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## On the interplay of Internet of Things and Cloud Computing: A systematic mapping study

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## ABSTRACT

The Internet of Things (IoT) is a novel paradigm relying on the interaction of smart objects (things) among each other and with physical and/or virtual resources through the Internet. Despite the recent advances that have made IoT a reality, there are several challenges to be addressed towards exploiting its full potential and promoting tangible benefits to society, environment, economy, and individual citizens. Recently, Cloud Computing has been advocated as a promising approach to tackle some of the existing challenges in IoT while leveraging its adoption and bringing new opportunities. With the combination of IoT and Cloud Computing, the cloud becomes an intermediate layer between smart objects and applications that make use of data and resources provided by these objects. On the one hand, IoT can benefit from the almost unlimited resources of Cloud Computing to implement management and composition of services related to smart objects and their provided data. On the other hand, the cloud can benefit from IoT by broadening its operation scope to deal with real-world objects. In spite of this synergy, the literature still lacks of a broad, comprehensive overview on what has been investigated on the integration of IoT and Cloud Computing and what are the open issues to be addressed in future research and development. The goal of this work is to fill this gap by systematically collecting and analyzing studies available in the literature aiming to: (i) obtain a comprehensive understanding on the integration of IoT and Cloud Computing paradigms; (ii) provide an overview of the current state of research on this topic; and (iii) identify important gaps in the existing approaches as well as promising research directions. To achieve this goal, a systematic mapping study was performed covering papers recently published in journals, conferences, and workshops, available at five relevant electronic databases. As a result, 35 studies were selected presenting strategies and solutions on how to integrate IoT and Cloud Computing as well scenarios, research challenges, and opportunities in this context. Besides confirming the increasing interest on the integration of IoT and Cloud Computing, this paper reports the main outcomes of the performed systematic mapping by both presenting an overview of the state of the art on the investigated topic and shedding light on important challenges and potential directions to future research.

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### 1. Introduction

The *Internet of Things* (IoT) has rapidly evolved in the last years as an umbrella term envisioning that every single object on Earth

(the so-called *things*) can be identified, controlled, and monitored through the Internet. These heterogeneous smart objects are typically endowed with sensing and/or actuation capabilities, thus being able to capture physical phenomena and translate them into data streams (thereby providing information about the environment where they are inserted into), as well as to affect the physical realm as a response to various stimuli. Furthermore, they can seamlessly collaborate with other physical and/or virtual resources also available in the Internet to provide value-added information and functionalities for end-users and applications with minimum human intervention.

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The dissemination of the IoT paradigm can produce a significant impact of the daily lives of human beings with the emergence of new applications and systems in several real-world domains. From the point of view of a user, IoT will play a leading role in application scenarios such as domotics, smart homes, healthcare, and enhanced learning. From the perspective of business users, the most apparent effects will be similarly visible in fields such as logistics, industry, energy, agriculture, and retail. Ultimately, IoT can foster the development of wide-scope applications in the contexts of smart cities, environmental monitoring, etc. [1–3].

Despite the advances in terms of information and communication technologies (ICTs) for making IoT a reality, this paradigm has a variety of open issues that require further research and development efforts, as discussed in several recent studies [1,4–6]. One of these challenges refers to the huge amount of physical devices interacting over the network. Indeed, previsions point out that billions of connected things will be in use by 2020 and hence IoT infrastructures need to be scalable enough to handle such a numerous amount of heterogeneous devices within the environment. An immediate consequence is the volume of data provided by these devices and transmitted through the network, which will also increase as well surely reaching an unprecedented level. In this context, challenges arise in terms of collecting, analyzing, managing, and storing such a large, non-structured, diverse volume of data.

Recently, the *Cloud Computing* paradigm has been advocated as a promising solution to meet some of the requirements of IoT. Cloud Computing can be defined as a model that allows accessing a set of shared and configurable computing resources (e.g., networks, servers, storage facilities, applications) offered as services [7]. These resources can be rapidly provisioned and released with minimum management effort or interaction with the service provider and are offered on-demand, so that users pay only the amount of the effective use of a resource, i.e., in a pay-per-use model. In addition, Cloud Computing promises reduced upfront investment, high availability, fault-tolerance, virtually infinite scalability, etc., characteristics that have notably attracted the attention from both academia and industry [8]. These features are appealing to IoT as they will allow any device to be a simple terminal without the need of possessing large computational resources since cloud services can be transparently, pervasively accessed via Internet. Envisioning the potential benefits brought by Cloud Computing to IoT, major cloud vendors such as Amazon and Google have recently started offering cloud services aiming to support IoT devices and applications in terms of computing capabilities, data analytics, resource elasticity, and scalability [9,10].

The alignment of IoT and Cloud Computing can take place through two different ways, namely (i) *bringing the cloud to the things* and (ii) *bringing the things to the cloud*. *Bringing the cloud to the things* refers to take advantage of the main features offered by cloud services to compensate technological constraints of IoT in terms of storage, processing, and energy, as well as to enhance the capabilities of IoT infrastructures and applications. As an example, most of the things in IoT are devices with low processing power and storage due to their limited energetic capabilities, a contrast to the several complex processing tasks that need to be performed. To overcome this limitation, such devices could play the role of simple data providers and send data to be processed and stored directly on the cloud, i.e., externally to the device itself. On the other side of the spectrum, *bringing the things to the cloud* stands for leveraging IoT-based capabilities and offering them as pay-per-use services on the cloud, leading to what has been sometimes referred to as *Cloud of Things* (CoT) [11]. This relies on the notion of Everything as a Service (XaaS or \*aaS), a flexible, customizable model that envisions offering not only computing resources, but also anything that can be consumed as utility cloud services [12]. An example of this approach is the

concept of *Sensing as a Service*, a model to virtualize, share, and reuse sensing devices and their respective data to be ubiquitously consumed from the cloud [13,14]. In this perspective, the cloud becomes an intermediate layer between the smart things and users/applications, thus hiding the inherent complexity of the former aiming to foster the development of the latter [15].

Given the increasing interest and the clear synergy on the integration of IoT and Cloud Computing, a comprehensive overview of the state of the art on this topic is quite relevant. Such an overview can enable both researchers and practitioners to critically reflect on the current state of the art and to identify important challenges requiring attention in future research and development. Nevertheless, to the best of our knowledge, there is currently no systematic study providing a broad panorama about what has been investigated so far and what are the open issues in this context. In the literature, a few studies have previously strived to summarize the existing work and some possible challenges on the integration of IoT and Cloud Computing [15–17]. However, they offer only a partial panorama of this relationship and they were performed in an ad-hoc way, thus not following a systematic, well-defined procedure. Furthermore, despite the combination of IoT and Cloud Computing can overcome some of the open issues to be individually solved in both sides, new, additional challenges arise from such a relationship and/or the existing ones can become more critical [16,18]. Therefore, the sought overview can play an important role in terms of providing a comprehensive agenda for future research towards tackling these challenges.

Aiming at addressing these issues, we have performed a *systematic mapping study* (or shortly *systematic mapping*) on the integration of the IoT and Cloud Computing paradigms. A systematic mapping is a form of secondary study performed to obtain a comprehensive overview of a given research topic, identify research gaps, and collect evidences to commission future research. In addition, it allows primary studies in a domain to be plotted at a high level of granularity, thus answering broad research questions regarding the current state of the research on a topic [19]. A systematic mapping follows a rigorous, methodological procedure that seeks to minimize bias while allowing other researchers to reproduce the same process when exploring the same research topic. To achieve this purpose, a well-defined protocol is established with research questions and explicit criteria to evaluate and select the primary studies. This protocol must be strictly followed throughout the whole process in order to provide scientific value to the obtained findings.

The goal of this systematic mapping is threefold: (i) to obtain a comprehensive understanding on the integration of the IoT and Cloud Computing paradigms; (ii) to provide an overview of the current state of research on this topic; and (iii) to identify important gaps in the existing approaches as well as promising research directions. To achieve this goal, studies published in recent journals, conferences, and workshops available at five relevant electronic databases were collected and analyzed. As a result, 35 primary studies presenting strategies and solutions on how to integrate IoT and Cloud Computing as well scenarios, research challenges, and opportunities in this context were selected. This paper reports the main outcomes of the performed systematic mapping by both (i) presenting an overview of the state of the art on the integration of IoT and Cloud Computing and (ii) shedding light on important current challenges and potential directions to future research.

The remainder of this paper is structured as follows. **Section 2** presents the systematic mapping methodology, research questions, search strategy, selection criteria, and data extraction method. **Section 3** details how the study was conducted to select relevant primary studies from the literature. **Section 4** provides a synthesis of the extracted data as well as answers to the research questions. **Section 5** elicits some challenges that can drive further

**Table 1**  
Research questions and respective goals.

| Research question  | Goal  |
|--|---|
| RQ1: What are the main topics investigated on the integration of the IoT and Cloud Computing paradigms?        | To obtain a comprehensive understanding on the integration IoT and Cloud Computing and to identify what has been investigated in this context |
| RQ2: What are the strategies for integrating IoT and Cloud Computing?  | To understand how and to which extent IoT and Cloud Computing have been effectively integrated  |
| RQ3: What are the existing architectures supporting the construction and execution of cloud-based IoT systems? | To characterize the architectures and infrastructures comprising the integration of IoT and Cloud Computing                                   |

**Table 2**  
Electronic databases used in the automated search process.

| Database            | URL   |
|---------------------|---|
| IEEEXplore          | <a href="http://ieeexplore.ieee.org">http://ieeexplore.ieee.org</a>       |
| ACM digital library | <a href="http://dl.acm.org">http://dl.acm.org</a>                         |
| ScienceDirect.com   | <a href="http://www.sciencedirect.com">http://www.sciencedirect.com</a>   |
| Scopus              | <a href="http://www.scopus.com">http://www.scopus.com</a>                 |
| Web of science      | <a href="http://www.webofknowledge.com">http://www.webofknowledge.com</a> |

research in this context. Section 6 enumerates threats to the validity of this systematic mapping. Finally, Section 7 contains some concluding remarks.

## 2. Research methodology

A *systematic mapping* is an evidence-based secondary study recently advocated as a useful mean to synthesize existing work from the literature by using a systematic, well-defined procedure. This type of study offers multiple benefits. First, systematic mappings are able to provide a comprehensive overview of the state of the art (or state of the practice) on the investigated research topic [20]. Second, they can identify relevant gaps in the literature and collect evidences to commission further research, thus avoiding effort duplication [21]. In this perspective, systematic mappings allow analyzing all available studies in a given domain at a high level thereby answering broad research questions regarding the current state of the art of the literature [19].

A systematic mapping typically comprises three basic steps, as depicted in Fig. 1. The *planning* step yields a protocol defining the research questions to be answered, the search strategy to be adopted, the criteria to be used for selecting primary studies, and the methods for extracting and synthesizing data. In the *conduction* (or *execution*) step, primary studies are identified, selected, and evaluated according to the previously established protocol. Finally, the *reporting* (or *analysis*) step aggregates information extracted from the relevant primary studies considering the research questions and outlines conclusions from them.

### 2.1. Research questions

Aiming at finding primary studies to understand and summarize evidences about the synergic relationship between IoT and Cloud Computing, we have proposed the research questions (RQs) outlined in Table 1.

### 2.2. Search strategy

In order to retrieve primary studies, we have used an automated search process performed over five electronic databases (see Table 2), which are among the most popular ones in Computer Science and Engineering and are able to ensure a high coverage

of potentially relevant studies [22,23]. In addition, we have taken into account other important criteria, such as: (i) coverage of the electronic database; (ii) content update, i.e., if the publications are regularly updated; (iii) availability of the full text of the primary study; (iv) easiness of building the search through fields and commands available at the electronic database; (v) quality of the results, which is related to the accuracy of the results obtained by the automated search procedure; and (vi) versatility to export results [24].

Based on the defined research questions, two main keywords were initially identified, namely *Internet of Things* and *Cloud Computing*. In addition, possible variations such as synonyms and singular/plural forms were considering, thus resulting in the following search string:

```
((internet of things OR iot OR web of things OR wot OR machine to machine OR m2m OR machine-to-machine) AND cloud computing) OR (cloud of things OR cloud-of-things)
```

in which the main keywords were connected by using the AND logical operator. In turn, the possible variations and synonyms were connected by using the OR logical operator.

### 2.3. Selection criteria

Selection criteria were used to evaluate each retrieved primary study according to the defined research questions. The main goal was to include studies that are potentially relevant to answer the research questions and to exclude the ones that do not contribute to answer them.

We have considered the following two inclusion criteria:

- IC1: The study presents and/or discusses scenarios, research challenges, and opportunities on the integration of the IoT and Cloud Computing paradigms.
- IC2: The study presents a strategy on how to integrate IoT and Cloud Computing (at either the architectural, platform or programming level, for example).

We have also established the following seven exclusion criteria:

- EC1: The study is not directly related to IoT.
- EC2: The study is not directly related to Cloud Computing.
- EC3: The study does not address the integration of the IoT and Cloud Computing paradigms.
- EC4: The study is a previous version of a more complete study about the same research.
- EC5: The study does not have an abstract or the full text is not available.
- EC6: The study is a table of contents, foreword, tutorial, editorial, keynote talk, or summary of conference/workshop.
- EC7: The study is not written in English, which is the most common language in scientific papers.

In this systematic mapping, a given primary study is considered as relevant if it does not meet any of the aforementioned exclusion criteria and it meets at least one inclusion criterion.

## 3. Selection process

This systematic mapping was undertaken from March to June 2015 and involved four researchers. The primary studies were searched, selected, and evaluated according to the established protocol, resulting in a set of possibly relevant studies. During the search process, the generic search string has undergone minor changes in order to make it compatible with the specificities of each electronic database engine. Afterwards, the automated search procedure was performed over the selected electronic databases by searching for all studies that have matched the adapted search

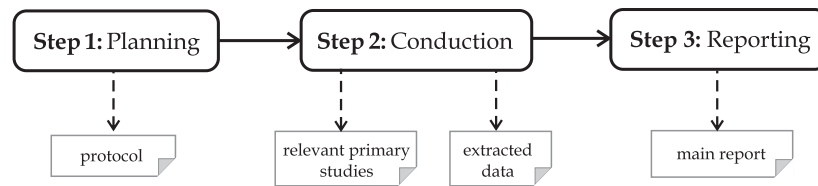


Fig. 1. Process for reviewing literature in systematic mappings [19].

243 string. The automated search was limited only to title, abstract, and  
244 keyword fields.

245 The selection of the studies was divided in two main phases.  
246 The preliminary selection encompassed reading title, abstract, and  
247 keywords of the studies retrieved from the electronic databases,  
248 whereas the final selection encompassed the full reading of the fil-  
249 tered studies. In order to minimize the effect of any bias or misin-  
250 terpretation, four researchers have individually performed the se-  
251 lection activities. After retrieving the studies from the electronic  
252 databases, studies indexed by more than one database were re-  
253 moved and then the researchers have performed the first selection  
254 to filter studies based on the selection criteria (see Section 2.3)  
255 against the information available in title, abstract, and keywords.  
256 Next, an agreement meeting was held in order to compare the re-  
257 sults and solve existing conflicts, thus resulting in a consensual  
258 preliminary selection. Afterwards, the full text of the filtered stud-  
259 ies was read and the selection criteria were applied in order to  
260 compose the final set of relevant studies. A new agreement meet-  
261 ing was carried out again in order to compare results and solve  
262 conflicts.

263 Fig. 2 depicts the steps for selecting the relevant primary stud-  
264 ies. After removing studies retrieved by two or more electronic  
265 databases, 842 studies were evaluated based on their title, abstract,  
266 and keywords against the selection criteria, resulting in a set of  
267 105 potentially relevant studies. The full text of the filtered stud-  
268 ies was read and the selection criteria were applied again. As a re-  
269 sult, 35 primary studies were selected as relevant to this system-  
270 atic mapping (see Appendix A).

## 271 4. Results

272 In this section, we summarize the results of the performed  
273 systematic mapping considering the research questions and the  
274 extracted/synthesized data. We first present a brief overview of  
275 the selected primary studies (Section 4.1) and then the answers  
276 to each research question based on the analysis of such studies  
277 (Sections 4.2–4.4).

### 278 4.1. Overview of selected primary studies

279 *Distribution along the years.* As both IoT and Cloud Computing  
280 represent relatively new research fields, it is interesting to observe  
281 when studies on the convergence of these paradigms were pub-  
282 lished. In particular, this useful to identify when such areas have  
283 begun to converge and when the research on this convergence has  
284 gained momentum. Fig. 3 shows the evolution of the selected pri-  
285 mary studies along the years. It is possible to observe that almost  
286 all studies were published in the last three years (no study was  
287 published before 2012), thus confirming that the integration of IoT  
288 and Cloud Computing has been target of a very recent interest  
289 from the scientific community. We have not plotted data regard-  
290 ing year 2015 due to its incompleteness: as the automated search  
291 procedure was performed in March 2015, only one study had been  
292 published so far. Furthermore, it is possible to notice the clear in-  
293 creasing trend on the number of publications along the years, de-

Table 3

Distribution of application domains targeted by the se-  
lected studies.

| Application domains           | Studies            |
|-------------------------------|--------------------|
| Healthcare                    | S11, S12, S15, S20 |
| Smart cities                  | S13, S20           |
| Ambient Assisted Living (AAL) | S8                 |
| Smart homes                   | S29                |
| Mobile applications           | S10, S27           |
| Intelligent business services | S33                |
| Supply chain management       | S34                |

noting that more relevant publications on this topic may be ex-  
pected for the next years.

294  
295  
296 *Publication venues.* We have also observed the venues where  
297 the selected primary studies were published. As shown in Fig. 4,  
298 29 studies (83%) were published in conferences and workshops,  
299 a possible indicative that they are at an early development stage  
300 and/or not fully validated. In addition, the selected studies were  
301 published in 29 different venues (see Appendix B), thus indicating  
302 that there is still no leading conference/journal addressing the in-  
303 tegration of IoT and Cloud Computing possibly due to the novelty  
304 of this topic in the scientific community. Another consequence of  
305 the lack of such a leading venue is that some studies were pub-  
306 lished in venues addressing either IoT or Cloud Computing or even  
307 correlated topics of interest.

308 *Validation/evaluation methods.* After examining the 29 studies  
309 that present a concrete approach on the integration of IoT and  
310 Cloud Computing, we have noticed that a quarter of these stud-  
311 ies did not present any method for validating or evaluating their  
312 proposal (see Fig. 5). In addition, more than half of these stud-  
313 ies are limited to either present an example illustrating a poss-  
314 ible scenario in which the proposed approach could be applied  
315 (4 of 29 studies, i.e., 13.79%) or show a simple proof of concept  
316 (prototype) aimed to demonstrate the implementation and/or ap-  
317 plication of the approach (12 of 29 studies, i.e., 41.38%). There-  
318 fore, it is possible to clearly conclude that stronger methods are  
319 required to validate/evaluate the proposed approaches aiming at  
320 providing solid evidences about their efficiency, effectiveness, and  
321 feasibility.

322 *Application domains.* As shown in Table 3, 12 studies (34.28%)  
323 present proposals targeting seven specific application domains,  
324 which represent typical scenarios that can benefit from the adop-  
325 tion of IoT and can also potentially benefit from the integration  
326 of this paradigm with Cloud Computing. In general, these studies  
327 take advantage of cloud services to: (i) provide scalability upon  
328 a myriad of heterogeneous physical devices; (ii) increase and  
329 decrease the use of infrastructure resources (e.g., CPU, storage, ap-  
330 plication services, etc.) according to the demands of applications,  
331 thus taking advantage of the elasticity of cloud; and (iii) allow  
332 ubiquitous access to data available on the cloud. Additionally, the  
333 cloud is used as a promising solution for challenges faced by some  
334 specific application domains. For example, healthcare and smart  
335 city applications are typically characterized by large amounts of  
336 data produced by heterogeneous sensors and that need to undergo

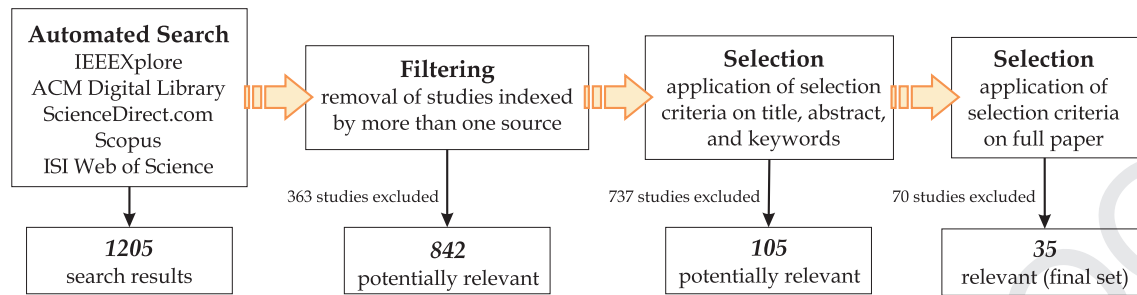


Fig. 2. Steps for selecting the relevant primary studies.

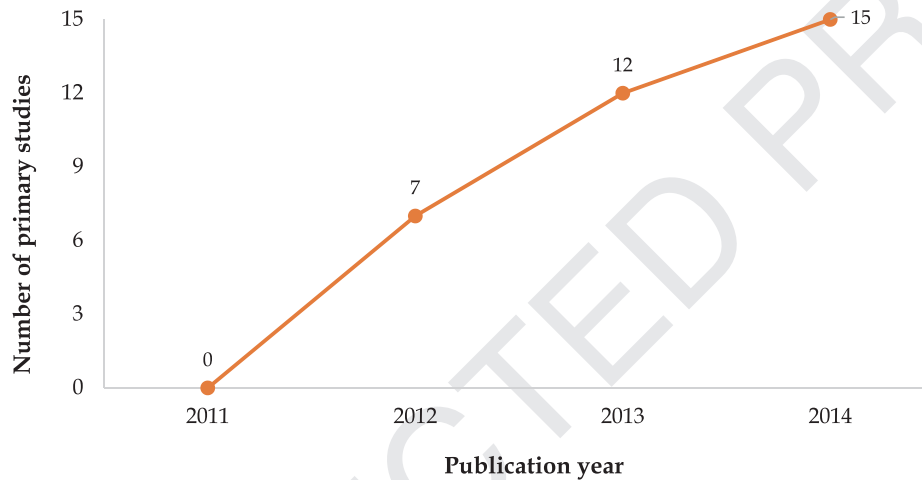


Fig. 3. Evolution of the selected primary studies along the years.

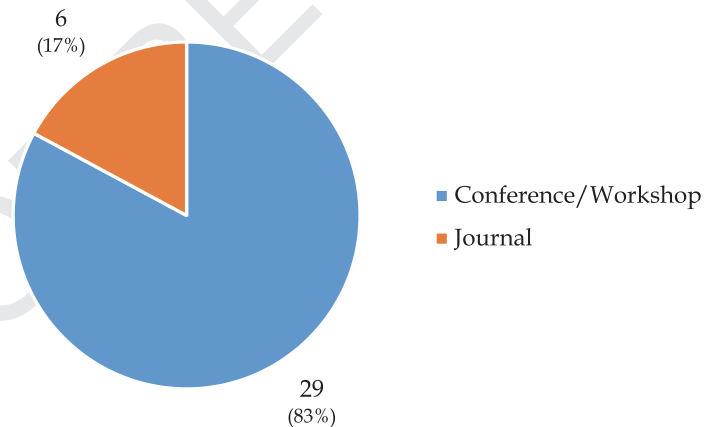


Fig. 4. Venue types in which the selected primary studies were published.

337 complex processing and analysis tasks. Healthcare applications  
 338 have important requirements to be fulfilled, such as high availabil-  
 339 ity and the need of performing tasks at real-time, under penalty  
 340 of threats to life, health, and safety of human beings. Smart city  
 341 applications rely on ICTs for collecting data from several urban  
 342 sensors to be used in management, education, transportation, and  
 343 public safety services in a city in order to provide value-added ser-  
 344 vices to the citizens and government entities. Therefore, processing  
 345 such a volume of data and providing efficient means of sharing  
 346 them are relevant concerns that can exploit both cloud and IoT  
 347 capabilities. Moreover, the adoption of Cloud Computing can play  
 348 a significant role in these and other scenarios towards meeting the  
 349 aforementioned requirements and being a distributed stratum for  
 350 management, analysis, dissemination, and control of data from IoT  
 351 devices.

#### 4.2. RQ1 – main investigated topics

352  
 353 Aiming to obtain a comprehensive understanding on the inte-  
 354 gration of IoT and Cloud Computing and to identify what has been  
 355 investigated in this context, we have analyzed the primary goals  
 356 of each study. 29 of the 35 selected studies propose concrete so-  
 357 lutions that can be broadly classified into four categories, namely:  
 358 (i) *architecture*, which refers to a high-level structure targeting the  
 359 integration of IoT and Cloud Computing and defining components,  
 360 their respective roles, and how they should interact for concretiz-  
 361 ing such an integration; (ii) *platform*, i.e., a hardware and/or soft-  
 362 ware infrastructure providing APIs to support the development and  
 363 execution of applications, as well as the integration, management,  
 364 and real-time monitoring of IoT devices; (iii) *framework*, i.e., a soft-  
 365 ware infrastructure providing reusable components to foster the

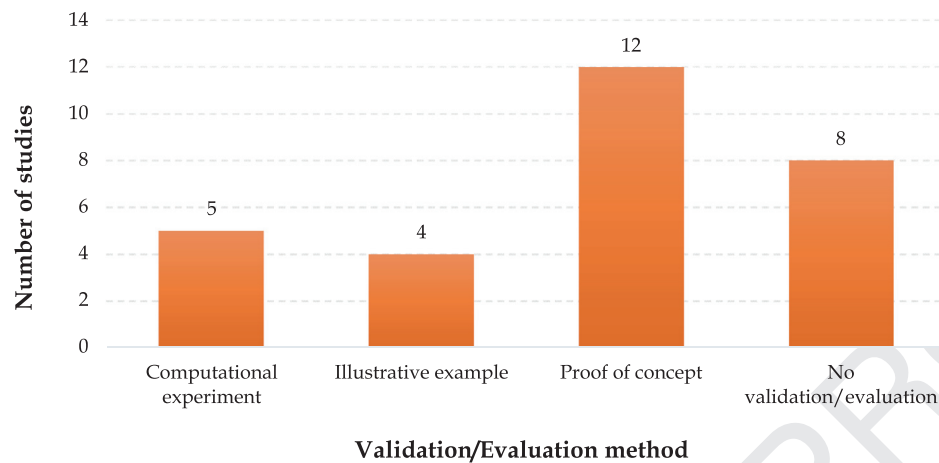


Fig. 5. Methods for validating/evaluating proposed approaches.

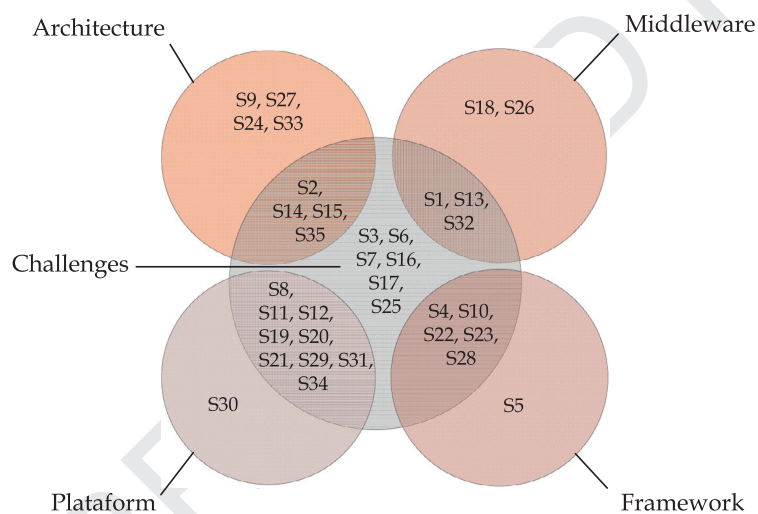


Fig. 6. Main topics addressed in the selected primary studies.

366 development of applications; and (iv) *middleware* providing both  
 367 generic and domain-specific services to applications and/or end-  
 368 users while abstracting away underlying IoT devices and cloud re-  
 369 sources. In addition, it is worth observing that 27 of 35 selected  
 370 studies present and/or discuss challenges related to the integra-  
 371 tion of IoT and Cloud Computing. Fig. 6 depicts a separation of the  
 372 studies according to these topics, each one briefly described in the  
 373 following.

374 *Architectures.* Eight studies have proposed high-level architec-  
 375 tures for the integration between IoT and Cloud Computing, in par-  
 376 ticular focusing on sharing data and providing monitoring and ac-  
 377 tuation capabilities by using the cloud infrastructure. In spite of  
 378 the existing differences among these architectures and their re-  
 379 spective purposes, they present some common elements such as:  
 380 (i) sensors responsible for gathering information from people and  
 381 the environment (e.g., health data, location, luminosity, etc.) to  
 382 be made available on the cloud; (ii) software components run-  
 383 ning over the cloud to perform tasks such as processing and mak-  
 384 ing information obtained from IoT devices; and (iii) components  
 385 within a network layer through which data are transmitted to the  
 386 software components on the cloud. In general, the components of  
 387 the proposed architectures are organized into layers according to  
 388 their features, e.g., (i) sensors, RFID tags, and other devices res-  
 389 ponsible for capturing and providing information about the physical  
 390 world; (ii) network devices responsible for mediating the commu-  
 391 nication between sensors and the cloud; and (iii) software compo-

392 nents responsible for consuming information provided by sensors. 392  
 393 The communication between devices and these software compo- 393  
 394 nents that consume information takes place either through drivers 394  
 395 tailored to each device or via smart gateways mediating such a 395  
 396 communication. In both cases, we found the use of the REST (*REp-* 396  
 397 *resentational State Transfer*) architectural style [25,26] as the pre- 397  
 398 dominant choice for transmitting data to the cloud. As an exam- 398  
 399 ple, the communication between devices and the cloud platform 399  
 400 in study S2 is performed through a smart gateway responsible 400  
 401 for receiving data from devices, temporarily storing them to per- 401  
 402 form preprocessing tasks on such data, and finally transmitting 402  
 403 the received data to the cloud. In study S33, drivers installed in 403  
 404 a gateway allow devices to communicate with it, which in turn 404  
 405 forwards information received from devices to the cloud via REST. 405  
 406 Additionally, the proposed architectures were designed to bene- 406  
 407 fit from high performance, availability, and data storage capabili- 407  
 408 ties offered by cloud platforms in order to support a large num- 408  
 409 ber of devices and users. In study S15, which targets the health- 409  
 410 care domain, vital signs, motion, and contextual information are 410  
 411 collected from wearable sensors and sent via Bluetooth to mobile 411  
 412 devices (e.g., smartphones), which play the role of a gateway and 412  
 413 forward such data via Internet to a server on the cloud. Before 413  
 414 sending data to be stored and analyzed on the cloud, these mo- 414  
 415 bile devices can perform tasks such as data preprocessing, priori- 415  
 416 tization of data according to their criticality, temporary storage, and 416  
 417 aggregation. 417

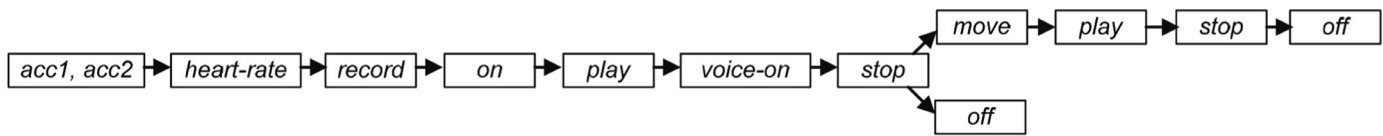


Fig. 7. Example of device orchestration performed in the platform proposed by study S8.

418 *Platforms.* As shown in Fig. 6, 10 studies have proposed plat- 476  
 419 forms targeting the integration of the IoT and Cloud Computing 477  
 420 paradigms. In particular, the main purpose of these platforms is 478  
 421 to allow integrating, managing, and monitoring IoT devices at real- 479  
 422 time through browsers or mobile devices. In these platforms, the 480  
 423 amount of cloud resources required to support the integrated de- 481  
 424 vices is provided on-demand. They also aim to support the devel- 482  
 425 opment of applications using IoT devices by providing APIs for data 483  
 426 storage, data retrieval and analysis, and deployment and execution 484  
 427 of applications. In addition, there are other relevant features of- 485  
 428 fered by some platforms in the selected studies, such as support 486  
 429 for the orchestration of devices (study S8) and sharing of virtual- 487  
 430 ized IoT devices (study S34). In study S8, device orchestration is 488  
 431 performed with the assistance of a dashboard through which users 489  
 432 can specify a profile describing a sequence of high-level tasks that 490  
 433 must be performed by the integrated devices to realize a given ac- 491  
 434 tivity. Afterwards, a gateway realizes the orchestration of the de- 492  
 435 vices according to the previously defined tasks against the capabil- 493  
 436 ities of each device (known a priori by the gateway). Fig. 7 shows 494  
 437 an example of an orchestration designed to monitor a patient with 495  
 438 cardiac problems. This orchestration is composed of actions per- 496  
 439 formed by the devices used in the environment, namely two ac- 497  
 440 celerometers, a heart rate sensor, a video camera, a voice trans- 498  
 441 mission device, and a media player. 499

442 *Frameworks.* Six of the selected studies have proposed frame- 500  
 443 works that offer APIs and reusable features to support the devel- 501  
 444 opment of applications benefiting from the convergence of IoT and 502  
 445 Cloud Computing. Some of the offered facilities are: (i) high-level 503  
 446 interfaces to access heterogeneous devices; (ii) communication be- 504  
 447 tween smart-objects via REST and WebSockets [27]; and (iii) de- 505  
 448 vice virtualization to encapsulate and access IoT resources and 506  
 449 capabilities by using a uniform API. An example of these frameworks 507  
 450 supporting the development of applications based on both IoT and 508  
 451 cloud is PatRICIA, proposed in study S23. PatRICIA provides a pro- 509  
 452 gramming model defining high-level constructs and operators to 510  
 453 enable developers to implement applications without being con- 511  
 454 cerned about the complexity of low-level device services and raw 512  
 455 sensor data streams. Some examples of the operators available at 513  
 456 the PatRICIA framework are: (i) *send*, for communicating with de- 514  
 457 vices; (ii) *notify*, for subscribing to and receiving information about 515  
 458 events; (iii) *poll*, for synchronization; and (iv) *delimit*, which refers 516  
 459 to tasks to be performed under satisfaction of certain conditions. In 517  
 460 turn, study S10 aims to provide a set of development kits (SDKs) 518  
 461 for developing context-aware mobile applications that make use 519  
 462 of smart objects (that communicate among each other by using 520  
 463 WebSockets) while taking advantage of scalability provided by the 521  
 464 cloud. 522

465 *Middleware.* Five studies have proposed using a middleware 523  
 466 layer to allow integrating IoT and Cloud Computing. In general, 524  
 467 they address important concerns such as: (i) abstractions to vir- 525  
 468 tualize devices, i.e., representing (groups of) sensors as virtual 526  
 469 entities that expose their capabilities at a high-level and abstract 527  
 470 away details related to protocol and data formats, thus facilitating 528  
 471 their use by applications; (ii) abstractions aimed to uniquely iden- 529  
 472 tify devices; (iii) communication based on the publish-subscribe 530  
 473 model [28] in which applications register interest in some infor- 531  
 474 mation and are immediately notified when such an information 532  
 475 is updated, thereby allowing to monitor information obtained by

sensors at real-time, for example; (iv) standardized interfaces to 476  
 enable applications to access resources such as information, sen- 477  
 sors, and cloud platforms towards minimizing interoperability is- 478  
 sues; (v) security mechanisms to ensure information integrity and 479  
 privacy; (vi) services for controlling and managing devices; and 480  
 (vii) discovery services aimed to identify new IoT devices. The pro- 481  
 posals reported in two studies (S1 and S13) have been designed 482  
 targeting the domain of smart cities, in particular to create an 483  
 ecosystem that allows developing smart applications based on both 484  
 IoT and Cloud Computing with capabilities for device discovery, 485  
 efficient data processing, access to information provided by the 486  
 integrated devices, etc. In study S13, a middleware called ClouT 487  
 provides services such as device discovery, (meta)data storage and 488  
 retrieval, and analysis and extraction of data stored on the cloud 489  
 for identifying events that occur in the city. In turn, the middle- 490  
 ware presented in study S1 provides sensor discovery capabilities 491  
 and allows large data processing in the cloud infrastructure as well 492  
 as real-time delivery of notifications to mobile devices through the 493  
 use of a publish/subscribe communication mechanism. In addition, 494  
 devices and their respective capabilities are semantically described 495  
 by using ontologies, thus supporting for semantic reasoning and 496  
 querying for mobile devices. 497

*Challenges on the integration of IoT and Cloud Computing.* Finally, 498  
 we have observed that the majority of the selected studies devotes 499  
 some attention to present and/or discuss challenges related to the 500  
 integration of IoT and Cloud Computing besides presenting solu- 501  
 tions to some of them. As previously mentioned, the combination 502  
 of these paradigms can overcome some of the open issues individ- 503  
 ually observed in each side, e.g., resource limitations of IoT devices 504  
 remedied by the use of cloud services. However, other challenges 505  
 persist and become more critical despite such an integration, in 506  
 particular interoperability (i.e., standardization of data and proto- 507  
 cols) and security in terms of data integrity and confidentiality. 508  
 With a smaller frequency, other challenges were also found, such 509  
 as the need of processing large volumes of data coming from IoT 510  
 devices (studies S7 and S23), the lack of standardization of data 511  
 from IoT devices, virtualization of IoT devices on the cloud (studies 512  
 S9 and S13), device orchestration (study S8), and energy efficiency 513  
 (studies S3 and S20). Furthermore, it is worth mentioning the exis- 514  
 tence of six studies (S3, S6, S7, S16, S17 and S25) whose main goal 515  
 is to elicit and discuss relevant challenges in this context. We pro- 516  
 vide a more detailed discussion about these and other challenges 517  
 on the integration of IoT and Cloud Computing in Section 5. 518

#### 4.3. RQ2 – strategies for integrating IoT and Cloud Computing 519

We have identified three broad categories in which the selected 520  
 studies can be classified with respect to integration strategies for 521  
 IoT and Cloud Computing, namely (i) *minimal integration*, (ii) *par-* 522  
*tial integration*, and (iii) *full integration*. This classification was made 523  
 based on an assessment of how both paradigms were integrated 524  
 to each other in 29 of 35 selected studies, which are the ones 525  
 presenting concrete solutions (i.e., architectures, frameworks, plat- 526  
 forms or middleware). In the minimal integration category, both 527  
 paradigms are used in their original purpose with almost no real 528  
 integration between them. In the partial integration category, both 529  
 paradigms are used in conjunction to provide at least one novel 530  
 service layer that would not exist otherwise. Finally, the solutions 531  
 532

**Table 4**  
Categories of integration strategies found in the selected primary studies.

| Integration strategy | Selected studies   |
|----------------------|--|
| Minimal integration  | S2, S4, S8, S11, S12, S14, S15, S19, S20, S21, S23, S27, S31 |
| Partial integration  | S1, S5, S9, S10, S18, S22, S24, S26, S28, S29, S30, S32, S34 |
| Full integration     | S13, S33, S35  |

in the full integration category provide a completely new perspective in which all service layers result from the use of both paradigms. Table 4 shows a categorization of the selected studies into these groups, each one detailed in the following.

The *minimal integration* strategy is characterized by the existence of an IoT middleware or platform deployed in a cloud environment (either in the IaaS – *Infrastructure as a Service* layer or in the PaaS – *Platform as a Service* layer) to make use of cloud services. Such a platform is usually related to the provision of visualization, computation, analytics, and storage of data collected from smart objects in a scalable way, capabilities achieved thanks to the adoption of the Cloud Computing paradigm in this context. As shown in Table 4, this category is represented by seven primary studies. For instance, the solution proposed in study S11 encompasses (i) collecting data from healthcare sensors, (ii) using a smart gateway to send these data to a Web platform deployed on a cloud infrastructure, and (iii) visualizing the collected data in such a platform. Similarly, study S14 presents an IoT Web portal that collects data from several types of sensors through an Internet gateway. Collected data are stored using a scalable cloud storage and the Web portal (also deployed in the cloud infrastructure) provides visualization and analytics services based on the underlying cloud infrastructure.

The *partial integration* strategy is characterized by a higher integration level in comparison to the minimal integration strategy. In this case, not only the IoT middleware or platform is deployed in a cloud environment, but the platform also provides new service models based on abstractions of smart objects. Therefore, service models such as SOaaS – *Smart Object as a Service* and SaaS – *Sensing as a Service* are provided to abstract away the heterogeneity of devices and virtualize their capabilities. These novel service models allow smart objects to be concurrently controlled by several users and/or applications (i.e., in a multi-tenant approach), including their actuation capabilities. For instance, study S10 proposes a framework that provides the notion of SOaaS, which envisions encapsulating and exposing smart objects as services. This approach allows smart objects to be accessed, actuate, and dynamically reconfigured at runtime. Similarly, study S18 presents a middleware that virtualizes physical smart things allowing isolation among multiple applications sharing them. The middleware can also allow efficient sharing and resource conflict resolution in the use of smart things.

Finally, in the highest level of integration, the *full integration* strategy stands for the emergence of new service models that extend all of the conventional Cloud Computing layers (i.e., IaaS, PaaS, and SaaS) to encompass services provided by physical objects. This means that physical devices are able to expose their functionalities as standardized cloud services. As first-class entities in the cloud, devices can also be used in conjunction with traditional cloud resources provided by third parties. For example, the developer of an environmental monitoring application can compose sensing capabilities of the so-called cloud of things along with computational and storage capabilities of traditional cloud. Based on the full integration strategy, study S13 proposes new service models in the context of smart cities. The first one is called ClaaS – *City Infrastructure as a Service*, which expands the traditional IaaS layer to offer virtualized resources by means of standard

interfaces and allows seamlessly accessing any device, data, and computing capabilities. The second layer is called CPaaS – *City Platform as a Service*, which expands the traditional PaaS layer to foster the development of applications as well as to provide a set of specialized middleware services, such as data processing for analysis and extraction of data stored in the cloud, service composition for aggregating data/services offered by deployed applications, and resource access for storing and retrieving (meta)data from the smart city. Finally, the uppermost level is called CSaaS – *City Software as a Service*, an expansion of the traditional SaaS layer represented by applications developed/deployed by using services provided at the ClaaS and CPaaS layers. Similarly, the framework proposed in study S33 is composed of three layers. The lowermost layer (called *WoT Infrastructure*) is built upon the traditional IaaS layer and provides a basic IoT infrastructure (comprising embedded devices and Web gateways) to bridge different types of sensors, actuators, and other devices, which are exposed as Web resources. In turn, the middle layer (*Service and Business Operation*) expands the traditional PaaS layer and provides service composition and business process management. At last, the uppermost layer (*Intelligent Services*) is an expansion of the traditional SaaS layer and it provides a collection of IoT applications and interfaces for end-users, which have a pay-per-use direct access to the provided services. Fig. 8 presents the architectures proposed in studies S13 and S33. Even though these studies use different names for their layers, both architectures have some similarities. For instance, the lowermost layer of studies S13 and S33 mainly focuses on the boundary between the physical and the digital realms, dealing with the virtualization of IoT devices. The middle layer is concerned with providing middleware services and supporting the execution of applications. The uppermost layer focuses on providing relevant applications that make use of the services made available by the lower layers.

#### 4.4. RQ3 – architectures and infrastructures for IoT and cloud

In order to answer this research question, we have striven to characterize the architectures and infrastructures supporting the construction, deployment, and execution of IoT applications in cloud platforms. We noticed that these architectures were designed considering particular features, thus making them significantly distinct from each other and hindering the conception of a general picture of what would be an architectural solution targeting the convergence of IoT and Cloud Computing. A reason for such a difference is the lack of standardized means for supporting the design of these architectures when integrating IoT and Cloud Computing. Nevertheless, we have observed that the majority of the solutions (23 studies, i.e., 66%) has adopted traditional approaches such as smart gateways, Web services relying upon SOAP (*Simple Object Access Protocol*) [29] or REST and drivers or APIs deployed in the SaaS cloud layer. On the other hand, the studies have typically adopted the PaaS layer to support the deployment of tools and services for developing applications, as well as the IaaS layer as underlying infrastructure for hosting and executing applications.

Despite several studies use the concepts of the three traditional cloud layers (IaaS, PaaS, SaaS) as a starting point for constructing their architectures, the proposed approaches are not limited to these service models, thus creating novel concepts and/or adapting



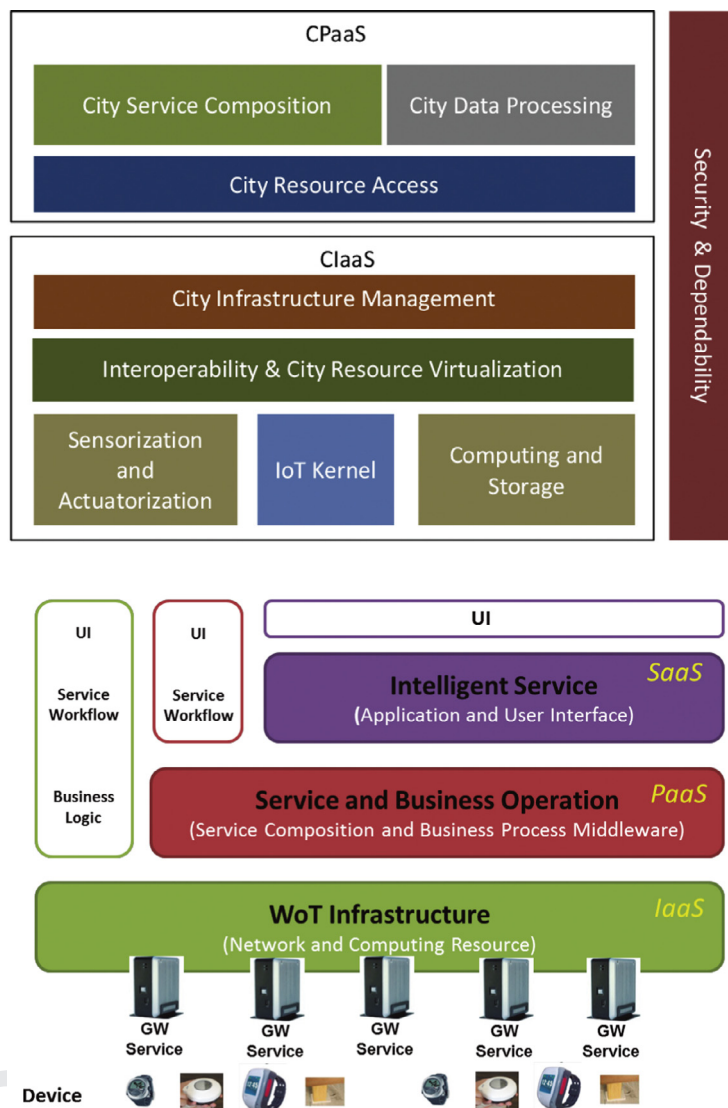


Fig. 8. Architectures of the solutions proposed in studies S13 (top) and S33 (bottom).

Table 5  
Traditional and novel cloud-based service models considered and/or proposed by the selected studies.

| Service model                              | Description  | Studies   |
|--|--|---|
| Software as a Service (SaaS)               | Applications are designed and deployed for end-users over the Web                            | S4, S5, S8, S9, S12, S13, S14, S28, S29, S30, S31, S32, S33, S34, S35 |
| Platform as a Service (PaaS)               | Tools and services to support the development of applications                                | S10, S11, S13, S14, S19, S21, S23, S29, S30, S32, S33, S34, S35       |
| Infrastructure as a Service (IaaS)         | Delivery of infrastructure (e.g., servers, storage, network, operating systems) as a service | S13, S14, S15, S22, S25, S27, S30, S33, S34, S35                      |
| Network as a Service (NaaS)                | Virtualization of devices in a wireless sensor network                                       | S26   |
| Sensing as a Service (SaaS)                | Ubiquitous access to sensor data   | S1, S5  |
| Sensing and Actuation as a Service (SAaaS) | Sensing and actuation capabilities provided as services                                      | S9, S24   |
| Smart Object as a Service (SOaaS)          | Virtualization physical smart objects (things) for multi-tenancy support                     | S18   |

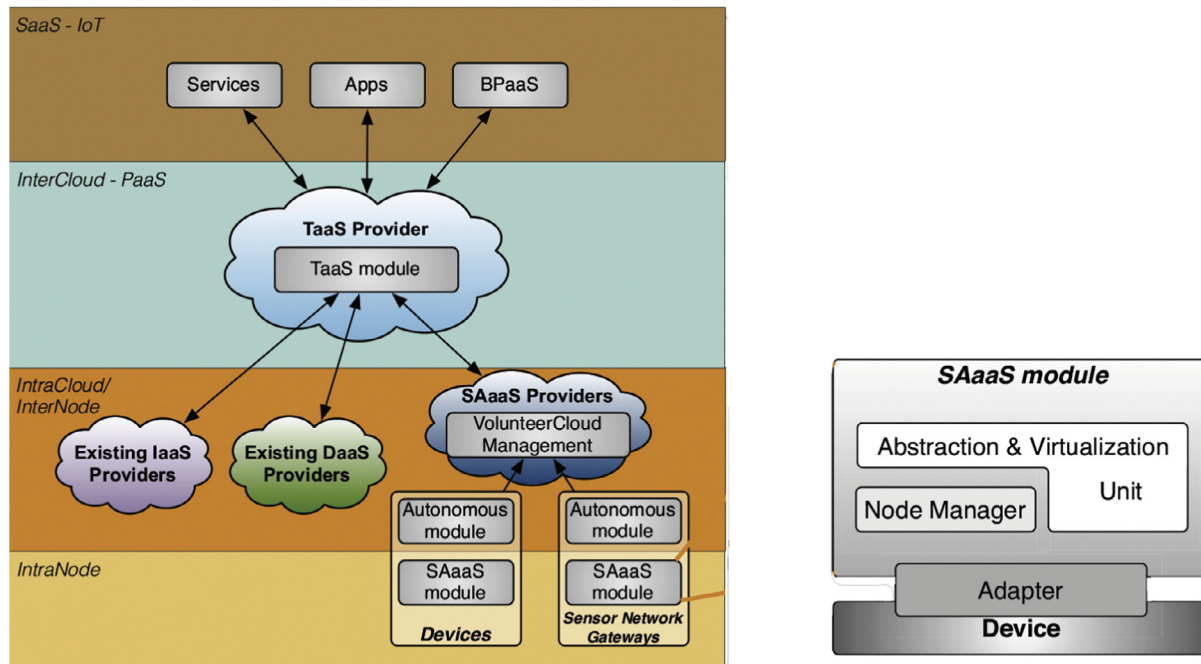


Fig. 9. Representation of the architecture presented in study S9 encompassing the SAaaS concept.

644 them to the IoT scenario. For instance, while the traditional SaaS  
 645 model is concerned with delivering software that can be used over  
 646 the Internet by using a thin client (browser), some studies have  
 647 adapted this concept to IoT for creating the SaaS – *Sensing as a*  
 648 *Service* model, characterized by accessing sensor data via Internet.  
 649 Novel service models were also proposed aiming at enhancing the  
 650 synergy between the IoT and Cloud Computing paradigms, as sum-  
 651 marized in Table 5. In the following, we present an overview of  
 652 these novel concepts.

653 Inspired on the utility model offered by cloud services, SAaaS –  
 654 *Sensing and Actuation as a Service* aims at exposing wireless sensor  
 655 networks (WSNs), smartphones or other devices endowed with  
 656 sensors/actuators as services accessible through the Internet. To  
 657 deal with this novel concept, studies S9 and S24 propose different  
 658 approaches. In the former (see Fig. 9), SAaaS modules at the bot-  
 659 tom of the architecture virtualize sensing and actuation capabilities  
 660 of IoT devices, expose such resources as services at the minor gran-  
 661 ularity level, and relay sensor events to upper layers. These SAaaS  
 662 modules in the proposed architecture also use adapter interfaces  
 663 to directly interact with the sensors/actuators, maintain connectiv-  
 664 ity control, and translate commands and relay them to the respec-  
 665 tive underlying physical endpoints, through device-native commu-  
 666 nication protocols. In turn, the approach found in the latter (see  
 667 Fig. 10) encompasses the concept of smart sensors, which are capa-  
 668 ble of performing an initial processing of the collected data besides  
 669 sensing tasks. The proposed architecture is based on a virtualized  
 670 and distributed infrastructure provided by a cloud IaaS environ-  
 671 ment as means of offering computational and flexible storage capa-  
 672 city with enhanced performance. Therefore, virtualization allows  
 673 seamlessly using information produced by heterogeneous sensor  
 674 networks and making it available to upper layers of the architec-  
 675 ture, such as the ones responsible for business logics and decision-  
 676 making, as well as end-user applications.

677 The NaaS – *Network as a Service* model is characterized by the  
 678 virtualization of devices in a wireless sensor network in which  
 679 hardware resources are exposed as services. Study S26 proposes  
 680 Serviceware, a service-oriented middleware that allows sensor net-  
 681 work infrastructure resources to be virtualized and exposed as ser-  
 682 vices, as depicted in Fig. 11. The middleware is directly deployed

into devices and it works as an encapsulating layer over the de-  
 vice’s operating system (OS) in order to virtualize its hardware re-  
 sources, which are accessed by using the API provided by the OS  
 layer.

687 Finally, SOaaS – *Smart Object as a Service* is a model that en-  
 688 visions virtualizing physical smart objects to be used by multi-  
 689 ple applications, i.e., in a multi-tenant way. Study S18 propos-  
 690 es ECO, a middleware compliant with this model (see Fig. 12).  
 691 The ECO middleware is organized in three layers, namely *Device*  
 692 *Framework Layer* (DFL), *Object Orchestration Layer* (OOL), and *Vir-*  
 693 *tualization Layer* (VL). Above VL, ECO Runtime APIs hide complex  
 694 operations performed at the virtualization layer and provide de-  
 695 velopers with abstractions for developing their applications atop  
 696 the ECO middleware. In turn, DFL aims to provide a unified ac-  
 697 cess to heterogeneous networks and capabilities of smart things.  
 698 OOL provides management of uniform resource objects (UROs),  
 699 which enable VL to access heterogeneous devices and services with  
 700 no knowledge about the underlying protocols. Finally, VL enables  
 701 multiple applications to use smart things at a higher abstraction  
 702 level.

## 5. Challenges on the integration of IoT and Cloud Computing

703  
 704 As previously mentioned, despite the combination of IoT and  
 705 Cloud Computing can overcome some of the open issues to be  
 706 individually solved in both sides, new, additional challenges arise  
 707 from such a synergistic relationship. In this section, we provide  
 708 a non-exhaustive list of some challenges identified from the ana-  
 709 lyzed studies and other ones that we regard as relevant to pave  
 710 the way for future research in this context.

711 *Reference architectures for cloud-based IoT solutions.* As reported  
 712 in Section 4.4, the existing solutions relying upon the convergence  
 713 of IoT and Cloud Computing are significantly different from each  
 714 other. Such a distinction is mainly due to the lack of standardized  
 715 means for supporting the design of these solutions, resulting in  
 716 an increased complexity and significant effort from architects and  
 717 developers. This problem becomes worse with the absence of a  
 718 proper guidance for achieving both functional and non-functional  
 719 requirements since the conception of these solutions up to their

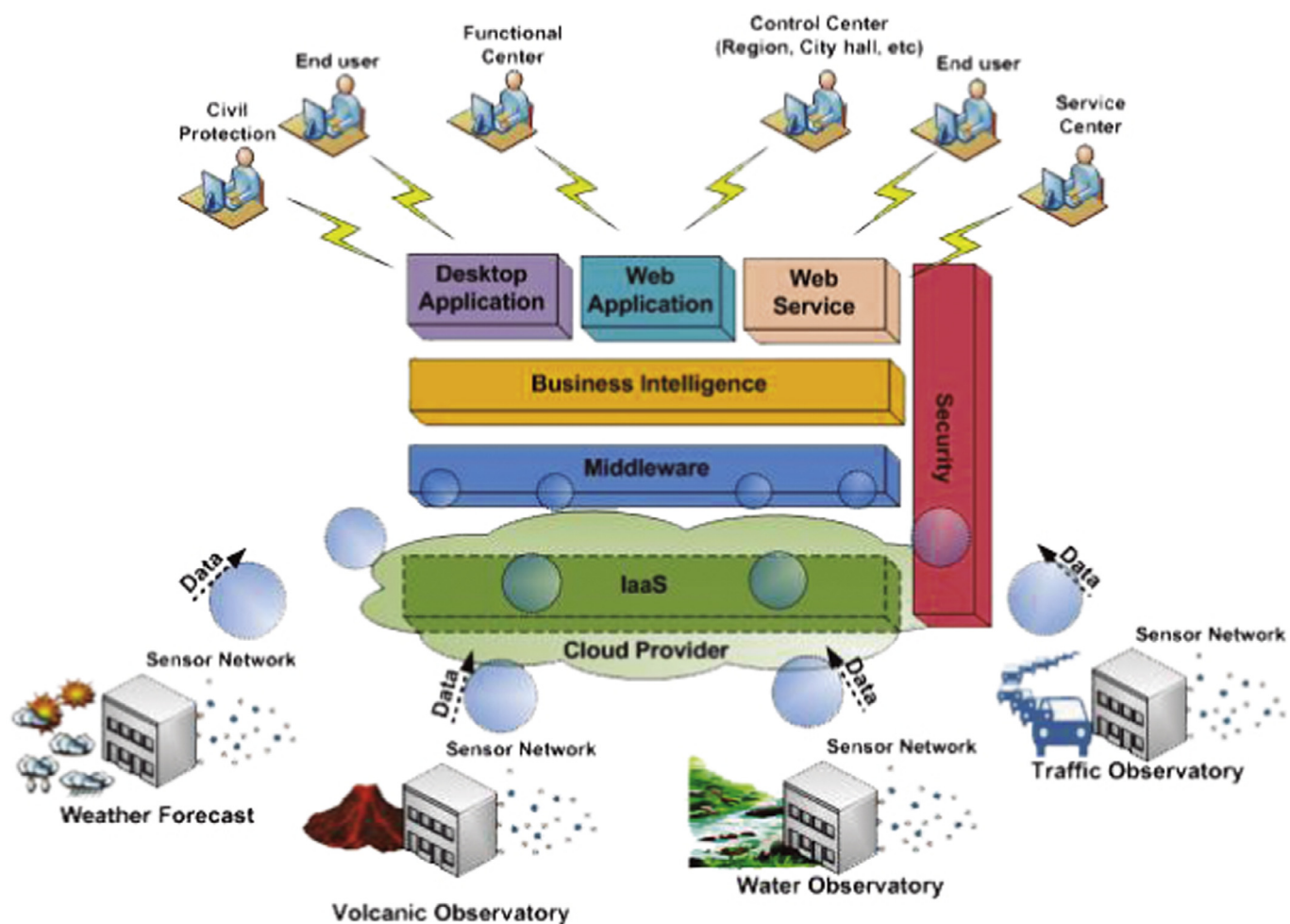


Fig. 10. Architecture presented in study S24 envisioning SaaS.

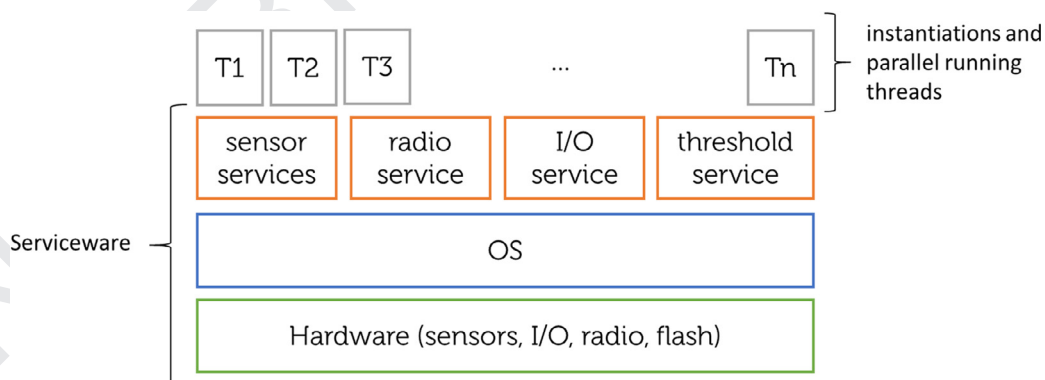


Fig. 11. Architecture of Serviceware middleware proposed in study S26.

720 implementation and deployment. To mitigate these issues, ref-  
 721 erence architectures could play a role in terms of defining the  
 722 essential building blocks for constructing architectural solutions  
 723 that take into account the integration of IoT and Cloud Computing.  
 724 Reference architectures can be understood as abstract architectures  
 725 encompassing knowledge and experiences in a given application  
 726 domain, thus being able to facilitate and guide development, stan-  
 727 dardization, interoperability, and evolution of software systems in  
 728 such a domain [30]. Therefore, directions provided by a reference  
 729 architecture can be important elements to guide and facilitate the  
 730 development of cloud-based IoT architectures coping with their  
 731 increasing scale and complexity. Furthermore, considering that

732 developing interoperable solutions is an important concern, such  
 733 an interoperability can be achieved by constructing architectural  
 734 solutions founded upon a reference architecture [31]. After analyz-  
 735 ing the solutions proposed in the selected studies, we have noticed  
 736 that the OpenIoT platform [32] seems to go in the direction of be-  
 737 coming a reference architecture in the integration of IoT and Cloud  
 738 Computing as it provides guidelines and mechanisms to construct  
 739 new architectures in this context. As an example, OpenIoT was  
 740 successfully used to develop the architectural solutions proposed  
 741 in studies S1 and S28 in different application domains.

742 *Efficient use of cloud resources by IoT devices/applications.* The  
 743 huge amount of physical devices connected to the Internet

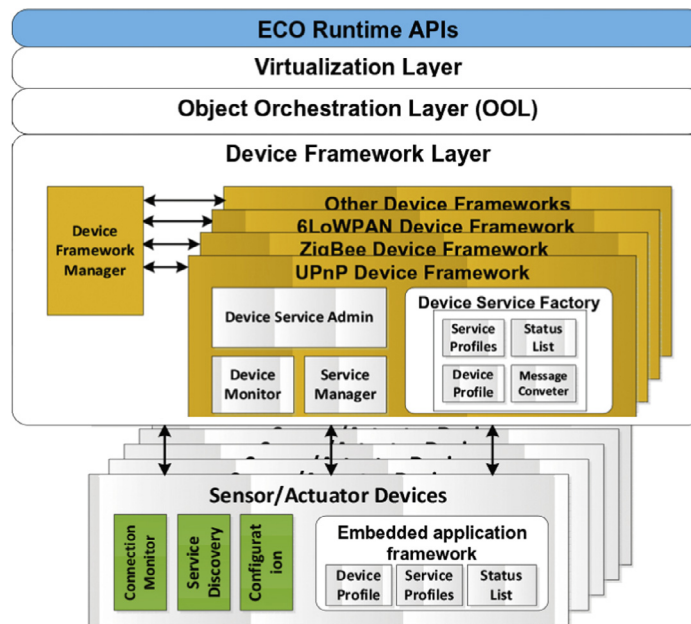


Fig. 12. Architecture of ECO middleware proposed in study S18.

744 generating and sending data to be processed on the cloud will  
 745 consume a significant amount of network, processing, and storage  
 746 resources. If all data generated by devices are sent to the cloud,  
 747 massive data streams transmitted across wide area networks may  
 748 cause network congestion and the communication latency may be  
 749 intolerable for applications that have real-time (or quasi-real-time  
 750 requirements). Even though current cloud platforms seem to be  
 751 prepared to provide as many resources as required, the exhaust-  
 752 ive use of such resources may bring expensive monetary costs.  
 753 As highlighted by studies S2, S3 and S6, it is necessary to pro-  
 754 pose strategies able to minimize the amount of cloud resources  
 755 to be consumed by IoT-based applications and to decide where  
 756 to process data generated by devices in an efficient way. One of  
 757 the strategies presented by these studies is applying the concept  
 758 of *Fog Computing*, which extends Cloud Computing to the edge of  
 759 the network towards providing processing, storage, and network-  
 760 ing services between devices and cloud platforms [33,34]. Such a novel  
 761 paradigm is appealing for IoT since it can reduce the consumption  
 762 of cloud resources while reducing latency for applications and end-  
 763 users, an essential requirement for some critical applications, such  
 764 as the ones in the healthcare domain. According to the aforemen-  
 765 tioned studies, this concept could be applied by using smart gate-  
 766 ways and local clouds, which will be responsible for mediating the  
 767 communication between IoT devices and cloud platforms and as  
 768 well as providing preprocessing and temporary storage of informa-  
 769 tion coming from the devices. Nonetheless, additional research ef-  
 770 forts are required for realizing these solutions and hence mitigate  
 771 the raised challenges. There is also a need of carefully evaluating  
 772 where to store and process sensed data to meet different types  
 773 of applications in order to obtain the best out of cloud resources  
 774 while exploiting the capabilities (albeit limited) of the devices  
 775 themselves.

776 *Standardization of data and services.* Data produced by the vari-  
 777 ous IoT devices do not follow any sort of standardization, i.e., they  
 778 are usually provided in different formats, units, etc. Moreover, cur-  
 779 rently there are no means for describing IoT devices and their ca-  
 780 pabilities in a standardized way, so that software agents cannot  
 781 easily perform tasks such as automatic discovery and orchestration  
 782 of devices, data, and services. These concerns are important to be  
 783 addressed in the context of IoT and Cloud Computing as means  
 784 of allowing applications to benefit from data provided by IoT de-

785 vices and from the scalability and availability promoted by cloud  
 786 services. As an attempt to mitigate this challenge, study S8 sug-  
 787 gests following the recent initiatives proposed by OASIS, in particu-  
 788 lar the DPWS (*Device Profile for Web Services*) standard [35]. DPWS  
 789 has been designed as a language-independent set of specifications  
 790 and guidelines to describe devices and their capabilities as ser-  
 791 vices, thus fostering interoperability among them. In addition, on-  
 792 tologies have been suggested for semantically describing devices.  
 793 A relevant example is the approach used in the OpenIoT middle-  
 794 ware [32]. OpenIoT extends the W3C's Semantic Sensor Network  
 795 (SSN) ontology [36] to provide a common model to semantically  
 796 describe both physical and virtual sensors while enhancing exist-  
 797 ing vocabularies for smart objects with additional concepts relevant  
 798 to the integration of IoT and Cloud Computing. The main premise  
 799 of adopting such an ontology is to offer an easy way to combine data  
 800 streams and services from diverse IoT applications that feature in-  
 801 compatible semantics (e.g., measurement units, raw sensor values,  
 802 etc.) as well as to support seamless discovery and monitoring of  
 803 sensors, which are not restricted to physical devices, but rather  
 804 represent anything that can calculate/estimate the value of a phe-  
 805 nomenon.

806 *Contextual information.* In IoT, sensor data are not the only infor-  
 807 mation to be manipulated, but also their context, i.e., any informa-  
 808 tion that can be used to characterize a person, location or object  
 809 considered as relevant to the environment [37]. Relevant contex-  
 810 tual information such as measurement properties, location, state  
 811 of entities, and data precision can be annotated along with the  
 812 raw data obtained by the sensing devices for a posteriori retrieval  
 813 and/or further processing aimed towards reacting to stimuli [38].  
 814 In spite of the fact that the insertion of contextual information  
 815 adds more value and semantics to sensor data, it also increases the  
 816 heterogeneity in terms of the used data formats. Indeed, the use of  
 817 different data representation models poses a barrier for informa-  
 818 tion exchange and retrieval. The way in which data generated by  
 819 the myriad of IoT devices are extracted and accessed must be re-  
 820 viewed to allow the extraction and easy access to the wide range  
 821 of data available to end-users and applications.

822 *Data security and privacy.* Data obtained from IoT devices can  
 823 contain sensitive information that others may be interested in,  
 824 not only the sensed data themselves, but also their corresponding  
 825 meta-information (e.g., location). As pointed out in study S16,

outsourcing these data to the cloud raises a number of issues related to privacy concerns mainly due to the lack of control or at least transparency over the access to data, which can be handled by third parties or misused for unintended purposes. Moreover, many devices able to share information and to be controlled via Internet may become vulnerable to several types of attacks. In this scenario, hackers or malicious users may try to remotely control devices, acquire confidential information or promote changes in the contents of messages while they are transmitted. As IoT devices often have limited processing capabilities, they usually do not have very complex security mechanisms such as data encryption, authentication, etc. Therefore, it is necessary to investigate mechanisms aiming at protecting critical and/or sensitive data from IoT devices and ensuring security when storing them on the cloud. At the same time, these actions must be lightweight and require a minimum of processing resources from IoT devices. Nonetheless, data security and privacy in the Cloud Computing context still suffers from several limitations and threats [39], thereby making the convergence of IoT and Cloud Computing more dramatic if these concerns are not properly addressed.

**High heterogeneity of IoT and cloud environments.** IoT and Cloud Computing are characterized by a high degree of heterogeneity, thus requiring solutions allowing interoperability among resources. On the one hand, it is necessary to handle a myriad of IoT devices from several manufacturers and with different capabilities/functionalities and network protocols, an issue that creates operational barriers to use such devices in a holistic, integrated way [40]. On the other hand, the heterogeneity and lack of standardization in Cloud Computing hampers the use of cloud services offered by multiple providers. This issue directly affects the development and deployment of applications since they become highly coupled to a single cloud provider and its services and constraints, leading to a problem known as cloud lock-in [41]. Despite middleware platforms have recently drawn attention from both academia and industry as promising solutions to integrate heterogeneous devices and provide high-level models for developing IoT applications, most of the existing proposals have not achieved a mature development state and often neglect important requirements in this context [42,43]. In turn, the literature still lacks of effective means of seamlessly using services provided by different cloud providers while considering parameters such as quality of service and cost efficiency [44]. Therefore, further research is necessary aiming at providing middleware platforms able to tackle the inherent heterogeneity of IoT and cloud environments while leveraging the development and deployment of applications that benefit from the integration of a plethora of IoT devices and can use multiple cloud services offered by different providers. In addition, it is important to provide users/developers with flexible mechanisms that allow easily shifting from a given service provider to another as well as selecting the cloud services that meet application requirements.

**Elasticity concerns.** In the Cloud Computing paradigm, elasticity allows dynamically increasing or decreasing the use of computational resources as response to the varying demands of users and applications. This principle aims at adjusting the amount of provisioned resources to exactly match the current needs, thus minimizing resource overprovisioning and avoiding unnecessary costs with idle or underutilized resources. Despite being one of the fundamental traits of Cloud Computing, elasticity has not yet received enough attention when integrating cloud services and IoT. As discussed in study S22, systems and solutions relying on the convergence of IoT and Cloud Computing are usually not tailored to incorporate elasticity concerns. For example, new types of resources such as data streams and devices delivered by an IoT infrastructure are not provided elastically, thereby preventing current IoT systems to fully take advantage of the benefits offered by the elastic na-

ture of the Cloud Computing paradigm. Therefore, new approaches are required to inherently incorporate elasticity concerns in cloud-based IoT systems.

**Dependability concerns.** Both IoT and cloud environments are highly dynamic. IoT devices can become unavailable for diverse reasons, e.g., failure, battery depletion, lack of network connectivity, user mobility, etc. In turn, cloud environments may experience typical situations such as unavailability or quality degradation of services used by applications. For both cases, it is necessary to provide means of adapting applications at runtime with minimal or no disruption, thus ensuring a proper response to several events at runtime as well as the satisfaction of non-functional requirements such as availability and quality. Although dependability is especially important for safety-critical applications, in which failures or quality degradation may be a threat to people or lead to economic losses and physical damages, none of the selected primary studies addresses this requirement.

**Support to the development of applications.** Currently, there is no de facto standard to programming languages for IoT devices and no standardized programming models for building IoT-based applications. IoT elements (sensors, actuators, gateways, applications, etc.) are often developed, deployed, and separately from cloud services such as storage and data processing. For example, most cloud services can reactively monitor the load from IoT and adjust their performance behavior, but they rarely communicate back to the IoT elements to steer it. Furthermore, it is hard to control and manage both IoT and cloud services as a uniform software layer. Truong and Dudstar [18] outline three essential phases that should be considered in the life cycle of a cloud-based IoT system: (i) *development phase*, including techniques for selecting, composing, and integrating components across the system for specifying and developing possible governance and control operations; (ii) *deployment and provisioning phase*, which includes techniques for deploying several types of software components at different levels of abstraction and capabilities to configure deployments and continuous resource provisioning; and (iii) *operation phase*, which includes capabilities to monitor end-to-end metrics, perform governance processes across the system, and control coordinated elasticity processes. Transversally, it is necessary to provide environments to support development of applications based on data streams generated by devices and available through the cloud.

**Models for device virtualization.** One of the first steps to realize the integration of IoT and Cloud Computing is to have a model for virtualizing devices (sensors, actuators, and other physical objects) and exposing their functionalities as services that can be requested by applications on-demand. By virtualizing devices, it is possible to extend the functionality of cloud platforms to encompass physical objects. However, the traditional service provisioning model must also be extended by creating a novel model of sensing/actuation as a service as well as novel business models to encompass the provision of this new type of service. Regardless of the adopted approach, lightweight virtualization techniques are required to accommodate the limited resources of IoT devices. In addition, physical devices and networks need to expose their functionalities to third parties, preferably using open Web standards. Therefore, the use of open interfaces can allow resources to be used in conjunction with well-established cloud tools and open cloud platforms and promotes interoperability with other existing systems and IoT applications.

## 6. Threats to validity

The conducted systematic mapping and its results may have been affected by some threats to validity. In the following, we discuss some of these limitations.

*Incompleteness of the study search.* The completeness of this systematic mapping may have been affected by missing relevant studies. In order to reduce this threat, we have used electronic databases (see Table 2) that are among the most relevant available sources in Computer Science and Engineering [22,23]. However, there are still limitations. First, some studies may have been missed due to technical limitations of the automated search engines, an issue that is out of our control. Second, the selected electronic databases do not represent an exhaustive list of publication sources, so that other databases might also be included. Third, we have not performed snowballing [45], a useful technique that consists of checking the reference lists of the read studies aiming at finding additional studies that were not retrieved in the automated search procedure. Fourth, we are aware about studies fitting the scope of our research (such as the ones focused on the integration of sensor networks and Cloud Computing), but that were not retrieved in the automated search process. This can be explained by the fact that they do not consider the terms used in the search string (e.g., IoT), which refer to a wider range of devices rather being limited to sensors. Therefore, other possibly relevant studies could have been identified and considered in this systematic mapping.

*Bias on study selection.* In order to make the results of this systematic mapping study reproducible, the protocol presented in Section 2 clearly established the search terms used in the automated search procedure, search sources, and criteria for selecting the primary studies. However, different researchers tend to have different understandings on these criteria, so that the results of the study selection performed by different researchers are likely to be varied. Even though the drawn conclusions may have been influenced by the researchers' opinions, we have striven to mitigate the effect of any personal bias or misinterpretation by adopting a multiple-revision strategy.

*Inaccuracy of data extraction.* Bias on data extraction may result in inaccuracy of the extracted data items, thus affecting the analysis of the selected studies. We have striven to reduce this bias by clearly defining the data items outlined in the data extraction spreadsheets. In addition, the data items to be extracted in this systematic mapping were discussed among the researchers and agreed upon their meaning.

*Bias on data synthesis.* Not all studies sufficiently and clearly describe the details of information to be extracted as data items aiming at supporting the answers to the defined research questions. Therefore, we have had to infer certain pieces of information regarding data items during data synthesis. In order to minimize the inaccuracy of such inferences, we have conducted discussions aiming at solving any disagreement and clarifying potential ambiguities.

## 7. Conclusion

The IoT paradigm has increasingly received attention from both academia and industry as a key enabler for the emergence of new applications and systems with a significant influence in the daily lives of human beings. Recently, we have witnessed the convergence of this paradigm with Cloud Computing largely motivated by the need of IoT infrastructures and applications to be enhanced in terms of computational resources, scalability, and performance. Furthermore, the cloud arises as a promising solution to overcome or alleviate several challenges faced by the IoT paradigm, which is essentially characterized by heterogeneous physical devices with technological constraints. In spite of this synergy, the literature still lacks of a wide, comprehensive overview on what has been investigated on the integration of IoT and Cloud Computing and the open issues in this context. To tackle this gap, we have performed a systematic mapping study aimed to provide a broad panorama of the

current state of the art on this topic and identify important challenges that can give directions for future research. We have analyzed 35 primary studies retrieved from five major electronic Computer Science and Engineering publication databases in order to: (i) obtain a comprehensive understanding on the convergence of IoT and Cloud Computing; (ii) identify what has been investigated in this context; (iii) understand how and to which extent IoT and Cloud Computing have been effectively integrated; and (iv) characterize the proposed solutions comprising the integration of IoT and Cloud Computing.

Our analyses of the selected studies have resulted in several findings about the current state of the art on the convergence of the IoT and Cloud Computing paradigms. First, almost all analyzed studies were published in the last three years, thus confirming an increasing interest of the scientific community on this topic and denoting that more relevant publications may be expected for the next years. Second, despite the several initiatives investigating the integration of IoT and Cloud Computing, existing proposals are still in infancy, i.e., they are at an initial development state and/or not fully validated. Indeed, we have found a significant number of studies lacking from any method to validate/evaluate their approaches and solutions or presenting simple proofs of concept, thereby indicating the lack of solid evidences about their efficiency, effectiveness, and feasibility. Third, we mapped four main types of concrete solutions proposed in the selected studies, namely: (i) *architectures*, which refer to high-level infrastructures targeting the integration of IoT and Cloud Computing and defining components and their respective roles; (ii) *platforms*, i.e., hardware and/or software infrastructures providing APIs to support the development and execution of applications as well as the management and monitoring of IoT devices at real-time; (iii) *frameworks*, which are software infrastructures providing reusable elements to foster the development of applications; and (iv) *middleware* providing services to applications and/or end-users while abstracting away underlying heterogeneous IoT devices and cloud resources. It is important to highlight that these solutions are significantly distinct from each other mainly due to the absence of standardized means for supporting their design, hindering a general picture of what would be an architectural solution targeting the integration of IoT and Cloud Computing. Fourth, the convergence of these paradigms is still limited, i.e., they are used either in their original purpose with almost no effective integration between them or in conjunction to provide new service models aiming at enhancing the synergy between IoT and Cloud Computing.

Finally, we have outlined a non-exhaustive list of some challenges identified from the analyzed studies and other ones that we considered as relevant to pave the way for future research in this context. These challenges refer to the need of: (i) standardizing cloud-based IoT solutions, data, and services; (ii) making IoT devices/applications to use cloud resources in an efficient way; (iii) dealing with massive and real-time data; (iv) handling contextual information; (v) providing effective solutions that consider data security and privacy; (vi) abstracting away the high heterogeneity of both IoT and cloud environments; (vii) considering elasticity and dependability concerns; (viii) supporting the development and deployment of cloud-based IoT applications; and (ix) providing models for device virtualization. In summary, the panorama presented in this paper provides not only a comprehensive overview on what has been investigated and developed in the integration of IoT and Cloud Computing, but it also points directions to future research in this context. With these analyses, such an overview can contribute to a more effective development of IoT infrastructures and applications taking advantage of the benefits and capabilities offered by the Cloud Computing paradigm.

## 1086 Appendix A. Selected studies

| Study ID | Reference   | Citation count in Dec. 2015 (Google scholar) |
|----------|---|--|
| S1       | I. P. Žárko, K. Pripuzić, M. Serrano, M. Hauswirth, IoT data management methods and optimization algorithms for mobile publish/subscribe services in cloud environments, in: Proceedings of the 2014 European Conference on Networks and Communications, IEEE, USA, 2014, pp. 1–5. doi:10.1109/EuCNC.2014.6882657   | 4  |
| S2       | M. Aazam, E.-N. Huh, Fog computing and smart gateway based communication for cloud of things, in: Proceedings of the 2014 International Conference on Future Internet of Things and Cloud, IEEE Computer Society, Washington, DC, USA, 2014, pp. 464–470. doi:10.1109/FiCloud.2014.83   | 17   |
| S3       | M. Aazam, P. P. Hung, E.-N. Huh, Smart gateway based communication for cloud of things, in: Proceedings of the 9th IEEE International Conference on Intelligent Sensors, Sensor Networks and Information Processing, IEEE, USA, 2014, pp. 1–6. doi:10.1109/ISSNIP.2014.6827673  | 14   |
| S4       | N. Alhakhani, M. M. Hassan, M. A. Hossain, M. Alnuem, A framework of adaptive interaction support in cloud-based Internet of Things (IoT) environment, in: G. Fortino, G. D. Fatta, W. Li, S. Ochoa, A. Cuzzocrea, M. Pathan (Eds.), Proceedings of the 7th International Conference on Internet and Distributed Computing Systems, Vol. 8729 of Lecture Notes in Computer Science, Springer International Publishing, Switzerland, 2014, pp. 136–146. doi:10.1007/978-3-319-11692-1_12   | 3  |
| S5       | J. Barbarán, M. Díaz, B. Rubio, A virtual channel-based framework for the integration of wireless sensor networks in the cloud, in: Proceedings of the 2014 International Conference on Future Internet of Things and Cloud, IEEE Computer Society, Washington, DC, USA, 2014, pp. 334–339. doi:10.1109/FiCloud.2014.59   | 0  |
| S6       | F. Bonomi, R. Milito, J. Zhu, S. Addepalli, Fog computing and its role in the Internet of Things, in: Proceedings of the First Edition of the MCC Workshop on Mobile cloud computing, ACM, New York, NY, USA, 2012, pp. 13–15. doi:10.1145/2342509.2342513  | 252  |
| S7       | A. Botta, W. De Donato, V. Persico, A. Pescape, On the integration of Cloud Computing and Internet of Things, in: Proceedings of the 2014 International Conference on Future Internet of Things and Cloud, IEEE Computer Society, Washington, DC, USA, 2014, pp. 23–30. doi:10.1109/FiCloud.2014.14   | 27   |
| S8       | J. Cubo, A. Nieto, E. Pimentel, A cloud-based Internet of Things platform for ambient assisted living, Sensors 14 (8) (2014) 14070–14105. doi:10.3390/s140814070  | 5  |
| S9       | S. Distefano, G. Merlino, A. Puliafito, Towards the Cloud of Things: Sensing and actuation as a service, a key enabler for a new cloud paradigm, in: Proceedings of the 8th International Conference on P2P, Parallel, Grid, Cloud and Internet Computing, IEEE Computer Society, Washington, DC, USA, 2013, pp. 60–67. doi:10.1109/3PGCIC.2013.16  | 5  |
| S10      | C. Doukas, F. Antonelli, COMPOSE: Building smart & context-aware mobile applications utilizing IoT technologies, in: Proceedings of the 2013 Global Information Infrastructure Symposium, IEEE, USA, 2013, pp. 1–6. doi:10.1109/GIIS.2013.6684373   | 5  |
| S11      | C. Doukas, I. Maglogiannis, Bringing IoT and Cloud Computing towards pervasive healthcare, in: Proceedings of the 6th International Innovative Mobile and Internet Services in Ubiquitous Computing, IEEE Computer Society, Washington, DC, USA, 2012, pp. 922–926. doi:10.1109/IMIS.2012.26  | 32   |
| S12      | C. Doukas, T. Pliakos, P. Tsanakas, I. Maglogiannis, Distributed management of pervasive healthcare data through Cloud Computing, in: K. S. Nikita, J. C. Lin, D. I. Fotiadis, M.-T. A. Waldmeyer (Eds.), Proceedings of the Second International ICST Conference on Wireless Mobile Communication and Healthcare, Vol. 83 of Lecture Notes of the Institute for Computer Sciences, Social Informatics and Telecommunications Engineering, Springer Berlin Heidelberg, Germany, 2012, pp. 386–393. doi:10.1007/978-3-642-29734-2_53 | 1  |
| S13      | J. A. Galache, T. Yonezawa, L. Gurgun, D. Pavia, M. Grella, H. Maeomichi, ClouT: Leveraging Cloud Computing techniques for improving management of massive IoT data, in: Proceedings of the 7th IEEE International Conference on Service-Oriented Computing and Applications, IEEE Computer Society, Washington, DC, USA, 2014, pp. 324–327. doi:10.1109/SOCA.2014.47   | 1  |
| S14      | J. Gubbia, R. Buyya, S. Marusica, M. Palaniswami, Internet of Things (IoT): A vision, architectural elements, and future directions, Future Generation Computer Systems 29 (7) (2013) 1645–1660. doi:10.1016/j.future.2013.01.010   | 719  |
| S15      | M. M. Hassan, H. S. Albakr, H. Al-Dossari, A cloud-assisted Internet of Things framework for pervasive healthcare in smart city environment, in: Proceedings of the 1st International Workshop on Emerging Multimedia Applications and Services for Smart Cities, ACM, New York, NY, USA, 2014, pp. 9–13. doi:10.1145/2661704.2661707   | 1  |
| S16      | M. Henze, L. Hermerschmidt, D. Kerpen, R. Häußling, B. Rumpke, M. Wehrle, User-driven privacy enforcement for cloud-based services in the Internet of Things, in: Proceedings of the 2014 International Conference on Future Internet of Things and Cloud, IEEE Computer Society, Washington, DC, USA, 2014, pp. 191–196. doi:10.1109/FiCloud.2014.38   | 4  |
| S17      | S. Karnouskos, V. Somlev, Performance assessment of integration in the cloud of things via Web services, in: Proceedings of the 2013 IEEE International Conference on Industrial Technology, IEEE, USA, 2013, pp. 1988–1993. doi:10.1109/ICT.2013.6505983   | 8  |
| S18      | S. H. Kim, D. Kim, Multi-tenancy support with organization management in the Cloud of Things, in: Proceedings of the 2013 IEEE International Conference on Services Computing, IEEE Computer Society, Washington, DC, USA, 2013, pp. 232–239. doi:10.1109/SCC.2013.61   | 2  |
| S19      | F. Li, M. Vögler, M. Claeßens, S. Dustdar, Efficient and scalable IoT service delivery on cloud, in: Proceedings of the 6th International Conference on Cloud Computing, IEEE Computer Society, Washington, DC, USA, 2013, pp. 740–747. doi:10.1109/CLOUD.2013.64   | 15   |
| S20      | J. Mohammed, A. Thakral, A. F. Ocleanu, C. Jones, C.-H. Lung, A. Adler, Internet of Things: Remote patient monitoring using Web services and Cloud Computing, in: Proceedings of the 2014 IEEE International Conference on Internet of Things, Green Computing and Communications, and Cyber-Physical-Social Computing, IEEE Computer Society, Washington, DC, USA, 2014, pp. 256–263. doi:10.1109/iThings.2014.45  | 1  |
| S21      | S. Mosser, F. Fleurey, B. Morin, F. Chauvel, A. Solberg, I. Goutier, SENSAPP as a reference platform to support cloud experiments: From the Internet of Things to the Internet of Services, in: Proceedings of the 14th International Symposium on Symbolic and Numeric Algorithms for Scientific Computing, IEEE Computer Society, Washington, DC, USA, 2012, pp. 400–406. doi:10.1109/SYNASC.2012.71  | 20   |
| S22      | S. Nastic, S. Sehic, D.-H. Le, H.-L. Truong, S. Dustdar, Provisioning software-defined IoT cloud systems, in: Proceedings of the 2014 International Conference on Future Internet of Things and Cloud, IEEE Computer Society, Washington, DC, USA, 2014, pp. 288–295. doi:10.1109/FiCloud.2014.52   | 8  |
| S23      | S. Nastic, S. Sehic, M. Vögler, H.-L. Truong, S. Dustdar, PatRICIA – a novel programming model for IoT applications on cloud platforms, in: Proceedings of the 6th IEEE International Conference on Service-Oriented Computing and Applications, IEEE Computer Society, Washington, DC, USA, 2013, pp. 53–60. doi:10.1109/SOCA.2013.48  | 13   |

(continued on next page)

| Study ID | Reference   | Citation count in Dec. 2015 (Google scholar) |
|----------|---|--|
| S24      | A. Puliafito, SensorCloud: An integrated system for advanced multi-risk management, in: Proceedings of the 3rd IEEE Symposium on Network Cloud Computing and Applications, IEEE Computer Society, Washington, DC, USA, 2014, pp. 1–8. doi:10.1109/NCCA.2014.10  | 0  |
| S25      | P. Rao, B. B. P. Saluja, N. Sharma, A. Mittal, S. V. Sharma, Cloud Computing for Internet of Things & sensing based applications, in: Proceedings of the 6th International Conference on Sensing Technology, IEEE, USA, 2012, pp. 374–380. doi:10.1109/ICSenST.2012.6461705   | 31   |
| S26      | S. Rea, M. S. Aslam, D. Pesch, Serviceware - a service based management approach for WSN cloud infrastructures, in: Proceedings of the 2013 IEEE International Conference on Pervasive Computing and Communications Workshops, IEEE, USA, 2013, pp. 133–138. doi:10.1109/PerComW.2013.6529470   | 3  |
| S27      | D. Seo, C.-S. Jeong, Y.-B. Jeon, K.-H. Lee, Cloud infrastructure for ubiquitous M2M and IoT environment mobile application, Cluster Computing 18 (2) (2015) 599–608. doi:10.1007/s10586-014-0415-7  | 0  |
| S28      | J. Soldatos, M. Serrano, M. Hauswirth, Convergence of utility computing with the Internet-of-Things, in: Proceedings of the 6th International Conference on Innovative Mobile and Internet Services in Ubiquitous Computing, IEEE Computer Society, Washington, DC, USA, 2012, pp. 874–879. doi:10.1109/IMIS.2012.135   | 33   |
| S29      | M. Soliman, T. Abiodun, T. Hamouda, J. Zhou, C.-H. Lung, Smart home: Integrating Internet of Things with Web services and Cloud Computing, in: Proceedings of the 5th IEEE International Conference on Cloud Computing Technology and Science, Vol. 2, IEEE Computer Society, Washington, DC, USA, 2013, pp. 317–320. doi:10.1109/CloudCom.2013.155   | 16   |
| S30      | G. Suciu, A. Vulpe, S. Halunga, O. F. G. Todoran, V. Suciu, Smart cities built on resilient Cloud Computing and secure Internet of Things, in: Proceedings of the 19th International Conference on Control Systems and Computer Science, IEEE Computer Society, Washington, DC, USA, 2013, pp. 513–518. doi:10.1109/CSCS.2013.58  | 28   |
| S31      | K. Velusamy, D. Venkitaramanan, S. K. Vasudevan, P. Periasamy, B. Arumugam, Internet of Things in cloud, Journal of Engineering and Applied Sciences 8 (9) (2013) 304–313   | 0  |
| S32      | D. Vouyioukas, A. Moralis, M. Sardis, D. Drakoulis, G. Labropoulos, S. Kyriazakos, D. Dres, EPIKOUROS – Virtualized platforms using heterogeneous sensor services in Cloud Computing environment, in: Proceedings of the 4th International Conference on Wireless Communications, Vehicular Technology, Information Theory and Aerospace and Electronic Systems, IEEE, USA, 2014, pp. 1–5. doi:10.1109/VITAE.2014.6934461 | 1  |
| S33      | Z. Wu, T. Itälä, T. T. C. Zhang, Y. Ji, M. Hämmäläinen, Y. Liu, Gateway as a service: A Cloud Computing framework for Web of Things, in: Proceedings of the 19th International Conference on Telecommunications, IEEE, USA, 2012, pp. 1–6. doi:10.1109/ICTEL.2012.6221246   | 8  |
| S34      | J. Yan, S. Xin, Q. Liu, W. Xu, L. Yang, L. Fan, B. Chen, Q. Wang, Intelligent supply chain integration and management based on cloud of things, International Journal of Distributed Sensor Networks 2014. doi:10.1155/2014/624839  | 3  |
| S35      | J. Zhou, T. Leppanen, E. Harjula, M. Ylianttila, T. Ojala, C. Yu, H. Jin, CloudThings: A common architecture for integrating the Internet of Things with Cloud Computing, in: Proceedings of the 17th IEEE International Conference on Computer Supported Cooperative Work in Design, IEEE, USA, 2013, pp. 651–657. doi:10.1109/CSCWD.2013.6581037  | 32   |

## Appendix B. Distribution of the selected primary studies over venues

1088

| Acronym    | Venue name   | Type       | Studies              |
|------------|--|------------|----------------------|
| FiCloud    | International Conference on Future Internet of Things and Cloud  | Conference | S2, S5, S7, S16, S22 |
| IMIS       | International Conference on Innovative Mobile and Internet Services in Ubiquitous Computing  | Conference | S11, S28             |
| SOCA       | IEEE International Conference on Service-Oriented Computing and Applications   | Conference | S23                  |
| EMASC      | International Workshop on Emerging Multimedia Applications and Services for Smart Cities   | Workshop   | S15                  |
| -          | Sensors  | Journal    | S8                   |
| IDCS       | International Conference on Internet and Distributed Computing Systems   | Conference | S4                   |
| ICST       | International Conference on Sensing Technology   | Conference | S25                  |
| -          | Cluster Computing  | Journal    | S27                  |
| CSCWD      | International Conference on Computer Supported Cooperative Work in Design  | Conference | S35                  |
| GIIS       | Global Information Infrastructure Symposium  | Conference | S10                  |
| MobiHealth | International ICST Conference on Wireless Mobile Communication and Healthcare  | Conference | S12                  |
| CLOUD      | IEEE International Conference on Cloud Computing   | Conference | S19                  |
| VITAE      | International Conference on Wireless Communications, Vehicular Technology, Information Theory and Aerospace and Electronic Systems | Conference | S32                  |
| MCC        | ACM Mobile Cloud Computing Workshop  | Workshop   | S15                  |
| ICT        | International Conference on Telecommunications   | Conference | S33                  |
| -          | International Journal of Distributed Sensor Networks   | Journal    | S34                  |
| FGCS       | Future Generation Computer Systems   | Journal    | S14                  |
| -          | Journal of Engineering and Applied Sciences  | Journal    | S31                  |
| iThings    | IEEE International Conference on Internet of Things  | Conference | S20                  |
| EuCNC      | European Conference on Networks and Communications   | Conference | S1                   |
| SCC        | IEEE International Conference on Services Computing  | Conference | S18                  |
| ICIT       | IEEE International Conference on Industrial Technology   | Conference | S17                  |
| SYkASC     | International Symposium on Symbolic and Numeric Algorithms for Scientific Computing  | Conference | S21                  |
| NCCA       | IEEE Symposium on Network Cloud Computing and Applications   | Conference | S24                  |
| PerCom     | IEEE International Conference on Pervasive Computing and Communications Workshops  | Workshop   | S26                  |
| CSCS       | International Conference on Control Systems and Computer Science   | Conference | S30                  |
| ISSNIP     | International Conference on Intelligent Sensors, Sensor Networks and Information Processing  | Conference | S3                   |
| CloudCom   | International Conference on Cloud Computing Technology and Science   | Conference | S29                  |
| 3PGCIC     | International Conference on P2P, Parallel, Grid, Cloud and Internet Computing  | Conference | S9                   |



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