

1 Security and the Smart City: A Systematic Review

2 Author name and affiliation:

3 1. Julian Laufs (corresponding author) [✉]*julian.laufs@ucl.ac.uk*

4 2. Hervé Borrion ^[SEP]*h.borrion@ucl.ac.uk*

5 3. Ben Bradford ^[SEP]*ben.bradford@ucl.ac.uk*

6

7 ^a UCL Jill Dando Institute of Security and Crime Science

8 35 Tavistock Square

9 WC1H 9EZ

10 London

11 United Kingdom

12

13 Abstract

14 The implementation of smart technology in cities is often hailed as the solution to many urban
15 challenges such as transportation, waste management, and environmental protection. Issues of
16 security and crime prevention, however, are in many cases neglected. Moreover, when researchers
17 do introduce new smart security technologies, they rarely discuss their implementation or question
18 how new smart city security might affect traditional policing and urban planning processes. This
19 systematic review explores the recent literature concerned with new ‘smart city’ security
20 technologies and aims to investigate to what extent these new interventions correspond with
21 traditional functions of security interventions. Through an extensive literature search we compiled
22 a list of security interventions for smart cities and suggest several changes to the conceptual status
23 quo in the field. Ultimately, we propose three clear categories to categorise security interventions
24 in smart cities: Those interventions that use new sensors but traditional actuators, those that seek
25 to make old systems smart, and those that introduce entirely new functions. These themes are then
26 discussed in detail and the importance of each group of interventions for the overall field of urban
27 security and governance is assessed.

28 **Keywords:** Systematic review, smart cities, safe city, security technologies, security functions

29 **1. Introduction**

30 Rapid urbanisation and progress in information and communication technologies (ICT) are two
31 of the most important phenomena impacting urban security planning and governance today
32 (Cocchia, 2014; Zhu, Li, & Feng, 2019). The latter, especially, has shaped the concept of smart
33 cities, an increasingly popular idea in recent years (Albino, Berardi, & Dangelico, 2015; Naphade,
34 Banavar, Harrison, Paraszczak, & Morris, 2011; Ralko & Kumar, 2016). The implementation of
35 smart city technology is hailed as the solution to many urban challenges such as transportation,
36 waste management, and environmental protection (Alawadhi et al., 2012; Ankitha, Nayana,
37 Shravya, & Jain, 2017; Gohar, Muzammal, & Rahman, 2018; Lella, Mandla, & Zhu, 2017; Zhang,
38 Wan, Yang, & Yang, 2017a; Zhang et al., 2017b). While these issues are the focus of a growing
39 debate about smart city development, aspects of security and crime prevention are often neglected
40 (Ralko & Kumar, 2016).

41 As a result, the implications of new smart city security systems for crime reduction, security,
42 and urban governance are rarely discussed. This systematic review attempts to address this gap by
43 exploring the last ten years' worth of literature on new security technologies that can be considered
44 to fall under the smart city concept. It aims to investigate the extent to which these new
45 interventions correspond with traditional functions of security interventions, and how they affect
46 urban planning and governance. Through an extensive literature search and an analysis of 121
47 studies, this article compiles a list of security interventions for smart cities, discusses and contrasts
48 their functions with those of more traditional interventions, before ultimately proposing several
49 changes to the conceptual status quo in the field.

50 In the following, we provide background information on the role of security in urban planning
51 and smart cities. We then outline the core methodological principles and search strategy used in
52 this review before presenting and discussing the findings.

53

54

55 **2. Background: Smart cities and urban security defined**

56 More than half the world’s population lives in cities, and this unprecedented level urbanisation
57 is only set to increase in the future (Jalali, El-Khatib, & McGregor, 2015; Zhang et al., 2016).
58 While this shift has improved the life of many, the explosion of urban populations has led to far-
59 reaching problems in many cities around the globe (Zhang et al., 2017b). To meet these challenges,
60 governments are increasingly turning to the use of information and communications technologies
61 (ICTs) and are aiming (or at least claiming) to make cities ‘smart’.

62 While the term ‘smart city’ is today widely used, there is no coherent and broadly accepted
63 definition; in fact, Ramaprasad, Sánchez-Ortiz, and Syn (2017) , find 36 different definitions from
64 disciplines as diverse as urban studies, computers and information technology, sociology, and
65 public health. Because this paper focusses largely on the technological aspects of smart cities and
66 implications for urban planning and security, it adopts the definition suggested by Elmaghraby &
67 Losavio (2014) of a smart city as a city that uses “information and communication technologies to
68 increase operational efficiency, independently shares information within the system, and improves
69 overall effectiveness of services and the wellbeing of citizens”. Ramaprasad et al. (2017) provide
70 a holistic discussion on the term and discuss different definitional approaches.

71 The term ‘smart city’ has spread around the globe, affecting urban development programmes
72 and government strategies (Berry, 2018). Many government initiatives seek to create a broad range
73 of services, ranging from smart transport and smart energy to smart citizens and education (Hall et
74 al., 2000). Such future cities are heralded for their efficient use of ICTs embedded within the fabric
75 of urban environments that aim to improve and rationalise public services in the future (Berry,
76 2018). These futuristic scenarios, however, often fail to recognise safety and security as a focus
77 (Hartama et al., 2017). This is critical, as it is not only one of the most basic tenets of urban planning
78 and management but also of human wellbeing — after all, safety and security are on the second
79 bottom layer of the Maslow pyramid (McLeod, 2007). As such, safety and security constitute

80 factors that are integral parts of human well-being and as such also of any smart city design (Reddy,
81 Suresh, Phaneendra, Shin, & Odelu, 2018a).

82 While some articles acknowledge that rapid urbanisation leads to challenges for traditional
83 safety and security infrastructure in cities (Isafiade & Bagula, 2017) and that these are critical issues
84 for contemporary integrated urban developments (Benkő & Germán, 2016), this is only rarely
85 reflected in the literature. Most of the studies that explore the impact of emergent ICTs have done
86 so through critiques of governmental programmes, drawing largely upon insights of large scale
87 evaluative studies and, in many cases, avoiding the discussion of real technological developments
88 (Berry, 2015).

89 Slowly, however, the realisation that crime and security problems are not isolated but often
90 impact all other factors of city life, and as such should become a central issue in the creation of
91 smart cities, has gained traction (Borrion et al., 2019). Thus, this paper follows the notion that
92 proactively ensuring the security and safety of the public is a basic operation of smart cities
93 (Bourmpos, Argyris, & Syvridis, 2014).

94 One approach that aims to reconcile issues of crime prevention with new smart city
95 developments is the safe city concept (Hartama et al., 2017). While initially conceived as a
96 framework for safety for natural disasters, it quickly came to cover all aspects of safety within the
97 city. In particular, the concept seeks to reconcile urban growth with the need for security through
98 a variety of technological functions and by optimising the allocation of law enforcement resources
99 (Castelli, Sormani, Trujillo, & Popovič, 2017; Oatley, Crick, & Bolt, 2015).

100 Furthermore, a safe city describes integration of technology and the natural environment that
101 “enhances the effectiveness and efficiency of the process of handling the threat of crime and terror,
102 to enable the availability of a healthy environment for citizens, and access to health, rapid response
103 to emergencies” (Hartama et al., 2017). Effectiveness and efficiency in smart cities, however, imply
104 far more than only efficacy or financial concerns (i.e. whether the designated task has been
105 completed and how much it costs). They also include issues of citizen satisfaction and whether the

106 innovation has created a benefit to those subjected to the intervention and beyond. This is
107 imperative, as citizens are in the end at the centre of any urban safety intervention and central to
108 creating a safe environment (Cagliero et al., 2015). Thus, gauging the perceptions of citizens on
109 urban security is a key point in Smart City management, as it will ensure that cities not only prevent
110 or respond to safety risks and security threats but that they also remain an attractive place to live in
111 (Cagliero et al., 2015).

112 Beyond these definitional issues, it is important to understand what a smart city practically
113 entails and how it functions. It is difficult to conceive of a *general* architecture for smart cities
114 because of the extremely diverse range of devices, technologies, and services that may be
115 associated in such a system, and because of the high degree of interdependence between various
116 components (Jalali et al., 2015). As such, there are many different models that discuss what
117 components and infrastructures a smart city needs (Gaur, Scotney, Parr, & McClean, 2015). Most,
118 if not all, of these smart city architectures contain, however, three basic layers: A sensor layer, a
119 network or processing layer, and a service or actuator layer (Filipponi et al., 2010; Gaur et al.,
120 2015; Jalali et al., 2015; Zhang et al., 2017b).

121 The sensor layer consists of the various (often heterogeneous) data collection units (i.e.
122 sensors). These can be deployed to measure almost anything in the city landscape. Examples
123 include environmental factors like brightness or sound, cameras, RFID tags to monitor entire
124 objects, or even participatory sensing through social media (Jalali et al., 2015). Data from this
125 sensor layer is then delivered to the respective actuators via the network layer. This second layer
126 provides the communication infrastructure to transport the data but also aggregates data from
127 different sensors (Filipponi et al., 2010). The last layer then contains actuators, i.e. those units that
128 bring about a physical change in the environment or provide the required service (in our case fulfil
129 a crime prevention function) (Gaur et al., 2015). Most relevant for our discussion of smart security
130 interventions are the sensor and the actuator layer, which is why they will be highlighted in the
131 following discussion.

132 This distinction between the different layers is also useful to understand the smart city as a
 133 complex system made up of various components on different levels, reaching from single sensors
 134 to software and servers that integrate them and ensure communication between them (Zhang et al.,
 135 2017b). Crime prevention interventions in the sense of this paper are thus specific technological
 136 solutions that seek to address a distinct (crime-related) problem and make up one or several
 137 components of a smart city infrastructure either on one or on multiple layers.

138 The extensive literature search upon which our arguments are based sought to give an overview
 139 of the variety of functions new security technologies might fulfil. Overall, it aims to augment but
 140 also challenge the current conceptualisation of emergent technologies as crime prevention
 141 measures for smart cities. By switching the focus to the ‘functions’ of these technologies (i.e., their
 142 direct/proximal effects on the environment), this paper seeks to bridge the gap between the bigger
 143 picture of safe cities and security on one hand and deeply technological solutions on the other.

144 **Table 1:** *Examples of technological components on different layers of smart city infrastructure. A single*
 145 *intervention may combine different components from one or multiple layers.*

Sensor Layer	Network Layer	Actuator Layer
RFID sensor	Transmission technologies	Retractable barricade
CCTV camera	Processing/computing units	Police response
Facial recognition camera	Compression/analysis software	Streetlights
Microphones	...	Speakers
Motion detection		Adaptable signage
WIFI-access points		UAV swarm
Crowd-sourcing app		Alarm
Light sensor		...
...		

146

147

148

149

150

151 **3. Literature search**

152 *3.1 Search terms*

153 Two methods were followed to narrow down the search terms for this review. As the term
154 ‘smart city’ is contested and not consistently used throughout the literature, this article used the
155 results of Cocchia's (2014) study to supplement the search strategy. Cocchia found that there is no
156 coherent definition of the word 'smart' and that its use (along with other related labels) is often
157 arbitrary, while at the same time identifying several core terms that are frequently used
158 interchangeably. In addition, scoping searches were carried out to find appropriate search terms
159 related to security and crime prevention.

160 Wildcards were used to include variants of words with the same word stem (e.g. ‘offend*’
161 would identify terms such as offend, offender and offending). The terms ‘police’ and ‘policing’
162 were preferred over the wildcard ‘polic*’ which returned an abundance of results related to policy.
163 Thus, two categories of search terms were used:

164 1. Terms related to ‘smart city’, including ‘future city’, ‘intelligent city’, ‘digital city’

165 AND

166 2. Terms related to crime prevention, including ‘crim*’, ‘secur*’, ‘offend*’, ‘police’,
167 ‘policing’, ‘law enforcement’

168 *3.2 Inclusion/exclusion criteria*

169 The results were screened against the following pre-set inclusion and exclusion criteria:

170 • Only literature from the past ten years was included (2009 - 2018) to ensure that interventions
171 were most relevant to today's smart city environments.

172 • Only literature that was available in English and German was included for practical reasons.

173 • Literature that was otherwise unobtainable or that was missing full-text or abstract was also
174 excluded.

175 • To circumvent the pitfall of publication selection bias, grey literature was included in the
176 review (following Mlinarić, Horvat, & Šupak Smolčić, 2017; Wilson, 2009) a review based on a

177 biased collection of studies is likely to produce biased conclusions (Rohstein & Hopewell, 2009).
178 However, this does not mean that all studies, regardless of methodological quality, were included.
179 Instead, only those that met methodological quality criteria were included in the review to ensure
180 an unbiased collection of high-quality empirical work aimed at answering the research question.

181 After a first round of sifting with the above-mentioned criteria, the following hierarchically
182 layered selection criteria were employed:

183 • Articles must have thematic relevance (e.g. articles that mentioned either of the search terms
184 as part of an enumeration were not considered, e.g. '*smart city technology* encompasses advances
185 in transport management, *crime prevention* and other city services'.)

186 • Outputs had to have a focus on technology (e.g. articles should introduce or evaluate new
187 technologies). Because smart cities do to a large extent depend on the innovation of existing
188 systems, works that suggested improvements to currently existing security interventions were also
189 included.

190 • Outputs should be related to crime prevention or the improvement of public safety/security

191 • Outputs that focussed on new crime opportunities in smart cities rather than crime prevention
192 were excluded. This included literature on cybercrime opportunities or cybersecurity in smart cities
193 unless they also made reference to opportunities to prevent those crimes.

194 3.3 Search strategy for identification of studies

195 Searches were carried out on the following search engines:

196 • General databases: Scopus, Web of Science, Proquest, Zetoc

197 • Technology specific databases: IEEE Xplore, ACM Digital Library

198 • Grey Literature Databases: British Library EThOS; Open Grey did not return any results.

199 Backward and forward searches were carried out once relevant articles were identified. This,
200 however, did not yield any additional results.

201

202

203 3.4 Filtering stages

204 The reference list and sifting process were managed using the EPPI Reviewer 4 software. After
205 all duplicates and articles that did not meet the basic inclusion and exclusion criteria were removed,
206 the title and abstract of the remaining papers were scanned against the layered selection criteria.
207 For those studies that were included based on title and abstract, the full-text was reviewed against
208 the same criteria again to ensure that only relevant studies would be included in the final analysis.

209 3.5 Inter-Rater Reliability (IRR)

210 To ensure good inter-rater reliability and to avert personal biases in the selection of the studies,
211 the original coding results were verified by four other coders. Each of the coders was assigned a
212 random sample of 100 studies. The sample size was selected to ensure that coders became familiar
213 with the criteria (Belur, Tompson, Thornton, & Simon, 2018). When the results were compared,
214 there was a 94 per cent agreement between the four coders. In the case of most disagreements, the
215 ‘correct’ coding (or that which was the final agreed coding) was usually that which had been agreed
216 on by a majority of coders. Disagreements that remained were discussed in the group and brought
217 to a resolution by elaborating the overall aim of the review. The discussions highlighted a lack of
218 clarity on some aspects of the inclusion and exclusion criteria, especially on issues of research
219 design, methodology, and type of outcome measure, but also more fundamentally about how to
220 screen studies that did not meet the inclusion criteria but might nevertheless be relevant.

221 Following the suggestion by Feng (2014) to improve accuracy, chance agreement was removed
222 from the estimation of reliability by calculating the κ -statistic (see also Belur et al., 2018; Viera &
223 Garrett, 2005). With a κ -statistic of 0.81 and above in three of the four cases, we achieved near
224 perfect agreement between the coders (Landis & Koch, 1977). Only in one case, a κ -statistic of
225 0.72 was reported, which however, still indicated substantial agreement (Landis & Koch, 1977).
226 Overall the inter-rater reliability tests showed a high agreement between coders and thus
227 strengthened the validity of this review.

228

229 **4. Synthesis approach**

230 Though some authors such as Wilson (2009) suggest that the credibility of a systematic review depends
231 more on the number of studies used than on the method of synthesis, we will still briefly introduce the
232 approach taken for grouping and analysing the included studies.

233 While the aim of any synthesis is to generate new knowledge grounded in the information of the
234 individual research studies, the right methodological path to this new knowledge is not set in stone and
235 depends heavily on the individual review (Thomas, O'Mara-Eves, Harden, & Newman, 2017a). Since this
236 review spans across a variety of academic disciplines and fields, a thematic synthesis approach was chosen
237 as the modus of analysis as it is especially suitable for analysing multidisciplinary datasets (Thomas et al.,
238 2017a).

239 To address the research aims, we identified common themes across the included studies and analysed
240 them in detail. As a starting point for this process, we used conceptualisations of traditional security
241 functions for both the sensor layer and the actuator layer (Borrion, Tripathi, Chen, & Moon, 2014; Ekblom
242 & Hirschfield, 2014) but then employed an iterative and flexible approach (Gough, Oliver, & Thomas,
243 2017). This means that while the review builds on a foundation of open questions and some secure initial
244 concepts, it is equally thematically grounded in the studies it contains (Thomas et al., 2017a).

245 The initial concepts used in this review should be seen as a starting point that introduces a
246 common language to compare and contrast the identified intervention, rather than a rigid theoretical
247 framework. Their sole purpose was to provide a common denominator (i.e. the clustering of
248 security technologies by their function) for developing new themes from the included studies
249 (Boyatzis, 1998).

250 **5. Initial concepts**

251 In the following, we lay out key functions of security interventions both as sensors (i.e. for
252 threat detection) and actuators (i.e. for crime prevention). The functions on both the sensor and the
253 actuator layer are critical to the implementation of effective and efficient security systems. Table 2
254 brings together two conceptualisations to form a new set of initial concepts. The table merges the

255 functions contributing to threat detection as identified by Borrión et al. (2014) with the functions
 256 pertaining to crime prevention as identified by Ekblom & Hirschfield (Ekblom & Hirschfield,
 257 2014). The network layer was left out because there are no distinct frameworks that specify
 258 different functions on this layer and because they are not uniquely pertaining to crime detection or
 259 prevention technologies.

260 *Table 2: Security functions on different layers of smart city infrastructure*

Situation Awareness – Focus on Sensor Layer	Intervention – Focus on Actuator Layer(after Ekblom & Hirschfield, 2014)
Detect: e.g. determining the presence of certain anomalies, substances, individuals or behaviours (Hardmeier, Hofer, & Schwaninger, 2005)	Defeat: physically block access and movement or block/obscure the information that offenders want to collect
Authenticate: e.g. verifying that an individual is a member of staff or that they have the right to access (after Adey, 2002)	Disable/Deny: equipment helpful to offenders such as bugs or cameras
Identify: e.g. determining the name of a given chemical substance (Federici et al., 2005)	Direct/Deflect: offenders towards/away from place or behaviour
Locate: e.g. determining the location of individual passengers considered as potential threats to the infrastructure (Lee, Smeaton, O'Connor, & Murphy, 2005)	Deter-known offenders know what the risk of exposure is and judge it unacceptable so abandon/ abort attempt
Profile: e.g. classifying passengers who fit the profile of an offender for extra security checks (Sweet, 2008)	Deter-unknown: offenders uncertain what control methods they are up against, so again judge risk of exposure unacceptable
Track: e.g. following the movement of certain passengers through station premises (McCoy, Bullock, & Brennan, 2005)	Discourage: offenders perceive effort too great, reward too little, relative to risk, so abandon/abort attempt
	Demotivate: awakening, of offenders, emotions contrary to the mission, e.g. empathy with victims, removing excuses, coward image
	Deceive: offenders act on wrong information and are exposed to arrest or intelligence collection, frustrated, or mistakenly decide not to select this site as target
	Disconcert: causing offenders to make an overt involuntary movement or otherwise become startled

Detect¹: passive, and active exposure to make offenders self-expose by instrumental, expressive or involuntary action; by making legitimate presence/ behaviour distinctive; and by improving capacity of people exercising security role to detect

Detain: once offenders are detected, they must be caught and held (or credible identifying details obtained so they can be traced)

Inform (i.e. communicate): e.g. raising an alarm or calling in armed units in response to a detected threat (Kirschenbaum, Mariani, Van Gulijk, Rapaport, & Lubasz, 2012)

Manage: e.g. performing resource allocation, tasking and scheduling (Olive, Laube, & Hofer, 2009)²

261

262 **6. Results**

263 After the first rounds of sifting, 209 documents were included for full-text analysis (figure 1).

264 Out of these, 37 papers were not obtainable and a further 51 papers were excluded because their

265 full-text did not meet the predefined criteria. This left 121 studies to be included in the final

266 synthesis based on full-text screening.

267

268

269

270

271

272

273

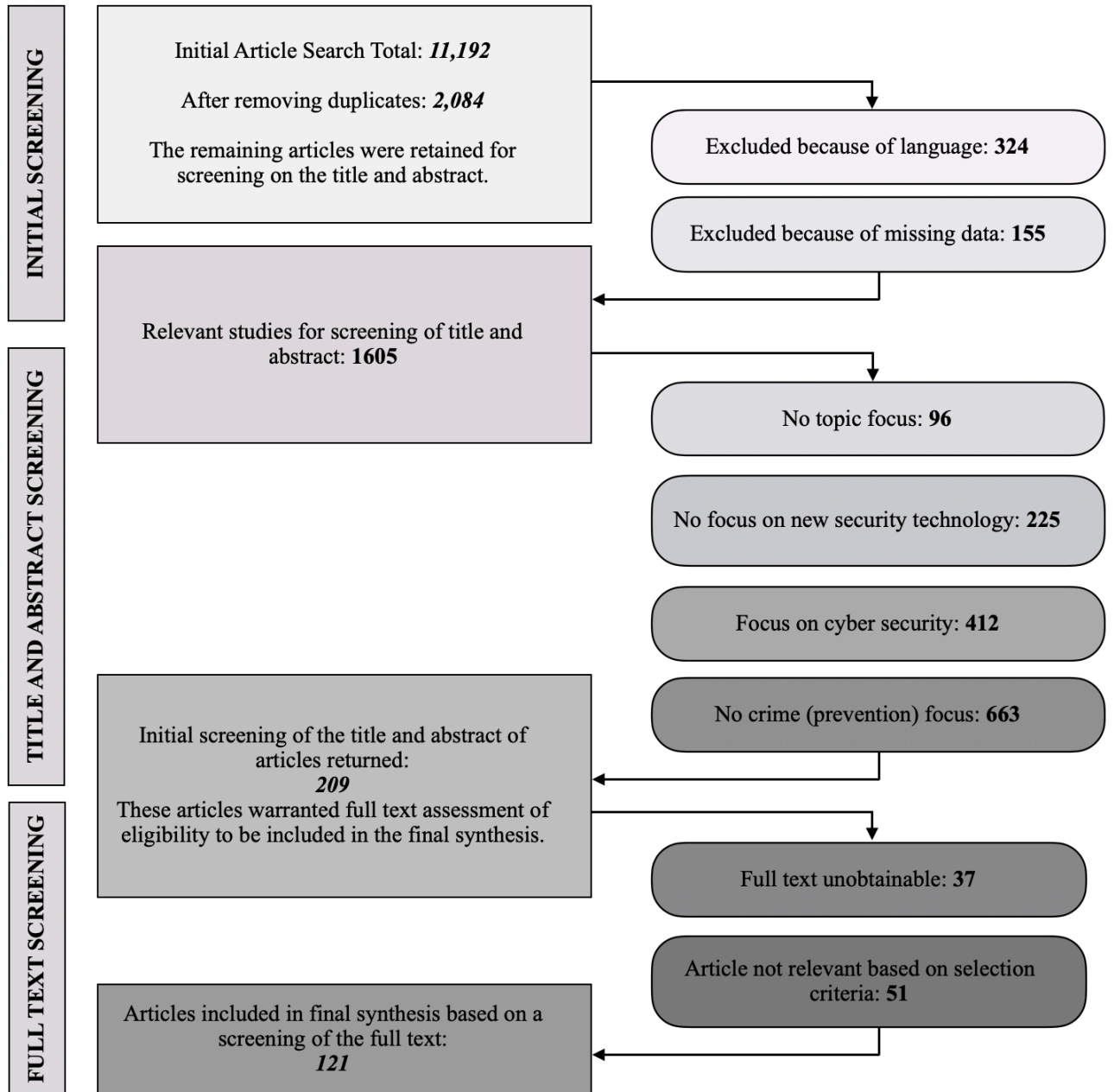
¹ Note that the function to detect on the actuator layer is distinct from that on the sensor layer. Actuators with the function to detect can – similarly to the detain function – be seen as an enforcement action with the goal of removing the offender presence, whereas sensors merely seek to detect anomalies or illicit action.

² While this function may in some cases be considered to refer to the network layer of an intervention, we categorise it as an actuator. This is because managing the interplay of different interventions has a much more direct impact on security and crime prevention in a smart city context.

274

275

Figure 1: Search stages and results of the systematic review



276

277

278

279

280

281

282

Table 3: Results of the systematic literature search by category

New sensors, traditional actuators	43
Detect and prevent unwanted or criminal behaviour	34
Identify, authenticate, defeat (potential) offenders	9
Making old systems smart	57
Improve/automate processes in order to adjust them to a smart city environment	32
Manage/Integrate the interplay of different existing security solutions	25
Entirely new functions:	21
(Mass) information and crowd-sourcing about criminal activity or public disorder	13
Predict potential threats	8

284

285 **7. Analysis**

286 Before detailing the content of the 121 included studies, it is worth making three general
287 observations. Firstly, the search showed that the literature on new crime prevention technologies
288 and smart cities is characterised by a disparity between highly technical studies on one hand and
289 conceptual studies on the other. While the former group of articles often neglects the bigger picture,
290 the latter focusses on conceptual aspects of large smart city systems, usually with no real
291 technological foundations. Only a few studies attempt to bridge this gap. Moreover, many works
292 that seek to predict the future use of a specific technology become quickly outdated due to the fast-
293 pace developments in the field.

294 Secondly, as smart cities are highly complex environments, there are many instances where
295 single interventions fulfil multiple functions. This can either be multiple sensing or actuator
296 functions or include a mix of both. The latter is especially the case for personal security systems
297 such as the portable safety device proposed by Mahajan, Reddy, & Rajput (2018). The device
298 comes in the form of a bracelet or small wearable item that automatically detects a threat to its
299 wearer or can be manually triggered to a range of defensive mechanisms. While some of these
300 functions were explicitly mentioned such as the raising of an alarm (inform), others were left
301 implicit, such as deterrence effects or the triggering of other actuators that the technological

302 solution may or may not have. Furthermore, the review identified a wide variety of technologies
303 that do not explicitly carry out security or crime prevention functions by themselves, but which
304 build upon and seek to improve existing technologies such as CCTV. With, 57 included studies,
305 this field makes up almost half of the identified interventions.

306 Thirdly, as already outlined in the background section of this paper, there is no clear definition
307 of smart cities or even of smart technologies. The definitional vagueness surrounding some of the
308 core concepts of this nexus is clearly reflected in the literature, often leading to less meaningful
309 conclusions and the lack of a common basis for discussion.

310 Despite these shortcomings of the overall field, three clear themes emerged from the
311 technological interventions examined in this review. The first theme concerns new security
312 technologies that fulfil clear traditional security functions such as to detect and prevent, or to
313 identify, authenticate, and defeat (7.1). The second theme includes studies that are focussed on the
314 process of improving and automating ‘traditional’ security functions (as outlined above), and those
315 that contribute to the management and integration of services to create the bigger picture of a smart
316 city (7.2.). The last theme this review found is concerned with those interventions that fulfil new
317 functions that as such did not really exist before, including disseminating mass information and
318 predicting trends or events (7.3.). Though many of these things may have been technically possible
319 before, they lacked technological solutions that made a wide-scale implementation possible and
320 feasible. In the following, these three themes will be described in more detail with regards to their
321 aim, shortcomings, and implications for urban security, planning, and governance as a whole.

322 *7.1 New sensors, traditional actuators*

323 *7.1.1 Detect and prevent*

324 The search identified 34 interventions (Table 4) that aim to detect anomalies, threats, or
325 unwanted behaviour. While some studies analysed human behaviour, facial expressions, or lip-
326 movement to identify threats in individual people (Anagnostopoulos, 2014; Byun, Nasridinov, &
327 Park, 2014; Rothkrantz, 2017b; Sajjad et al., 2018), others sought to detect fraudulent behaviour

328 through the analysis of big data and crowd movement patterns (Cemgil, Kurutmaz, Cezayirli,
329 Bingol, & Sener, 2017; Gupta, Chakraborty, & Mondal, 2017; Liu, Ni, & Krishnan, 2014; Rocher,
330 Taha, Parra, & Lloret, 2018; Sadgali, Sael, & Benabbou, 2018). Even though many of these
331 interventions operated to a large extent the sensor layer of the smart city and relied on already
332 existent actuators, they often did include secondary functions. This included automatically
333 informing the police if fraudulent or dangerous behaviour was detected (Venkatesan, Jawahar,
334 Varsha, & Roshne, 2017), and actuators aimed at de-escalating situations through environmental
335 modification such as changes in light, sound, or smell (Al-Anbuky, 2014; Schuilenburg & Peeters,
336 2018). Secondary functions were also included in the four interventions with the aim to track the
337 movement of persons, vehicles, or UAVs, which in case a threat was detected, could independently
338 contain it (Anees & Kumar, 2017; Brust et al., 2017; Reddy, Loke, Jani, & Dabre, 2018b;
339 Saravanakumar, Deepa, & Kumar, 2017) (Table 5).

340 **Table 4:** Interventions with the primary function to detect.

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Cho (2012)	Visual emergency detection	Detection and recognition method for emergency and non-emergency speech.	detect	
Liu et al. (2014)	(Charging) fraud in Taxis	Uses GPS speed and location data to compute the actual service distance on the city map, and detect fraudulent behaviours	detect	
Byun et al. (2014)	Offender identification and prediction	Detect crimes in real-time by analysing the human emotions	detect	
Bourmpos et al. (2014)	Prevention of crises (critical infrastructure)	Fibre sensing network to monitor diverse parameters of infrastructures, environmental conditions, and vehicle traffic	detect	
Giyenko and Im Cho (2016)	Faults of static CCTV	Intelligent IoT platform to facilitate the use of UAVs	detect	
Baba, Pescaru, Gui, and Jian (2016)	Stray dog attacks	Dangerous behaviour detection of group of stray dogs	detect	
Gupta et al. (2017)	Energy theft	Clustering based energy theft detection technique	detect	
Cemgil et al. (2017)	Fraud in meter readings	Fraud detection mechanism on the electricity consumption data	detect	

Welsh and Roy (2017)	Gunshot detection	Utilising 10 different sensors to detect gunshots	detect	
Bellini, Cenni, Nesi, and Paoli (2017)	Monitoring the flow of people	System to monitor the use of WiFi access points to determine how and where traffic is flowing	detect	
Baldoni et al. (2017)	Faults of static CCTV	Capillary video surveillance platform using plug-and-play that is flexible and scalable with the number of transmitting and receiving devices	detect	
Cubik et al. (2017)	Perimeter protection	Use of fibreoptic sensors in perimeter protection.	detect	
Ertugrul, Kocaman, and Sahingoz (2018)	Mapping and surveillance of buildings	Autonomous UAVs for indoor mapping of buildings and physical security control	detect	
Borges et al. (2017)	Detection of crime hot spots	Analyses characteristics of the urban environment to detect categories and hotspots of criminal activities	detect	
(Sajjad et al., 2018)	Facial expression recognition	Suspicious activity recognition based on facial expression analysis	detect	
Sadgali et al. (2018)	Credit card fraud	Detection of fraudulent transactions from big data using machine learning	detect	
Chackravarthy, Schmitt, and Yang (2018)	Backlog of video data created by traditional CCTV	Neural networks in combination with a Hybrid Deep Learning algorithm to analyse video stream data	detect	manage
Durga, Surya, and Daniel (2018)	Faults of static CCTV	Android application that obtains video feed, images and sound clips from the users and then uses cloud services for video enhancement and restoration of the content	detect	
Calavia, Baladrón, Aguiar, Carro, and Sánchez-Esguevillas (2012)	Faults of static CCTV	Intelligent video surveillance system able to detect and identify abnormal and alarming situations by analysing object movement	detect	authenticate
Datta and Sarkar (2017)	Faults of static CCTV	A flexible surveillance system using smart phones and existing sensors as well as home automation	detect	authenticate
Hu and Ni (2018)	Vehicle/object detection + license plate recognition	Automated object detection for urban surveillance systems	detect	authenticate
Manasa (2016)	Concealed explosives	Nanoscale technologies to find hidden explosives.	detect	identify
Agha, Ranjan, and Gan (2017)	Illegal racing/tail pipe modification	Automatic noisy vehicle surveillance camera	detect	identify
Rocher et al. (2018)	Fraudulent use of dyed fuels	IoT system to detect the presence of low-taxed fuels in the deposit of cars	detect	identify
Rothkrantz (2017a)	Sound recognition in CCTV	Lip-movements of a talking mouth can be recorded and understood, and aggressive behaviour detected	detect	improve
Venkatesan et al. (2017)	Theft	IoT based security system for homes, offices, banks. Sensors for theft and fire detection. Can automatically notify the user and automatically captures images of the intruder.	detect	inform

Ahir, Kapadia, Chauhan, and Sanghavi (2018)	Harassment, molestation	Smart device for women, including GPS/vital tracking, alarm, force sensor, and shock function	detect	inform
Eigenraam and Rothkrantz (2016)	Traffic rule violations/suspicious behaviour	Multi-camera surveillance systems designed as a Decision Support System (DSS)	detect	locate
de Diego, San Román, Montero, Conde, and Cabello (2018)	Faults of static CCTV	Distributed intelligent video surveillance architecture based on Wireless Multimedia Sensor Networks	detect	track
García, Meana-Llorián, G-Bustelo, Lovelle, and Garcia-Fernandez (2017)	Faults of static CCTV	Analysis of pictures through Computer Vision to detect people in the analysed pictures	detect	manage
Anagnostopoulos (2014)	Suicide in metro stations	Information system architecture which can predict whether an individual intends to commit a suicide	detect	predict
Al-Anbuky (2014)	Street crime	Sensor-actuator smart public lighting network	detect	prevent
Schuilenburg and Peeters (2018)	Crime in night-time economy	Sound, smell and lighting programming combined with data analysis is used to reduce violence and aggression	detect	prevent
Huang and Chu (2017)	Trapped people	Detect trapped-victims underneath fallen objects	detect	track

341

342

Table 5: Interventions with the primary function to track as well as an integrated containment function.

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Saravanakumar et al. (2017)	Vehicle theft, speeding	Track vehicles that commit crime and enable decision making at neighbouring traffic sites.	track	contain
Brust et al. (2017)	Malicious UAVs	UAV defence system for the purpose of intercepting and escorting a malicious UAV outside the flight zone.	track	contain
Reddy et al. (2018b)	Re-identification in CCTV	Facial recognition system to track or search a target person from a real time video feed	track	improve
Anees and Kumar (2017)	Crowd-density scanning	Key-point descriptors extracted from the scene are used to compute the dense areas which is further used to define the direction of the flow	track	manage

343

344

345

346

347

348

349

Technologies for the detection of threats through the collection and use of large amounts of data and technological measures to prevent crime long existed and are in widespread use today (e.g. CCTV). The automatic and local containment of unwanted behaviour or dangerous situations without the involvement of the broader security infrastructure (e.g. police services) or in some cases any human input is, however, new. Interventions that fall into this category often bring sensor and actuator layer closer together by creating a single intervention or by changing or adding new

350 actuators to the equation. This does not only have an impact on crime prevention but also on urban
351 planning and governance processes as a whole. Self-contained interventions pose fundamentally
352 different requirements to urban planning and governance than those that require external actuators
353 such as police interventions. An example of this are audio sensors that, if commotion is recognised,
354 turn up the streetlights rather than triggering more traditional actuators like a police response (Al-
355 Anbuky, 2014; de Kort et al., 2014). Because these interventions rely on the interplay of different
356 smart city components to alert authorities, self-contained security interventions rely on the broad
357 implementation of smart infrastructure across other realms such as lighting and the far-reaching
358 deployment of more elaborate sensors and actuators (de Kort et al., 2014). This is also emblematic
359 of the difficulties inherent in the retrofitting of existing cities with smart technologies brings about.
360 Because smart interventions rely so heavily on each other and because a broad implementation
361 across various realms opens up a variety of possibilities, it is inefficient to ‘divide and conquer’,
362 i.e. to modernise sector after sector (Rathore, Ahmad, Paul, & Rho, 2016; Zygiaris, 2013). Since
363 the usefulness of self-contained interventions is highly dependent on a holistic approach, it poses
364 significant challenges to current processes of urban governance and especially modernisation
365 efforts. Thus, interventions that are made up of not only sensor technology but also of actuators
366 that automatically contain a threat can potentially have a great effect on urban security as a whole.

367 In addition to these more practical requirements, crowdsensing and big data analytics promise
368 some degree of privacy for individuals, whereas facial or motion recognition technologies rely on
369 singling out persons from the larger group (Balla & Jadhao, 2018; Braun, Fung, Iqbal, & Shah,
370 2018). As such, the studies examined show that interventions that rely on motion or facial
371 expression recognition are especially controversial in terms of privacy, bringing many new ethical
372 considerations and requirements into the planning process for urban security (Marx, 1998; Parra &
373 Lopez, 2017). These considerations are not only important to ensure an inclusive and rigorous data-
374 protection regime in smart surveillance environments, but they also have operational significance
375 for the planning, implementation, and often functioning of these security measures (Patton, 2000).

376

377 *7.1.2 Identify, authenticate, defeat*

378 These initial findings tie in with the five included studies (Table 6) that aimed to authenticate
 379 individuals or vehicles attempting to access a restricted area (be it a private property or a congestion
 380 zone in a city). Operationally, this was done either through Near Field Communication (NFC)
 381 (Castella-Roca, Mut-Puigserver, Payeras-Capella, Viejo, & Angles-Tafalla, 2017) or through
 382 camera surveillance systems relying on automated license plate recognition (Balla & Jadhao, 2018;
 383 Boukerche, Siddiqui, & Mammeri, 2017; Hadjkacem, Ayedi, Abid, & Snoussi, 2017; Rothkrantz,
 384 2017a). While the latter to some extent often constituted an improvement or automation of an
 385 existing system, the interventions were considered distinct because they are independent systems
 386 for access control that could also be implemented without any prior interventions in place. As such,
 387 the systems posed a significantly lesser challenge to urban security planning than those mentioned
 388 in the previous section.

389 **Table 6:** *Interventions with the primary function to authenticate.*

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Castella-Roca et al. (2017)	Vehicle access to restricted zones	Driver’s smartphone is used to validate the access to a restricted zone	authenticate	
Balla and Jadhao (2018)	Unauthorised access	Intelligent security system using facial recognition	authenticate	
Sajjad et al. (2017)	Identification of suspects	Cloud assisted facial recognition framework	authenticate	identify
Rothkrantz (2017b)	CCTV does not operate in real time	Use of surveillance cameras to localise and recognise faces from suspect individuals	authenticate	improve
Boukerche et al. (2017)	Vehicle re-identification	Automated vehicle detection and classification system.	authenticate	profile

390

391 The effectiveness of these measures relies to a large extent on the use of physical barriers to
 392 ‘defeat’ intruders or the threat of repercussions if they are caught violating access rules (e.g. fines).
 393 Access control measures have, however, especially in a smart city far more use than the explicitly
 394 mentioned actuators might suggest. Holistic smart city architectures could for example not only

395 prevent vehicles from entering a controlled zone but could also track movement patterns and
396 impose automatic fines (Barba, Mateos, Soto, Mezher, & Igartua, 2012). This would alleviate the
397 need for controlling access to congestion or environmental protection zones in city centres by the
398 police and thus save resources in the long run. The implementation of such smart access control
399 measures could additionally help the expansion of ‘greener’ transportation and as such would
400 positively impact other realms of smart city development in the future (Barba et al., 2012).

401 In addition to these static access control measures, Sajjad et al. (2017) introduce a cloud-
402 assisted face recognition framework. They propose the use of nano-devices for a concealed and
403 secure face recognition system. Wearing a small-sized portable wireless camera and a small
404 processing unit for face detection and recognition on officer’s uniforms would allow for the
405 identification of anyone police interact with, without the need for manual identification. While this
406 is only an example, it is symbolic for a move to supplement current static CCTV systems through
407 mobile components. Whether this includes body worn cameras, cars, or drones, it has the potential
408 to severely change the way we think about urban surveillance. This has some clear benefits such
409 as the ability for cameras to follow crime and to surpass issues of re-identification between cameras
410 if suspects are on the move (Zhang & Yu, 2018).

411 Nevertheless, these benefits come at a cost. While most of the systems proposed in the literature
412 are often minimally intrusive and offer maximum amounts of privacy (Castella-Roca et al., 2017),
413 the use of wearable facial recognition devices, as proposed by Sajjad et al. (2017) should be seen
414 as problematic. Though the system may offer some use to the police, the potential downsides of its
415 implementation are grave. It would for example mean that police officers could not be approached
416 without citizens being subject to facial recognition, which in turn may dissuade many from
417 approaching the police. This has important implications for citizens in their relations and contacts
418 with police actors. This intervention in particular shows that privacy and data protection concerns
419 are not only important on a legal level but also raise the question to what extent an intervention like

420 this can have negative consequences for existing measures and in how far it can be reconciled with
421 the citizen focus of the smart city concept (Braun et al., 2018).

422 *7.1.3 Section summary*

423 Overall, this study has identified a substantial body of literature concerned with using new
424 sensors to detect criminal behaviour and identify individual perpetrators, often relying on already
425 existing actuators for deterrence and crime prevention. Many of the identified interventions could
426 transform urban security and the vision of a safe city. They reinforce the idea that in a smart city,
427 many new security interventions rely on the broad implementation of smart technologies across
428 different realms of the urban environment. Because security interventions no longer only rely on
429 input from the police or their own sensors but can draw from a broad array of data sources, they
430 become significantly more all-encompassing and holistic. Security measures no longer rely solely
431 on the police or a far-reaching security apparatus in a city but their effectiveness also relies on
432 smart technologies in other realms such as street lighting or traffic management (Vitalij, Robnik,
433 & Alexey, 2012). A lack of smartification in one realm can thus have impacts on the effectiveness
434 of interventions in all other realms, first and foremost security interventions. This has great
435 implications for the planning process of smart cities and their security infrastructure itself and
436 shows that future security infrastructures are not separate systems but both reliant on and a
437 prerequisite for the implementation of smart systems across other realms of city services.

438 This, however, does not mean that new interventions are uncontroversial. Privacy and data
439 protection issues are at the forefront of concerns that may arise with their implementation and that
440 need to be addressed in the planning and implementation of safe city concepts (Braun et al., 2018).
441 As such, the interventions clustered in this theme offer great potential, but also require a thorough
442 and far-reaching rethinking of the planning process itself because systems become significantly
443 more interconnected and the effectiveness of single components dependent on the broader
444 infrastructure (Mishra & Kumar, 2013).

445

446 7.2 Making old systems smart

447 7.2.1 Improve/automate

448 While many of the previously introduced measures sought to introduce entirely new systems,
 449 this is often neither necessary nor feasible. Instead, old systems that function well and are already
 450 in place can be improved and processes automated in order to adjust them to a smart city
 451 environment. This review identified 32 studies that address this issue (Table 7).

452 **Table 7: Interventions with the primary function to improve or automate.**

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Sudha (2015)	Faults of static CCTV	Parallel architecture for smart video surveillance	improve	
Sormani et al. (2016)	No datasets for training algorithms	Generation of datasets for training reasoning algorithms for predicting the likelihood of terrorist actions against specific assets and locations in urban environment	improve	
Xiong et al. (2017)	Re-identification in CCTV	Multiple deep metric learning method empowered by the functionality of person similarity probability measurement	improve	
Shi, Ming, Fan, and Tian (2017)	Facial recognition	Recognition algorithm based on multi-scale completed local binary pattern	improve	
Zheng, Sheng, Zhang, Zhang, and Xiong (2015)	Re-identification in CCTV	Weight-based sparse coding approach for person re-identification	improve	
Salmerón-García et al. (2017)	Faults of static CCTV	Cloud-based surveillance system.	improve	
Thomas, Gupta, and Subramanian (2017b)	Faults of static CCTV	Video summarisation to reduce amounts of data recorded	improve	
Singh, Patil, and Omkar (2018)	Computational cost in multiple object tracking.	Parallel solution which effectively handles the challenges of time-dependencies among the various sections of the video file processed during multiple object tracking	improve	
Saba (2017)	Latency issues in CCTV	Device to capture and compress images and mounted with PIR sensor to detect movement	improve	
Tian, Wang, Zhou, and Peng (2018)	Faults of static CCTV	Block-level background modelling (BBM) algorithm to support long-term reference structure for efficient surveillance video coding	improve	
Hadjkacem et al. (2017)	Re-identification in CCTV	New approach based-on the analysis of all the video data extracted from camera-networks	improve	manage
Zhang, Chowdhery, Bahl, Jamieson, and Banerjee (2015)	Re-identification in CCTV	Real-time distributed wire-less surveillance system that leverages edge computing to support real-time tracking and surveillance	improve	manage

Tan and Chen (2014)	Faults of static CCTV	Approach for fast and parallel video processing	improve	manage
Zhou, Saha, and Rangarajan (2015)	Faults of static CCTV	Using existing public bus transit system to collect data from the cameras and physically transport it to the bus terminus, to be uploaded to the data centre	improve	manage
Oza and Gohil (2016)	Faults of static CCTV	Cloud based surveillance system for live video streaming	improve	manage
Xu, Mei, Liu, Hu, and Chen (2016)	Faults of static CCTV	Semantic based cloud environment to analyse and search surveillance video data	improve	manage
Valentín et al. (2017)	Faults of static CCTV	Architecture for automated video surveillance based on cloud computing	improve	manage
Wang, Pan, and Esposito (2017)	Faults of static CCTV	IoT based elastic surveillance system using edge computing to perform data processing	improve	manage
Zhang et al. (2017a)	Human action recognition	Background modelling method from surveillance video	improve	manage
Mehboob et al. (2017)	Faults of static CCTV	3D conversion from traffic video content to Google Maps using time-stamped glyph-based visualisation	improve	manage
Zingoni, Diani, and Corsini (2017)	Moving object recognition	Algorithm capable of successfully recognising and classifying moving objects	improve	manage
García, Valentín, Serrano, Palacios-Alonso, and Sucar (2017)	Faults of static CCTV	Visualisation techniques for both local and global visualisation	improve	manage
Ramírez, Barragán, García-Torales, and Larios (2016)	Transmission latency of data (CCTV)	Wireless sensor network (WSN) using low-power devices for the transceiver process to improve the data management using both, storage and transmission data	improve	manage
Ma et al. (2018)	CCTV placement	Use of traffic patterns to improve the placement of CCTV cameras	improve	manage
Pereira et al. (2018)	Faults of static CCTV	Low-cost smart surveillance platform designed to create a ubiquitous environment and to adapt to the client's needs	improve	manage
Memos, Psannis, Ishibashi, Kim, and Gupta (2018)	Faults of static CCTV	Algorithm to use less memory at the wireless sensor nodes	improve	manage
Peixoto et al. (2018)	Faults of static CCTV	Gap filling algorithms to data missing problem in a smart surveillance environment	improve	manage
Kumar, Datta, Singh, and Sangaiah (2018)	Faults of static CCTV	Intelligent decision computing based paradigm for crowd monitoring	improve	manage
Miraftabzadch, Rad, Choo, and Jamshidi (2018)	Re-identification in CCTV	Algorithm to extract and administrate the crowd-sourced facial image features (e.g., social media platforms and multiple cameras in a dense crowd, such as a stadium or airport)	improve	manage
Zhang and Yu (2018)	Re-identification in CCTV	Deformable convolution module to the traditional baseline to enhance the transformation modelling capability without additional supervision	improve	manage
Al-Shami, Zekri, El-Zaart, and Zantout (2017)	Traffic rule violations	Parallelization processes that enables the online processing of images by an embedded system	improve	manage
Jun, Chang, Jeong, and Lee (2017)	Faults of static CCTV	Collaboration-based Local Search Algorithm (COLSA)	improve	manage

454 The key premise of these studies is that current surveillance systems need improvements to be
455 useful in the future. The scalability and cost-effectiveness of current systems depends largely on
456 these improvements as increased amounts of data and the need for faster processing, drive demand
457 for innovation (Valentín et al., 2017). The most prominent example of this are many video
458 surveillance platforms in use today, which are presented with severe problems of efficiency and
459 scalability when the numbers of data flow senders and receivers increase (Baldoni et al., 2017).

460 In addition, the scalability of modern surveillance systems is often limited by the human factor,
461 driving the demand for automation (human operators can watch ten cameras, but will not be able
462 to monitor 10 000 deployed sensors). Many studies that sought to automate processes that currently
463 require manual input, focus on human re-identification in multi-camera surveillance networks
464 (Hadjkacem et al., 2017; Zhang et al., 2017b; Zheng et al., 2015) or even introduce a wholistic
465 automated system architecture that do entirely without human operators (Valentín et al., 2017). The
466 latter, in particular, is needed to realise the complex system that is a smart city because it does not
467 tackle the issue on merely one layer but improves sensors, processing, and actuators alike. These
468 developments are also problematic when examining current planning processes for security
469 infrastructure. In many instances, there is a disparity between private developments and security
470 agencies. And even where security and crime prevention are considered as factors, developments
471 are often planned with already or soon-to-be outdated systems (Morton, Horne, Dalton, &
472 Thompson, 2012; Sandborn, 2007).

473 Due to steadily improving camera and sensor technology and their large-scale deployment,
474 data streams are exploding in urban surveillance. This impacts the scalability of current systems
475 massively as they ‘outgrow’ the current infrastructure (Brayne, 2017). These issues of scalability
476 of older systems are tackled by interventions on the processing layer of the smart city, aimed at
477 making the transmission, storage, and processing of data cheaper, easier, and faster (Memos et al.,
478 2018; Saba, 2017; Singh, Majumdar, & Rajan, 2017; Thomas et al., 2017b; Zhou et al., 2015).
479 While in this case, the processing layer plays a significant role as the key variable limiting the

480 growth and the flexibility of the systems, it is also sensors and actuators where innovation has a
481 relevant impact on crime prevention in the future.

482 Future systems aim to analyse data in real-time using artificial intelligence (AI) to allow for a
483 quicker response in case of danger (Reddy et al., 2018b; Zhang et al., 2015). Because in many cases
484 not enough historical data exists to train AI, or because the data has gaps that could affect the
485 machine learning, some studies introduce approaches to generate dummy data that can be used for
486 training (Peixoto et al., 2018; Sormani et al., 2016). Such approaches are especially noteworthy
487 because they do not only address shortcomings of current crime prevention technologies but rather
488 provide practical solutions to aid the implementation of other interventions.

489 Similarly, studies such as those of Ma et al. (2018) and Jun et al. (2017) highlight the need for
490 improving not only existing software and hardware but also the methods and procedures by which
491 the deployment of technologies is determined.. Ma et al. (2018) discuss new metrics for the sensible
492 deployment of surveillance cameras but the essence of their research is transferable to many other
493 contexts; if the urban landscape changes significantly, parameters for the allocation of security
494 technologies will also change. Unless this is considered along the way, the planning of urban
495 security runs danger of missing crucial developments and ultimately failing in the future.

496 In terms of urban security as a whole and implications for its planning, interventions that seek
497 to improve and automate current security measures fulfil one of the most important functions. This
498 is because in practice, only in few cases smart cities are built from the ground up. Thus, when
499 speaking about building smart cities, we often mean the retrofitting and improvement of existing
500 systems with smart technologies (Habibzadeh, Soyata, Kantarci, Boukerche, & Kaptan, 2018). As
501 such it is crucial that we approach the smartification of cities holistically but at the same time
502 maintain an eye for existing infrastructures as the basis for these developments.

503 7.2.2. Manage/integrate

504 A truly safe (smart) city is defined by increased integration of different systems and the
505 boundary-less coordination of measures across all fields. This review identified 25 interventions

506 that sought to integrate or to manage the interplay of different existing security solutions in urban
 507 environments (Table 8). The scope and focus of these interventions differed greatly, reaching from
 508 single-layer solutions tackling the complex interplay of different sensors (Camboim, Neto,
 509 Rodrigues, & Zhao, 2017; Chen, Xu, & Guo, 2013) to holistic integrated framework architectures
 510 that work to connect sensors and actuators across the city (Bartoli, Fantacci, Gei, Marabissi, &
 511 Micciullo, 2015; Dbouk, Mcheick, & Sbeity, 2014; Fernández et al., 2013; Khan, Azmi, Ansari, &
 512 Dhalvelkar, 2018; Liu et al., 2017b; Vitalij et al., 2012). The aim of the interventions is in many
 513 cases the more efficient use of resources (Al-Muaythir & Hossain, 2016; Hochstetler, Hochstetler,
 514 & Fu, 2016) but also the improvement of services through management and integration of different
 515 measures (Kunst, Avila, Pignaton, Bampi, & Rochol, 2018).

516 *Table 8: Interventions with the primary function to manage or integrate.*

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Khan et al. (2018)	Large quantities of call data records	Use of the call data records (CDRs) of various suspects and victims in order to extract significant evidence	manage	investigate
Dbouk et al. (2014)	Terrorist attacks	Surveillance system architecture	manage	
Bartoli et al. (2015)	Growing populations and demand to respond	Integrated platform for new generation professional mobile radio system, wireless sensor networks, social networks, and a data gathering and analysis system able to collect and elaborate heterogeneous information coming from different sources	manage	
Al-Muaythir and Hossain (2016)	Limited resources/inflexible systems	Parametric subscriptions/cloud-based publish/subscribe framework	manage	
Hochstetler et al. (2016)	Limited resources	Network of clusters to efficiently assign patrols based on informational entropy in order to minimise police time-to-arrival and the overall numbers of police on patrol	manage	
Lohokare, Dani, Sontakke, Apte, and Sahni (2017)	Response time	Capturing live location of the emergency services to connect them directly to nearest citizen in need	manage	
Duan, Lou, Wang, Gao, and Rui (2018)	Faults of static CCTV	AI oriented large-scale video management. Person/vehicle re-identification, facial recognition before coding	manage	
Hartama et al. (2017)	Emergency traffic management	Strategy related to efforts to improve the distribution of space and time based on traffic volume	manage	

Hosseini, Salehi, and Gottumukkala (2017)	Oversubscription of servers with relevant video feeds	Stream-priority aware resource allocation mechanism to enable interactive video prioritisation without a major impact on the flow of non-prioritised video streams	manage	
Patel, Wala, Shahu, and Lopes (2018)	Inefficient police records	Proposed online system for police stations which will help the policeman and user to digitalise their work.	manage	
Dey, Chakraborty, Naskar, and Misra (2012)	Faults of static CCTV	Multimedia surveillance backend system architecture based on the Sensor Web Enablement framework and cloud-based “key-value” stores	manage	improve
Chen et al. (2013)	Faults of static CCTV	New architectures integrated with Hadoop to resolve the urgent pressure of overloaded and to put the whole system into the computer cluster	manage	improve
Khorov, Gushchin, and Safonov (2015)	Faults of static CCTV	Easy implementation strategy to drop the smallest (in bytes) video frame whenever queue overflows	manage	improve
Lei et al. (2016)	Large quantities of data	K-means algorithm that can automatically split and merge clusters which incorporates the new ideas in dealing with huge scale of video data	manage	improve
Chen et al. (2016)	Faults of static CCTV	Dynamic video stream processing scheme to meet the requirements of real-time information processing and decision making	manage	improve
Pribadi, Kumiawan, Hariadi, and Nugroho (2017)	CCTV placement	Algorithm for improved camera placement.	manage	improve
Rametta, Baldoni, Lombardo, Micalizzi, and Vassallo (2017)	Faults of static CCTV	Smart CCTV platform to exploit the facilities offered by full SDN-NFV networks.	manage	improve
Liu and Lin (2017)	Automated license plate recognition	Hierarchical architecture combining supervised K-means and support vector machine.	manage	improve
Wu et al. (2017)	Data management in geographic information systems (GIS)	Hybrid database organization and management approach with SQL relational databases (RDB) and not only SQL (NoSQL) databases	manage	improve
Kunst et al. (2018)	High amounts of data traffic lessen the quality of service.	Multi-purpose real time video surveillance application using resource sharing	manage	improve
Duan et al. (2018)	Faults of static CCTV	Computational methodology for reorienting, repositioning, and merging camera positions within a region under surveillance.	manage	improve
Camboim et al. (2017)	Vehicle theft, violent crime on public transport	Smart surveillance system to recognise security threats in real time	manage	improve
Vitalij et al. (2012)	Lack of integration between different parts of smart cities.	Integrated framework with intelligent video surveillance, emergency communication, public address, general alarm/local notification systems, environmental monitoring and forecasting, local fire/chemical control systems, spotting, position location / eCall, ERA-GLONASS services, communications and mass media.	manage	improve

Liu et al. (2017b)	Integration of different data sources	Community safety oriented public information platform	manage	integrate
Fernández et al. (2013)	Vandalism prevention, perimeter security	Intelligent surveillance platform based on the use of large numbers of cheap sensors.	manage	detect

517

518

519 While it may at first seem as if the interventions collected in this category are not as relevant
520 to security because they do not directly introduce new sensors or actuators (i.e. do not execute
521 crime prevention tasks as such), they, in fact, take a central role in the security aspect of safe cities.
522 This is especially relevant for safety and crime prevention planning and urban governance because
523 larger quantities of information are transported and processed faster than before. This means not
524 only that policies and decisions can rely on a more larger evidence base but also that decision
525 making processes may need to change.

526 The integration of different security measures and their improved management through the
527 implementation of connected systems is a prerequisite for the smart city (Ralko & Kumar, 2016).
528 And because urban trends are heading in this direction, it is imperative that planners embrace the
529 opportunities that come with it in all administrative procedures and planning processes to maintain
530 the ability to solve urban problems in the future.

531 *7.2.3 Section summary*

532 Overall, many of the interventions clustered in this theme aim to enable smart city
533 developments through the increased improvement and integration of city service infrastructure and
534 its technological components. Despite this clear aim, the approaches taken in the literature differ
535 substantially. While some studies approach smart city efforts on a micro-level (i.e. single layer),
536 others propose holistic systems for the management of different services from sensors and
537 processing units to actuators. This variety of approaches highlights the fact that smart city security
538 infrastructure depends on integration on all levels, between and within the different parts of the
539 surveillance and security apparatus (Hall et al., 2000).

540 This category of interventions is also crucial because it is most likely to be realised in practice.
541 Only rarely are smart cities built from the ground up, and a more realistic path is the gradual
542 improvement of existing systems (Mishra & Kumar, 2013). In this context, it is important to
543 remember that smart security measures and the concept of the safe city are not born from the
544 overwhelming failure of existing interventions but rather from the wish to improve existing efforts
545 and to make them more efficient and manageable in the future (Truntsevsky, Lukiny, Sumachev,
546 & Kopytova, 2018). As such, the interventions mentioned in this theme are not only practically
547 appropriate, but they are also closest to the reality of financial and resource constraints in cities
548 today. Given this it is surprising that only few studies (Al-Muaythir & Hossain, 2016; Hochstetler
549 et al., 2016; Jun et al., 2017; Pereira et al., 2018) consider the economic implications or the financial
550 efficiency of their interventions as a relevant factor in their deployment and evaluation. Despite the
551 fact that efficiency and effectiveness are crucial factors in a smart city environment, we found many
552 studies discussing operational efficiency in terms that were far from today's urban realities.

553

554 *7.3 Entirely new functions*

555 *7.3.1 (Mass) information and crowdsourcing*

556 While the original framework suggested an 'inform'-function limited to sounding alarm or
557 alerting security services, this paper suggests that this definition should be revised. In total, this
558 paper reviewed 13 studies that aimed to inform (i.e. communicate information about a specific
559 situation) (Table 9). Only three of the interventions, however, functioned to automatically trigger
560 actuators like alarming security services of a crime (Liu, Warade, Pai, & Gupta, 2017a; Mahajan
561 et al., 2018; Nasui, Cernian, & Sgarciu, 2014). The other interventions were either user focussed
562 on providing information about crime and crime prevention to the population (Ballesteros,
563 Rahman, Carbanar, & Rishe, 2012; Kagawa, Saiki, & Nakamura, 2017; Mata et al., 2016; Peng,
564 Xiao, Yao, Guan, & Yang, 2017; Truntsevsky et al., 2018) or fulfilled a hybrid role. To distinguish

565 these two different groups, we will refer to the latter as 'mass information', while the former will
 566 be labelled as interventions with the aim to 'inform'. All of the studies are listed in table 9.

567 **Table 9:** *Interventions with the primary function to crowdsource inform or provide mass-information.*

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Ballesteros et al. (2012)	Lack of awareness in dangerous situations	Combined use of personal mobile devices and social networks to make users aware of the safety of their surroundings.	(mass) inform	
Mata et al. (2016)	Mobile applications do not show safe routes	Approach to provide estimations defined by crime rates for generating safe routes in mobile devices.	(mass) inform	
Peng et al. (2017)	Safety indices for local areas are often inaccurate	Urban safety analysis system to infer safety index by leveraging multiple cross-domain urban data	(mass) inform	
Kagawa et al. (2017)	No information about security	PRISM (Personalized Real-time Information with Security Map)	(mass) inform	
Truntsevsky et al. (2018)	Street crime	Exploring the possible application of modern digital technologies in the evaluation and prevention of crime.	(mass) inform	
Arauz, Moreno, Nancalres, Pérez, and Larios (2017)	Corruption	Platform to integrate user orientation, application of standards for the development of the city and citizen participation.	inform	
Carreño, Gutierrez, Ochoa, and Fortino (2015)	Personal security	A mobile application which implements participatory sensing to help people be aware of the risks that appear to exist in a certain place at a certain time.	(mass) inform	
Moreira, Cacho, Lopes, and Cavalcante (2017)	Inefficient information for citizens	Mobile application as an alternative communication channel between public safety agencies and population.	inform	
Ferreira, Visintin, Okamoto, and Pu (2017)	Assaults and street crime	Smart surveillance cameras system, a back office system with a workflow engine and a mobile application within a collaborative concept	inform	
Bonatsos, Middleton, Melas, and Sabeur (2013)	Fear of crime/lack of awareness/misconceptions	Decision-support system integrating information, data and software modules representing city assets, hazards and processing models that simulate exposures to risks and potential compromise to safety and security.	inform	
Mahajan et al. (2018)	Assault, violence, attacks against women	Wearable/portable system that creates a sense of safety among women with a range of different features (automatic alarm, shock, audio streaming, location)	inform	
Nasui et al. (2014)	Transportation safety	Cloud based student transportation safety system is a location aware mobile asset management solution for operators of commercial fleets, having a cloud based platform at its core.	inform	detect
Liu et al. (2017a)	Street crime	Fine-grained location-aware smart campus security systems that leverages hybrid localisation approaches with minimum deployment cost.	inform	detect

568

569 Despite their different foci, both types of intervention increasingly involve the use of mobile
570 applications to crowdsource information about criminal activity or public disorder. While some of
571 these applications create a knowledge base that in turn aims to inform users (Carreño et al., 2015;
572 Ferreira et al., 2017; Moreira et al., 2017), other applications, such as the online platform developed
573 by Arauz et al. (2017), seek to tackle specific problems such as corruption by allowing users to
574 report criminal activity directly to the authorities.

575 When assessing the effect these interventions have on the larger picture of urban security, it is
576 important to distinguish between their different functions. While on one hand, mobile applications
577 may be highly useful for mass information, i.e. to reach a large part of the population and to create
578 broad awareness about crime and crime prevention, they also have downsides.

579 The most obvious issue of mobile applications is that their functionality and their ability to
580 crowdsource information relies heavily on an active user base — without a crowd, no
581 crowdsourcing. Even (or especially) if they are actively used, however, user-centric applications
582 are open to misuse (Yang, Zhang, Ren, & Shen, 2015). Malicious actors may report false crimes to
583 purposefully waste police resources or to put someone else in the crosshairs of security services.
584 Another concern is that criminals could use apps just like the genuine user but to determine where
585 victims might move to in order to avoid crime (Monahan & Mokos, 2013).

586 As discussed above, smart security technologies are aimed at making public services more
587 efficient and effective and ultimately freeing up resources. This, however, is a double-edged sword,
588 as 'inform'-functions make especially clear. While crowdsourcing information about crime with the
589 goal of recording more crimes of a certain type can be considered an innovation on the sensor layer
590 and may in some cases be desirable, it may in other cases put an unnecessary strain on already tight
591 resources and overwhelm existing actuators. For example, an increased report rate for domestic
592 abuse may very well save lives, but an app that floods police with hundreds of reports of anti-social
593 behaviour or noise complaints may in the end take up disproportional amounts of resources (Elliott-

594 Davies, Donnelly, Boag-Munroe, & Van Mechelen, 2016). While interventions might be able to
595 create a large network of 'eyes on the street' (Cozens & Davies, 2013; Hillier & Cozens, 2012),
596 they may also create a flood of information that could overwhelm many public institutions.

597 Nevertheless, these interventions do offer some potential benefits. Especially crowd sourcing
598 and mass information platforms can bring citizens and governments closer together (Kim & Lee,
599 2012). Not only can this help to streamline city services, but things like e-participation can also
600 allow citizens to interact more directly with the administration of the place they live. This in turn
601 can help to more directly include public opinion in planning processes and democratise the design
602 and management of urban spaces (Macintosh, 2004).

603 Taking all of the above into account, it is difficult to assess the usefulness and impact of these
604 interventions in terms of urban planning and governance as a whole. While elaborate measures of
605 harm and police demand may give some indication of the usefulness of these interventions in terms
606 of crime prevention, they largely ignore the overall usefulness across other realms (Greenfield &
607 Paoli, 2013; Ratcliffe, 2015).

608 *7.3.2 Predict*

609 Predictive policing is in itself nothing new and has in the past grown to become one of the most
610 well-researched realms in the field of policing. More recently, however, the wide-scale use of
611 predictive policing has also come under intense scrutiny from both academics and practitioners
612 (Brantingham, Valasik, & Mohler, 2018; Degeling & Berendt, 2017). Whether new technologies
613 can revolutionise current approaches enough to make it a viable tool for policing without
614 compromising privacy and data protection too much remains to be seen.

615 Nevertheless, this study has identified eight interventions that sought to provide security
616 services with some form of predictive capabilities (Table 10).

617

618

619

Table 10: Interventions with some form of predictive capabilities.

Author (Year)	(Crime) Problem	Solution	Primary security function	Secondary security function
Noor, Nawawi, and Ghazali (2013)	Prediction of situational crime factors	New tool that uses decision support system (DSS) and fuzzy association rule mining (FARM), in which it can extract the factors of situational (opportunity) crime	predict	
Oatley et al. (2015)	Faults of static CCTV	Utilising CCTV as a sensor to accurately model or give feedback on the reality of occurrences in digital space	predict	
Castelli et al. (2017)	Growing amounts of data	AI system for predicting violent crimes in urban areas starting from socio-economic and law-enforcement data	predict	
Garg, Malik, and Raj (2018)	Street crime	Gain insights into historical crime data to predict crimes	predict	
Catlett, Cesario, Talia, and Vinci (2018)	Forecasting inefficient	Predictive approach based on spatial analysis and auto-regressive models to automatically detect high-risk crime regions in urban areas and forecast crime	predict	
Isafiade and Bagula (2017)	Street crime	Crime series pattern detection	predict	
Araujo, Cacho, Thome, Medeiros, and Borges (2017)	Robbery and homicide	Smart city platform aimed at integrating several information systems from law enforcement agencies	predict	integrate
Kagawa, Saiki, and Nakamura (2018)	No information about crime in nearby area	Analyse street crimes according to users' living area using personalised security information service. Output is a crime map that helps citizens to avoid crime areas.	predict	warn

621

622

623

624

625

626

627

628

629

630

631

632

633

The extent and scope of these capabilities varied, however, greatly between the different interventions and reached from more traditional uses of historical crime data (Catlett et al., 2018; Garg et al., 2018; Noor et al., 2013) to the detection of psychopathy and potentially dangerous behaviour through CCTV and agent-based simulation through friendship networks on social media platforms (Oatley et al., 2015). What is new about many of these interventions is that their predictive capabilities include the real-time analysis of data as well as mechanisms for subsequent resource allocation, i.e. actuators. This separates them from current predictive policing tools which have been criticised for not being more accurate than an experienced police officer.

In addition, the growing importance of the online realm is reflected in a growing number of approaches. The model introduced by Oatley et al. (2015) emphasises that many people no longer express themselves actively in urban spaces but rather online, and that surveillance systems scanning crowds for suspicious behaviour only see half the picture (Oatley et al., 2015). This not

634 only adds social media as a new dimension of urban surveillance, but it also forces a fundamental
635 change in how we think about and plan for urban security.

636 *7.3.3 Section summary*

637 This section has introduced various interventions with functions that are not, or only to some
638 extent, currently in use in policing and crime prevention. As such, they do not correspond to
639 traditional functions of security interventions. While many of these interventions certainly offer
640 great potential for transforming safe city designs and urban security landscapes, it is hard to
641 evaluate the extent to which they will impact urban security as a whole due to the fast-paced nature
642 of technological development. In addition, a lack of implementation cases and evaluative studies
643 makes it impossible to predict what side-effects they may have (Siregar, Syahputra, Putra, &
644 Wicaksono, 2018).

645

646 **8. Conclusion**

647 Our review introduced three categories of security interventions in smart cities. While some of the
648 examined interventions did correspond with the traditional functions of security interventions both
649 as sensors and actuators, we proposed a new classification for smart security interventions based
650 on their functions.

651 Our classification distinguishes between three main categories, each with two sub-categories. The
652 first category focussed on those interventions that combined new sensors with traditional actuators.
653 This included interventions to detect and prevent unwanted criminal behaviour, and those aimed at
654 identifying, authenticating, and defeating offenders. The second category included those
655 interventions that sought to make old systems smart by either improving/automating processes or
656 by managing and integrating the interplay between existing security solutions. The third category
657 entailed those interventions that introduced entirely new functions such as (mass) information and
658 crowd-sourcing as well as threat or crime prediction.

659 While this classification can help to group and compare interventions, they can also be useful to
660 explore the distinct set of opportunities and challenges that they bring about. The proposed
661 classification highlights that not all systems need to be fundamentally new to become smart and
662 that building on existing infrastructure is crucial for a successful smartification. In addition, our
663 analysis emphasises that the implications of the implementation of new security technologies in
664 urban spaces are far-reaching with regards to urban planning and governance. Throughout, we
665 show that future security infrastructures are not separate systems but reliant on and a prerequisite
666 for the implementation of smart systems across other realms of city services. Especially the latter
667 is important to consider for future smart city planning. Instead of treating security and crime
668 prevention as the cherry on top of any smart city development, urban planners should consider it
669 as a foundation. Not only do safety and security significantly impact if and how citizens interact
670 with urban spaces but as shown in the discussion above, there are a variety of tools that can be used
671 for citizen engagement across different realms of city services.

672 Overall, it is important to remember that smart security measures and the concept of the safe city
673 are not born from the overwhelming failure of existing interventions but rather from the wish to
674 improve existing efforts and to make them more efficient and manageable in the future. As such,
675 they should be seen as a part of a larger holistic system that offers opportunities across all realms
676 of city administration.

677 These opportunities do come, however, at a cost. The far-reaching implementation of smart
678 technologies brings about new ethical considerations as well as implications for the planning
679 process itself. Questions of data ownership and privacy rights grow in importance and need to be
680 reflected in contemporary planning processes. In this review, we highlighted the importance of
681 discussing these issues and criticised the lack of attention they have received in the smart city
682 debate.

683 The question remains whether the use of such technologies will undermine individual privacy
684 needs in the long run. Some authors stipulate that “surveillance technologies are a key component

685 of smart and networked cities preventing or detecting crime and giving the residents a sense of
686 safety” (van Heek, Aming, & Ziefle, 2016), while others such as Oatley et al. (2015) go as far as to
687 describe CCTV networks as the fifth utility in smart cities. Yet while many innovations might
688 create more efficient city services or effectively reduce crime, they might at the same time make
689 people feel less secure because they have a sense that ‘Big Brother’ is watching. Particularly in
690 authoritarian (or at least not fully democratic) regimes, the implementation of these new security
691 measures can exponentially increase state power and control over its citizens. There is thus
692 significant tension, as yet unresolved, between issues connected with these new technologies,
693 especially with regards to privacy and data protection, and the importance of urban surveillance
694 and security infrastructure for providing safety and security in the 21st century city.

695

696 **Bibliography**

- 697 Adey, P. (2002). Secured and Sorted Mobilities: Examples from the Airport. *Surveillance &*
698 *Society*, 1(4).
- 699 Agha, A., Ranjan, R., & Gan, W.-S. (2017). Noisy vehicle surveillance camera: A system to deter
700 noisy vehicle in smart city. *Applied Acoustics*, 117, 236-245.
- 701 Ahir, S., Kapadia, S., Chauhan, J., & Sanghavi, N. (2018). *The Personal Stun-A Smart Device For*
702 *Women's Safety*. Paper presented at the 2018 International Conference on Smart City and
703 Emerging Technology (ICSCET).
- 704 Al-Anbuky, A. (2014). *Sensor-actuator smart lighting system: system organizational concept and*
705 *challenges*. Paper presented at the ICT for Sustainability 2014 (ICT4S-14).
- 706 Al-Muaythir, A., & Hossain, M. A. (2016). *Cloud-based parametrized publish/subscribe system*
707 *for public safety applications in smarter cities*. Paper presented at the 2016 IEEE/ACM 9th
708 International Conference on Utility and Cloud Computing (UCC).
- 709 Al-Shami, S., Zekri, A., El-Zaart, A., & Zantout, R. (2017). *On the parallelization of closed-set*
710 *patterns classification for an automatic license plate recognition system*. Paper presented
711 at the 2017 Sensors Networks Smart and Emerging Technologies (SENSET).
- 712 Alawadhi, S., Aldama-Nalda, A., Chourabi, H., Gil-Garcia, J. R., Leung, S., Mellouli, S., . . .
713 Walker, S. (2012). *Building understanding of smart city initiatives*. Paper presented at the
714 International conference on electronic government.
- 715 Albino, V., Berardi, U., & Dangelico, R. M. (2015). Smart cities: Definitions, dimensions,
716 performance, and initiatives. *Journal of Urban Technology*, 22(1), 3-21.
- 717 Anagnostopoulos, T. (2014). *A Surveillance System for Preventing Suicide Attempts in Urban*
718 *Metro Stations*. Paper presented at the Proceedings of the 18th Panhellenic Conference on
719 Informatics.
- 720 Anees, V. M., & Kumar, G. S. (2017). *Direction estimation of crowd flow in surveillance videos*.
721 Paper presented at the 2017 IEEE Region 10 Symposium (TENSymp).
- 722 Ankitha, S., Nayana, K., Shravya, S., & Jain, L. (2017). *Smart city initiative: Traffic and waste*
723 *management*. Paper presented at the Recent Trends in Electronics, Information &
724 Communication Technology (RTEICT), 2017 2nd IEEE International Conference on.
- 725 Araujo, A., Cacho, N., Thome, A. C., Medeiros, A., & Borges, J. (2017). *A predictive policing*
726 *application to support patrol planning in smart cities*. Paper presented at the 2017
727 International Smart Cities Conference (ISC2).
- 728 Arauz, M. R., Moreno, Y., Nancalres, R., Pérez, C. V., & Larios, V. M. (2017). *Tackling corruption*
729 *in urban development through open data and citizen empowerment: The case of "visor*
730 *urbano" in guadalajara*. Paper presented at the 2017 International Smart Cities Conference
731 (ISC2).
- 732 Baba, M., Pescaru, D., Gui, V., & Jian, I. (2016). *Stray dogs behavior detection in urban area video*
733 *surveillance streams*. Paper presented at the 2016 12th IEEE International Symposium on
734 Electronics and Telecommunications (ISETC).

- 735 Baldoni, G., Melita, M., Micalizzi, S., Rametta, C., Schembra, G., & Vassallo, A. (2017). *A*
736 *dynamic, plug-and-play and efficient video surveillance platform for smart cities*. Paper
737 presented at the 2017 14th IEEE Annual Consumer Communications & Networking
738 Conference (CCNC).
- 739 Balla, P. B., & Jadhao, K. (2018). *IoT Based Facial Recognition Security System*. Paper presented
740 at the 2018 International Conference on Smart City and Emerging Technology (ICSCET).
- 741 Ballesteros, J., Rahman, M., Carbanar, B., & Rishe, N. (2012). *Safe cities. A participatory sensing*
742 *approach*. Paper presented at the 37th Annual IEEE Conference on Local Computer
743 Networks.
- 744 Barba, C. T., Mateos, M. A., Soto, P. R., Mezher, A. M., & Igartua, M. A. (2012). *Smart city for*
745 *VANETs using warning messages, traffic statistics and intelligent traffic lights*. Paper
746 presented at the 2012 IEEE Intelligent Vehicles Symposium.
- 747 Bartoli, G., Fantacci, R., Gei, F., Marabissi, D., & Micciullo, L. (2015). A novel emergency
748 management platform for smart public safety. *International Journal of Communication*
749 *Systems*, 28(5), 928-943.
- 750 Bellini, P., Cenni, D., Nesi, P., & Paoli, I. (2017). Wi-Fi based city users' behaviour analysis for
751 smart city. *Journal of Visual Languages & Computing*, 42, 31-45.
- 752 Belur, J., Tompson, L., Thornton, A., & Simon, M. (2018). Interrater reliability in systematic
753 review methodology: exploring variation in coder decision-making. *Sociological methods*
754 *& research*, 0049124118799372.
- 755 Benkő, M., & Germán, T. (2016). Crime prevention aspects of public space renewal in Budapest.
756 *Journal of Place Management and Development*, 9(2), 191-209.
- 757 Berry, B. J. (2015). *The human consequences of urbanisation*: Macmillan International Higher
758 Education.
- 759 Berry, M. (2018). Technology and organised crime in the smart city: an ethnographic study of the
760 illicit drug trade. *City, Territory and Architecture*, 5(1), 16.
- 761 Bonatsos, A., Middleton, L., Melas, P., & Sabeur, Z. (2013). *Crime open data aggregation and*
762 *management for the design of safer spaces in urban environments*. Paper presented at the
763 International Symposium on Environmental Software Systems.
- 764 Borges, J., Ziehr, D., Beigl, M., Cacho, N., Martins, A., Sudrich, S., . . . Etter, M. (2017). *Feature*
765 *engineering for crime hotspot detection*. Paper presented at the 2017 IEEE SmartWorld,
766 Ubiquitous Intelligence & Computing, Advanced & Trusted Computed, Scalable
767 Computing & Communications, Cloud & Big Data Computing, Internet of People and
768 Smart City Innovation (SmartWorld/SCALCOM/UIC/ATC/CBDCOM/IOP/SCI).
- 769 Borrion, H., Ekblom, P., Alrajeh, D., Borrion, A. L., Keane, A., Koch, D., . . . Toubaline, S. (2019).
770 The Problem with Crime Problem-Solving: Towards a Second Generation Pop? *The British*
771 *Journal of Criminology*.
- 772 Borrion, H., Tripathi, K., Chen, P., & Moon, S. (2014). Threat detection: A framework for security
773 architects and designers of metropolitan rail systems. *Urban, Planning and Transport*
774 *Research*, 2(1), 173-194.

- 775 Boukerche, A., Siddiqui, A. J., & Mammeri, A. (2017). Automated Vehicle Detection and
776 Classification: Models, Methods, and Techniques. *ACM Computing Surveys (CSUR)*,
777 50(5), 62.
- 778 Bourmpos, M., Argyris, A., & Syvridis, D. (2014). Smart city surveillance through low-cost fiber
779 sensors in metropolitan optical networks. *Fiber and Integrated Optics*, 33(3), 205-223.
- 780 Boyatzis, R. E. (1998). *Transforming qualitative information: Thematic analysis and code*
781 *development*: Sage.
- 782 Brantingham, P. J., Valasik, M., & Mohler, G. O. (2018). Does predictive policing lead to biased
783 arrests? Results from a randomized controlled trial. *Statistics and Public Policy*, 5(1), 1-6.
- 784 Braun, T., Fung, B. C., Iqbal, F., & Shah, B. (2018). Security and privacy challenges in smart cities.
785 *Sustainable cities and society*, 39, 499-507.
- 786 Brayne, S. (2017). Big data surveillance: The case of policing. *American Sociological Review*,
787 82(5), 977-1008.
- 788 Brust, M. R., Danoy, G., Bouvry, P., Gashi, D., Pathak, H., & Gonçalves, M. P. (2017). *Defending*
789 *against intrusion of malicious uavs with networked uav defense swarms*. Paper presented at
790 the 2017 IEEE 42nd Conference on Local Computer Networks Workshops (LCN
791 Workshops).
- 792 Byun, J.-Y., Nasridinov, A., & Park, Y.-H. (2014). Internet of things for smart crime detection.
793 *Contemporary Engineering Sciences*, 7(15), 749-754.
- 794 Cagliero, L., Cerquitelli, T., Chiusano, S., Garino, P., Nardone, M., Pralio, B., & Venturini, L.
795 (2015). *Monitoring the citizens' perception on urban security in Smart City environments*.
796 Paper presented at the 2015 31st IEEE International Conference on Data Engineering
797 Workshops.
- 798 Calavia, L., Baladrón, C., Aguiar, J. M., Carro, B., & Sánchez-Esguevillas, A. (2012). A semantic
799 autonomous video surveillance system for dense camera networks in smart cities. *Sensors*,
800 12(8), 10407-10429.
- 801 Camboim, H. B., Neto, A. J. V., Rodrigues, J. J., & Zhao, Z. (2017). *Applying Fog Computing to*
802 *Improve Crime Assistance in Smart Transportation Safety Systems*. Paper presented at the
803 2017 IEEE First Summer School on Smart Cities (S3C).
- 804 Carreño, P., Gutierrez, F. J., Ochoa, S. F., & Fortino, G. (2015). Supporting personal security using
805 participatory sensing. *Concurrency and Computation: Practice and Experience*, 27(10),
806 2531-2546.
- 807 Castella-Roca, J., Mut-Puigserver, M., Payeras-Capella, M. M., Viejo, A., & Angles-Tafalla, C.
808 (2017). *Secure and Anonymous Vehicle Access Control System to Traffic-Restricted Urban*
809 *Areas*. Paper presented at the 2017 26th International Conference on Computer
810 Communication and Networks (ICCCN).
- 811 Castelli, M., Sormani, R., Trujillo, L., & Popovič, A. (2017). Predicting per capita violent crimes
812 in urban areas: an artificial intelligence approach. *Journal of Ambient Intelligence and*
813 *Humanized Computing*, 8(1), 29-36.

- 814 Catlett, C., Cesario, E., Talia, D., & Vinci, A. (2018). *A Data-Driven Approach for Spatio-*
815 *Temporal Crime Predictions in Smart Cities*. Paper presented at the 2018 IEEE
816 International Conference on Smart Computing (SMARTCOMP).
- 817 Cemgil, T., Kurutmaz, B., Cezayirli, A., Bingol, E., & Sener, S. (2017). *Interpolation and fraud*
818 *detection on data collected by automatic meter reading*. Paper presented at the 2017 5th
819 International Istanbul Smart Grid and Cities Congress and Fair (ICSG).
- 820 Chackravarthy, S., Schmitt, S., & Yang, L. (2018). *Intelligent Crime Anomaly Detection in Smart*
821 *Cities Using Deep Learning*. Paper presented at the 2018 IEEE 4th International Conference
822 on Collaboration and Internet Computing (CIC).
- 823 Chen, N., Chen, Y., You, Y., Ling, H., Liang, P., & Zimmermann, R. (2016). *Dynamic urban*
824 *surveillance video stream processing using fog computing*. Paper presented at the 2016
825 IEEE second international conference on multimedia big data (BigMM).
- 826 Chen, X., Xu, J.-B., & Guo, W.-Q. (2013). *The research about video surveillance platform based*
827 *on cloud computing*. Paper presented at the 2013 International Conference on Machine
828 Learning and Cybernetics.
- 829 Cho, Y. I. (2012). *Designing smart cities: Security issues*. Paper presented at the IFIP International
830 Conference on Computer Information Systems and Industrial Management.
- 831 Cocchia, A. (2014). Smart and digital city: A systematic literature review *Smart city* (pp. 13-43):
832 Springer.
- 833 Cozens, P., & Davies, T. (2013). Crime and residential security shutters in an Australian suburb:
834 Exploring perceptions of 'Eyes on the Street', social interaction and personal safety. *Crime*
835 *prevention and community safety*, 15(3), 175-191.
- 836 Cubik, J., Kepak, S., Nedoma, J., Fajkus, M., Zboril, O., Novak, M., . . . Vasinek, V. (2017). *Fiber*
837 *optic perimeter system for security in smart city*. Paper presented at the Electro-Optical
838 Remote Sensing XI.
- 839 Datta, S., & Sarkar, S. (2017). *Automation, security and surveillance for a smart city: Smart, digital*
840 *city*. Paper presented at the 2017 IEEE Calcutta Conference (CALCON).
- 841 Dbouk, M., Mcheick, H., & Sbeity, I. (2014). CityPro; An integrated city-protection collaborative
842 platform. *Procedia Computer Science*, 37, 72-79.
- 843 de Diego, I. M., San Román, I., Montero, J. C., Conde, C., & Cabello, E. (2018). Scalable and
844 flexible wireless distributed architecture for intelligent video surveillance systems.
845 *Multimedia Tools and Applications*, 1-23.
- 846 de Kort, Y., IJsselsteijn, W., Haans, A., Lakens, D., Kalinauskaite, I., & Schietecat, A. (2014). *De-*
847 *escalate: Defusing escalating behaviour through the use of interactive light scenarios*.
848 Paper presented at the Proc. Experiencing Light.
- 849 Degeling, M., & Berendt, B. (2017). What is wrong about Robocops as consultants? A technology-
850 centric critique of predictive policing. *AI & SOCIETY*, 1-10.
- 851 Dey, S., Chakraborty, A., Naskar, S., & Misra, P. (2012). *Smart city surveillance: Leveraging*
852 *benefits of cloud data stores*. Paper presented at the 37th Annual IEEE Conference on Local
853 Computer Networks-Workshops.

- 854 Duan, L., Lou, Y., Wang, S., Gao, W., & Rui, Y. (2018). AI oriented large-scale video management
855 for smart city: Technologies, standards and beyond. *IEEE MultiMedia*.
- 856 Durga, S., Surya, S., & Daniel, E. (2018). *SmartMobiCam: Towards a New Paradigm for*
857 *Leveraging Smartphone Cameras and IaaS Cloud for Smart City Video Surveillance*. Paper
858 presented at the 2018 2nd International Conference on Trends in Electronics and
859 Informatics (ICOEI).
- 860 Eigenraam, D., & Rothkrantz, L. (2016). *A smart surveillance system of distributed smart multi*
861 *cameras modelled as agents*. Paper presented at the 2016 Smart Cities Symposium Prague
862 (SCSP).
- 863 Ekblom, P., & Hirschfield, A. (2014). Developing an alternative formulation of SCP principles—
864 the Ds (11 and counting). *Crime Science*, 3(1), 2.
- 865 Elliott-Davies, M., Donnelly, J., Boag-Munroe, F., & Van Mechelen, D. (2016). ‘Getting a
866 battering’ The perceived impact of demand and capacity imbalance within the Police
867 Service of England and Wales: A qualitative review. *The Police Journal*, 89(2), 93-116.
- 868 Elmaghraby, A. S., & Losavio, M. M. (2014). Cyber security challenges in Smart Cities: Safety,
869 security and privacy. *Journal of advanced research*, 5(4), 491-497.
- 870 Ertugrul, E., Kocaman, U., & Sahingoz, O. K. (2018). *Autonomous aerial navigation and mapping*
871 *for security of smart buildings*. Paper presented at the 2018 6th International Istanbul Smart
872 Grids and Cities Congress and Fair (ICSG).
- 873 Federici, J. F., Schulkin, B., Huang, F., Gary, D., Barat, R., Oliveira, F., & Zimdars, D. (2005).
874 THz imaging and sensing for security applications—explosives, weapons and drugs.
875 *Semiconductor Science and Technology*, 20(7), S266.
- 876 Feng, G. C. (2014). Intercoder reliability indices: disuse, misuse, and abuse. *Quality & Quantity*,
877 48(3), 1803-1815.
- 878 Fernández, J., Calavia, L., Baladrón, C., Aguiar, J., Carro, B., Sánchez-Esguevillas, A., . . .
879 Smilansky, Z. (2013). An intelligent surveillance platform for large metropolitan areas with
880 dense sensor deployment. *Sensors*, 13(6), 7414-7442.
- 881 Ferreira, J. E., Visintin, J. A., Okamoto, J., & Pu, C. (2017). *Smart services: A case study on smarter*
882 *public safety by a mobile app for University of São Paulo*. Paper presented at the 2017 IEEE
883 SmartWorld, Ubiquitous Intelligence & Computing, Advanced & Trusted Computed,
884 Scalable Computing & Communications, Cloud & Big Data Computing, Internet of People
885 and Smart City Innovation (SmartWorld/SCALCOM/UIC/ATC/CBDCom/IOP/SCI).
- 886 Filipponi, L., Vitaletti, A., Landi, G., Memeo, V., Laura, G., & Pucci, P. (2010). *Smart city: An*
887 *event driven architecture for monitoring public spaces with heterogeneous sensors*. Paper
888 presented at the 2010 Fourth International Conference on Sensor Technologies and
889 Applications.
- 890 García, C. G., Meana-Llorián, D., G-Bustelo, B. C. P., Lovelle, J. M. C., & Garcia-Fernandez, N.
891 (2017). Midgar: Detection of people through computer vision in the Internet of Things
892 scenarios to improve the security in Smart Cities, Smart Towns, and Smart Homes. *Future*
893 *Generation Computer Systems*, 76, 301-313.

- 894 García, R. O., Valentín, L., Serrano, S. A., Palacios-Alonso, M. A., & Sucar, L. E. (2017).
895 GEOVISUALIZATION FOR SMART VIDEO SURVEILLANCE. *ISPRS Annals of*
896 *Photogrammetry, Remote Sensing & Spatial Information Sciences*, 4.
- 897 Garg, R., Malik, A., & Raj, G. (2018). *A Comprehensive Analysis for Crime Prediction in Smart*
898 *City Using R Programming*. Paper presented at the 2018 8th International Conference on
899 Cloud Computing, Data Science & Engineering (Confluence).
- 900 Gaur, A., Scotney, B., Parr, G., & McClean, S. (2015). Smart city architecture and its applications
901 based on IoT. *Procedia Computer Science*, 52, 1089-1094.
- 902 Giyenko, A., & Im Cho, Y. (2016). *Intelligent UAV in smart cities using IoT*. Paper presented at
903 the 2016 16th International Conference on Control, Automation and Systems (ICCAS).
- 904 Gohar, M., Muzammal, M., & Rahman, A. U. (2018). SMART TSS: Defining transportation
905 system behavior using big data analytics in smart cities. *Sustainable cities and society*, 41,
906 114-119.
- 907 Gough, D., Oliver, S., & Thomas, J. (2017). *An introduction to systematic reviews*: Sage.
- 908 Greenfield, V. A., & Paoli, L. (2013). A framework to assess the harms of crimes. *British Journal*
909 *of Criminology*, 53(5), 864-885.
- 910 Gupta, A., Chakraborty, N., & Mondal, S. (2017). *CETD: An efficient clustering based energy theft*
911 *detection technique in smart grid*. Paper presented at the 2017 IEEE Region 10 Symposium
912 (TENSymp).
- 913 Habibzadeh, H., Soyata, T., Kantarci, B., Boukerche, A., & Kaptan, C. (2018). Sensing,
914 communication and security planes: A new challenge for a smart city system design.
915 *Computer Networks*, 144, 163-200.
- 916 Hadjkacem, B., Ayedi, W., Abid, M., & Snoussi, H. (2017). *A new method of video-surveillance*
917 *data analytics for the security in camera networks*. Paper presented at the 2017 International
918 Conference on Internet of Things, Embedded Systems and Communications (IINTEC).
- 919 Hall, R. E., Bowerman, B., Braverman, J., Taylor, J., Todosow, H., & Von Wimmersperg, U.
920 (2000). *The vision of a smart city*.
- 921 Hardmeier, D., Hofer, F., & Schwaninger, A. (2005). *The X-ray object recognition test (X-ray*
922 *ORT)-a reliable and valid instrument for measuring visual abilities needed in X-ray*
923 *screening*. Paper presented at the Security Technology, 2005. CCST'05. 39th Annual 2005
924 International Carnahan Conference on.
- 925 Hartama, D., Mawengkang, H., Zarlis, M., Sembiring, R. W., Furqan, M., Abdullah, D., & Rahim,
926 R. (2017). *A research framework of disaster traffic management to Smart City*. Paper
927 presented at the 2017 Second International Conference on Informatics and Computing
928 (ICIC).
- 929 Hillier, D., & Cozens, P. (2012). Revisiting Jane Jacobs's 'Eyes on the Street' for the Twenty-First
930 Century: Evidence from Environmental Criminology *The Urban Wisdom of Jane Jacobs*
931 (pp. 202-220): Routledge.
- 932 Hochstetler, J., Hochstetler, L., & Fu, S. (2016). *An optimal police patrol planning strategy for*
933 *smart city safety*. Paper presented at the 2016 IEEE 18th International Conference on High
934 Performance Computing and Communications; IEEE 14th International Conference on

- 935 Smart City; IEEE 2nd International Conference on Data Science and Systems
936 (HPCC/SmartCity/DSS).
- 937 Hosseini, M., Salehi, M. A., & Gottumukkala, R. (2017). *Enabling interactive video streaming for*
938 *public safety monitoring through batch scheduling*. Paper presented at the 2017 IEEE 19th
939 International Conference on High Performance Computing and Communications; IEEE
940 15th International Conference on Smart City; IEEE 3rd International Conference on Data
941 Science and Systems (HPCC/SmartCity/DSS).
- 942 Hu, L., & Ni, Q. (2018). IoT-driven automated object detection algorithm for urban surveillance
943 systems in smart cities. *IEEE Internet of Things Journal*, 5(2), 747-754.
- 944 Huang, P.-R., & Chu, E. T.-H. (2017). *Indoor trapped-victim detection system*. Paper presented at
945 the 2017 IEEE SmartWorld, Ubiquitous Intelligence & Computing, Advanced & Trusted
946 Computed, Scalable Computing & Communications, Cloud & Big Data Computing,
947 Internet of People and Smart City Innovation
948 (SmartWorld/SCALCOM/UIC/ATC/CBDCom/IOP/SCI).
- 949 Isafiade, O. E., & Bagula, A. B. (2017). *Fostering smart city development in developing nations:*
950 *A crime series data analytics approach*. Paper presented at the 2017 ITU Kaleidoscope:
951 Challenges for a Data-Driven Society (ITU K).
- 952 Jalali, R., El-Khatib, K., & McGregor, C. (2015). *Smart city architecture for community level*
953 *services through the internet of things*. Paper presented at the 2015 18th International
954 Conference on Intelligence in Next Generation Networks.
- 955 Jun, S., Chang, T.-W., Jeong, H., & Lee, S. (2017). Camera Placement in Smart Cities for
956 Maximizing Weighted Coverage with Budget Limit. *IEEE Sensors Journal*, 17(23), 7694-
957 7703.
- 958 Kagawa, T., Saiki, S., & Nakamura, M. (2017). *Developing personalized security information*
959 *service using open data*. Paper presented at the 2017 18th IEEE/ACIS International
960 Conference on Software Engineering, Artificial Intelligence, Networking and
961 Parallel/Distributed Computing (SNPD).
- 962 Kagawa, T., Saiki, S., & Nakamura, M. (2018). Analyzing street crimes in Kobe city using PRISM.
963 *International Journal of Web Information Systems*.
- 964 Khan, E. S., Azmi, H., Ansari, F., & Dhalvelkar, S. (2018). *Simple Implementation of Criminal*
965 *Investigation using Call Data Records (CDRs) through Big Data Technology*. Paper
966 presented at the 2018 International Conference on Smart City and Emerging Technology
967 (ICSCET).
- 968 Khorov, E., Gushchin, A., & Safonov, A. (2015). *Distortion Avoidance While Streaming Public*
969 *Safety Video in Smart Cities*. Paper presented at the International Workshop on Multiple
970 Access Communications.
- 971 Kim, S., & Lee, J. (2012). E-participation, transparency, and trust in local government. *Public*
972 *Administration Review*, 72(6), 819-828.
- 973 Kirschenbaum, A. A., Mariani, M., Van Gulijk, C., Rapaport, C., & Lubasz, S. (2012). Airports at
974 risk: the impact of information sources on security decisions. *Journal of Transportation*
975 *Security*, 5(3), 187-197.

- 976 Kumar, S., Datta, D., Singh, S. K., & Sangaiah, A. K. (2018). An intelligent decision computing
977 paradigm for crowd monitoring in the smart city. *Journal of Parallel and Distributed*
978 *Computing, 118*, 344-358.
- 979 Kunst, R., Avila, L., Pignaton, E., Bampi, S., & Rochol, J. (2018). Improving network resources
980 allocation in smart cities video surveillance. *Computer Networks, 134*, 228-244.
- 981 Landis, J. R., & Koch, G. G. (1977). The measurement of observer agreement for categorical data.
982 *biometrics, 159-174*.
- 983 Lee, H., Smeaton, A. F., O'Connor, N., & Murphy, N. (2005). User-interface to a CCTV video
984 search system.
- 985 Lei, J., Jiang, T., Wu, K., Du, H., Zhu, G., & Wang, Z. (2016). Robust K-means algorithm with
986 automatically splitting and merging clusters and its applications for surveillance data.
987 *Multimedia Tools and Applications, 75(19)*, 12043-12059.
- 988 Lella, J., Mandla, V. R., & Zhu, X. (2017). Solid waste collection/transport optimization and
989 vegetation land cover estimation using Geographic Information System (GIS): A case study
990 of a proposed smart-city. *Sustainable cities and society, 35*, 336-349.
- 991 Liu, K., Warade, N., Pai, T., & Gupta, K. (2017a). *Location-aware smart campus security*
992 *application*. Paper presented at the 2017 IEEE SmartWorld, Ubiquitous Intelligence &
993 Computing, Advanced & Trusted Computed, Scalable Computing & Communications,
994 Cloud & Big Data Computing, Internet of People and Smart City Innovation
995 (SmartWorld/SCALCOM/UIC/ATC/CBDCOM/IOP/SCI).
- 996 Liu, S., Ni, L. M., & Krishnan, R. (2014). Fraud detection from taxis' driving behaviors. *IEEE*
997 *Transactions on Vehicular Technology, 63(1)*, 464-472.
- 998 Liu, W.-C., & Lin, C.-H. (2017). *A hierarchical license plate recognition system using supervised*
999 *K-means and Support Vector Machine*. Paper presented at the 2017 International
1000 Conference on Applied System Innovation (ICASI).
- 1001 Liu, Y., Yu, L., Chi, T., Yang, B., Yao, X., Yang, L., . . . Cui, S. (2017b). *Design and*
1002 *implementation of community safety management oriented public information platform for*
1003 *a smart city*. Paper presented at the 2017 Forum on Cooperative Positioning and Service
1004 (CPGPS).
- 1005 Lohokare, J., Dani, R., Sontakke, S., Apte, A., & Sahni, R. (2017). *Emergency services platform*
1006 *for smart cities*. Paper presented at the 2017 IEEE Region 10 Symposium (TENSYPMP).
- 1007 Ma, X., He, Y., Luo, X., Li, J., Zhao, M., An, B., & Guan, X. (2018). Camera Placement Based on
1008 Vehicle Traffic for Better City Security Surveillance. *IEEE Intelligent Systems, 33(4)*, 49-
1009 61.
- 1010 Macintosh, A. (2004). *Characterizing e-participation in policy-making*. Paper presented at the 37th
1011 Annual Hawaii International Conference on System Sciences, 2004. Proceedings of the.
- 1012 Mahajan, M., Reddy, K., & Rajput, M. (2018). *A Switch Triggered Rescue Assistance System for*
1013 *Safety of Women*. Paper presented at the 2018 International Conference on Smart City and
1014 Emerging Technology (ICSCET).

- 1015 Manasa, N. (2016). *Nano Sensors and Pattern Recognition for Detection of Hidden Explosives*.
 1016 Paper presented at the Proceedings of the Second International Conference on Information
 1017 and Communication Technology for Competitive Strategies.
- 1018 Marx, G. T. (1998). Ethics for the new surveillance. *The Information Society*, 14(3), 171-185.
- 1019 Mata, F., Torres-Ruiz, M., Guzmán, G., Quintero, R., Zagal-Flores, R., Moreno-Ibarra, M., & Loza,
 1020 E. (2016). A mobile information system based on crowd-sensed and official crime data for
 1021 finding safe routes: A case study of Mexico City. *Mobile Information Systems*, 2016.
- 1022 McCoy, T., Bullock, R., & Brennan, P. (2005). RFID for airport security and efficiency.
- 1023 McLeod, S. (2007). Maslow's hierarchy of needs. *Simply psychology*, 1.
- 1024 Mehboob, F., Abbas, M., Rehman, S., Khan, S. A., Jiang, R., & Bouridane, A. (2017). Glyph-based
 1025 video visualization on Google Map for surveillance in smart cities. *EURASIP Journal on*
 1026 *Image and Video Processing*, 2017(1), 28.
- 1027 Memos, V. A., Psannis, K. E., Ishibashi, Y., Kim, B.-G., & Gupta, B. B. (2018). An efficient
 1028 algorithm for media-based surveillance system (EAMSuS) in IoT smart city framework.
 1029 *Future Generation Computer Systems*, 83, 619-628.
- 1030 Miraftebzadeh, S. A., Rad, P., Choo, K.-K. R., & Jamshidi, M. (2018). A Privacy-Aware
 1031 Architecture at the Edge for Autonomous Real-Time Identity Reidentification in Crowds.
 1032 *IEEE Internet of Things Journal*, 5(4), 2936-2946.
- 1033 Mishra, D., & Kumar, M. (2013). Role of Technology in Smart Governance: 'Smart City, Safe City'.
 1034 *Safe City* (August 15, 2013).
- 1035 Mlinarić, A., Horvat, M., & Šupak Smolčić, V. (2017). Dealing with the positive publication bias:
 1036 Why you should really publish your negative results. *Biochemia medica: Biochemia*
 1037 *medica*, 27(3), 1-6.
- 1038 Monahan, T., & Mokos, J. T. (2013). Crowdsourcing urban surveillance: The development of
 1039 homeland security markets for environmental sensor networks. *Geoforum*, 49, 279-288.
- 1040 Moreira, B., Cacho, N., Lopes, F., & Cavalcante, E. (2017). *Towards civic engagement in smart*
 1041 *public security*. Paper presented at the 2017 International Smart Cities Conference (ISC2).
- 1042 Morton, P. J., Horne, M., Dalton, R. C., & Thompson, E. M. (2012). Virtual city models: Avoidance
 1043 of obsolescence. *Education and Research in Computer Aided Architectural Design in*
 1044 *Europe-eCAADe*, 1, 213-224.
- 1045 Naphade, M., Banavar, G., Harrison, C., Paraszczak, J., & Morris, R. (2011). Smarter cities and
 1046 their innovation challenges. *Computer*, 44(6), 32-39.
- 1047 Nasui, D., Cernian, A., & Sgarciu, V. (2014). *Cloud based Student Transportation Safety System*.
 1048 Paper presented at the Proceedings of the 2014 6th International Conference on Electronics,
 1049 Computers and Artificial Intelligence (ECAI).
- 1050 Noor, N. M. M., Nawawi, W. M. F. W., & Ghazali, A. F. (2013). *Supporting decision making in*
 1051 *situational crime prevention using fuzzy association rule*. Paper presented at the 2013
 1052 International Conference on Computer, Control, Informatics and Its Applications
 1053 (IC3INA).

- 1054 Oatley, G., Crick, T., & Bolt, D. (2015). *CCTV as a smart sensor network*. Paper presented at the
1055 2015 IEEE International Conference on Computer and Information Technology;
1056 Ubiquitous Computing and Communications; Dependable, Autonomic and Secure
1057 Computing; Pervasive Intelligence and Computing.
- 1058 Olive, E. W., Laube, R., & Hofer, F. (2009). *A comparison between two leadership models for*
1059 *security checkpoints*. Paper presented at the Security Technology, 2009. 43rd Annual 2009
1060 International Carnahan Conference on.
- 1061 Oza, N., & Gohil, N. (2016). *Implementation of cloud based live streaming for surveillance*. Paper
1062 presented at the 2016 International Conference on Communication and Signal Processing
1063 (ICCSP).
- 1064 Parra, J., & Lopez, R. (2017). Application of predictive analytics for crime prevention: The case of
1065 the City of San Francisco *Police: Global Perceptions, Performance and Ethical Challenges*
1066 (pp. 85-109): Nova Science Publishers, Inc.
- 1067 Patel, J., Wala, H., Shahu, D., & Lopes, H. (2018). *Intellectual and Enhance Digital Solution For*
1068 *Police Station*. Paper presented at the 2018 International Conference on Smart City and
1069 Emerging Technology (ICSCET).
- 1070 Patton, J. W. (2000). Protecting privacy in public? Surveillance technologies and the value of
1071 public places. *Ethics and Information Technology*, 2(3), 181-187.
- 1072 Peixoto, M. L., Souza, I., Barbosa, M., Lecomte, G., Batista, B. G., Kuehne, B. T., & Leite Filho,
1073 D. M. (2018). *Data Missing Problem in Smart Surveillance Environment*. Paper presented
1074 at the 2018 International Conference on High Performance Computing & Simulation
1075 (HPCS).
- 1076 Peng, Z., Xiao, B., Yao, Y., Guan, J., & Yang, F. (2017). *U-safety: Urban safety analysis in a smart*
1077 *city*. Paper presented at the 2017 IEEE International Conference on Communications (ICC).
- 1078 Pereira, R., Correia, D., Mendes, L., Rabadão, C., Barroso, J., & Pereira, A. (2018). *Low-Cost*
1079 *Smart Surveillance System for Smart Cities*. Paper presented at the International Conference
1080 on Universal Access in Human-Computer Interaction.
- 1081 Pribadi, A., Kumiawan, F., Hariadi, M., & Nugroho, S. M. S. (2017). *Urban distribution CCTV for*
1082 *smart city using decision tree methods*. Paper presented at the 2017 International Seminar
1083 on Intelligent Technology and Its Applications (ISITIA).
- 1084 Ralko, S., & Kumar, S. (2016). Smart City Security.
- 1085 Ramaprasad, A., Sánchez-Ortiz, A., & Syn, T. (2017). *A unified definition of a smart city*. Paper
1086 presented at the International Conference on Electronic Government.
- 1087 Rametta, C., Baldoni, G., Lombardo, A., Micalizzi, S., & Vassallo, A. (2017). S6: a Smart, Social
1088 and SDN-based Surveillance System for Smart-cities. *Procedia Computer Science*, 110,
1089 361-368.
- 1090 Ramírez, C. A., Barragán, R., García-Torales, G., & Larios, V. M. (2016). *Low-power device for*
1091 *wireless sensor network for Smart Cities*. Paper presented at the 2016 IEEE MTT-S Latin
1092 America Microwave Conference (LAMC).
- 1093 Ratcliffe, J. H. (2015). Towards an index for harm-focused policing. *Policing: A journal of policy*
1094 *and practice*, 9(2), 164-182.

- 1095 Rathore, M. M., Ahmad, A., Paul, A., & Rho, S. (2016). Urban planning and building smart cities
1096 based on the internet of things using big data analytics. *Computer Networks*, 101, 63-80.
- 1097 Reddy, A. G., Suresh, D., Phaneendra, K., Shin, J. S., & Odelu, V. (2018a). Provably secure
1098 pseudo-identity based device authentication for smart cities environment. *Sustainable cities
1099 and society*, 41, 878-885.
- 1100 Reddy, K. B. S., Loke, O., Jani, S., & Dabre, K. (2018b). *Tracking People In Real Time Video
1101 Footage Using Facial Recognition*. Paper presented at the 2018 International Conference
1102 on Smart City and Emerging Technology (ICSCET).
- 1103 Rocher, J., Taha, M., Parra, L., & Lloret, J. (2018). *IoT Sensor to Detect Fraudulent Use of Dyed
1104 Fuels in Smart Cities*. Paper presented at the 2018 Fifth International Conference on
1105 Internet of Things: Systems, Management and Security.
- 1106 Rohstein, H., & Hopewell, S. (2009). Grey Literature. In H. Cooper, L. V. Hedges, & J. Valentine
1107 (Eds.), *Handbook of Research on Adult Learning and Development* (2nd ed., pp. 184-202):
1108 Routledge.
- 1109 Rothkrantz, L. (2017a). *Lip-reading by surveillance cameras*. Paper presented at the 2017 Smart
1110 City Symposium Prague (SCSP).
- 1111 Rothkrantz, L. (2017b). *Person identification by smart cameras*. Paper presented at the 2017 Smart
1112 City Symposium Prague (SCSP).
- 1113 Saba, A. (2017). *IOT based energy efficient security system*. Paper presented at the 2017 3rd
1114 International Conference on Applied and Theoretical Computing and Communication
1115 Technology (iCATccT).
- 1116 Sadgali, I., Sael, N., & Benabbou, F. (2018). Detection of credit card fraud: State of art.
1117 *INTERNATIONAL JOURNAL OF COMPUTER SCIENCE AND NETWORK SECURITY*,
1118 18(11), 76-83.
- 1119 Sajjad, M., Nasir, M., Muhammad, K., Khan, S., Jan, Z., Sangaiah, A. K., . . . Baik, S. W. (2017).
1120 Raspberry Pi assisted face recognition framework for enhanced law-enforcement services
1121 in smart cities. *Future Generation Computer Systems*.
- 1122 Sajjad, M., Nasir, M., Ullah, F. U. M., Muhammad, K., Sangaiah, A. K., & Baik, S. W. (2018).
1123 Raspberry Pi assisted facial expression recognition framework for smart security in law-
1124 enforcement services. *Information Sciences*, 479, 416-431.
- 1125 Salmerón-García, J. J., van den Dries, S., Díaz-del-Río, F., Morgado-Estevez, A., Sevillano-
1126 Ramos, J. L., & van de Molengraft, M. (2017). Towards a cloud-based automated
1127 surveillance system using wireless technologies. *Multimedia Systems*, 1-15.
- 1128 Sandborn, P. (2007). *Designing for technology obsolescence management*. Paper presented at the
1129 IIE Annual Conference. Proceedings.
- 1130 Saravanakumar, K., Deepa, K., & Kumar, N. S. (2017). *A study on possible application of RFID
1131 system in different real-time environments*. Paper presented at the 2017 International
1132 Conference on Circuit, Power and Computing Technologies (ICCPCT).
- 1133 Schuilenburg, M., & Peeters, R. (2018). Smart cities and the architecture of security: pastoral power
1134 and the scripted design of public space. *City, Territory and Architecture*, 5(1), 13.

- 1135 Shi, J., Ming, Y., Fan, C., & Tian, L. (2017). *Face recognition algorithm based on multi-scale*
 1136 *CLBP*. Paper presented at the 2017 IEEE SmartWorld, Ubiquitous Intelligence &
 1137 Computing, Advanced & Trusted Computed, Scalable Computing & Communications,
 1138 Cloud & Big Data Computing, Internet of People and Smart City Innovation
 1139 (SmartWorld/SCALCOM/UIC/ATC/CBDCCom/IOP/SCI).
- 1140 Singh, A., Patil, D., & Omkar, S. (2018). Eye in the Sky: Real-time Drone Surveillance System
 1141 (DSS) for Violent Individuals Identification using ScatterNet Hybrid Deep Learning
 1142 Network. *arXiv preprint arXiv:1806.00746*.
- 1143 Singh, G., Majumdar, S., & Rajan, S. (2017). *MapReduce-based techniques for multiple object*
 1144 *tracking in video analytics*. Paper presented at the 2017 IEEE SmartWorld, Ubiquitous
 1145 Intelligence & Computing, Advanced & Trusted Computed, Scalable Computing &
 1146 Communications, Cloud & Big Data Computing, Internet of People and Smart City
 1147 Innovation (SmartWorld/SCALCOM/UIC/ATC/CBDCCom/IOP/SCI).
- 1148 Siregar, A. H., Syahputra, D., Putra, D. A., & Wicaksono, B. (2018). *Policy Evaluation of Security*
 1149 *System Based on Security Camera Technology in Batam City*. Paper presented at the IOP
 1150 Conference Series: Earth and Environmental Science.
- 1151 Sormani, R., Soldatos, J., Vassilaras, S., Kioumourtzis, G., Leventakis, G., Giordani, I., & Tisato,
 1152 F. (2016). A serious game empowering the prediction of potential terrorist actions. *Journal*
 1153 *of Policing, Intelligence and Counter Terrorism, 11*(1), 30-48.
- 1154 Sudha, N. (2015). *Enabling Seamless Video Processing in Smart Surveillance Cameras with*
 1155 *Multicore*. Paper presented at the 2015 International Conference on Advanced Computing
 1156 and Communications (ADCOM).
- 1157 Sweet, K. (2008). *Aviation and airport security: terrorism and safety concerns*: CRC Press.
- 1158 Tan, H., & Chen, L. (2014). *An approach for fast and parallel video processing on Apache Hadoop*
 1159 *clusters*. Paper presented at the 2014 IEEE International Conference on Multimedia and
 1160 Expo (ICME).
- 1161 Thomas, J., O'Mara-Eves, A., Harden, A., & Newman, M. (2017a). Synthesis Methods for
 1162 Combining and Configuring Quantitative Data. In D. Gough, S. Oliver, & J. Thomas (Eds.),
 1163 *An Introduction to Systematic Reviews* (2nd ed., pp. 181-211). London: SAGE Publications.
- 1164 Thomas, S. S., Gupta, S., & Subramanian, V. K. (2017b). *Smart surveillance based on video*
 1165 *summarization*. Paper presented at the 2017 IEEE Region 10 Symposium (TENSYPMP).
- 1166 Tian, L., Wang, H., Zhou, Y., & Peng, C. (2018). Video big data in smart city: Background
 1167 construction and optimization for surveillance video processing. *Future Generation*
 1168 *Computer Systems, 86*, 1371-1382.
- 1169 Truntsevsky, Y. V., Lukiny, I., Sumachev, A., & Kopytova, A. (2018). *A smart city is a safe city:*
 1170 *the current status of street crime and its victim prevention using a digital application*. Paper
 1171 presented at the MATEC Web of Conferences.
- 1172 Valentín, L., Serrano, S. A., García, R. O., Andrade, A., Palacios-Alonso, M. A., & Sucar, L. E.
 1173 (2017). A CLOUD-BASED ARCHITECTURE FOR SMART VIDEO SURVEILLANCE.
 1174 *International Archives of the Photogrammetry, Remote Sensing & Spatial Information*
 1175 *Sciences, 42*.

- 1176 van Heek, J., Aming, K., & Ziefle, M. (2016). "How fear of crime affects needs for privacy &
1177 safety": Acceptance of surveillance technologies in smart cities. Paper presented at the
1178 Smart Cities and Green ICT Systems (SMARTGREENS), 2016 5th International
1179 Conference on.
- 1180 Venkatesan, S., Jawahar, A., Varsha, S., & Roshne, N. (2017). *Design and implementation of an*
1181 *automated security system using Twilio messaging service*. Paper presented at the 2017
1182 International Conference on Smart Cities, Automation & Intelligent Computing Systems
1183 (ICON-SONICS).
- 1184 Viera, A. J., & Garrett, J. M. (2005). Understanding interobserver agreement: the kappa statistic.
1185 *Fam med*, 37(5), 360-363.
- 1186 Vitalij, F., Robnik, A., & Alexey, T. (2012). " Safe City"-an Open and Reliable Solution for a Safe
1187 and Smart City. *Elektrotehniski Vestnik*, 79(5), 262.
- 1188 Wang, J., Pan, J., & Esposito, F. (2017). *Elastic urban video surveillance system using edge*
1189 *computing*. Paper presented at the Proceedings of the Workshop on Smart Internet of
1190 Things.
- 1191 Welsh, D., & Roy, N. (2017). *Smartphone-based mobile gunshot detection*. Paper presented at the
1192 2017 IEEE International Conference on Pervasive Computing and Communications
1193 Workshops (PerCom Workshops).
- 1194 Wilson, D. B. (2009). Missing a critical piece of the pie: simple document search strategies
1195 inadequate for systematic reviews. *Journal of experimental criminology*, 5(4), 429-440.
- 1196 Wu, C., Zhu, Q., Zhang, Y., Du, Z., Ye, X., Qin, H., & Zhou, Y. (2017). A NOSQL–SQL hybrid
1197 organization and management approach for real-time geospatial data: A case study of public
1198 security video surveillance. *ISPRS International Journal of Geo-Information*, 6(1), 21.
- 1199 Xiong, M., Chen, D., Chen, J., Chen, J., Shi, B., Liang, C., & Hu, R. (2017). Person re-identification
1200 with multiple similarity probabilities using deep metric learning for efficient smart security
1201 applications. *Journal of Parallel and Distributed Computing*.
- 1202 Xu, Z., Mei, L., Liu, Y., Hu, C., & Chen, L. (2016). Semantic enhanced cloud environment for
1203 surveillance data management using video structural description. *Computing*, 98(1-2), 35-
1204 54.
- 1205 Yang, K., Zhang, K., Ren, J., & Shen, X. (2015). Security and privacy in mobile crowdsourcing
1206 networks: challenges and opportunities. *IEEE communications magazine*, 53(8), 75-81.
- 1207 Zhang, C., Ni, B., Song, L., Zhai, G., Yang, X., & Zhang, W. (2016). *BEST: benchmark and*
1208 *evaluation of surveillance task*. Paper presented at the Asian Conference on Computer
1209 Vision.
- 1210 Zhang, F., Wan, M., Yang, G., & Yang, Z. (2017a). *Background modeling from surveillance video*
1211 *via transformed L 1 function*. Paper presented at the 2017 International Smart Cities
1212 Conference (ISC2).
- 1213 Zhang, K., Ni, J., Yang, K., Liang, X., Ren, J., & Shen, X. S. (2017b). Security and privacy in
1214 smart city applications: Challenges and solutions. *IEEE communications magazine*, 55(1),
1215 122-129.

- 1216 Zhang, S., & Yu, H. (2018). Person Re-Identification by Multi-Camera Networks for Internet of
1217 Things in Smart Cities. *IEEE Access*, 6, 76111-76117.
- 1218 Zhang, T., Chowdhery, A., Bahl, P. V., Jamieson, K., & Banerjee, S. (2015). *The design and*
1219 *implementation of a wireless video surveillance system*. Paper presented at the Proceedings
1220 of the 21st Annual International Conference on Mobile Computing and Networking.
- 1221 Zheng, Y., Sheng, H., Zhang, B., Zhang, J., & Xiong, Z. (2015). Weight-based sparse coding for
1222 multi-shot person re-identification. *Science China Information Sciences*, 58(10), 1-15.
- 1223 Zhou, W., Saha, D., & Rangarajan, S. (2015). *A system architecture to aggregate video surveillance*
1224 *data in Smart Cities*. Paper presented at the 2015 IEEE Global Communications Conference
1225 (GLOBECOM).
- 1226 Zhu, S., Li, D., & Feng, H. (2019). Is smart city resilient? Evidence from China. *Sustainable cities*
1227 *and society*, 101636.
- 1228 Zingoni, A., Diani, M., & Corsini, G. (2017). A flexible algorithm for detecting challenging moving
1229 objects in real-time within IR video sequences. *Remote Sensing*, 9(11), 1128.
- 1230 Zygiaris, S. (2013). Smart city reference model: Assisting planners to conceptualize the building
1231 of smart city innovation ecosystems. *Journal of the knowledge economy*, 4(2), 217-231.
- 1232