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Chapter 28

Introducing Predictive Policing Technologies (PPT): An Action Research–Oriented Approach for EBOCD Initiatives

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ABSTRACT

This reflective case-history presents the findings of a 12-week pilot study of a collaborative organizational change project which oversaw the implementation of predictive policing technology (PPT) into a territorial police force in the North of England. Based on the first year of a two-year initiative, the reflections consider the impact on the future of the project and their potential future application and cultural embeddedness, beyond the organizational and time-bound specifics of this case.

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INTRODUCTION

This reflective case-history presents the case of a 12-week pilot-study of a collaborative organizational change project that sought to oversee the successful implementation of a ‘predictive policing technology’ (PPT) into a territorial police force in the North of England (West Yorkshire Police, referred to henceforth as ‘the Force’). We critically reflect on the process of this evidence-based organizational change and development (EBOCD) initiative, the immediate impact of the initiative, and the research findings. In doing so we provide observations regarding the implementation and use of such technologies and the challenges they represent in relation to organizational change and culture. The question underpinning this research was, ‘How can predictive policing technology be culturally embedded?’ Our hope is that the findings from this pilot-study can be applied more widely as other districts move to engage with similar technologies as part of further Home Office and policing initiatives (Grierson, 2016).

Context and Drivers of the EBOCD Initiative

The 12-week pilot that forms the focus of this case-history was the first phase of an ongoing ‘action research’ oriented organizational change project between the Force and a team of academics from a range of backgrounds, disciplines and institutions. The project was funded by the UK Home Office and forms part of a wider strategy that sees academic expertise aligned with a range of challenges and crime-prevention initiatives identified in territorial police forces throughout England and Wales. The overarching aim of the collaboration was to facilitate the successful implementation of PPT, based on a number of regional and criminological factors.

The rationale for these initiatives was premised on statistical evidence which correlates the numbers of police officers available for deployment and patrol in relation to the statistics of reported crime. According to UK Home Office figures, in 2009 the number of police officers in England and Wales stood at 143, 769. Following progressive cuts to the public sector, by 2016 this figure fell to 124, 066; a fall of 14% (Harrison, 2015; BBC News, 2016). Even though reported crime-rates tended to fall during this period, by December 2016 this trend had stalled and in some areas (such as fraud) it had reversed.¹ Consequently, police forces in England and Wales have come under increasing amounts of pressure to deliver more with less. A central challenge is that many forms of crime prevention are based around officers’ presence preventing crimes being committed (Farrington, MacKenzie, Sherman, & Welsh, 2003). Therefore, with decreasing numbers of “bobbies on the beat” (Hopkins, 2015), the successful direction and presence of resources to the “right place at the right time” has positioned ‘predictive policing technology’ as a cornerstone for preventative crime measures in the new digital age of policing on both sides of the Atlantic (Bachner, 2013; Holt, 2017).

One response has been to consider the use of predictive analytical software to aid the efficient and effective deployment of ‘visible’ patrols. In recent years, a significant increase in the volume, velocity, veracity, variety, and value (referenced in Rahman, 2016; Rahman & Aldhaban, 2015) of data (‘big data’) has meant that organisations in a range of sectors have sought to leverage the data available to enable ‘probabilised’ decision-making processes (Allenby, Bradlow, George, Liechty, & McCulloch, 2014).

The foremost expectation held by organizations regarding big data’s potential is based around ‘predictability’ (Agarwal & Dhar, 2014; Bughin, Chui, & Manyika, 2010; Hashem et al., 2015). In crude terms, the size of data available is in positive correlation to the leverage against risk. In other words, the

accrual and analysis of big data will provide the opportunity to move on from ‘present action’ and ‘past reflection’ towards a ‘calculable future’, derived from evidence-based data incorporated into an algorithm.

Predictive Policing

Predictive policing has become a key feature of what is known as ‘Intelligence-led Policing’ (ILP, (Ratcliffe, 2016). According to Beck and McCue (2009, p. 22) predictive policing is concerned with ‘[t]he ability to anticipate the time, the location, and the nature of the crime’ in order to ‘[support] the police manager’s ability to proactively allocate resources – preventing or deterring crime through targeted police presence and enabling response by pre-positioning police assets when and where they are likely to be needed.’ This approach is based on predictions derived when large data-sets are processed by an algorithm. The logic behind the algorithm is that certain types of crime follow predictable patterns and therefore resources can be deployed to prevent a crime or act as a deterrent through their presence in specific locations. The use of PPT requires a mediated delivery of the output of reported crime as intelligence through a variety of digital technologies (ICT, smart phones and apps etc.).

However, the success of PPT can be inhibited by two types of factors: factors relating to the predictive ability of the technology to identify the appropriate time and place of the officers’ deployment, and factors relating to the technology’s uptake, which are based on elements such as the views of the officers, their beliefs in evidence-based practice, the supporting systems that inform and direct them on patrol, and the management of compliance with these requirements. Often best intentions are thwarted through the unintended selection of inappropriate techniques (College of Policing, 2015) or a failure of operational officers to be convinced of the value of patrol plans that are based on prediction, leading to a failure to comply with requirements and implement evidence-based crime prevention (Farrington et al., 2003). Likewise, the perceived or real lack of supporting systems and infrastructure is seen as a factor inhibiting effective and efficient operational delivery. The focus of this reflective case history is primarily in relation to the second type of factors; i.e., relating to the acceptance of the technology and the change in working practices that it entailed.

Organizational Analysis: Making Sense of Technology Acceptance

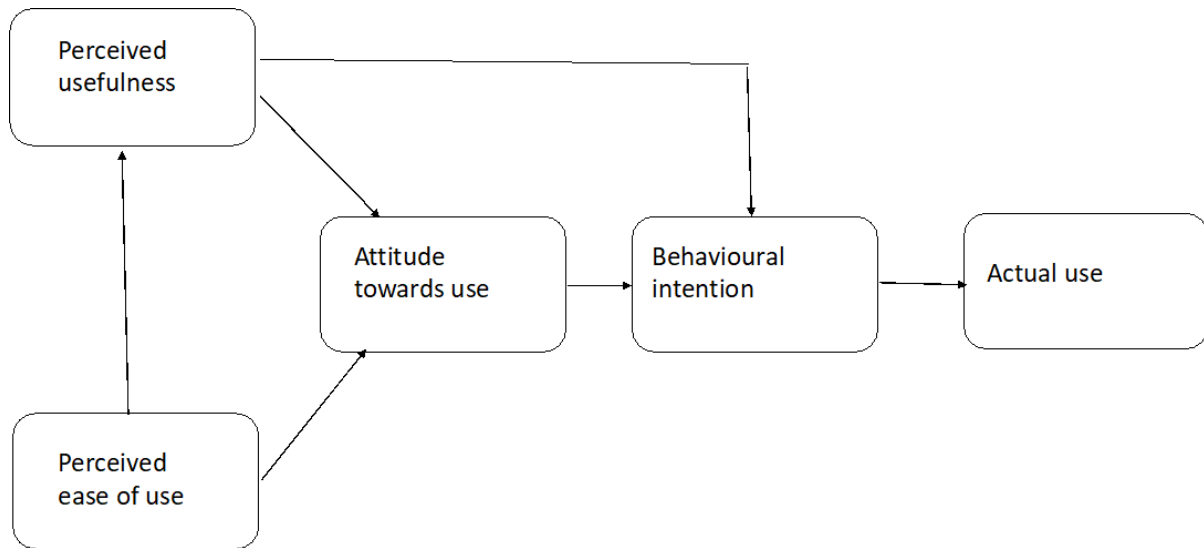
It is a commonplace observation that any attempt to bring about change in organisations may be greeted by a range of responses, from enthusiastic support to indifference or opposition (Beckhard & Harris, 1987). A change may have different potential impacts on those affected by it, and thus will be perceived as a positive development by some stakeholders, and as a negative imposition by others. Attitudes towards a change may be influenced by the manner in which it is introduced, and the extent to which those affected by it accept this method of introduction (Balogun, Hailey & Gustafsson, 2016).

In specific relation to changes concerning the adoption of new technology, the Technology Acceptance Model (TAM) (Davis, 1986, 1989) proposes key factors that are likely to affect the behavioural response to the requirement to change (see Figure 1). Key factors influencing the take-up of information technology by users are the ‘perceived usefulness of the technology’, and its ‘perceived ease of its use’. Later developments of TAM included social influence factors (from colleagues and bosses) that affected perceptions of usefulness and behavioural intention to use the technology (Holden & Karsh, 2010; Turner, Kitchenham, Brerton, Charters & Budgen, 2010).

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Figure 1. The TAM

Source: from Holden & Karsh 2010, p. 161)



There is a lack of evidence-based accounts of TAM in police settings, however Lindsey, Jackson and Cooke's model for 'mobile policing' (2011, 2014) was deemed sufficient to provide a framework for a model to apply in our case, adapted to fit the specific requirements and challenges presented by PPT in this context.

We employed TAM in the first instance as an analytical framework, which allowed us to embark on a process of organizational sense-making. This is different from our methodology for reasons we outline below. It was through this analytical framing that we could employ action modes of research (Raelin, 2009) to help gain acceptance of the technology through the production of actionable knowledge (Argyris, 1996) for its progressive implementation throughout the Force.

EVIDENCE-BASED OCD INITIATIVE

An algorithm was developed by a university-based technology team that processed reported crime data for a specific area and generated maps of streets deemed to be at higher risk of further crime, thus indicating priority areas to patrol. The technology was designed to update the maps every 24 hours and to communicate them to the teams of officers responsible for patrolling the area via a tailored application on their hand-held devices (smartphones).

The project was funded by the Home Office, and a project manager within the Force liaised with the range of officers who would be involved, and with academics from three universities who were active in developing and analysing the project. A specific inner city area was chosen to pilot the technology. The focus of the first pilot was chosen by the senior leadership team for the division, and was the crime of domestic burglary.

Action research provided an overall framework for understanding and guiding the project: a collaborative approach was taken to working with police officers at all levels of the Force, to design the

pilot, to monitor progress and to address emergent problems as they arose, at the same time generating knowledge as evidence for use in a progressive and cyclical process, starting with diagnosis, questions and planning, leading to action taking (Lewin, 1946) and generating actionable knowledge for wider dissemination (Argyris, 1996).

Meetings were held with senior officers (superintendent and inspectors) responsible for the pilot area, to discuss the working of the project, and to acquire from them a statement of their aims in taking part. From these meetings, three key indicators of success were named:

- A reduction in burglary from dwellings;
- A reduction in calls for service;
- An increase in public confidence.

A meeting was also held with a group of sergeants, to discuss the project and to emphasise the importance of their role. What little research there is on police acceptance of mobile technology (Lindsay et al., 2011, 2014) highlights the influential role of sergeants. Further briefing meetings were held between the project manager, a member of the university research group, and individual sergeants and the police constables (PCs) and the police community support officers (PCSOs) who would use the maps as they patrolled the streets.

The pilot was scheduled to last for 12 weeks. The PCs, PCSOs and the sergeants were asked to use the maps and to report back on their experiences on a regular basis. During the pilot period, 16 meetings were held with sergeants and the officers in their teams at the start of their shift, allowing learning about the working of the technology to be shared with the group and the university researchers. The articulation of what was being learned enabled evidence to emerge of what was working, and what was helping or hindering. The experiences and perceptions of the PCs and PCSOs were sought in a group setting, and then some further information was gathered from each sergeant in a separate interview. Some meetings were also attended by an inspector or the superintendent, to gather information and to contribute to the evaluation.

In the first part of the pilot phase, selected officers in each team were asked to use the technology, and their time was protected ('ring-fenced'). In the second part of the pilot, this ring-fencing was relaxed, and the technology was used by a wider range of officers in the team. Both approaches were evaluated to assess which was more effective.

As the pilot proceeded, different aspects of the use of the technology were discussed and evaluated. Action was taken to make improvements and deal with difficulties, and longer-term actions were identified for future implementation. Information from the team meetings, and from meetings with other officers, was analysed through thematic analysis (Braun & Clarke, 2013), allowing the emergent evidence to be thematised. From this the research team augmented the basic elements of a Technology Acceptance Model (TAM) for use in future phases of the project.

FINDINGS

Certain themes emerged from our discussions with the users of the new system. Of particular interest were factors that appeared to help the system to work, and factors that hindered the system, and thus influenced adoption by the users. Throughout the pilot, officers engaged in processes of sensemaking

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regarding the new technology, and made adjustments to the ways in which the technology impacted on the work of officers.

Minimal requirements for adoption were the support of senior managers (the Chief Constable, the divisional superintendent and inspectors) and the availability of resources. The early meeting with the divisional superintendent and inspectors, where measures of success for the pilot were discussed and agreed, was an important step. The support of this group of senior managers was also signaled by individuals attending some of the feedback meetings with sergeants and officers.

Resourcing challenges were partly eased by funding from the Home Office for the project, which enabled the project manager to provide the necessary smartphones for PCSOs, and enabled the divisional superintendent to authorize overtime during the pilot period. Resourcing still remained an issue, however, with operational calls on the time of PCs and PCSOs potentially conflicting with the requirements of the system, and some officers seeking time off rather than more overtime. There is a tension between spending time on crime prevention – ‘proactive policing’ (Clarke, 2006) – and spending time reacting – ‘reactive policing’ (Scott, 1998) – to reported incidents. This tension between ‘proactive’ and ‘reactive’ strategies is central to evidence-based policing matrices (Lum, Koper, & Telep, 2011).

Beyond these minimal requirements of support from senior managers and sufficient resources, major themes arising from discussions with officers concerned the perceived value of the system, which was closely linked to its perceived credibility. Credibility concerned a) the perceived likelihood of the accuracy of the system, b) the perceived plausibility of actual predictions, and c) the effectiveness and reliability of the technology.

In initial briefings for the sergeants and officers, and in subsequent meetings with them, the project manager explained the theory behind patterns of domestic burglaries, using practical examples and demonstrating a good understanding of police approaches to this type of crime. She was also able to explain the volume and type of data that was used by the algorithm, and could give examples of similar systems being used elsewhere. She and the university researchers also emphasized that they wished to get feedback from officers as the pilot progressed, in order to learn how the system could be improved, accepting that adjustments would probably need to be made.

In the early stages of the pilot, queries were raised by some officers about specific predictions, where maps showed locations that did not appear to contain likely premises, or conversely where maps did not indicate a location close to a recent reported crime. In one meeting, officers openly queried the value of the algorithm as compared to the knowledge they had gained through practical experience. The project manager regularly explained that the information from the algorithm should be used alongside the assessment of experienced officers as to which areas to patrol. The question of whether this technology should complement the judgement of experienced officers rather than over-ride it is an important issue in acceptance and operation. As a result of feedback, some adjustments were made to how the algorithm generated maps.

A limitation of the system in the pilot phase was that very little additional information was provided – such as the reason why a particular location had been highlighted for preventive policing. Some officers said that more up-to-date intelligence reports should accompany the patrol plans, to give them a sense of why they were being asked to patrol certain locations, so that the technology could complement their ‘craft’ and ‘beat knowledge’. During the pilot phase, however, it was not possible to link the patrol plans with intelligence in a timely manner.

The effectiveness and reliability of the technology was a third aspect of credibility. Technological ‘teething problems’ in the early days of the pilot resulted in the maps not changing every 24 hours, as promised, but remaining the same for two weeks. The perceived credibility of the system suffered early damage as a result.

While the perceived value of the system was affected by perceptions of its credibility, another factor of concern was whether it was or would actually be effective in deterring crime.

Deterrence of a crime such as burglary through showing police presence is normally only evidenced in retrospect. During the pilot, officers patrolled the streets indicated by the maps, in addition to their other duties. They were encouraged to be observant and to point out potential risks – such as windows left open – to householders. However, it was not clear to them at the time they were patrolling whether they were being effective in deterring crime. As one sergeant said: ‘If you are wandering around and nothing’s happening, it’s hard for people to see that they are doing a good job’.

Figures on burglaries and incidents were gathered and analyzed at the end of the pilot period. The results were:

- In the pilot area, burglary dwellings (i.e., domestic burglaries) had fallen relative to the same period the previous year, from 274 to 202 crimes: a reduction of 35.64%
- In the pilot area, burglary dwellings had fallen relative to the previous three months, from 253 to 202 crimes: a reduction of 25.25%
- In the whole district, these crimes had risen relative to the same period the previous year, from 986 to 998: an increase of 1.2%.
- In the whole district, these crimes had risen relative to the preceding three months, from 989 to 998: an increase of 0.9%

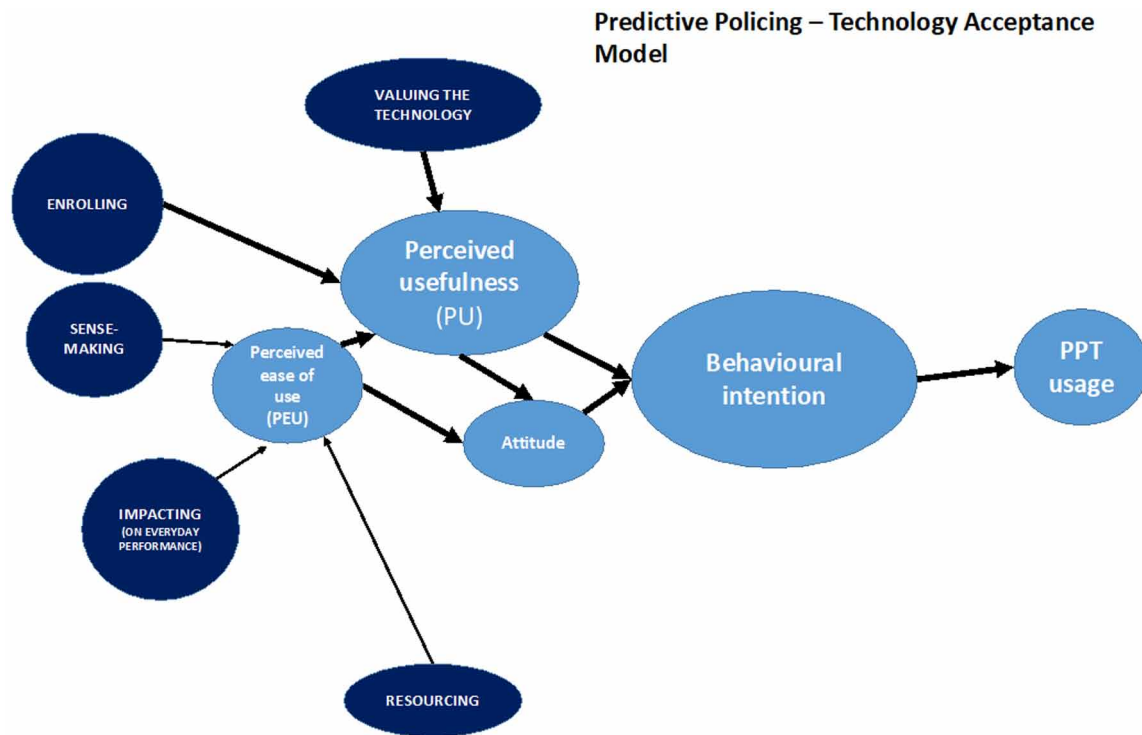
Of course, these figures need to be treated cautiously: we are not able to draw firm conclusions on causal links between the project and the figures, but these were at least some positive signs, and they reinforced senior managers’ support for the project. Within the teams led by the sergeants, ways of working with the technology had evolved and become accepted. For example, it was found more effective to allocate PPT maps throughout the whole team, rather than ring-fence the time of particular individuals. As one sergeant said at the end of the pilot period: ‘It’s become part of daily business now.’ The review by senior officers at the conclusion of the pilot suggested that the briefing of officers against the maps had become considered ‘normal business’ with discretion left to sergeants and officers, which is continuing.

DISCUSSION

A variation of the Technology Acceptance Model, the PP-TAM Version 1, was developed in order to analyze the process in this case (see Figure 2). Sensemaking was a constant process, influenced by previous experiences and by attitudes expressed by colleagues, as well as by communications with the project manager and more senior officers. The perception of the extent to which the technology impacted on everyday performance in a positive way, or a negative way, affected perceptions of ease of use, as did resourcing. A key influence on perceived usefulness is the extent to which those involved in the project value the technology, and a critical process is the extent to which various actors, technologies and maps can be enrolled into the activity.

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Figure 2. A PP-TAM



Evaluation

It is difficult to determine the extent to which the theory underpinning TAM was successful in our attempt to implement the predictive technology. As stated, the principles underpinning the core of the model have been adapted by others to meet the specific organizational formations and challenges of the Force's culture and structure. Indeed, the quantitative success story that can be told by the comparative figures are only indicators of success. Moving forward, what we have is a working PP-TAM that can continue to be adapted as new data related to the new organizational, methodological, technological and criminological challenges are revealed. This is a crucial basis on which the success of future phases of the project can be based.

Old Insights, New Applications

A central challenge that the pilot uncovered was the view of the technology as something that should *complement* rather than *determine* the conduct of the officers on the beat. At first this might appear to be an obvious and mundane observation. There are certainly more than enough accounts of technologically-oriented organizational change initiatives taking a technological determinist view to development needs. However, moving forward, this challenge is a factor to be reiterated and considered further due to its organizational pertinence regarding the reception to the technology by the Force. This is principally due to the hierarchical structure of the Force and the officer's conduct being channeled by response to 'commands' and the direction of patrol-plans based on intel. As we move forward with the next phases

of the project, there are specific challenges as we transferred to different districts, crime-focuses, organizational dynamics (non-ring-fenced provisions) etc. However, underpinning all of these is the essential factor of complementarity in the success of the technology in this pilot phase. This is something that others can learn from, both within the specific Force we are engaged with and more widely as others move forward with similar technologies in other regions as part of wider Home Office and policing initiatives (Grierson, 2016).

CONCLUDING REFLECTIONS

In this chapter we have provided a case history to show how action modes of research, principally action research and action learning research could both support learning of participants and provide evidence to develop a PP-TAM. While the model is based around the conceptualization of abstracted themes, it is made meaningful in the context of continuing work within the Force and beyond through its connection to the evidence. The collaborative nature of this project (between academic researchers and serving police personnel in West Yorkshire) combined with the action modes of research has allowed the benefits of regular access to project participants to develop this actionable knowledge into this early formation discussed in this chapter. The evidence-based nature of this inquiry, and the consolidative nature of the project is deemed of value by both academics and practitioners.

REFERENCES

- Agarwal, R., & Dhar, V. (2014). Editorial—big data, data science, and analytics: The opportunity and challenge for IS research. *Information Systems Research*, 25(3), 443–448. doi:10.1287/isre.2014.0546
- Allenby, G. M., Bradlow, E. T., George, E. I., Liechty, J., & McCulloch, R. E. (2014). Perspectives on Bayesian Methods and Big Data. *Customer Needs and Solutions*, 1(3), 169–175. doi:10.100740547-014-0017-9
- Argyris, C. (1996). Actionable knowledge: Design causality in the service of consequential theory. *The Journal of Applied Behavioral Science*, 32(4), 390–406. doi:10.1177/0021886396324004
- Bachner, J. (2013). *Predictive policing: preventing crime with data and analytics*. Johns Hopkins University, IBM Center for the Business of Government.
- Beck, C., & McCue, C. (2009). Predictive policing: What can we learn from Wal-Mart and Amazon about fighting crime in a recession? *The Police Chief*, 76(11), 18–25.
- Braun, V., & Clarke, V. (2013). *Successful qualitative research: a practical guide for beginners*. London: Sage.
- Bughin, J., Chui, M., & Manyika, J. (2010). Clouds, big data, and smart assets: Ten tech-enabled business trends to watch. *The McKinsey Quarterly*, 56(1), 75–86.

Introducing Predictive Policing Technologies (PPT)

Davis, F. D. (1986). *A technology acceptance model for empirically testing new end-user information systems: Theory and results*. Massachusetts Institute of Technology. Retrieved from https://www.researchgate.net/profile/Fred_Davis2/publication/35465050_A_technology_acceptance_model_for_empirically_testing_new_enduser_information_systems__theory_and_results_/links/0c960519fbaddf3ba7000000.pdf

Davis, F. D. (1989). Perceived usefulness, perceived ease of use, and user acceptance of information technology. *Management Information Systems Quarterly*, 13(3), 319–340. doi:10.2307/249008

Farrington, D. P., MacKenzie, D. L., Sherman, L. W., & Welsh, B. C. (2003). *Evidence-based crime prevention*. London: Routledge. doi:10.4324/9780203166697

Grierson, J. (2016, February 24). Police and academics developing system to map crime hotspots. *The Guardian*.

Harrison, J. (2015, May 20). Are bobbies on the beat “endangered”? *BBC News*. Retrieved from <http://www.bbc.co.uk/news/uk-32807677>

Hashem, I. A. T., Yaqoob, I., Anuar, N. B., Mokhtar, S., Gani, A., & Khan, S. U. (2015). The rise of “big data” on cloud computing: Review and open research issues. *Information Systems*, 47, 98–115. doi:10.1016/j.is.2014.07.006

Holt, J. (2017, June 23). “Right place at right time” nets burglary suspects. *The Signal*. Retrieved from <https://signalscv.com/2017/06/right-place-right-time-nets-burglary-suspects/>

Hopkins, N. (2015, October 27). Bobbyes on the beat coming to an end, senior officers say. *BBC News*. Retrieved from <http://www.bbc.co.uk/news/uk-34651119>

Lewin, K. (1946). Action research and minority problems. *The Journal of Social Issues*, 2(4), 34–46. doi:10.1111/j.1540-4560.1946.tb02295.x

Lindsay, R., Jackson, T. W., & Cooke, L. (2011). Adapted technology acceptance model for mobile policing. *Journal of Systems and Information Technology*, 13(4), 389–407. doi:10.1108/13287261111183988

Lindsay, R., Jackson, T. W., & Cooke, L. (2014). Empirical evaluation of a technology acceptance model for mobile policing. *Police Practice and Research*, 15(5), 419–436. doi:10.1080/15614263.2013.829602

News, B. B. C. (2016, July 21). *Police officer numbers drop by nearly 20,000 since 2009*. Retrieved from <http://www.bbc.co.uk/news/uk-36857326>

Raelin, J. A. (2009). *Seeking conceptual clarity in the action modalities (SSRN Scholarly Paper No. ID 1558086)*. Rochester, NY: Social Science Research Network.

Rahman, N. (2016). *Factors affecting big data technology adoption*. Student Research Symposium. Retrieved from <http://pdxscholar.library.pdx.edu/studentssymposium/2016/Presentations/10/>

Rahman, N., & Aldhaban, F. (2015). Assessing the effectiveness of big data initiatives. In *Management of Engineering and Technology (PICMET), 2015 Portland International Conference on* (pp. 478–484). IEEE. Retrieved from <http://ieeexplore.ieee.org/abstract/document/7273189/>

Ratcliffe, J. H. (2016). *Intelligence-led policing*. London: Routledge. doi:10.4324/9781315717579

Scott, J. (1998). 'Performance culture': The return of reactive policing. *Policing and Society*, 8(3), 269–288. doi:10.1080/10439463.1998.9964792

Turner, M., Kitchenham, B., Brereton, P., Charters, S., & Budgen, D. (2010). Does the technology acceptance model predict actual use? A systematic literature review. *Information and Software Technology*, 52(5), 463–479. doi:10.1016/j.infsof.2009.11.005

ENDNOTE

- ¹ For more information on crime statistics and trends in the England and Wales, see <https://www.ons.gov.uk/peoplepopulationandcommunity/crimeandjustice/bulletins/crimeinenglandandwales/yearendingdec2016>, accessed June 2017.