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A Study on Security and Privacy Guidelines, Countermeasures, threats: IoT Data at Rest Perspective

Hezam Akram Abdulghani , Niels Alexander Nijdam, Anastasija Collen, Dimitri Konstantas

Geneva School of Economics and Management, Geneva University, Switzerland;
{mohammed.akram | niels.nijdam | anastasija.collen | dimitri.konstantas}@unige.ch

* Correspondence: mohammed.akram

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Abstract: Internet of Things (IoT) makes our lives much easier, more valuable, and less stressful due to the development of so many applications around us like smart city, smart car, and smart grid that offer endless services and solutions. Protecting IoT data at rest either on the objects or in the cloud of such applications is an indispensable requirement for achieving a symmetry in the handling and protection of IoT, as we do with data created by persons and applications. This is because unauthorised access to such data may lead to harmful consequences such as linkage attacks, loss of privacy, and data manipulation. Such undesired implications may jeopardise the existence of IoT applications, if protection measures are not taken, and they stem from two main factors. One is IoT objects have limited capabilities in terms of memory capacity, battery life, and computational power that hamper the direct implementation of conventional Internet security solutions without some modifications like traditional symmetric algorithms. Another factor is the absence of widely-accepted IoT security and privacy guidelines for IoT data at rest and their appropriate countermeasures, which would help IoT stakeholders like developers and manufacturers to develop secure IoT systems and, therefore, enhance IoT security and privacy by design. To this end, we first briefly describe the main IoT security goals and identify IoT stakeholders. Moreover, we briefly discuss the most well-known data protection frameworks such as GDPR and HIPAA. Second, we highlight potential attacks and threats against data at rest and show their violated security goals (e.g., confidentiality and integrity). Third, we review a list of protection measures by which our proposed guidelines can be accomplished. Fourth, we propose a framework of security and privacy guidelines for IoT data at rest that can be utilised to enhance IoT security and privacy by design and establish a symmetry with the protection of user created data. Our framework also presents the link between the suggested guidelines, mitigation techniques, and attacks. Moreover, we state those IoT stakeholders like manufacturers and developers who will benefit most of these guidelines. Finally, we suggest several open issues required further investigation in the future, and we also discuss the limitations of our suggested framework.

Keywords: Internet of Things; security guidelines; privacy guidelines; countermeasures; security goals; attacks, IoT data at rest

1. Introduction

The Internet of Things (IoT) is a network of objects equipped with sensors, actuators, electronics and connectivity protocols enabling object interaction with each other without human intervention [1]. IoT is involved in the creation of a variety of applications and services around us, for example in smart cities, smart cars, smart grids, quality of life applications, and electronic gadgets, all of which make our lives more productive and less stressful. For instance, the authors in [2] have proposed an IoT system

33 which can be used to manage students' stress. It is unquestionable that such IoT applications and
34 services provide a huge benefit for human's life, yet they may come with a massive cost for individual's
35 privacy and security protection. This is because IoT inherits most of the issues of Internet associated
36 with location awareness, security, quality of service [3] and, most likely, amplifies them due to direct
37 connection with physical objects [4,5].

38 As IoT applications may store their data locally on objects or remotely in the cloud, based on their
39 storage capabilities, protecting their data at rest is of paramount importance. Several IoT applications
40 may cooperate with each other to accomplish specific tasks or services. If data integrity of a single IoT
41 application at rest has been compromised, then there is a huge risk of dealing with a cascading effect of
42 the data compromise. For instance, the authors in [6] state that a thermostat deployed in a smart home
43 relies heavily on a smoke detector's data to shut a heating system down in case of danger. However,
44 accessing the smoke detector's data by unauthorised objects may put the entire smart home at risk.
45 The countermeasure on ensuring the data integrity across multiple Cloud Storage Services (CSSs) was
46 proposed by [7], where reliance on the Third-Part Auditor (TPA) for data verification was eliminated
47 by use of a decentralised blockchain based Integrity Management Service. Furthermore, once an IoT
48 system stores its data in the cloud, there is no assurance that only authorised objects or users will have
49 access to their data. An example was given by European Union Agency for Network and Information
50 Security (ENISA)¹, where an employee at the SharpLocks company, attacker, was capable (due to given
51 access rights) to send a malicious update from the company's server to all connected IoT objects [8].

52 However, most security and privacy issues of IoT data at rest, such as unauthorised access
53 and weak or absent encryption schemes, arise from two principal reasons. IoT objects have limited
54 capabilities in terms of computational power, memory, and bandwidth [9]. Because of these limitations,
55 a direct implementation of traditional security mechanisms into IoT objects tends to be very hard
56 without some modifications. This is why a new breed of lightweight IoT security techniques and
57 protocols (e.g., a secure system of uploading and replicating IoT data suggested in [10]) has been
58 developed [11,12]. The second reason, which motivates us to conduct this work, is the lack of
59 widely-accepted security and privacy guidelines for IoT at data at rest, along with their appropriate
60 mitigation techniques. The main objective of such guidelines and countermeasures is to improve IoT
61 security and privacy by design by giving IoT stakeholders, such as manufacturers and developers,
62 a chance to embrace such guidelines and countermeasures from the early stages of IoT system
63 development[13]. If IoT stakeholders overlook these guidelines when dealing with IoT data at rest, the
64 appearance of several attacks and threats is inevitable.

65 It is clear that unauthorised access and privacy violations of individuals, associated with data at
66 rest, will appear again and again in IoT systems, unless the mindset of all IoT stakeholders shifts to
67 properly integrate at early stages the security and privacy guidelines for IoT data at rest, along with
68 their corresponding countermeasures. The first step towards this paradigm shift is the development of
69 a comprehensive set of security and privacy guidelines. However, there is a complete lack of research
70 efforts conducted specifically to such objective. To the best of authors' knowledge, there is no paper
71 or document clearly addressing security and privacy guidelines for IoT data at rest along with their
72 mitigation strategies. The existing research proposals [14–21] focus primarily on IoT guidelines in
73 general. They neither provide a comprehensive guidelines for IoT data at rest, nor discuss their proper
74 countermeasures. A detailed explanation of such efforts along with a brief comparison between our
75 suggested guidelines for IoT data at rest and theirs is presented in Section 2.

76 The main contributions of this work are the following:

- 77 1. To highlight IoT security goals as well as IoT stakeholders.
- 78 2. To summarise the attacks and threats against IoT data at rest.

¹ <https://www.enisa.europa.eu/>

- 79 3. To review a set of implementation techniques by which our suggested guidelines can be
80 implemented and also states those IoT stakeholders who will benefit from these guidelines.
- 81 4. To propose a framework of security and privacy guidelines for IoT data at rest that can be utilised
82 to reinforce IoT security and privacy by design.
- 83 5. To discuss open issues, limitations and future work.

84 The rest of the article is structured as follows. In section 2, we present the current research studies
85 on IoT guidelines with focus on data at rest, describe the most popular data protection frameworks
86 and highlight the IoT security goals and distinguish IoT stakeholders. Section 3 discusses threats and
87 attacks on IoT data at rest. We identify the appropriate techniques for mitigating the identified attacks
88 on IoT data at rest in Section 4. A proposed framework on security and privacy guidelines is presented
89 in Section 5. Finally, we discuss open issues for further investigation for future work in Section 6.

90 2. Related work

91 In this section, we outline the current state of the art related to IoT security guidelines, existing
92 frameworks on data protection and identify the involved stakeholders specific for the IoT environment.
93 We limit our discussion to the guidelines on protecting IoT data at rest, as the main topic of this work.

94 2.1. Research efforts on IoT guidelines

95 In [14], the authors have proposed a set of security and privacy guidelines for IoT data at rest,
96 such as minimise data storage, minimise data retention, encrypt data storage, and time-period data
97 aggregation. Furthermore, attacks and threats against IoT data at rest are analysed. Having said that,
98 the authors do not provide a comprehensive set of guidelines for IoT data at rest, nor do they state the
99 required implementation techniques to achieve their guidelines.

100 In [15], the BITAG suggests a list of security and privacy guidelines for IoT data at rest like
101 minimise data storage, encrypt data storage, remove the sensitive data, and ensure data availability.
102 BITAG, however, does not offer a thorough set of guidelines for IoT data at rest, nor does it identify
103 the countermeasures required to carry out its guidelines. Furthermore, threats and attacks against IoT
104 data at rest remain unchecked.

105 In [16], the Open Web Application Security (OWASP) suggests different security and privacy
106 guidelines for IoT data at rest, such as minimise data storage, minimise data retention, ensure
107 authorised access, and prevent physical access. In addition, the OWASP states those IoT stakeholders
108 like manufacturer, developer, and customer who may use its guidelines to protect IoT data at rest.
109 Nevertheless, the OWASP neither recognised the required countermeasures to implement its guidelines,
110 nor distinguishes possible attacks and threats against IoT data at rest.

111 In [17], the ENISA proposes several security and privacy guidelines for IoT data at rest, such
112 as minimise data retention, encrypt data storage, define recovery strategies, inform customers, and
113 proper data destruction. The ENISA, however, does not recognise those IoT stakeholders who may
114 utilise its guidelines, nor does it identify proper solutions to apply its guidelines. Moreover, the ENISA
115 uncovers attacks and threats against IoT data at rest.

116 In [18], the IoTA suggests a list of security and privacy guidelines for IoT data at rest, example
117 of which are encrypt data storage, inform customers, and remove sensitive data. That said, IoTA
118 neither distinguishes the required implementation techniques to fulfil its guidelines, nor points out
119 IoT stakeholders who may use its guidelines. Attacks and threats against IoT data at rest also are left
120 unidentified.

121 In [19], IoT Security Foundation (IoTSF) proposes a set of security and privacy guidelines for IoT
122 data at rest, such as distribute data storage, define recovery strategies, proper data destruction, and
123 search on encrypted data. IoTSF, however, neither recognised suitable countermeasures to realise its
124 guidelines, nor distinguishes attacks and threats against IoT data at rest.

125 In [21], the authors proposes a comprehensive set of security and privacy guidelines for the
126 first two levels of CISCO'S reference model (edge nodes and communication). Even though their

127 guidelines are not meant specifically for protecting IoT data at rest, only three of these guidelines
128 (ensure authorised access, remove or hide sensitive data, and search on encrypted data) can be used
129 to do so. It is worth mentioning that the authors identify all possible threats and attacks against edge
130 nodes and communication, state those IoT stakeholders who may use their guidelines, and recognise
131 suitable implementation techniques to implement them.

132 2.2. Data protection frameworks

133 The lack of efficient standards, regulation, and weak governance is a cause of security and privacy
134 issues of IoT. However, some initiatives at a national level have been proposed, which we briefly
135 present in the next paragraphs.

136 **General Data Protection Regulation (GDPR):** In December of 2016, European Union (EU) has
137 voted to use GDPR as a replacement of outdated Data Protection Directive (DPD) proposed in 1996.
138 The main goal of DPD was to preserve personal information of individuals within EU from being
139 misused and allow the individuals of EU to have better control over their personal data. The GDPR
140 is intended to substitute the DPD as a regulation, and it will cover the whole EU as unified law. The
141 GDPR has included six major changes (e.g., personal data redefined, individuals rights, data controller
142 and processor, and global impact) compared to the DPD. The detailed explanation of each one of
143 them, for interested readers, can be found in [22]. To preserve personal data, the GDPR has imposed
144 six fundamental principles, the detail of which can be found in [23]. It is worth mentioning that not
145 all IoT applications are dealing with personal information, for instance Industrial Internet of Things
146 (IIoT) applications. However, this may be the case in the majority of IoT systems (e.g., healthcare). It is
147 obvious that the market of IoT solutions is increasing worldwide including Europe and it is a relevant
148 topic which needs thorough management. One example is a digital transformation of healthcare due
149 to the fast growth of wearable and interconnected medical objects which provide a remote health
150 monitoring. Therefore, healthcare data is highly sensitive data, and it attracts the attention of attackers.
151 To this end, authors in [24] identified different attacks associated with IoT multi-cloud e-Healthcare
152 environment, such as side-channel attacks and malicious insider attacks. Healthcare data, thus, must
153 be covered under the scope of the GDPR. Other important IoT applications, which deal with personal
154 data like smart metering and smart home applications, must also be covered under the scope of GDPR.
155

156 **Health Insurance Portability (HIPAA):** The main goal of HIPAA is to protect individuals from
157 losing their health insurance if they have pre-existing health problems or they change their jobs.
158 HIPAA, however, has been extended over the years to minimise the administrative and the cost
159 burdens of healthcare processes. Most recently, HIPAA concentrates on developing standards as well
160 as requirements to ensure security and privacy of Personal Health information (PHI) which can be
161 created, stored, or transferred in several formats (e.g., written documents and verbal conversations).
162 PHI may contain anything in patient health records like images, names, email addresses, and other
163 information. As patients demand their data to be secure, HIPAA's security and privacy rules, discussed
164 in details in [25], require healthcare organisations to embrace a set of processes and procedures to assure
165 the highest level of patient confidentiality. Under HIPAA, a covered organisation may not utilise or
166 reveal PHI unless it has received an explicit consensus from a patient to do so. It is unquestionable that
167 IoT will change the healthcare experience. To this point, several examples are available on the market to
168 illustrate how IoT has simplified the process of care management. For instance, collecting information
169 in users' homes will assist healthcare providers to comprehend user's health in a comprehensive way,
170 choose suitable treatment plans, alter the plans as time moves, and most importantly anticipate future
171 health actions. It is clear that IoT can be utilised by healthcare organisations to reduce costs and at
172 the same time enhance health outcomes in patients. Hence, healthcare IoT solutions must be covered
173 under the scope of HIPAA.

174 **Industrial Internet Consortium (ICC):** It is an organisation developed in 2014 to improve the
 175 growth of interconnected objects. The primary goal of such organisation is to build an alliance of
 176 companies, academia, and governments and cooperate on the development of test beds for real-world
 177 systems. Furthermore, it has actively involved in supporting the necessities of standards in the IoT
 178 industry. To this end, ICC in 2016 has proposed a security framework developed specifically for IIoT.
 179 The main purpose of this security framework is to establish global industry acceptance on how to
 180 develop secure IoT systems [26].

181 **IoTSF:** It is a non-profit company, which has been reinforcing the IoT industry since 2015. This
 182 reinforcement includes developing security and privacy guidelines, courses, and training. IoTSF also
 183 has addressed the issues in the industry, and more importantly it has struggled to cover the gap via
 184 a cooperative initiative with many companies dealing with IoT. This kind of cooperation attempts
 185 to share expertise, knowledge, and enhancing best practices. To contribute to such objectives, IoTSF
 186 in [19], for example, has proposed an IoT framework as a checklist which can be used by the IoT
 187 manufacturers to simplify their compliance to IoTSF's framework.

188 2.3. IoT security goals and stakeholders

189 In this section, we first discuss security goals specific to IoT environment. Traditional security
 190 goals, in the literature, are broken down into three primary sets: (i) Confidentiality, (ii) Integrity and
 191 (iii) Availability, referred to as the CIA-triad. Confidentiality assures that only authorised objects or
 192 users can get access to sensitive data. As several IoT objects might deal with sensitive data, such as
 193 medical records and credit cards, the confidentiality of such data must be preserved. The impact of
 194 unauthorised access to medical objects, which may reveal personal data or lead to life-threatening
 195 situations, have been illustrated in [27]. In the IoT context, integrity is also crucial, since it ensures IoT
 196 data has not been tampered. If the integrity of IoT data has been compromised, undesired consequences
 197 may take place, for instance revealing a patient's privacy as a result of compromising his/her insulin
 198 pump [28]. IoT availability is essential, as it guarantees that IoT data is available and accessible to its
 199 users. Even though the Confidentiality, Integrity and Availability triad (CIA-triad) is popular, it fails
 200 to address new threats which appear in a cooperative environment, according to [29]. To tackle this
 201 issue, the authors in [29] propose a complete list of security goals called Information, Assurance, and
 202 Security octave (IAS) octave, known as the IAS octave, by studying a huge number of information in
 203 the state of the art in terms of security. Table 1 summarises the security goals suggested by the IAS
 204 octave, along with their definitions and abbreviations in link with IoT environment.

Table 1. IoT Security goals[4]

Security Requirements	Definition	Abbreviations
Confidentiality	Only authorised objects or users can get access to the data	CONF
Integrity	Data completeness and accuracy is preserved	INTG
Non-repudiation	IoT system can validate the occurrence of any event	NREP
Availability	Ensuring accessibility of an IoT system and its services	AVAL
Privacy	Presence of privacy rules or policies	PRIV
Auditability	Monitoring of the IoT object activity	AUDI
Accountability	End users can take charge of their actions	ACNT
Trustworthiness	Reliability on IoT object identity	TRST

205 In order to build the framework of security and privacy guidelines, suitable to all aspects of IoT
 206 environment life-cycle, we first propose a classification of identified IoT stakeholders into four groups,
 207 depicted in Table 2. It relates the main stakeholders with their associated role, in order to dictate the
 208 degree of guideline adaptation and stakeholder impact.

Table 2. IoT stakeholders [21]

Stakeholders	Roles	Abbreviations
Manufacturer	Building IoT hardware products	MAN
Developer	Developing IoT software solutions or services	DEV
Provider	Providing services for IoT products to customers	PRV
Consumer	Using IoT objects in different aspects of their daily lives	CNS

209 3. Attacks on IoT data at rest

210 In this section, we describe attacks and threats applicable for IoT data at rest and correlate them
 211 with IoT security goals, identified in Table 1. More specifically, we annotate with ‘▲’ when security
 212 goal in question is violated by the described attack.

213 **(AT1) Misuse of data remnants:** This type of attack takes place when IoT objects are taken in
 214 possession by other end users either by voluntarily life cycle transitions (second hand, recycling) or
 215 due to loss, theft, malfunctioning and other involuntarily causes. Even though there is possibility that
 216 data of such objects may be deleted before selling them, it is not always the case for all sold objects.
 217 Those objects which store valuable information during their entire life cycle (e.g., personal photos
 218 and passwords) could be a primary target for many attackers, since they are not discarded properly
 219 during their end-of-life period or their data is not purged completely during attempted data cleanup
 220 [30]. This attack directly violates all security goals as the attacker has full control and access to the
 221 physical storage.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲	▲	▲	▲	▲	▲	▲

222 **(AT2) Linkage attack:** The probability of unauthorised access, as well as leaks of sensitive
 223 information, grows significantly with the data sources of IoT systems linked, where each additional
 224 link creates exponential exposure. When intercepting and cross referencing between multiple data
 225 sources leads to partial data identification it is called linkage attack [31]. This attack violates CONF,
 226 INTG and PRIV security goals, as the attacker is manipulating the intercepted data without directly
 227 interfering with the actual IoT object(s). Therefore, other security goals are not applicable.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲			▲			

228 **(AT3) Data manipulation:** illegitimate modification of data at rest can be accomplished in two
 229 methods: (i) exploiting several vulnerabilities in application program interface such as SQL injection,
 230 and cross site scripting, and (ii) taking advantage of weak security mechanisms like small or weak
 231 passwords[32], [33]. Similarly to the “Misuse of data remnants” attack, all security goals are violated
 232 as the attacker operates directly on the IoT object.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲	▲	▲	▲	▲	▲	▲

233 **(AT4) Side-channel attacks:** This attack is based on the discovery of information by analysing
 234 exposed side properties of the algorithmic implementation, such as processing timing, power
 235 consumption, or even associated sounds. This type of attacks may take place due to the absence
 236 of secure methods of processing and storing IoT data, for instance storing unencrypted data either in

237 the cloud or on IoT objects. The authors in [34] have discussed several data leakage attacks on CSSs,
 238 such as a confirmation of a file and learning the content of files. In the case of confirmation of a file, an
 239 adversary who already knows the plain text content of a file, can examine if a duplicate of the file has
 240 been stored elsewhere in the CSS. In the case of learning the contents of the files, the adversary can
 241 reveal highly sensitive data, since the attacker already recognises most the of file and attempts to guess
 242 or identify the unknown segments of file by examining whether the output of the encryption meets
 243 the observed cipher text. Similarly to the “Linkage attack”, the *CONF*, *INTG* and *PRIV* security goals
 244 are violated as the attacker is indirectly revealing the private data, already generated and processed by
 245 the IoT object.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲			▲			

246 **(AT5) Denial of Service attacks:** Denial of Service (DoS) attacks in cloud computing prevent
 247 CSSs from offering their normal services or solutions for a period of time. The authors in [35] stated
 248 that DoS attacks which compromise the availability of such CSS stem from many contributing factors,
 249 the most notable of which are resource exhaustion, process disruption, physical disruption, and data
 250 corruption. For instance, the authors in [31] have stated that an attacker could flood a CSS with fake
 251 data at high frequency, which in turn makes such storage service spending most of its time validating
 252 the authenticity of such data and, therefore, is not able to timely reply to any valid requests. The
 253 inability of timely response may cause a delay which is not preferable for most IoT applications,
 254 specifically real-time applications like air traffic system and NFC Payment. For interested readers, the
 255 recently published survey of DoS attacks in the cloud can be found in [35]. First of all the *AVAL* is
 256 affected by this attack, as implied by the definition of the attack. *ACNT* is also no longer guaranteed
 257 due to the slow response times of the system. For *INTG* the guaranteed transmission and/or storage
 258 can be compromised, especially for real-time applications. *AUDI* is also violated, since the system can
 259 not perform continuous of the objects activity

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
	▲		▲		▲	▲	

260 **(AT6) Insider attack:** An insider threat takes place when either a former or a current user, who
 261 has authorised access to an object’s data or CSSs, misuses such access rights to compromise security
 262 goals of IoT, such as confidentiality, integrity, and availability. Malicious insiders can be considered
 263 as a considerable threat to many organisations. They could adversely influence a company’s mission
 264 and reputation, and therefore they could pose a major impact to any business. The cyber security
 265 intelligence index [36] in 2016 has stated that 60 percent of all attacks are derived by the insiders. From
 266 this percentage, the malicious insiders had a great contribution, since the majority of such attacks (
 267 44.5 percent) was caused by them. EY in its recently published survey in [37] has identified several
 268 types of insider threats such as fraud, infrastructure sabotage, and unauthorised trading. In the context
 269 of IoT, the whole IoT ecosystem, starting from objects located in different environments and their
 270 data and applications in the cloud, may be vulnerable to insider threats. To this end, the authors in
 271 [24] have illustrated the applicability of malicious insider attacks in all the layers of IoT multi-cloud
 272 based e-healthcare architecture composed of 4 layers. In layer 1 (physical) where several sensors are
 273 deployed to collect the health data of different patients, an insider attacker, in this case, could alter the
 274 settings to send wrong data to the healthcare companies. Not only that, the attacker could obtain and
 275 reveal the patients’ information. Likewise, in layer 2 (network) where many connectivity protocols
 276 (e.g., Bluetooth low energy) are utilised to transfer the patient data to the next layer, the malicious
 277 insider could carry out many unwanted activities like redirecting the packets to a vicious network

278 and compromising the availability of health data by initiating DoS attack on the network. In layer 3
 279 (cloud), the malicious insiders could perform a set of malicious activities like gaining unauthorised
 280 access to patients' data, altering the e-health applications, modifying data stored in the storage, and
 281 executing collision attacks. Also layer 4 (application) is susceptible to malicious insider attacks. This
 282 is because any authorised entity from lower to upper level could uncover or alter the patients' health
 283 data, which, for sure, will impact the level of trust among patients, doctors, and health organisations
 284 [24]. Similarly to the "Misuse of data remnants" attack, all security goals are violated as the attacker
 285 operates directly on the IoT object.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲	▲	▲	▲	▲	▲	▲

286 **(AT7) Homogeneity attack:** As several anonymisation-based solutions have been proposed,
 287 such as k-anonymity and t-closeness techniques, some of them lack a method in which the diversity
 288 of their sensitive attributes is not supported, making such techniques like *k*-anonymity vulnerable to
 289 homogeneity attacks. This attack is applicable on the cases where there are identical records within
 290 data sets. Similarly to the "Linkage attack", the *CONF*, *INTG* and *PRIV* security goals are violated as
 291 the attacker is indirectly capable of attributing of the private data to specific identities.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲			▲			

292 **(AT8) Unauthorised access:** IoT data is vulnerable to different attacks because of storing it either
 293 in IoT objects or remotely in the cloud with no supervision of their holders. It is also expected that
 294 the number of threats and attacks will be intensified, since attackers can get access to such data once
 295 it is not properly protected due to the absence of strong encryption techniques. Furthermore, data
 296 might be placed in several data centres located at different countries, and such countries may have
 297 a high power to access such data without permission of their holders[38,39]. Another example of
 298 unauthorised access can be found in [40]. The authors stated that an adversary may get access to
 299 IoT data illegitimately during the migration procedure of virtual machine to a untrusted host which
 300 might reveal its sensitive data. Similarly to the "Misuse of data remnants" attack, all security goals are
 301 violated as the attacker has a direct access to the data on the IoT object or in the CSS.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲	▲	▲	▲	▲	▲	▲

302 **(AT9) Identification:** An identification attack can be considered one of the most common threat
 303 against IoT data in which an attacker could link some identifier attributes (e.g., name and address)
 304 with some individuals. Similarly to the "Linkage attack", the *CONF*, *INTG* and *PRIV* security goals are
 305 violated as the attacker is indirectly capable of attributing of the private data to specific identities.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲			▲			

306 **(AT10) Hash collision:** The key goal of the collision attack is to reveal two input strings of a
 307 hash function that gives the same hash value. Due to the fact that a hash function has variable input
 308 lengths and a short fixed length output, there is a possibility that two different inputs generate the same
 309 output and this case is known as a collision[41,42]. As a consequence, an attacker can compromise

310 the encryption key and therefore intercept or have an access to the IoT object's data. Similarly to the
 311 "Linkage attack", the *CONF*, *INTG* and *PRIV* security goals are violated as the attacker is indirectly
 312 revealing the private data, already generated and processed by the IoT object.

CONF	INTG	NREP	AVAL	PRIV	AUDI	ACNT	TRST
▲	▲			▲			

313 4. Mitigation techniques for protecting data at rest

314 In this section, we analyse existing methods on IoT data protection and attribute these mitigation
 315 techniques for the attack vectors, identified in Section 3.

316 **(MT1) Deduplication schemes:** Attributed to attack *AT4* and *AT8*. Data deduplication is a
 317 method in which only a unique copy of redundant IoT data is stored, and links (not actual data) to
 318 the copies are provided. This is why such technique can be used as a backup strategy. Therefore, the
 319 development of secure deduplication schemes, capable of detecting identical data copies and storing
 320 them once, is a need and challenge at the same time. To this end, several data deduplication techniques
 321 in literature have been proposed which can be classified broadly into two categories (server-side and
 322 client-side) based on location at which data deduplication is accomplished[31].

323 It is worth noting that despite the benefits of deduplication schemes in saving disk space,
 324 minimising network bandwidth, and preventing unauthorised access, such techniques are susceptible
 325 to side-channel attacks. For instance, the authors in [34] have stated that implementing deduplication
 326 techniques in cloud storage may cause side-channel attacks like identifying files, learning the contents
 327 of files, and a covert channel. The authors also have illustrated several practical solutions like
 328 encryption and proof of ownership to mitigate such attacks. To this point, a few secure deduplication
 329 techniques have been proposed, described below:

330 In[43], the authors propose a novel deduplication algorithm in which a given file is broken into
 331 multiple segments. Each segment is encrypted by a user, and the encryption process involves both a
 332 secure hash function and a block encryption technique. These segments have an index tree, which
 333 is composed of hash values of such segments. The index tree is generated and encrypted by the
 334 user using an asymmetric algorithm. The authors claim that their approach will prevent storage
 335 providers, if it is implemented, from getting access to users' data or their decryption keys. In [44], the
 336 authors propose a new deduplication technique based on *Attribute-Based Encryption (ABE)* to encrypt
 337 data stored in the cloud and at the same time provide a secure data access to such data. According
 338 to authors' evaluation, this technique is suitable for practical deployment due to its effectiveness,
 339 scalability, and efficiency. Other research proposals associated with this topic can be found in [45,46]

340 For interested readers, the recently published survey regarding this topic can be found in [47].

341 **(MT2) Secure storage schemes:** Attributed to attack *AT3*, *AT4*, *AT8* and *AT10*. Secure storage
 342 techniques can be used to prevent IoT data breaches. To this end, several research proposals
 343 have been introduced classified broadly into two categories: (i) cryptographic-based scheme and
 344 (ii) non-cryptographic-based scheme. An example of cryptographic-based scheme can be found In[48]
 345 .The authors have proposed a secure IoT storage technique in which IoT aggregated data can be stored
 346 securely on an object based on Shamir's secret sharing method. To protect IoT data on such system,
 347 Shamir's secret sharing algorithm has been used, along with internal padding. Data in this approach,
 348 prior to storing it, is divided into many segments and each segment will be stored in different storage
 349 objects. Another example of cryptographic-based scheme is presented in [49]. The authors proposed
 350 a new technique based on an elliptic curve algorithm, which allows different users to access and
 351 store their data securely from the cloud. Moreover, it assures that neither cloud storage provider nor
 352 unauthorised users can get access the data. This approach is also capable of protecting individuals'
 353 data, even when the cloud provider is compromised due to its data encryption.

354 In [50], the authors propose a novel architecture in which individuals and companies can
355 upload and store securely their data in the cloud. Such architecture is composed of things, gateway,
356 network infrastructure, and cloud. To collect data in such architecture, IoT objects or things are
357 deployed in physical environment. The need of gateways in this architecture, which uses as an
358 intermediate layer between objects and cloud, stems from that not all IoT objects are equipped with
359 connectivity protocols (e.g., Wi-Fi) that allow them to connect to the Internet and transmit their data.
360 The administrator in such system plays a key role in defining responsibilities according to the job
361 functionality described in the organisation. For the interested readers, other research efforts associated
362 with this topic can be found [51–54].

363 An example of non-cryptographic-based scheme can be found in [55]. The authors have proposed
364 a new storage schema called POST-SHAREDS which offers long-term security for IoT data without
365 involving any encryption techniques. The security of such scheme derives from dividing data into so
366 many segments (each segment has its own pointers) and scattering them into different storage. If
367 an attacker wants to get data of one segment, he needs to get all its pointers, which are distributed in
368 several storage.

369 **(MT3) Access control:** *Attributed to attack AT6 and AT8. Attributed to attack AT3.* To control
370 access to stored IoT data by customers or companies, many research efforts have been proposed.
371 Such efforts can be divided broadly into four categories: (i) **Mandatory Access Control (MAC)**,
372 (ii) **Discretionary Access Control model (DAC)**, (iii) **Role Based Access Control model (RBAC)** and
373 (iv) **ABE**. Having integrated MAC into an IoT system, the system administrator will have privileges to
374 manage the customers' roles as well as rights. In MAC, it is also possible that the system administrator
375 manipulates access policies resulting in preventing customers to access the system. This type of access
376 method can be added to sensitive systems like military and research centres [56]. If DAC is integrated
377 into an IoT system, the customers will have rights to manipulate the access rules for any objects. This
378 approach is extremely dangerous if an attacker gains access rights over a customer account. Thus, it is
379 not wise to give one customer full rights to the IoT system.

380 If RBAC is integrated into an IoT system, customers can gain access to resources based on
381 their roles and responsibilities in the system. Several research proposals have been conducted in
382 relation to this topic [57,58,58,59]. For instance, the authors in [59] have suggested new five principles
383 known as **Abstraction, Separation, Containment, Automation and Accountability (ASCAA)** for the
384 next-generation RBAC. They claimed that such principles are applied to access control in general
385 despite the fact that they are developing specifically to RBAC. ABE provides adaptable one-to-many
386 encryption without prior information who will be accessing such information. It also draws attention
387 for fine-grained access technique over outsourced data. The identification of a customer in ABE is
388 accomplished by a set of attributes which can be used to define access policy of the customer [60].
389 Recently, several research proposals have been attempted to implement ABE in fog computing [61–63]

390 **(MT4) Recovery Strategy:** *Attributed to attack AT3 and AT5.* Despite the importance of
391 providing high availability and disaster recovery for IoT storage, a few research proposals have
392 been found in the state-of-the-art. In [10], authors have investigated the problem of uploading IoT data
393 from a set of several sensors and the creation of different replicas of this data on distributed storage
394 in the cloud. The applicability of such approach depends on the existence of several distributed data
395 centres known as mini-clouds. In [64], authors propose a new replication approach to minimise power
396 consumption, delay, and the cost of uploading a huge amount of data send by several IoT applications.
397 Each application is composed of too many small objects. To reduce time latency, the authors deployed
398 local cloud computing resources. Other research proposals related to this topic can be found in [65–67]

399 **(MT5) Anonymisation schemes:** *Attributed to attack AT2, AT7, AT8 and AT9.* Such solutions
400 can fall broadly into three categories: (i) **K-anonymity**, (ii) **L-diversity** and (iii) **(iii) T-closeness**.

401 k -anonymity is a technique in which the privacy of data holders will be preserved when they issue
402 their data, preventing threats associated with subject identification. This technique assures that the
403 information for each person cannot be identified from a set of at least $k(-1)$ individuals who belongs
404 to. The concept of k -anonymity represents data as a table composing of set of rows and columns. Each
405 row indicates the insertion of new information related to specific entity and it should not be unique
406 [68], whereas each column represents an attribute for entity. Two techniques have been used to achieve
407 k -anonymity. First is suppression in which the values of some attributes are substituted by an asterisk
408 *. Second is generalisation in which the personal values of attributes are changed by values in a wider
409 range. For instance, if the attribute *age* is used, the value of 35 can be substituted by the term < 40 .

410 In **IoT**, k -anonymity can be used for the localisation of smart objects to enhance location privacy.
411 This can solve security issue related to the need of a third entity for managing different k -anonymity
412 sets for several queries, inapplicability of using universal GPS regulates indoor, and obfuscation. In
413 [69], the authors proposed a tree based location privacy technique against multi-precision attacks
414 using a new location query technique in which multi-precision queries are fully supported. In [70],
415 the authors propose another k -anonymity technique in which data can be released based on concrete
416 generalisation.

417 L -diversity is suggested to mitigate k -anonymity weakness which is its inability to prevent
418 homogeneity attack as well as background attack. In [71], the authors propose a new and powerful
419 privacy technique known as l -diversity which can be used to prevent several attacks (e.g., homogeneity
420 attack). Moreover, they perform an experimental evaluation to show that the proposed technique is
421 practical, and it can be implemented effectively.

422 T -closeness was first coined in [72] to overcome the shortcomings of k -anonymity and l -diversity
423 associated with attribute inspiration. The authors in [72] propose that a distribution of sensitive
424 information in any set must be close or connected to their scattering in the whole database. To
425 summarise the value of such work, the authors have used different real examples as well as experiments.
426 In [73], the authors suggest a decomposition technique with $(n - 1)$ closeness, the main purpose of
427 which is to preserve privacy in case of several sensitive attributes by reducing the amount of sensitive
428 information which may be elicited from the published data in the t -closeness situation.

429 **(MT6) Transient data storage:** **Attributed to attack AT1 and AT4.** The existing research proposals
430 have been focused on managing the persistent data in **IoT** systems. In this case, data may be stored
431 even after such systems have finished their executions. Nevertheless, a handful of research works
432 have been concentrated on managing transient **IoT** data generated during systems executions. The
433 importance of transient data stems from processing data during system execution to generate new
434 visions of data, which may be stored in storage for users' needs or may be purged, and therefore it
435 may reduce threats associated with such data. In [74], the authors have proposed a new system of
436 managing transient **IoT** data in which such data can be processed, placed and managed. This system
437 is composed of several components like resource estimator, transient data characterisation, and data
438 manager. Other research proposals related to this topic can be found in [75–77].

439 **(MT7) Searchable Encryption (SE):** **Attributed to attack AT5 and AT6.** Another way to protect
440 data in **IoT** storage is to perform information retrieval on encrypted data which is known as a **SE**
441 boomed in 2000. The main idea behind such technique can be summarised as follows: An object
442 should index and encrypt its data, and then it sends its encrypted data along with an index to a server.
443 In order to search for given data, the object needs to generate a trapdoor thought which the server
444 can execute search operations directly on encrypted data, and the output will be encrypted too. This
445 field is known as a homomorphic encryption. In this regard, **Fully Homomorphic Encryption (FHE)**
446 has been proposed in 2009 by Gentry [78]. That said, the key distribution as well as user revocation in
447 a multi-user search setting is a need and a challenge at the same time. To this end, some traditional
448 technologies like broadcast encryption proposed in [79] and secret sharing suggested in [80] can be

449 used to cope with the key distribution issue. While user revocation can be solved using either trusted
450 third party proposed in [81] or semi-trusted third party suggested in [82].

451 It is worth mentioning that **ABE** proposed in [83] has been used by Sun et al. to develop the
452 **Attribute-Based Keyword Search (ABKS)** scheme to offer fine-grained search authorisation in the cloud,
453 proposed in [84,85].

454 **(MT8) Distributed data repositories:** **Attributed to attack AT5.** Several research studies have
455 been proposed related to this topic. In [86], authors have proposed a new secure storage scheme for
456 sharing data in public storage (in the cloud) known as a Shield. Both authentication and access control
457 in such approach are granted by a proxy server. A new version of Merkel Hash Tree was introduced
458 to achieve integrity check and file content update. Moreover, both key management and effective
459 permission revocation can be accomplished using a hierarchical key organisation. Another example of
460 integrating access control into secure storage can be found in [87]. The authors propose a new secure
461 storage repository called Cryptonite for sharing a huge amount of scientific data in the cloud. This
462 approach provides an easy way for its users to securely store and share their data in the cloud without
463 revealing their sensitive data not only for unauthorised users or attackers but also for the cloud storage
464 provider and system itself.

465 It is also worth mentioning that a secure version of **Hadoop Distributed File System (HDFS)** can
466 be used to achieve such objective, and research proposals associated with this topic are described
467 below:

468 In [88], the authors have proposed a secure version of **HDFS** in which two security
469 countermeasures are involved to prevent hackers getting data in the cloud. The first countermeasure
470 is a trust mechanism between name node used to manage data nodes and end user is established.
471 This type of trust mechanism requires that end user must be authenticated in order to access name
472 node. To achieve such objective, the end user first sends hash function, and then name node compares
473 hash functions, which are **Secure Hash Algorithm 2 (SHA-2)**, generated by both the end user and
474 the name node. The end use is only authorised to access system if compare result is correct. The
475 other countermeasure is random encryption methods like **Rivest–Shamir–Adleman (RSA)**, **Advanced**
476 **Encryption Standard (AES)** and **Rivest Cipher 6 (RC6)** are used on data to prevent an adversary to
477 get access such data. The encryption and decryption process is accomplished by MapReduce which
478 allows data aggregation and parallel processing of a huge amount of data.

479 Another example of secure **HDFS** which is equipped with three countermeasures can be found in
480 [89].

481 **(MT9) Introspection:** **Attributed to attack AT5 and AT6.** Another technique which can be used
482 to conserve users' sensitive information is the introspection by checking all the activities on **virtual**
483 **machines (VMs)** in which **IoT** data is stored. The main idea behind such technique is to inspect the state
484 of **Central Processing Unit (CPU)** for each **VM**, detect the malicious software on **VM**, and check **Input**
485 **& Output (IO)** for records or files. Nevertheless, users' privacy may be compromised as a consequence
486 of losing one object's integrity by malicious software.

487 **(MT10) Blockchain:** **Attributed to attack AT2, AT6 and AT8.** The use of blockchain technology
488 in **IoT** has several advantages, the most dominant of which are decentralisation, trust, and
489 non-repudiation. It is clear that previously mentioned countermeasures can be used to solve different
490 security and privacy issues (e.g., unauthorised access and data leakage) associated with **IoT** data at
491 rest. Yet there is a need of blockchain technology to address other important issues like untrusted
492 **TPAs** and data integrity across different cloud storage. A handful of research proposals have been
493 proposed to contribute to such objectives, the recently published of which can be found in [7,90–92]. In
494 [7], the authors have proposed a blockchain-based solution to provide decentralised process in which
495 data integrity for **IoT** data stored in the semi-trusted clouds is verified and checked. Furthermore, the

496 authors have illustrated the feasibility of their approach by applying a proof of concept on a personal
497 (private) blockchain system. In [90], the authors first proposed three different requirements to allow IoT
498 systems to share and store their data in such an untrustworthy environment like untrusted TPAs. Such
499 requirements are divided into three categories: (i) trusted trading, (ii) trusted privacy and (iii) trusted
500 data access. Moreover, they proposed a decentralised architecture based on blockchain technology
501 to accomplish above mentioned requirements. The authors also demonstrated the feasibility of such
502 technique by implementing a proof of concept on Ethereum blockchain. In [91], the authors have
503 suggested a data-centric approach based on blockchain technology which concentrates on sharing,
504 resilience, and auditable preservation of data. However, the authors have only presented the initial
505 design of their approach, which is blockchain-based end-to-end encrypted data storage system. A
506 secure and permanent data management is achieved as a result of using blockchain as an auditable
507 access control level to a distributed storage level. In [92], the author have proposed a blockchain-based
508 storage system called Sapphire. This system is designed specifically for data analytic in IoT. In
509 such system IoT data coming from different IoT applications like smart home, smart grid, and smart
510 city is classified into two categories, namely text data and media data. This kind of classification is
511 accomplished by a data classifier. The collected data in both formats either text or media is stored
512 into blockchain-based a large scale of storage via customer process. Each IoT object, in Sapphire,
513 is represented as an Object-based Storage Device (OSD). Sapphire links the system interface model
514 through the Put/Get application program interface. The main building block of Sapphire is an EB-scale
515 storage system which uses the hash-based mapping approach to divides the key address space into
516 OSDs. The OSDs are used as a technique to enhance load balancing and more importantly to simplify
517 the cooperative caching. The number of OSDs may be scaled up or down in size based on the number
518 of physical objects which may joint or depart the system. To investigate the issue of fault tolerance
519 caused by storage nodes failure, several data replicas are used.

520 **(MT11) Physical security:** *Attributed to attack AT8.* IoT data may be scattered in different
521 physical locations making them susceptible to physical attacks despite the the existence of
522 previously-mentioned solutions. Therefore, there is a need of physical security measures for protecting
523 IoT data at rest. This is because the above mentioned solutions can not prevent the physical damage
524 of IoT objects along with their storage as well as data centres. To date, several physical security
525 solutions can be used to protect IoT data at rest including, but not limited to, security guards, physical
526 barriers, video surveillance, and locks. It is also wise to improve the efficiency of such physical
527 security measures by integrating them with IoT technology due to the use of connected sensors and
528 actuators. Intelligent monitoring, tampering alerts, perimeter protection, and facial recognition are
529 some examples of this kind of integration.

530 **(MT12) Monitoring and auditing:** *Attributed to attack AT8.* Monitoring activities in storage of
531 IoT data in the cloud is of paramount importance to prevent data breaches [93]. To this end, several
532 research efforts have been conducted, some of the most recent of which can be described here. In [94],
533 the authors have proposed a centralised monitoring technique for cloud applications used to monitor
534 server, agents, and files along with their configurations. To overcome the limitations of a centralised
535 monitoring approach which include scalability and most importantly single point of failure, this
536 technique provides multi-level notifications, redundancy, and automatic healing. In [95], the authors
537 have proposed a scalable distributing monitoring solution for clouds. Such solution depends heavily
538 on a scattered management tree which involves a set of parameters along with their protocols for
539 data collection. Moreover, the authors have reviewed the shortcomings of current intrusion detection
540 solutions and also investigate the use of one of the emerging fields for securing virtual machines (VMs)
541 in the cloud known as the virtual machine level intrusion detection. In [96], the authors proposed
542 a novel architecture in which the virtualisation technology can be integrated into the heart of cloud
543 computing to carry out intrusion detection security utilising hypervisor performance metrics like

544 packets transmitted/receive, CPU utilisation, and read/write requests. The authors also illustrate
 545 and validate that malicious activities could be happened, even when the attackers lack the knowledge
 546 of the operating system which operates within the VMs. For the interested readers, other research
 547 proposals related to this topic can be found in [97,98].

548 **(MT13) Decommissioning:** *Attributed to attack AT1.* The process of proper decommissioning
 549 of IoT objects along with their data in the cloud is a fundamental requirement in IoT security, and
 550 its solutions can be broadly classified into two categories: (i) object-based solutions which focus
 551 on decommissioning of IoT objects and their on board data and (ii) cloud-based solutions which
 552 concentrate on destruction IoT data in the cloud storage. Despite the importance of object-based
 553 decommission techniques for addressing some security and privacy concerns like personal data
 554 breaches, there is a lack of research works conducted in the state-of-the-art in this regard. Nevertheless,
 555 Smart card alliance in [99] has suggested two choices for decommissioning. Firstly, the objects can
 556 be reset to factory default mode. In this option, all data in such objects will be deleted except the
 557 basic security parameters. These objects can come back to life later. Secondly, a blacklist technique
 558 implemented on a server will be used to prevent blocked objects to re-join a network unless their
 559 statuses on the server have been changed.

560 **(MT14) Secure data migration:** *Attributed to attack AT4 and AT8.* Despite the importance of
 561 secure data migration solutions in preventing some threats and attacks (e.g., unauthorised access and
 562 linkage attacks), a few research works in literature have been proposed. Next, we briefly discuss them.

563 In [100], the authors have proposed a secure data migration solution for migrating or transporting
 564 IoT data from a cloud storage service to another. To assure pre-migration authentication, such solution
 565 is equipped with mutual authentication composed of key splitting and sharing approaches. Having
 566 used a symmetric algorithm(RSA) to encrypt migrated data, several security goals like confidentiality,
 567 integrity and authenticity are fulfilled. Two OpenStack servers have been used to implement and
 568 validate the feasibility of such technique. In [101], the authors have proposed a simple and effective
 569 solution for migrating securely data between different cloud storage services. This technique depends
 570 heavily on the use of cryptography and steganography, and it is known as *Secure Cloud Migration*
 571 *Architecture using Cryptography and Steganography (SCMACS)*. To encrypt and decrypt migrated
 572 data in such technique, a shared key generated by a symmetric algorithm is used by both sender and
 573 receiver. The advantage of this approach stems from generating a dynamic value for the private key.
 574 The authors also develop a prototype to illustrate the feasibility of their technique based on HDFS.
 575 Other techniques for migrating data securely among different cloud storage can be found in[102–104].

576 An overview of countermeasures proposed IoT data at rest is presented in Table 3

Table 3. A summary of the mitigation techniques proposed for IoT data at rest

Implementation Techniques	Research proposal	Year	Mechanism used
	[48]	2015	Suggests a secure and scalable IoT storage technique that meets different non-functional requirements (e.g. flexibility, liability, and security)
	[50]	2016	Proposes a technique in which IoT data can be stored securely using cryptographic algorithms and several access control policies
Secure storage schemes	[51]	2018	Proposes a flexible framework to address storing IoT data securely by merging cloud computing used to store non-time-sensitive data and fog computing used to store time-sensitive data.
	[53]	2016	Provides possible techniques to address some cloud computing issues like data breaches, unavailability, and reliability

	[52]	2014	Proposes a file system based on Ciphertext-Policy Attribute-Based Encryption (CP-ABE) in which secure deletion as well as access to encrypted files in the cloud can be achieved
	[60]	2017	Suggests a hybrid method composed of data encryption with fine-grained access technique and index encryption, which is suitable for a fog computing ecosystem
	[61]	2016	Proposes security model called Chosen Ciphertext Attack (CCA) of attribute-based encryption with outsourced decryption, which is the best technique for IoT data protection in the cloud
Access control	[63]	2018	Proposes a general framework in which a fully secure leakage-resilient function encryption technique is constructed to prevent several attacks (e.g., side-channel attacks)
	[57]	2012	Discusses different solutions like encryption, authentication interface, and multi-level virtualisation, which can be used to protect user data
	[62]	2018	Suggests a novel technique to reinforce CP-ABE solutions to offer immunity against key-delegation abuse concern.
	[86]	2014	Proposes a new type of the Merkle Hash Tree to enhance effective integrity checking, secure file sharing without any alteration, and file content update.
	[105]	2011	Suggests a novel approach for a secure data repository service developed on a public cloud infrastructure to allow customers to securely share and store their sensitive data in the cloud
Distributed data storage	[88]	2015	Describes HDFS and its difficulty in preserving the security and privacy of Big data
	[89]	2015	Represents three techniques (Kerberos, Name node, and Algorithm) used to enhance HDFS security
	[10]	2017	Suggests distributed cloud storage system based on mini-clouds assumption used to replicate and upload IoT data among different data centres
	[64]	2016	Suggests the distribution of local cloud computing resources to investigate the long-term evolution architecture shortcomings within radio access network and proposes a new protocol for a memory replication
Recovery Strategy	[65]	2013	Proposes a divide-and-conquer technique which uses to replicate the content in Wireless Mesh Networks (WMN)
	[67]	2015	Suggests different local search algorithms and represents a new technique called Aurora used to apply such algorithms in HDFS with lower overhead
	[66]	2013	Suggests a lightweight and scalable approach in which object replication and placement technique in WMN is achieved
	[74]	2018	Proposes a technique for managing mass transient data in IoT applications, which consists of several components (e.g., transient data characterisation, data manager, and resource estimator)
	[75]	2014	Proposes a software architecture that facilitates the collection of sensors' data in IoT environment
Transient data storage	[77]	2015	Proposes a distributed cloud-based storage system developed specifically for IoT data
	[76]	2014	Propose a new approach known as IoT4S used to collect environmental data from different heterogeneous objects to be stored in a data storage system that supports SQL and NoSQL technologies
	[69]	2012	Suggests an anonymity-tree based technique suitable for IoT , which reinforces multi-precision queries

	[70]	2012	Proposes an enhanced fine-grained algorithm FK-anonymity which re-evaluates the generalisation scale corresponding to application domain
Anonymisation based solutions	[72]	2008	Proposes a new privacy notation known as t-closeness in which the distribution of a feature in any class should be very close to the distribution of the feature in the whole table
	[73]	2010	Proposes a privacy procedure based on information theory, the implementation of which depends on the postrandomization method.
Searchable Encryption	[79]	2011	Suggests some improvements in searchable symmetric encryption in terms of effective constructions and definitions
	[80]	2008	Addresses the notation threshold privacy preserving keyword search (TPPKS), identifies its security goals, and develop a TPPKS approach.
	[82]	2011	Proposes a multi-user searchable encryption approach which is more practical, and it has a set of benefits over the popular techniques
	[84]	2016	Proposes the first attribute-based keyword search approach equipped with an effective user revocation technique, offering fine-grained search authorization
Blockchain based solutions	[90]	2018	Proposes a decentralised architecture based on block-chain technology in which the trust and integrity of IoT data among TPAs can be achieved
	[7]	2017	Proposes a blockchain-based architecture to provide the integrity of IoT data among different TPAs
	[91]	2017	Proposes a block-chain IoT data storage system which provides a secure and permanent IoT data management
	[92]	2018	Proposes a new IoT data storage system (known as Sapphire) based on block-chain technology. IoT data coming from different IoT devices constitutes objects with methods, IDs, features, and policies
Monitoring and auditing	[97]	2014	Proposes an approach to deal with a huge amount of data to investigate for security monitoring point of views
	[98]	2015	Proposes a new public auditing technique based on Merkle Hash Tree known as MuR-DPA which integrates a new authenticated data structure(ADS)
	[96]	2014	Suggests a hypervisor-based cloud intrusion detection technique that does not need extra software installed in VMs and , at the same time, provides more benefits compared to host-based intrusion detection techniques
	[95]	2013	Proposes a scalable monitoring approach for clouds to supervise their data in a distributed manner
Deduplication schemes	[34]	2010	Suggests simple solutions which allow cross-user deduplication and, at the same time, minimize the danger of data leakage.
	[43]	2012	Proposes a secure data deduplication architecture for cloud storage
	[45]	2013	Proposes a secure and effective storage service known as ClouDedup, which enables block-level deduplication as well as data confidentiality
	[44]	2016	Proposes an approach based on attribute-based encryption used not only to deduplicate encrypted data stored in the cloud storage, but also to provide a secure data access control

	[100]	2017	Proposes a secure data migration approach to move the data from one cloud storage service to another
	[102]	2015	Proposes a new architecture which enables a secure data transportation from users to the cloud server providers
Secure data migration techniques	[104]	2018	Proposes a new framework for multi-tenant cloud migration used to fulfill data integrity and confidentiality
	[103]	2016	Proposes an inter-cloud data migration technique which provides better security goals and quicker response time for transferring large files into cloud storage service

577 5. Analysis on security and privacy guidelines for IoT data at rest

578 This section first describes our derived guidelines in link with the involved stakeholders.
579 Consecutively, the overall guideline framework is presented with the linking between guidelines,
580 mitigation techniques and attacks.

581 **(G1) Minimise data storage :** The [GDPR](#) has proposed six principles for processing of personal
582 data, among which is data minimisation. [CSSs](#), under [GDPR](#), should only store personal data required
583 to achieve their processing purposes [106]. Two benefits are associated with this type of principle. One
584 is data breach will be minimised, as unauthorised users will have access to a restricted amount of data.
585 The other benefit is that data accuracy will be improved [107]. This guideline, hence, suggests that the
586 amount of data stored either on objects or in the cloud should be minimised, and any segment of data
587 that is not needed to execute a specific task should be removed from [IoT](#) storage [108]. For example,
588 the authors in [14] have stated that raw data can be removed from storage once secondary contexts are
589 extracted and, more importantly, all data must be de-identified. **Three countermeasures, namely *MT1*,**
590 ***MT5*, and *MT8*, can be used to accomplish this guideline.**

MAN	DEV	PRV	CNS
✓	✓	✓	✓

591 *Reasoning:* This guideline is proposed based on one of the [Privacy by design](#) as well as [Security](#)
592 [by design](#) principles, which is [minimisation data](#), proposed by [Hoepman](#) in [109] and [OWASP](#) [110],
593 respectively. We do believe this guideline can be used by manufacturers, developers, and providers as
594 they are directly involved in the production, deployment, and development of [IoT](#) objects. This can be
595 happened by allowing such objects to minimize the amount of data on them by deleting any segment
596 of data that is not needed. It is also applicable to be used by customers, since in the future [IoT](#) objects
597 may be armed with dashboard settings, permitting to minimise data collection.

598 **(G2) Minimise data retention:** In [111], the authors have stated that retaining data for a long
599 time is associated with data breaches, since it gives an attacker an opportunity to try all his/her
600 hacking techniques to compromise it. A part from data breaches, privacy risks may also be increased,
601 according to [14]. This is because long retention periods may cause unauthorised secondary usage.
602 This is why the [GDPR](#) has stated that sensitive data must be stored "no longer than is necessary for
603 the purposes for which the personal data are processed"[106]. This guideline, therefore, suggests that
604 data retention on [IoT](#) objects or in the cloud should be minimised as possible. **This guideline can be**
605 **implemented by *MT6*.**

MAN	DEV	PRV	CNS
✓	✓	✓	✓

606 *Reasoning:* This guideline proposes based on a minimisation principle suggested Privacy by design
 607 (Hoepman in [109]) as well as Security by design (OWASP in [110]) frameworks. It can be utilised by
 608 manufacturers to ensure that their objects are equipped with data retention rules. For providers and
 609 developers, this guideline can be implemented to ensure that IoT applications are engineered from
 610 start to avoid keeping data longer than it is required. Customers can also benefit from this guideline
 611 by deleting unnecessary data on their objects.

612 **(G3) Distributed data storage:** To prevent a single point of failure in IoT applications and
 613 enhance their availability, this guideline suggests that IoT data should be stored in a distributed
 614 manner. However, using such guideline has trade-offs. On one hand, it improves the availability of
 615 IoT applications and also reduces some privacy risks. For instance, the authors in [14] have stated
 616 that the use of distributed data storage can be used to minimise privacy violations, since it prevents
 617 unauthorised access and secondary knowledge discovery. On the other hand, it opens doors for
 618 several attacks and threats like data leakage, as it increases the attack surface of IoT, according to [112].
 619 Therefore, this guideline should be investigated with caution, and it can be implemented by MT8.

MAN	DEV	PRV	CNS
✓	✓	✓	✗

620 *Reasoning:* This guideline is suggested based on aggregation principle proposed in privacy by
 621 design framework (Hoepman in [109]). It can be utilised by manufacturers to assure that their IoT
 622 products have capabilities to store data in distributed environments. For providers and developers, it
 623 can be utilised to assure that their IoT applications and services are designed to store data in distributed
 624 environments by supporting distributed IoT architectures.

625 **(G4) Encrypt data storage:** In [17], the ENISA has expressed the importance of encrypted IoT
 626 data at rest to minimise privacy violations in IoT. It is also worth mentioning that the Payment
 627 Card Industry Data Security Standard (PCI DSS) has forced all companies dealing with credit card
 628 information like Visa and Master Card to implement encryption techniques when storing data [113].
 629 Moreover, PCI DSS explicitly prevents the use of storage encryption as provided by operating systems.
 630 This guideline, thus, suggests that data of IoT applications should be stored in encrypted manner
 631 either on IoT objects or in the cloud. Two protection measures (MT2 and MT7) can be used to achieve
 632 such guideline.

MAN	DEV	PRV	CNS
✓	✓	✓	✓

633 *Reasoning:* This guideline is stated based on hide principle proposed in privacy by design
 634 framework (Hoepman). It can be utilised by manufacturers to assure that their IoT products have
 635 capabilities to encrypt their stored data. Both provider and developers can integrate such guideline
 636 from start into their IoT applications so that they always store their data in encrypted format.
 637 Customers can also benefit from this guideline by enabling this feature if it comes with IoT products.

638 **(G5) Prevent data leakage:** Even though if IoT data is stored in an encryption form, it is still
 639 vulnerable to side-channel attacks. This can be happen due to many reasons such as weak encryption,
 640 hardware failure, and human error[53]. Therefore, this guideline suggests that IoT stakeholders like
 641 manufacturers and providers should implement suitable data leakage prevention techniques (e.g.,
 642 SE techniques) by taking into their considerations governments rules and industry standards. Six
 643 implementation techniques (MT1, MT2, MT5, MT6, MT6, and MT12) can be used to carry out this
 644 guideline.

MAN	DEV	PRV	CNS
✓	✓	✓	✗

645 *Reasoning:* This guideline is suggested based on hide principle proposed in privacy by design
 646 framework (Hoepman). Manufacturers can integrate such guidelines into their IoT products to be
 647 more resistant to side channel attacks. It can also be utilised by providers and developers to assure
 648 that their IoT applications are engineered from ground up to prevent data leakage.

649 **(G6) Minimise data granularity:** The granularity here refers to the level of details available and
 650 to be utilised. A concrete representation of data (e.g., a full address) is known as high granularity,
 651 while a summary view of data or a high level representation of data is called low granularity (e.g., a
 652 dissemination of location), according to [114]. In this context, storing concrete data is always associated
 653 with high privacy risks compared to storing data in a abstract level, since it is composed of more data.
 654 Therefore, this guideline suggests that IoT systems should only store minimal level of data in which
 655 their functions can be maintained [14]. This guideline can be implemented by MT1 and MT7 .

MAN	DEV	PRV	CNS
✓	✓	✓	✗

656 *Reasoning:* This guideline is stated based on a minimisation principle suggested in Privacy by
 657 design (Hoepman in [109]) as well as Security by design (OWASP in [110]) frameworks. It can be
 658 implemented by manufacturers to ensure that their products always collect and store information
 659 about their customers in a high level. Both providers and developers can also integrate such guideline
 660 into their IoT applications from ground up to prevent identification attacks as they store data in a
 661 abstract level .

662 **(G7) Ensure data availability:** With the growth of IoT, data will be generated at an
 663 unprecedented rate, as billions of objects will be connected to the Internet. Since most of these
 664 objects like actuators, sensors, and thermostats do not have on-board storage, their data must be stored
 665 in cloud data centres [10].The availability of such data is a crucial requirement for so many applications
 666 to achieve their tasks. This guideline, therefore, suggests that CSSs as well as IoT objects should
 667 implement efficient techniques (e.g., recovery strategy and DoS prevention) by which the availability
 668 of their data is guaranteed in case of natural disasters or some attacks. For instance, an attacker could
 669 flood a storage server with invalid data at very high rate in such a way that the storage server wastes
 670 most of its time validating the authenticity of data and, therefore, fails to timely respond to valid
 671 network traffic [31]. This guideline can be implemented by two protection measures (MT4 and MT1).

MAN	DEV	PRV	CNS
✓	✓	✓	✗

672 *Reasoning:* This guideline can be utilised by developers, providers, and manufacture to ensure that
 673 their applications, services, and objects are always accessible by their authorized users.

674 **(G8) Location-based aggregation:** This guideline suggests IoT applications should aggregate
 675 their data based on geographical boundaries. For instance, a query would be "how many medical
 676 objects used in each city in Switzerland". The response to this question would be an aggregated value
 677 which is unique to each city". However, it is not needed to gather details about individuals medical
 678 objects, as it may lead to privacy breaches [115]. This guideline can be implemented by MT8.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

679 *Reasoning:* This guideline is suggested based on aggregation principle proposed in privacy by
 680 design framework (Hoepman in [109]). Manufacturers and providers can implement this guideline
 681 into their products and services so that they can aggregate their data based on geographical boundaries.
 682

683 **(G9) Post-inform customers:** A recently published survey by Eurobarometer has shown that 67
 684 percent of Europeans indicated that they were worried about their information and personal data that
 685 they offer online, since they lack control over it [22]. To this end, GDPR was developed to give the
 686 individuals of EU a control over their personal data. This guideline, thus, suggests that IoT applications
 687 should always inform their users before storing or sharing data related to them. For instance, GDPR
 688 states that processing the personal data of individuals in business enterprises requires an explicit
 689 opt-in or consent from them, and several kinds of data will necessitate distinct consent [106]. This
 690 guideline can be implemented by MT2.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

691 *Reasoning:* This guideline is stated based on inform principle proposed in privacy by design
 692 framework (Hoepman in [109]). This guideline can be implemented by manufacturers to assure that
 693 their objects have capabilities to inform their customers about their data. Providers also can implement
 694 this guideline into their services so that they can notify users, when they store personal data or sensitive
 695 data, and give users control over their data

696 **(G10) Chain data aggregation:** This guideline suggests to accomplish data aggregation
 697 on-the-go as data transfers from one object to another[14] so that each object will have an opportunity
 698 to respond. In this case, if there is a query (e.g., requires a count), any object can respond to it without
 699 the need of a centralised entity. This will give all objects chance to respond to such query, and, at the
 700 same time, it will improve availability of IoT [116]. This guideline can be implemented by MT8.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

701 *Reasoning:* This guideline is proposed based on aggregation principle proposed in privacy
 702 by design framework (Hoepman in [109]). Both manufacturers and providers can implement this
 703 guideline into their products and services so that they can aggregate their data while transferring data
 704 from one object to another.

705 **(G11) Time-Period data aggregation:** This guideline suggests that IoT applications should store
 706 their data over a long period of time like days and months. This guideline will minimise the granularity
 707 of IoT data which, in turn, reduces data breaches. For instance, it is wise to report power consumption
 708 of a given building in aggregate form per week rather than a daily basis [117]. This guideline can be
 709 implemented by MT8.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

710 *Reasoning:* This guideline is stated based on aggregation principle proposed in privacy by design
 711 framework (Hoepman in [109]). It can be implemented by manufacturers to assure that their products
 712 are armed with techniques in which customers can decide when they want such product to aggregate
 713 data related to them. Providers also can integrate this guideline into their services so that they can
 714 aggregate over a long period of time.

715 **(G12) Ensure authorised access to IoT data:** The importance of providing solid techniques to
 716 control access to IoT data at rest derives from two primary factors. One is an IoT object may require
 717 to communicate with other objects in order to share and access their data. Such object, therefore,
 718 must only interact with authorised objects. The other factor is that different IoT objects may store
 719 their data in the CSSs which are only logically isolated, but in reality such data may be physically
 720 kept in the same data centre [118]. Overlooking access control mechanisms to IoT data at rest may
 721 lead to harmful consequences. For instance, the authors in [6] show that a thermostat deployed in a
 722 smart home depends heavily on a smoke detector's data to turn a heating system off in case of danger.
 723 However, sharing and accessing the smoke detector's data by unauthorised objects may place the
 724 whole smart home at risk. This guideline, thus, suggests that each IoT object or CSS should be armed
 725 with authorisation techniques (e.g., a role-based technique) through which all unauthorised requests
 726 are blocked or prevented. Three implementation techniques, namely MT11, MT9, and MT3, can be
 727 used to fulfil this guideline.

MAN	DEV	PRV	CNS
✓	✓	✓	✗

728 *Reasoning:* This guideline is suggested based on principle of defense in depth proposed in security
 729 by design framework (OWASP [110]). This guideline can be implemented by manufacturers to ensure
 730 that their products are equipped with measures via which only authorised users can get access to them.
 731 Providers as well as developers also can integrate this guideline into their services and applications so
 732 that only legitimate users can gain access to their stored data.

733 **(G13) Remove or hide sensitive data:** This guideline suggests that IoT applications must first
 734 get rid of a personally identifiable information and, then, store it. It is obvious that storing data set
 735 along with its personally identifiable information will significantly increase the risk of privacy losses.
 736 For instance, the authors in [119] have illustrated that a retailer company, called Target, once received
 737 a complaint from a customer who was really disappointed, as company sent coupons for kids' clothes
 738 to his teenage daughter. Nevertheless, the Target intentionally sent such coupons to the daughter since
 739 she was pregnant at that time. This type of inference may happen as a consequence of storing data
 740 along with its personally identifiable information, which, in turn, helps such company to conduct data
 741 mining on its customers' data. This guideline can be implemented by MT5.

MAN	DEV	PRV	CNS
✓	✓	✓	✗

742 *Reasoning:* This guideline is proposed based on hide principle proposed in privacy by design
 743 framework (Hoepman in [109]). Manufacturers can integrate such guideline into their products
 744 to ensure that they have capabilities to de-identify personal data before storing it. It can also be
 745 implemented by both provider and developers to assure their applications and services are developed
 746 from ground up so that they are capable of identifying a personally identifiable information and more
 747 importantly de-identifying it before storing it.

748 **(G14) Search on encrypted data:** As several research efforts have expressed the importance of
 749 performing information retrieval on encrypted data to prevent data linkage attacks [17,18,21], this
 750 guideline suggests that IoT applications should be shielded with techniques (e.g., SE) that allow IoT
 751 applications to respond to any queries by searching on encrypted data without revealing sensitive
 752 information. This guideline can be implemented by MT7.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

753 *Reasoning:* This guideline is stated based on two principles, the first of which is hide principle
 754 proposed in privacy by design framework (Hoepman in [109]). The second principle is defense in
 755 depth suggested in Security by design framework (OWASP in [110]). Manufacturers and Providers
 756 can implement such guideline into their objects and services to assure that such objects and services
 757 have capabilities to search on encrypted data to alleviate linkage attacks.

758 **(G15) Provide Data Integrity across different platforms:** With the emergence of IoT, different
 759 IoT applications, which may store their data on several platforms, may cooperate with each other
 760 to accomplish specific tasks or services. If data integrity of a single IoT application in the cloud has
 761 been tampered with, there is a risk to deal with unsecured applications [7]. This guideline, therefore,
 762 suggests that IoT objects as well as CSSs should be equipped with techniques in which data integrity
 763 across different platforms must be checked. This guideline can be implemented by MT10.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

764 *Reasoning:* This guideline is stated based on not trust principle suggested in Security by design
 765 framework (OWASP in [110]). This guideline can be implemented by manufacturers and providers to
 766 ensure that their objects and services are capable of checking the integrity of data they deal with.

767 **(G16) Secure share data with untrusted TPA:** Secure share data among untrusted TPA is of
 768 paramount importance for two reasons. One is IoT systems will store their data on different CSSs due
 769 to their limited capabilities as well as dynamic nature of such technology. The other reason is that it
 770 is unrealistic to assume that all cloud service providers are reliable as expected. Thus, this guideline
 771 suggests that IoT systems should be armed with techniques in which such systems could store and
 772 share their data securely in different CSSs, even when some of which are not untrusted[92]. The
 773 guideline can be implemented by MT10.

MAN	DEV	PRV	CNS
✓	✗	✓	✗

774 *Reasoning:* This guideline is stated based on two principles, namely not trust and defense in depth
 775 suggested in Security by design framework (OWASP in [110]). Both manufacturers and providers can
 776 equip their objects and services with techniques through which such objects and services can share
 777 their data securely among untrusted TPA.

778 **(G17) Prevent physical access to data storage:** In [120], the authors stated that physical attacks
 779 against traditional computers have become more easily to carry out compared to logical attacks, since
 780 logical security measures have been significantly improved. Likewise, IoT will suffer from physical
 781 attacks. This is because IoT inherits most of the issues of the existing Internet and, most probably,

782 increases them due to direct association with physical objects [4,21]. Hence, physical security in IoT
 783 is crucial due to the fact that logical security measures like firewall, intrusion detection systems, and
 784 encryption can not prevent physical attacks against IoT objects as well as data centres in the cloud.
 785 This guideline, thus, suggests that a barrier around IoT objects and data centres should be placed to
 786 prevent unauthorised physical access. **This guideline can be implemented by MT11.**

MAN	DEV	PRV	CNS
✓	✗	✓	✗

787 *Reasoning:* This guideline is proposed based on control principle suggested in Privacy by design
 788 framework (Hoepman in [109]). This guideline can be utilised by manufacturers to equip their products
 789 with techniques to prevent physical tampering. Providers can also benefit from such guideline by
 790 storing data collected by their provided services in different data storage which may be located in
 791 environments, over which they have control.

792 **(G18) Take precautions in case of natural disaster:** In [121], the authors have stressed the
 793 importance of taking precautions in case of natural catastrophe. To this end, they have used three
 794 backup servers, the data of which must be stored in encrypted format. If something unusual occurs to
 795 the server, the secret key used to encrypt data will be used again to decrypt it. This guideline suggests
 796 that CSSs should have recovery strategies used to get their data back in case of unusual situations.
 797 **This guideline can be implemented by MT4.**

MAN	DEV	PRV	CNS
✓	✓	✓	✓

798 *Reasoning:* This guideline is proposed based on fail securely principle proposed in Security
 799 by design framework (OWASP in [110]). This guideline can be utilized by all stakeholders by
 800 implementing recovery techniques so that they have a copy of their data in case of natural disasters.

801 **(G19) Minimise duplicated copies:** Unlike minimising data storage which focuses on removing
 802 unnecessary segments of data not required to carry out a specific task before storing phase, this
 803 guideline concentrates on minimising duplicate data in the cloud. This kind of data replication could
 804 occupy network bandwidth and may be stored in different data storage, increasing the attack surface.
 805 Data duplication in the cloud derives from two main factors. One is HDFS generates a lot of duplicate
 806 data due to its a replication mechanism [31]. The other factor is different IoT objects may be deployed to
 807 monitor the same environment which may generate duplicated copies of IoT data [122]. This guideline,
 808 therefore, suggests that cloud-based storage services should be equipped with a technique in which
 809 only a unique copy of duplicate data is stored [47]. **This guideline can be implemented by MT1.**

MAN	DEV	PRV	CNS
✓	✗	✓	✗

810 *Reasoning:* This guideline is stated based on minimising attack surface area principle suggested
 811 in Security by design framework (OWASP in [110]). it can be utilised by both manufacturers and
 812 providers to guarantee that their products and services only have a distinct copy of duplicate data,
 813 and most importantly they store in different locations from its origin.

814 **(G20) Proper data destruction:** The secure destruction of IoT data either in IoT objects or in the
 815 cloud is a vital requirement to prevent different security and privacy issues (e.g., data leakage). In

816 one hand, the destruction of IoT objects along with their sensitive data is inevitable, since each IoT
 817 object will reach its end of life, and therefore its data must be destroyed properly [99]. In this case, this
 818 guideline suggests that each object should be equipped with a clear end-of-life technique in which
 819 such object can be disposed or destroyed without exposing its sensitive data [123]. On the other hand,
 820 the destruction of IoT data in the cloud stems from many reasons. One is due to the termination of a
 821 contract with a provider in which a secure deletion of customers' data must be accomplished [124].
 822 Another reason is because of the compliance to GDPR in which people have the right not only to
 823 access their personal data, but also to demand the destruction of their data [106]. In this case, this
 824 guideline suggests that data providers should delete and stop further use of users' personal data as
 825 they ask for their data to be forgotten.

MAN	DEV	PRV	CNS
✓	✓	✓	✓

826 *Reasoning:* This guideline is suggested based on control principle suggested in Security by design
 827 framework (OWASP in [110]). It can be utilised by manufactures to ensure that their objects have
 828 capabilities to destroy their data in a secure manner when they reach an end-of-life stage. Providers
 829 and developers can also integrate this guideline to their services and applications in order to give their
 830 users control over their data, among which the right of their data to be forgotten.

831 **(G21) Secure data migration:** Although most IoT systems transfer users' data to the cloud
 832 storage services due to their limited capabilities in terms of memory and storage space, such systems
 833 may decide to migrate or transfer their data from one cloud storage to another due to many factors
 834 (e.g., the lack of security and availability)[125]. This process of migrating or transporting data is known
 835 as data migration, and it must be carried out in a secure manner. That said, the lack of secure data
 836 migration when moving data from one CSS to another may open a door for many attacks and threats.
 837 For example, the authors in [126] have stated that if data migration is not accomplished in a systematic
 838 and proper manner, such data is susceptible to many attacks like unauthorised access. Hence, this
 839 guideline suggests that CSSs should be armed with techniques in which data migration among them
 840 should be carried out in a secure way.

MAN	DEV	PRV	CNS
✓	X	✓	X

841 *Reasoning:* This guideline is stated based on not trust services principle proposed in Security by
 842 design framework (OWASP in [110]). It can be utilised by providers when they transfer customers'
 843 data from one data storage to another. Manufacturers also can benefit from such guideline when
 844 moving their data (e.g., objects update) from one serve to another.

845 **(by G22) Manage encryption keys:** To protect sensitive data, strong encryption is of paramount
 846 importance. Although several research proposals have been proposed to encrypt IoT data, the lack
 847 of strong technique for managing the encryption keys is a common issue among them, which may
 848 lead to several vulnerabilities like unauthorised access. Moreover, the process of managing thousands
 849 of encryption keys within an IoT company is a challenge. This guideline, therefore, suggests that
 850 encryption keys must be kept on separate devices from the data they are used to encrypt. This kind of
 851 separation makes it harder for attackers to compromise data and its encryption keys at the same time
 852 [127].

MAN	DEV	PRV	CNS
✓	✗	✓	✗

853 *Reasoning:* This guideline is stated based on defense in depth principle proposed in Security
854 by design framework (OWASP in [110]). This guideline can be implemented by both providers
855 and manufacturers to ensure the protection of users' data, since they separate encryption keys from
856 encrypted data.

857 Fig. 1 summarises the relationship between our proposed guidelines for IoT data at rest, followed
858 by their suitable mitigation techniques, and associated attacks vectors.

859 6. Discussion and future work

860 A summary of the previously mentioned research efforts is presented in Table 4, along with
861 our intended objectives. It is not hard to observe many limitations while going through them. Our
862 research, therefore, is directed to overcome those shortcomings that can be categorised as follows:
863 (i) The absence of a comprehensive list of security and privacy guidelines for IoT data at rest, followed
864 by to whom these guidelines are intended for. (ii) The lack of suitable implementation techniques to
865 implement such guidelines. (iii) The necessity of attack investigations related to IoT data at rest.

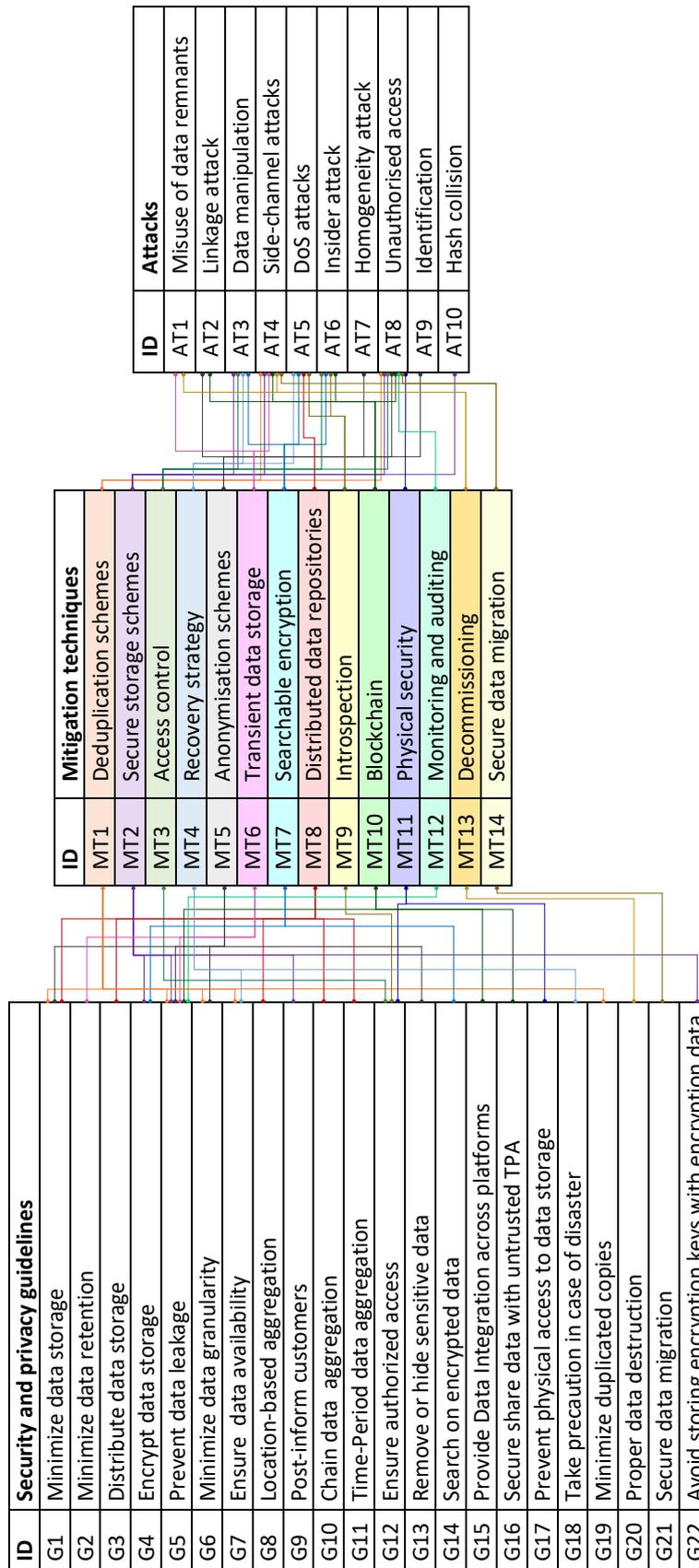


Figure 1. A summary of guidelines, stakeholders, attacks and countermeasures for IoT data at rest

Table 4. Comparison of research efforts presented in the literature.

	Addressed Features	State-of-the Art Work							
		[14]	[15]	[16]	[17]	[18]	[19]	[21]	This Work
IoT security and privacy Guidelines	Minimise data storage	✓	✓	✓	✓	✓	✓	✗	✓
	Minimise data retention	✓	✗	✓	✓	✓	✗	✗	✓
	Encrypt data communication	✓	✓	✓	✓	✓	✓	✓	✗
	Secure boot process	✗	✗	✓	✓	✓	✓	✓	✗
	Hide data routing	✗	✗	✗	✓	✓	✓	✓	✗
	reduce interference	✗	✓	✓	✓	✓	✓	✓	✗
	Distribute data storage	✓	✗	✗	✗	✗	✓	✗	✓
	Encrypt data storage	✓	✓	✓	✓	✓	✓	✗	✓
	Minimise data granularity	✓	✗	✗	✗	✗	✗	✗	✓
	Location-based aggregation	✓	✗	✗	✗	✗	✗	✗	✓
	Inform customers	✓	✓	✗	✓	✓	✓	✗	✓
	Chain data aggregation	✓	✗	✗	✗	✗	✗	✗	✓
	Time-Period data aggregation	✓	✗	✗	✗	✗	✗	✗	✓
	Ensure authorised access	✗	✓	✓	✓	✓	✓	✓	✓
	Remove or hide sensitive data	✓	✓	✓	✓	✓	✓	✓	✓
	Search on encrypted data	✓	✗	✗	✓	✗	✓	✓	✓
	Provide Data Integrity across CSPs	✗	✗	✗	✗	✗	✗	✗	✓
	Secure share data with untrusted CSPs	✗	✗	✗	✗	✗	✗	✗	✓
	Minimise duplicated copies	✗	✗	✗	✗	✗	✗	✗	✓
	Prevent physical access	✗	✗	✓	✓	✗	✓	✗	✓
Ensure data availability	✗	✗	✓	✓	✓	✓	✗	✓	
Proper data destruction	✗	✗	✗	✓	✗	✓	✗	✓	
secure data migration between CSS	✗	✗	✗	✗	✗	✗	✗	✓	
void storing enc. keys with enc. data	✗	✗	✗	✗	✗	✗	✗	✓	
Types of Guidelines	Privacy	✓	✓	✓	✓	✓	✓	✓	✓
	Security	✗	✓	✓	✓	✓	✓	✓	✓
Guidelines Intended for	Manufacturer	✗	✗	✓	✗	✓	✗	✓	✓
	Developer	✓	✓	✓	✓	✓	✓	✓	✓
	Customer	✗	✗	✓	✗	✗	✗	✓	✓
	Provider	✗	✗	✗	✗	✓	✗	✓	✓
Threats Mitigated by Guidelines		✗	✗	✗	✗	✗	✗	✓	✓
Technique to implement Guidelines		✗	✗	✗	✗	✗	✗	✓	✓

866 6.1. Recommendations for future work

867 In spite of considerable research efforts devoted to the IoT security domain, we can still suggest
868 many issues that require to be addressed.

869 6.1.1. Protection against insider threats

870 Although some of previously-mentioned protection measures like monitoring and incident
871 response can be used to mitigate malicious insiders threats, such techniques can not fully protect IoT
872 data at rest from such threats as they are not designed specifically for such purpose. Therefore, there

873 is a need to suggest a mechanism which can defend against insider threats in many IoT domains
874 (e.g., e-healthcare organisations). This can be achieved by implementing preventive, detective, and
875 reactive measures, such as user behavior analysis, policy-based frameworks, detection of anomalies
876 in data access and context-based access. Such techniques must be taken into account to avoid the
877 impacts of malicious insiders in IoT data at rest. To this end, user behavior analysis techniques have
878 been proposed by a few number of researchers [128]. There are, however, many shortcomings in all
879 the suggested approaches and therefore can not be implemented in all IoT domains like multi-cloud
880 e-healthcare environment to prevent malicious insiders, according to [24]. Hence, it is a challenge for
881 researchers to design approaches which are suitable to investigate user behavior and implement them
882 to alleviate this threat particularly in the e-healthcare environment. It is also possible that individuals
883 who are in contact patients could accidentally or intentionally alter the settings of IoT objects collecting
884 health data, which may adversely impact the reply of healthcare organisation relating to patients care.
885 It is, therefore, essential to develop an approach which could detect such a modification and at the
886 same time notify health organisations about it so that they could turn back the changes in the settings.

887 Detection of anomalies in data access has been investigated in [129]. The authors propose and
888 design a novel technique which is capable of detecting, notifying, and responding to any anomalies
889 inside Relational Database Management Systems (RDBMS). It is also capable of automatically creating
890 and maintaining profiles of normal applications as well as end users. This process of keeping track
891 of normal users and applications depend heavily on their communication with monitored RDBMS
892 during training stage. Then, it uses such profiles to identify malicious behaviour that sidetracks from
893 normality. Nevertheless, several privacy issues may stem from monitoring the behavior of end users.
894 Therefore, a lot of research is required to balance between security and privacy threats and end users'
895 privacy.

896 6.1.2. Need for Legislation

897 In the literature, a huge number of issues related to IoT data are described poorly, and they need
898 more investigation. For instance, customers would often like to know what type of data is collected
899 and stored by their smart objects before purchasing them, which is not possible at this time. Moreover,
900 customers may also want to know how their data stored either in the objects or in the cloud is protected.
901 This kind of information is not generally offered. Therefore, it is also wise to give customers chance to
902 reconfigure their privacy choices or preferences in IoT objects, in similar way as Smartphones. For
903 instance, Google, in the mobile context(e.g., Android), has designed a dashboard technique that gives
904 users more control over their personal data. The appropriate implementation of such dashboard
905 techniques will provide customers more precise ways to regulate what data they want to share
906 and when and how their personal information is gathered and used [130]. Nevertheless, further
907 investigation is required, since the the current approaches are not mature enough for standardization,
908 nor do they design specifically for IoT.

909 6.1.3. The necessity of common intercloud architecture

910 Despite the benefit of common intercloud architecture(e.g., OpenStack that offers APIs and
911 a framework for cloud systems) in which different clouds can coordinate, share and manage their
912 functionalities to provide services, it still lacks a common standards which impede its interoperability.
913 To this end, a handful research proposals have been conducted in this regard [131–133]. In [132],
914 the authors propose a model that integrate different services to ease intercloud interaction between
915 different platforms so as to display all available services. However,a lot of effort is required to create a
916 graphical user interface to offer a common management platform.

917 6.1.4. The distribution of the cloud infrastructure over the edge computing

918 A new emerging technology known as edge computing has been proposed by Cisco [134] as
919 a intermediate layer between IoT objects and cloud computing. The main objective behind such

920 technology is to scatter the cloud infrastructure over the edge layer, making it so closer to IoT objects
921 and users. Several advantages in terms of bandwidth and latency that improve service quality come as
922 a result of this closeness. It is expected that IoT may take benefit from edge cloud computing so as
923 to achieve some desired requirements like performance and security. To this point, several research
924 proposals are required to develop efficient security techniques based on edge computing technology.
925 A part from required security solutions, a few questions are still raised up and need to be addressed,
926 for instance, how to develop a trust model between IoT objects and fog nodes in such highly-scattered
927 IoT environment.

928 6.2. Limitations of the study and threats-to-validity

929 In this paper, we propose a framework of security and privacy guidelines for IoT data at rest
930 followed by to whom these guidelines are intended for, and their appropriate countermeasures.
931 However, such guidelines and their protection measures are not absolute, nor they can be guaranteed
932 the protection of IoT data at rest for three reasons. One is that new vulnerabilities are continuously
933 being disclosed, which indicates there is a necessity to monitor, review, and maintain IoT security and
934 privacy guidelines for IoT data at rest as well as best practices developed for particular environments
935 and use cases (e.g., healthcare) on a regular basis. Another reason is IoT paradigm is enabled by several
936 technologies(e.g., middleware, , sensors, and communication protocols such as a secure video steaming
937 in [135]), and, for sure, new emerging technologies related to how either to store or process IoT data at
938 rest will be introduced. New security and privacy guidelines, therefore, are of paramount importance
939 to avoid IoT data breaches. Unfortunately, such rules are breakable due to the advancements of hacking
940 tools as well as level of knowledge used by the adversaries. The other reason is that the successful
941 of our suggested framework of security and privacy guidelines for IoT data at rest depends heavily
942 on their implementation. Thus, the poor implementation of such countermeasures may lead to data
943 breaches.

944 6.3. Conclusion

945 Because of unexpected growth of connected sensors and network infrastructure, the dawn of IoT
946 in which several applications like smart car, smart building, and smart grid can interact with each
947 other to make our lives simpler and more productive is approaching. IoT is a beneficial ecosystem
948 that offers different solutions such as amazon echo; nevertheless, at the same time, risk associated
949 with data breaches can be enormous too. Therefore, in this paper we conduct an in-depth analysis
950 on IoT data at rest to identify its possible attacks and alleviate its associated risks. To this end, we
951 propose a framework of security and privacy guidelines for IoT data at rest which can be used by IoT
952 stakeholders who may utilise such guidelines to build secure IoT systems from the start, and, therefore,
953 enhance security and privacy by design. This framework also shows the link between guidelines,
954 mitigation techniques, and attacks. More importantly, we discuss our derived guidelines in link with
955 involved stakeholders, and we also give “reasoning” under which each guideline is stated based on
956 one or two principles of either security by design or privacy by design frameworks. Furthermore, we
957 briefly discuss several limitations of our framework, an example of which is poor implementation of
958 protection measures. Finally, we suggest some open challenges need further investigation.

959 In the future, we will propose a step-by-step methodology of how our suggested guidelines can be
960 implemented by developers from the early stages of their IoT systems so that the poor implementation
961 of such guidelines could be mitigated.

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967 information contained therein.

968 **Abbreviations**

ABE	Attribute-Based Encryption	
ABKS	Attribute-Based Keyword Search	
AES	Advanced Encryption Standard	
ASCAA	Abstraction, Separation, Containment, Automation and Accountability	
CCA	Chosen Ciphertext Attack	
CIA-triad	Confidentiality, Integrity and Availability triad	
CP-ABE	Ciphertext-Policy Attribute-Based Encryption	
CPU	Central Processing Unit	
CSS	Cloud Storage Service	
DAC	Discretionary Access Control model	
DoS	Denial of Service	
DPD	Data Protection Directive	
ENISA	European Union Agency for Network and Information Security	
EU	European Union	
FHE	Fully Homomorphic Encryption	
969	GDPR	General Data Protection Regulation
HDFS	Hadoop Distributed File System	
HIPAA	Health Insurance Portability	
IAS	Information, Assurance, and Security octave	
ICC	Industrial Internet Consortium	
IIoT	Industrial Internet of Things	
IO	Input & Output	
IoT	Internet of Things	
IoTSF	IoT Security Foundation	
MAC	Mandatory Access Control	
OSD	Object-based Storage Device	
OWASP	Open Web Application Security	
PCI DSS	Payment Card Industry Data Security Standard	
PHI	Personal Health information	
RBAC	Role Based Access Control model	

RC6	Rivest Cipher 6
RDBMS	Relational Database Management Systems
RSA	Rivest–Shamir–Adleman
SCMACS	Secure Cloud Migration Architecture using Cryptography and Steganography
970 SE	Searchable Encryption
SHA-2	Secure Hash Algorithm 2
TPA	Third-Part Auditor
VM	virtual machine

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1288 **Sample Availability:** Samples of the compounds are available from the authors.

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