b>TABLE</b>	<b>TITLE</b>	<b>PAGE</b>
Table 1.1	Research questions and hypotheses for quantitative phase	9
Table 1.2	List of definitions	14
Table 2.1	Restatement of the research hypotheses and key literature support	49
Table 3.1	Measurement of variable	60
Table 3.2	Sampling size table (Saunders et al., 2012)	63
Table 3.3	Expert and pilot of data collection instruments	65
Table 3.4	Statistical reliability of data collection instrument	66
Table 4.1	Independent test for difference among early and late response	75
Table 4.2	Variable summary	78
Table 4.3	Model summary – normality assessment	80
Table 4.4	ANOVA – normality assessment	81
Table 4.5	Coefficients - normality assessment	81
Table 4.6	Cook's distance scores for extreme cases	83
Table 4.7	Correlation between constructs	85
Table 4.8	Multicollinearity statistic	86
Table 4.9	Total Variance Explained –Common Method Bias Assessment	88
Table 4.10	Gender	90
Table 4.11	Age group	91
Table 4.12	Level in organisation	93
Table 4.13	Time with the Abu Dhabi Crime Scene Department	94
Table 4.14	Descriptive statistics	96
Table 4.15	KMO and Bartlett's Test – initial EFA	99

Table 4.16	Goodness of fit Test – Initial EFA	100
Table 4.17	Communalities Table- Initial EFA	100
Table 4.18	Total Variance Explained – Initial EFA	102
Table 4.19	Rotated Component Matrix – Initial EFA	105
Table 4.20	Main model – R squared Change to test common method variance	108
Table 4.21	Outer Loadings and Significance Levels	111
Table 4.22	Model Fit Indices	112
Table 4.23	R-Squared Statistic – test for significance	113
Table 4.24	Adjusted R-Squared Statistic – test for significance	113
Table 4.25	Cronbach’s Alpha test for reliability	114
Table 4.26	Composite Reliability test results	114
Table 4.27	rho A test for reliability	115
Table 4.28	Average Variance Extracted	116
Table 4.29	HTMT Analysis of Discriminant Validity (Inter-factor)	116
Table 4.30	Latent Variable Correlations	117
Table 4.31	Path Coefficients	118
Table 4.32	Total Indirect Effects	119
Table 4.33	Specific Indirect Effects	119
Table 4.34	Model Controlled for level in organisation	127
Table 4.35	Model Controlled for time spent in organisation	128
Table 4.36	Hypotheses summary	133

## LIST OF FIGURES

FIGURE	TITLE	PAGE
Figure 1.1	Abu Dhabi Police GHQ strategy for moving from reactive to predictive	5
Figure 1.2	Contribution of AI to GDP	29
Figure 2.1	An illustration of the relationships between automation, Digital revolution, and AI technologies	28
Figure 2.2	Activity Theory Model	37
Figure 2.3	Operational predictive Policing Model	46
Figure 2.4	Conceptual framework	50
Figure 3.1	Research onion	53
Figure 3.2	United Arab Emirates	54
Figure 3.3	Research Flow	56
Figure 4.1	Summary of missing values	77
Figure 4.2	Missing data pattern analysis – initial analysis	79
Figure 4.3	Histogram – regression standardized residual	82
Figure 4.4	Normal P-P plot of regression standardized residual	82
Figure 4.5	Initial Cooks Distance	84
Figure 4.6	Cooks Distance - Final	84
Figure 4.7	Gender	91
Figure 4.8	Age group	92
Figure 4.9	Level in organisation	93
Figure 4.10	Time with the Abu Dhabi Crime Scene Department	95
Figure 4.1	Scree Plot – Initial EFA	104
Figure 4.12	Main Model (Bootstrapping Results – With Marker)	109



## LIST OF ABBREVIATIONS

AVE	-	Average variance extracted
ADP	-	Abu Dhabi Police
AV	-	Analysis of Variance
CD	-	Standard deviation
CFA	-	Confirmatory factor analysis
CR	-	Composite reliability
EFA	-	Exploratory factor analysis
AI	-	Artificial Intelligence
GCC	-	Gulf Corporation Council
GoF	-	Goodness of fit
PS	-	Participative style
SDG	-	sustainable development goals
SEM	-	Structural equation modelling
SPSS	-	Statistical package for the social science
UAE	-	United Arab Emirates
VIF	-	Variance inflation factor

## LIST OF APPENDICES

APPENDIX	TITLE	PAGES
A	Information sheet	167
B	Informed consent form	168
C	Survey questionnaire	169
D	Quantitative data summary	172





## LIST OF PUBLICATIONS

Hind Rashed Saleh Al Shamsi, Su'aidi Safei. 2023. PLS-SEM Model of Crime Mitigation Performance of Abu Dhabi Police Predictive Policing. *Tropical Scientific Journal* (ISSN: 2710-5997). Vol 2, Issue 2, 2023. URL: <https://scientificacademic.com/index.php/tsj/article/view/22>.

Hind Rashed Saleh Al Shamsi, Su'aidi Safei. 2023. Artificial Intelligence Adoption in Predictive Policing to Predict Crime Mitigation Performance. *International Journal of Sustainable Construction Engineering and Technology*. Vol. 14 No. 3 (2023) 289-298. DOI: <https://publisher.uthm.edu.my/ojs/index.php/IJSCET/article/view/15323/5911>.



# CHAPTER 1

## INTRODUCTION

### 1.1. Introduction

This chapter provides an insight into this dissertation titled “Use of Artificial Intelligence in Abu Dhabi Police to Mitigate Crime”. This chapter begins with the background of the study that discusses the overview of the issues in this area. In section 1.3, the study presents problem statements that show the gaps in the study. To address the problem statement, the study presents the research questions and research objectives in sections 1.4 and 1.5. Next, the significance of the study was presented to show the relevance to conduct the study in section 1.6. This chapter continues with the scope of the study, and the limitation of the research in sections 1.7 and 1.8. The next section presents the operation definitions to indicate the terminological terms of the variables. In section 1.10, the study shows the structure of the whole thesis. Lastly, in section 1.11 the study presents the summary of Chapter 1.

### 1.2 Background

The future of policing is no doubt inter-locked with the application of computers to complement traditional policing techniques (Adams, 2012). Predictive policing constitutes the use of technology to predict potential criminal activity using various forms of intelligence algorithm, simulation, and big data analytics of huge amounts of data (Ferguson, 2019). Predictive policing has recently been used by law enforcement agencies to direct patrol officers, target expected crimes, and conduct real-time analysis in the critical response division of the law enforcement organisation (Saunders et al., 2016). This next level of smart

police innovation builds on voluminous data with historical and real-time properties to create a smart policing weapon with the potential to mark crimes before they occur (Mohler et al., 2015; Ferguson, 2019).

Despite the potential of predictive policing through the use of AI and big data for the purpose of crime mitigation in law enforcement organisations, significant challenges have been highlighted (Ekblom, 2013; Meijer and Wessels, 2019; Richardson et al., 2019). The use of such systems has remained questionable in the wake of jeopardised public safety, the use of "dirty data" and "bad predictions", with increasing rights violations (Richardson et al., 2019). Meijer and Wessels (2019) in a systematic literature review, highlighted that predictive policing is being hit by a number of drawbacks, including crime increase and the lack of proper implementation. Ultimately, strong tension exists between predictive policing and the police officer's craft (Ratcliffe et al., 2019). The need to resolve these tensions is critical to the effective application of predictive policing to realise the inherent benefits (Ferguson, 2019).

With the need for training, substantial costs, and the lack of willingness of police officers to change their role by adopting the new technology systems, seriously threaten predictive technology operationalisation (Ferguson, 2019). Ferguson (2016) asserts that training is critical to release the burden of effective and efficient application of predictive policing technology. This leads to the ultimate conclusion that the adoption of predictive technologies necessitates that officers go through the right training. Nonetheless, evidence exists that officers are often uncomfortable with paperwork and technology systems; the overall commitment to the use of such systems has remained low (Kirschner et al., 2018).

Effective training is essential to the accurate input of data, proper use of predictive policing methodologies, overcoming of system vulnerabilities and the correct interpretations of predictive policing system responses (Mohler et al., 2015; Ferguson, 2016; Ferguson,

2019). The adoption of innovation training and learning mechanisms have as well encouraged to realise the long-anticipated benefits of predictive policing (Saunders et al., 2016; Kirschner et al., 2018; Meijer and Wessels, 2019; Ratcliffe et al., 2019; Ferguson, 2019).

In the bid to mitigate and prevent crime in the UAE, the region has witnessed an increased role of artificial surveillance, predictive policing, and the use of big data officers to curb crime and facilitate crime investigation (MENA Herald, 2018; Shouk 2019a). AI and big data have already proven effective in ending crime (zero crime) in dedicated communities in Dubai (Shouk. 2019b). Through AI data analytics in a special AI-driven program called the "Oyoon" (meaning Eyes in English), the Dubai Police was able to solve crimes and identify security gaps across, residential, industrial, shopping and all community centres, in an integrated data-driven AI system that operates independently of human intervention across thousands of cameras within the city (Shouk. 2019b). According to Col Suleiman Al Kaabi, director of innovation and foresight at Abu Dhabi Police:

*"AI is capable of changing our ideas and vision thanks to a new system that will govern humanity... when borders no longer exist between what is normal and artificial, real and superficial, and human and artificial, AI will put in place mechanisms that can learn, think and take decisions," Malek (2018).*

The Abu Dhabi Emirate has prepared its citizens to thrive in a future dominated by artificial intelligence, and the Abu Dhabi Police and other law enforcement agencies in the other Emirates are playing their individual and unique roles towards the realisation of this vision (Malek, 2018; Ramahi, 2018; Larsen, 2017; Harrison, 2019). However, the area of officer training has been ignored, especially regarding newly introduced automated

intelligent tools and how to make the best use of big data (Babuta, 2017). Babuta (2017) observes that law enforcement officials and detectives manually trawl through vast amounts of data, often unaware that they have access to automated tools that could save them a significant amount of time. Evidence exists that thorough technology training is often absent. However, analytical tools are only as effective as the individual operating them, and investment must be directed at ensuring that the officers are able to use any new technology systems effectively.

As part of the Abu Dhabi Police AI strategy, the Department plans to advance from the current proactive policing to the full operationalisation of predictive policing (Figure 1.1) (Abu Dhabi Police GHQ). Starting from the reactive stage, the police department would often react to crime events after they are reported. No form of actions to prevent crime happen before this stage. Here, the efficiency and effectiveness of crime prevention lie in the responsiveness of the police to the crime call, and the time it takes the police personnel to reach the crime scene. Proactive policing represent a stage where the police take some pre-emptive action towards crime mitigation. These actions are not often based on intelligence but on general data available on crime hotspots. Frequent patrols and surveillance, police engagement, alongside other proactive moves to curb crime.

The final stage of policing strategy that the Department is working towards is predictive policing. At this stage, the police apply various forms of technology to facilitate and create the intelligence necessary to react before a crime occurs. This ensures that the police remains ahead of crime and is more capable of controlling crime seamlessly. In this quest, the use of Big Data and AI have gained popularity, as argued in the context of the present study. Given this strategic path, the time is right that closer attention is dedicated to the adoption and proper application of predictive policing through effective training and collaborative learning (Kirschner et al., 2018).

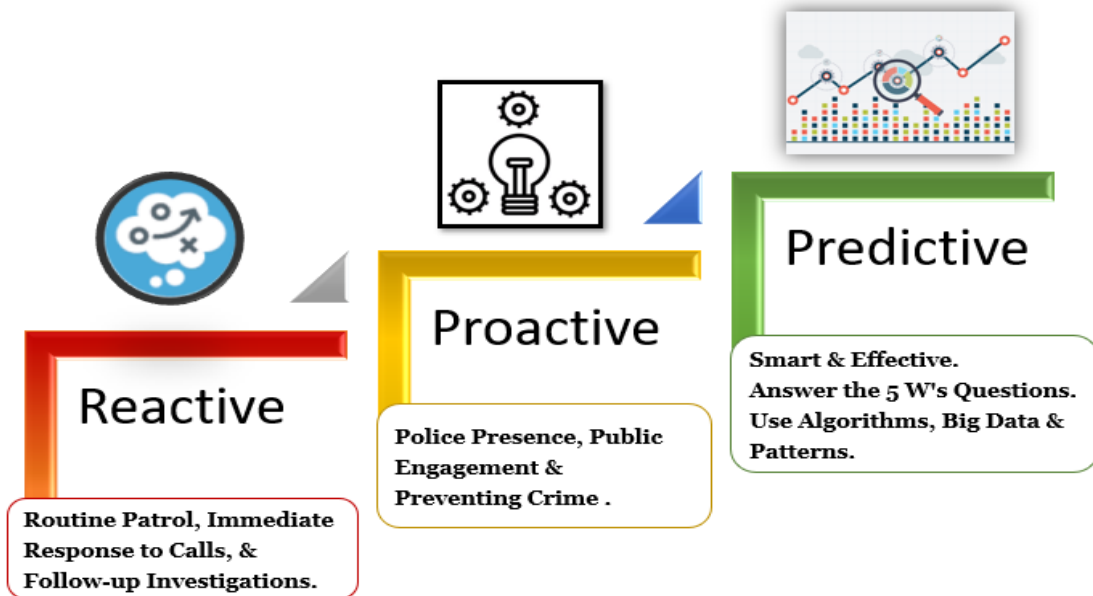


Figure 1.1: Abu Dhabi Police GHQ strategy for moving from reactive to predictive

Source: Abu Dhabi Police GHQ (2020)

### 1.3 Problem statement

The rapid evolution of technology, particularly Artificial Intelligence (AI), is disrupting established norms and ushering in an era of uncertainty. AI, as a pioneering technology, has demonstrated its efficacy in managing intricate tasks involving real-time data processing, signal interpretation, and knowledge accumulation (Krasadakis, 2018).

In the United Arab Emirates (UAE), the government is harnessing innovation to reshape perspectives, enhance industrial prospects, and bolster emergent sectors like healthcare (Halaweh, 2018; Wehbe and Svetinovic, 2018). A standout approach involves deploying AI across vital domains, deviating from incremental improvements and leading to significant advancements (Halaweh, 2019). Research by Bessen (2018) underscores how AI streamlines work processes, optimizes resource allocation, and elevates citizen services. Moreover, AI's potential to create sophisticated roles, such as AI web programmers, artificial engineers, and robotics experts, highlights its impact on workforce augmentation. Bessen

(2018) adds that AI addresses data challenges, enhances cognitive processes, and advances predictive capabilities, enabling governments to make informed policy decisions.

The UAE government's utilization of AI is evident in driving business efficiency, enhancing employee productivity, and revolutionizing customer experiences through technology-driven sales recommendations and fraud detection (Shah and Shaheen, 2016). The UAE's AI strategy (2017) underscores the imperative of AI in achieving a customer-centric approach, guiding service representatives in delivering effective solutions. Additionally, AI offers cost-saving potentials, with estimated annual savings of billions through automation (UAE AI strategy, 2016). The strategy highlights the possibility of redirecting resources towards enhancing service quality and accommodating evolving needs.

The UAE government's commitment to innovation extends to public services, industrial opportunities, and national well-being (UAE AI strategy, 2016). The strategic employment of AI in public services, particularly predictive policing, seeks to expedite service delivery and bolster overall service quality (Dave and Sharma, 2019). To ensure alignment with national visions (UAE vision 2021; UAE vision 2030), AI technology is harnessed to enhance citizen satisfaction.

Despite these advancements, the effective operationalization of predictive policing technology faces challenges related to alignment with police officers' expertise (Ratcliffe et al., 2019). The opacity of predictive models leads to errors that hinder effective application (Meijer and Wessels, 2019; Ferguson, 2019). Meijer and Wessels (2019) emphasize the need for officers to comprehend predictive algorithms for optimal decision-making. Addressing the lack of evidence on specialized training for AI and big data usage in crime mitigation, this study aims to shed light on this area (Ferguson, 2019). Collaborative learning, informed by educational psychology theory, is also crucial for optimizing officer performance (Kirschner et al., 2018).

The adoption of predictive policing, innovative officer performance, and crime mitigation initiatives within law enforcement agencies presents a multifaceted landscape marked by several challenges and issues (Meijer, and Wessels, 2019). In the rapidly evolving technological environment, law enforcement agencies, including the Abu Dhabi Police, encounter substantial hurdles related to the integration of predictive policing technologies (Kaufmann et al., 2019).

These challenges encompass issues such as technology integration complexities, potential resistance among law enforcement personnel, and critical concerns regarding data privacy and security (Hardyns, and Rummens, 2018). As predictive policing relies heavily on advanced technologies like artificial intelligence and big data analytics, the transformative impact of these tools necessitates a careful consideration of the ethical, legal, and operational implications surrounding their implementation (Alikhademi et al., 2022).

Furthermore, the effectiveness of innovative officer performance, a key component in the predictive policing paradigm, faces its own set of challenges (Galiani and Jaitman, 2023). Overcoming barriers related to training requirements and the adaptation to new methodologies emerges as a crucial aspect (Galiani and Jaitman, 2023). Officers need to acquire new skills and competencies to effectively utilize predictive technologies, and this transition may encounter resistance or reluctance within the existing law enforcement culture (Ugwudike, 2022). The successful integration of predictive policing technologies also demands a reevaluation of traditional policing practices, necessitating a cultural shift and comprehensive training programs (Birks et al., 2023).

Predictive policing, driven by artificial intelligence and big data, is gaining global prominence due to its potential to enhance crime surveillance and aid law enforcement agencies in staying ahead of criminal activities (Ajil and Staubli, 2023). However, the ambitious goals of these technological advancements are met with skepticism and concerns