Understanding methylamphetamine drug markets in an Australian context

Ву

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A thesis submitted for the

Degree of Doctor of Philosophy (Science)

University of Technology Sydney

Certificate of authorship and originality

I, Ana Popovic declare that this thesis, is submitted in fulfilment of the requirements

for the degree of Doctor of Philosophy, in the Faculty of Science at the University of

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Acknowledgements

I would like to express my gratitude to my supervisors Assoc. Prof. Alison Beavis, Prof. Claude Roux and Dr. Marie Morelato for their continuous support during my PhD. Without their guidance and encouragement, the goals of this project would not have been achieved.

I am also grateful for the support of the Australian Federal Police and staff, specifically Dr. Mark Tahtouh and Dr. Adrian De Grazia. I would like to thank Dr. Mark Tahtouh for providing me with data that was pivotal to giving this project substance. Additionally, I would like to acknowledge that this work was supported by an Australian Research Council grant (LP160100352).

I would like to express my thanks to Bojana Popovic and Sara Dawood for their continuous editorial services. Without their input this thesis would be a sea of red and blue squiggly lines.

Nobody has been more important to me in the pursuit of this project than the members of my family. I would like to thank my parents and siblings for supporting me throughout writing this thesis and my life in general. I wish to thank Sandra for her continuous love and encouragement. Additionally, I would like to thank Daniel for his inspiring chats, especially during late nights and early mornings. They have all kept me going and this work would not have been possible without their involvement.

Publications and presentations

Publications:

A. Popovic, M. Morelato, C. Roux, A. Beavis, Review of the most common chemometric techniques in illicit drug profiling, Forensic Science International 302 (2019) 109911.

A. Popovic, M. Morelato, C. Roux, A. Beavis, Interpreting the link value of similarity scores between illicit drug specimens through a dual approach, featuring deterministic and Bayesian frameworks, Forensic Science International, 319 (2021) 10651.

Presentations:

A. Popovic, M. Morelato, C. Roux, A. Beavis (2019). *Chemometrics in illicit drug intelligence*. FoSTER, University of Technology Sydney, Australia, 29 November 2019.

A. Popovic, C. Roux, A. Beavis (2018). *Chemometrics in illicit drug intelligence*. 24th International Symposium on the Forensic Sciences, ANZFSS, Perth, Australia, 9-13 September 2018.

A. Popovic, C. Roux, A. Beavis (2018). *Chemometrics in illicit drug intelligence*. 8th European Academy of Forensic Science (EAFS), Lyon, France, 27-31 August 2018 (poster).

A. Popovic, C. Roux, A. Beavis (2017). *Illicit drug profiling in forensic intelligence (street seizures)*. Doctoral school, Les Diableretes, Switzerland, 28-31 August 2017.

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Abbreviations

2PG 2-Path Graph

ACT Australian Capital Territory

AFP Australian Federal Police

AIDIP Australian Illicit Drug Intelligence Program

AM Amphetamine

ATS Amphetamine-Type Stimulants

AUC Area Under the Curve

CA Clustering Analysis

CAN Canberra

CC Chemical Class

CE Capillary Electrophoresis

CHAMP Collaborative Harmonisation of Methods for Profiling of

Amphetamine-Type Stimulants

CMP 1-(1',4'-cyclohexadienyl)-2-methylaminopropane

CNS Central Nervous System

DAD Diode Array Detector

d-MA *d*-Methylamphetamine

d-MA-E d-Methylamphetamine Enriched

DNV Dynamic Network Visualisation

ELSD Evaporative Light Scattering Detector

ENIPID Enhanced National Intelligence Picture on Illicit Drugs

EPH Ephedrine

EUC Euclidean

FID Flame Ionisation Detector

FN False Negative

FP False Positive

Abbreviations

FR Fruchterman-Reingold

GC Gas Chromatography

HCA Hierarchical Clustering Analysis

HS Headspace

ICP Inductively Coupled Plasma

IDDR Illicit Drug Data Report

ILP Intelligence Led Policing

IRMS Isotopic Ratio Mass Spectrometry

KK Kamada-Kawai

L Logarithm

LC Liquid Chromatography

LLE Liquid-Liquid Extraction

l-MA *l*-Methylamphetamine

I-MA-E *I*-Methylamphetamine Enriched

LPAC L-Phenylacetylcarbinol

LR Likelihood Ratio

MA Methylamphetamine

MAN Manhattan

MCF Modified Cosine Function

MDMA 3,4-Methylenedioxymethylamphetamine

MLP Multi-Profile

MOP Mono-Profile

MS Mass spectrometry

N Normalisation

NA Not Applicable

NIM National Intelligence Model

NMI National Measurement Institute

NSW New South Wales

NT Northern Territory

PCA Principal Component Analysis

PCC Pearson Correlation Coefficient

PSE Pseudoephedrine

QLD Queensland

RAC Racemic

RIS Regulation Impact Statement

ROC Receiver Operating Characteristic

RT Reingold-Tilford

SA South Australia

SCF Squared Cosine Function

SNA Social Network Analysis

SPME Solid-Phase Microextraction

TAS Tasmania

TN True negative

TP True positive

VIC Victoria

VRR Variability Reduction Rules

WA Western Australia

WDR World Drug Report

WH Working Hypothesis

Abstract

Decisions by law enforcement agencies regarding crime disruption, prevention and reduction rely on significant volumes of information from various sources. There is a desire to use this information for generating intelligence to support proactive policing rather than reactively responding to crime. Utilising an existing proof of concept, this research explored the application of chemometric techniques to chemical profiles of state-level methylamphetamine seizures, acquired from the Australian Federal Police through the Enhanced National Intelligence Picture on Illicit Drugs capability. The main aim was to create and deliver a methodology that would expand the use of illicit drug profiling for strategic and operational intelligence purposes, to be more effective in the fight against illicit drug trafficking.

The use of comparison metrics and clustering analysis to determine links between illicit drug specimens was evaluated and automated. The scores resulting from comparison metrics were evaluated through two established approaches, i.e. deterministic and Bayesian approaches. Results showed that using the two approaches in combination provided more information about linkages than when either approach was used independently.

Relational, temporal, spatial and quantitative analyses were subsequently used to gain an insight into illicit drug markets. Relational analysis identified clusters of seizures central to the network. Temporal analysis then provided insights into the behaviour of distribution markets, specifically the emergence and extinction of certain groups of seizures over time. Finally, spatial analysis aided the understanding of the inter-jurisdictional nature of illicit drug markets. These analyses allowed for the generation of strategic intelligence relating to when and where the Australian illicit drug market was the most active. Additionally, the strategic level trends identified clusters of seizures which were worth investigating further. These clusters were explored through a case study to provide drug market knowledge at an operational level.

The presentation forensic case data, for intelligence or court purposes, typically involve the preparation of static reports. The final aim of this project was the creation of a visualisation tool, which was created to enhance the way processed data was conveyed. This tool was produced in the form of a web application and was used to aid the exploration of drug markets. It provided an automated way of analysing the forensic case data and producing relevant visualisations in an interactive and timely manner.