



Article

Light Node Communication Framework : a new way to communicate inside a Smart Home

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Abstract: The Internet of Thing has profoundly changed the way we imagine information science and architectures and Smart Homes are an important part of this domain. Created a decade ago, the few existing prototypes use the technologies of the day forcing designers to create centralized and costly architectures that raise some issues concerning the reliability, the scalability and ease of access which cannot be tolerated in an assistance context. In this paper we briefly introduce a new kind of architecture where the focus was placed on the distribution and especially. More specifically, we answer the first issue we met by proposing a lightweight and portable messaging protocol. After running several tests, we observed a maximized bandwidth, no packets were lost and a good encryption was obtained. These results tend to prove that our innovation may be employed in a real context of distribution on small entities.

Keywords: Messaging protocol; IoT; SmartHome; Distributed Computing

1. Introduction

The evolution of our society towards the all-digital of the Internet of Things (IoT) profoundly remodeled our relationship with the science of information. In this new one, the smart home became the subject of numerous researches [1–3] and joins the recent current of thought stemming from the Ambient Intelligence (Amb. I). This last one refers to a tendency that wants us to miniaturize a set of electronic devices (sensors and effectors) in order to integrate them into any object of everyday life (lamp, refrigerator, etc.) in a transparent way for the person. The aim behind this idea is to supply punctual assistance to the occupants according to the gathered information and to the history of the accumulated data.

The vast majority of work in the smart home domain focuses on the activity recognition problem in order to assist the inhabitant with a potential dementia often caused by an advanced age[4–6]. Nevertheless, none of them seems to propose a standard architecture which provides both high-reliability and scalability capabilities at a relative low-cost. And still, high-reliability has to be a mandatory feature of such architecture since the assistance is vital for the inhabitant with a potential dementia. Moreover, as the disease can stay for decades, any work on architecture must take the scalability parameter into account since many sensors or improvements can be realized during the illness evolution. Finally, the low-cost aspect has to be taken into account as the vast majority of the aging population will be located in poor or developing countries by 2050 [7]. As far as we know, this paper is the first focusing on those three points in particular.

Here, we briefly introduce a new kind of smart home architecture providing both reliability and scalability based on low-cost smart sensors. To achieve this objective, the first issue we ran into is the difference between all the possible entities in the environment. In fact, our solution has to integrate

34 different operating systems (e.g. Linux or FreeRTOS) running on different hardware (e.g. computer or
35 microcontroller) and using different communication technologies (e.g. Wi-Fi, ZigBee or 6LowPan). To
36 answer these dissimilarities, we have to use a highly portable communication protocol using a broker
37 less architecture to provide the highest reliability. This point will be the main concern of this paper.
38 Even if many protocols exist like MQTT, RabbitMQ or ZeroMQ [8–10] none of them fully answer our
39 requirements since the first two require brokers to work and the last one is based on POSIX sockets
40 and cannot be embedded in some light systems. Consequently, the contribution we make in this paper
41 is a new way to communicate that can be embedded in every system as soon as they implement an
42 IP stack. Our solution provides discovery mechanism, security via AES encryption and two different
43 channels in order to address the difference between configuration messages and data messages.

44 This paper is divided in four sections. The first one will present a state of the art about existing
45 smart homes and their architectures. The second part will cover the technological breakthroughs that
46 the embedded computing has experienced since the creation of the first Smart Homes. Then, the
47 proposed solution will be explained and some tests on the messaging protocol will be presented in
48 the third section. Finally, a conclusion and some future works will end this paper.

49 2. Existing architectures

50 Many smart habitations have been implemented in laboratories since the creation of the ambient
51 intelligence [1,3,11,12]. Each of these projects use the technology of its day to create a testing
52 environment in which the data accessibility was the main challenge. Here, we depict three of those
53 starting with the LIARA and DOMUS, which share the same architecture [12]. We continue this
54 review with the Gator Tech house [3] and CASAS [1]. Finally, we end this part by describing Software
55 Defined Smart Home [13,14], a recently released architecture based on software defined networks.

56 2.1. LIARA and DOMUS

57 LIARA and DOMUS laboratories aim to study how smart homes can assist people with cognitive
58 deficiencies. They both created a very similar architecture, represented in Figure 1, to test their
59 algorithms and solutions [15]. Inherited from the industry, they use some islands, made of industrial
60 grade hardware, to agglomerate transducers. Then, an automate is in charge of getting back, from
61 the islands, the values of the sensors or changing the values of the effectors. To end this process,
62 the automate will update a relational database hosted on a SQL Server in order to provide a simple
63 interface for other systems like, in this case, an artificial intelligence.

64 These two smart homes present some interesting features we have to discuss. First, the use of
65 industrial grade hardware means that all the components have been tested for a continuous use in
66 a far much harder environment than just a house (e.g. production line in a factory). Therefore it
67 demonstrates an excellent reliability even if the smart home has to operate at all times. The second
68 main advantage of these environments comes from the highly centralized architecture itself. Indeed,
69 all the values coming from sensors and all the actuator controls end up in the same database. As a
70 result, it facilitates interaction with the home since this kind of storage offers an easy way to retrieve
71 sensors values or interact with actuators.

72 Nevertheless, industrial material suffers from two main drawbacks. The first one is the
73 introduction of black boxes in a research environment. As a matter of fact, the communications
74 between all these pieces of hardware often rely on proprietary libraries that can impact future
75 evolution. The second main disadvantage is the price of such an architecture. Based on the hardware
76 presented by Bouchard *et al.* and the price of it, we were able to compute the total price of the chain
77 Island-Automate-Main Server. With 2000 dollars each island [16], 1500 dollars the automate [16] and
78 4000 the server [17], the architecture reaches 13,500 dollars without any transducers or backbone
79 structure (e.g. networking, cooling for the server, maintenance). Finally, the highly centralized
80 architecture presented here creates many Single Points of Failure (SPoFs) like the automate, the
81 Islands and the main server hosting both the AI and the SQL server. So if one of these SPoFs fails,

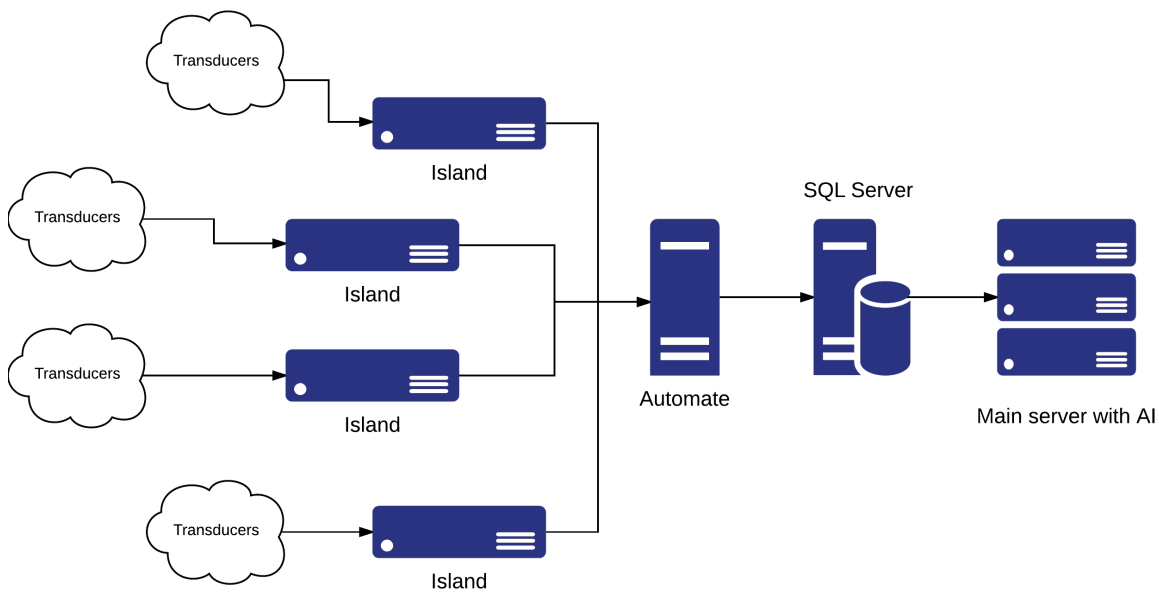


Figure 1. The LIARA and DOMUS architecture.

82 at least a quarter of the environment and its assistance will fail too. Moreover, these points represent
 83 some serious bottlenecks in the architecture preventing a real scalability.

84 2.2. Gator Tech

85 Gator Tech [3] is a project funded in Florida. Its main goal is to prove the feasibility of a low-cost
 86 smart home where the integration of new transducers will be easy. In order to accomplish that the
 87 authors present an OSGI [18] based architecture sums up in Figure 2. In this last one, each transducer
 88 has a simple EEPROM memory containing the driver to communicate with it. Once powered, the
 89 transducer registers itself by sending its driver to an OSGI service definition. This last one will act
 90 as an abstraction layer to create basic services that allow the consumption of highly abstracted data
 91 (e.g. "Sunny" instead of 10 000 lumen for a light sensor) or the combination of basic services to a
 92 composite service (e.g. create a voice recognition service on all the different microphone services). All
 93 this architecture allows developers to create applications without any knowledge of the underlying
 94 communication and with only highly abstracted data which simplify the development.

95 This environment has some really good advantages. First of all, the automatic transducer
 96 registration really helps the scalability of such an environment (e.g. add new sensors or replace
 97 some of them). Secondly, the high abstraction of the data generated by this system greatly helps
 98 the application development. For example, it is straightforward to enable the air conditioning when
 99 the temperature is "Hot." However, it is more complicated when the decision is only based on the
 100 microcontroller value since this one depends on the hardware (e.g. the microcontroller itself, the
 101 temperature sensor or even the analog to digital chips). Finally, the price of such infrastructure is as
 102 low as possible as every transducer is designed to be the most affordable possible by using Atmega128
 103 as the main processor unit which is a low-cost platform [19]. Moreover, because every transducer is
 104 wireless, there is no need of Islands or Automate as in LIARA and Domus homes. Despite all these
 105 great advantages, the use of OSGI on a unique server create a SPoF which is a big problem in a high
 106 reliability architecture.

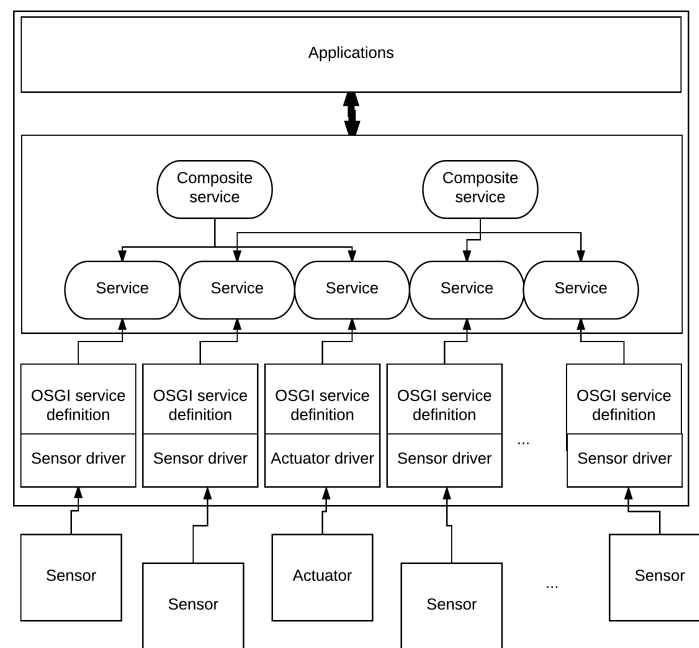


Figure 2. The Gator Tech architecture.

107 2.3. CASAS

108 CASAS [1], or the Smart Home in a box, is an infrastructure where the accent was put on the price
 109 and the ease of installation. Depicted in Figure 3, the architecture is divided in four main elements.
 110 The first one is the ZigBee mesh which represents all the transducers communicating between them in
 111 a network where every node relay the information to its neighbors by using the ZigBee protocol. This
 112 mesh sends events on a Publish/Subscribe (Pub/Sub) messaging service through a ZigBee bridge
 113 in charge of converting events to higher-level XMPP messages. The messaging service allows other
 114 applications to easily integrate the infrastructure and use the transducers. By default, there are two
 115 services which are the Storage and the Intelligence. The first archives all the events occurring in the
 116 environment by using the Scribe Bridge. As for the intelligence, it is in charge of energy monitoring
 117 and the discovery and recognition of any activity that can happen in the house.

118 CASAS has two main benefits : the price and the ease of installation. For the first, the authors
 119 present a detailed summary of the cost. They state that their solution cost only 2,765 dollars which
 120 is really a great achievement. Regarding the ease of installation, they demonstrate it by conducting a
 121 test on people aged from 21 to 62 and it requires only an hour to set up the whole environment. In
 122 spite of these qualities, CASAS, as the other architecture, suffer from the existence of many SPoFs like
 123 the ZigBee bridge or the Application Bridge which can stop the assistance or the Pub/Sub-messaging
 124 service which is a sensitive component.

125 2.4. Software Defined Smart Home

126 The works previously described are the old founders of the Smart Home architecture. However,
 127 some more recent papers exist in this particular domain [13,14]. One of them introduce the idea
 128 of Software Defined Smart Home or SDSH for short [13]. This concept is a new way to integrate
 129 heterogeneous smart appliances (e.g. Smart Light, Smart Flowerpot, etc.) in one homogeneous
 130 platform creating a Smart Home from the chaos of the different hardwares and communication
 131 protocols implemented by the companies who create these devices. To achieve such a goal, the
 132 authors propose a three layers architecture derived from software defined networks [20]. First, the
 133 Smart Devices layer includes all the different kinds of smart hardware in a home (a.k.a. the smart

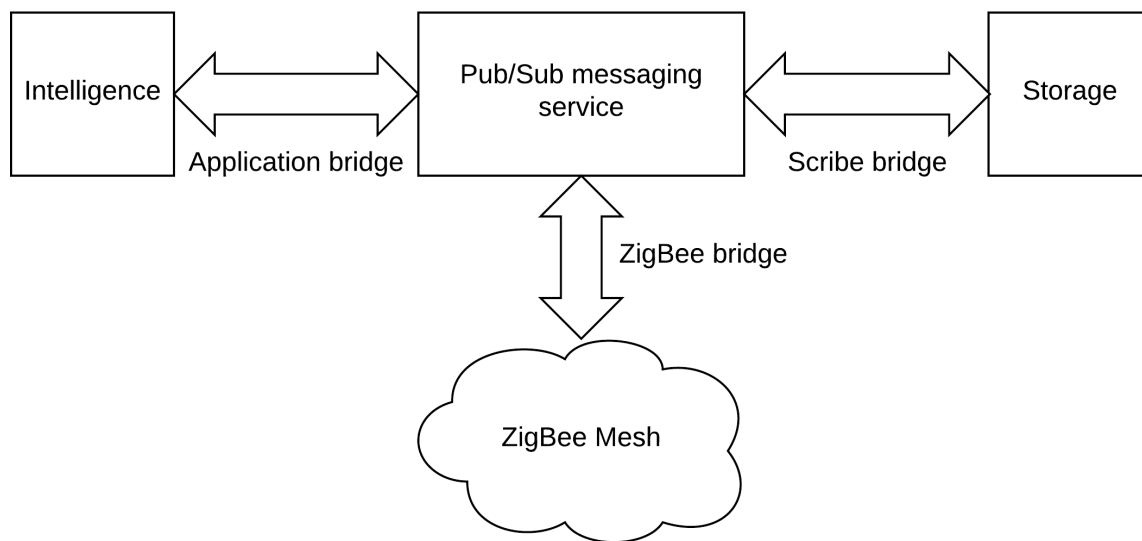


Figure 3. The CASAS architecture.

134 appliances). Next, the controller layer is a centralized management service locally implemented or
 135 deployed in the cloud. Its main goal is to hide the implementation details of the hardware layer,
 136 retrieve and analyze the user demands and manage the whole smart home. Moreover, it is in charge
 137 of encapsulates information extracted from the smart home and provide them to the last layer : the
 138 external service layer. This last one uses the smart home resources to provide some smart services
 139 like home security or medical attention.

140 SDSH offers some great features. First of all, as Gator Tech, this architecture offers a strong
 141 separation between raw sensors and final services via its controller layer. Next, it uses OpenFlow
 142 [21] as its main protocol which is a well-known protocol widely implemented in software defined
 143 network. Finally, it uses smart appliances already in the market and standardize the access to their
 144 data. Unfortunately, the centralized controller depicted as one of the main advantages of this work
 145 is also a great default because, if it allows to configure the whole Smart Home at the same place,
 146 it represents a severe single point of failure. To answer this problematic, the authors introduced
 147 visualization techniques but these one require either an Internet cloud connection (which cannot be
 148 tolerated in some applications) or a server strong enough to deal with many systems started on the
 149 same hardware which can be very expensive for a single house.

150 2.5. Conclusion on existing architectures

151 All these existing architectures have some common points. First the majority of their components
 152 are transducers. Moreover, according to the Gator Tech and CASAS cases, it seems that embed
 153 some intelligence and communication abilities in them helps to reduce costs and ease installation
 154 and scalability. Finally, we have to point out the common problem in all these architectures: the
 155 centralization. This weakness creates single points of failure which can lead to a complete stop of the
 156 assistance. In corporate computing and Web domains, this particular issue has been solved ten years
 157 ago by using redundancy, clusters and distributed computing [22–25].

158 The ideal architecture appears to be composed of many smart transducers easing the scalability
 159 of such architecture. This specific attribute bring our environment closer to another computer science
 160 domain which is the Internet of Things (IoT). In this last one, already used in the Smart Home [26,27],
 161 a multitude of smart objects communicate in a uniform manner and generates a huge amount of data
 162 often associate to "Big Data" [28]. In this last case, it is not conceivable to handle the information in a
 163 centralized way any more, even if we use server clusters. It is more appropriate to use decentralized

164 methods relocating the intelligence as close as possible to the units composing this huge data pool
165 [28,29].

166 3. Technological breakthroughs

167 The transducers (i.e. sensors and actuators) are the essential basis of every Smart Home
168 architecture. CASAS and Gator Tech case studies proved that embed intelligence and communication
169 in these entities allow to reduce costs while improving the ease of implementation and the scalability.
170 In this part, we are going to study the concept of a smart transducer such as designed by the
171 standards. Then, we will review some hardware evolution realized since the creation of the first
172 Smart Homes.

173 A smart transducer is clearly defined in the IEEE 1451.2 standard [30]. To sum up, it is an entity
174 providing more features than the one's mandatory to generate a good representation of the controlled
175 quantity. Some of these attributes can be sensor identification, a process to simplify the installation or
176 the maintenance, network interfaces or the coordination and synchronization with other entities [31].
177 In order to guide the community, Lewis [31] proposed three objectives for these transducers. The first
178 one is to move the intelligence closest to the sensing point. The second one is to make the installation
179 easier, and the maintenance of massive distributed sensor networks less expensive. The last one is to
180 facilitate the interfacing of many different sensors. Now that the concept of a smart transducer is well
181 defined, we can work at the different technological breakthroughs that occur during the last ten years
182 and allow us to finally design and build inexpensive smart sensors for the smart environments.

183 Many prototypes of smart environment have emerged during the last decade (e.g. Gator Tech in
184 2005 or LIARA/DOMUS in 2009 [3,12]). They have been built on top of existing technologies, which
185 for the most part are anterior to great innovations made recently. One of these is the apparition and
186 especially the democratization of System on Chip (SoC) which are full systems integrated on a single
187 substrate providing all the elements to run an application (e.g. processor, memory, radio). The SoCs
188 are the cutting edge of the modern electronic and can be found everywhere from smart sensors to
189 nano-computers and drive the price and power consumption reduction in all the modern devices
190 [32].

191 To illustrate the growth in power and integration, we propose to make a quick comparison
192 between two microcontrollers platforms and some nano computers. The first two are the Arduino
193 USB, easily accessible at the time of the creation of the first smart homes, and the latest released
194 ESP 32 from the Espressif company. Concerning the nano computers, we chose the evolution of
195 the Raspberry Pi since its creation. The attributes we compare are the released year, the processor
196 frequency, the memory available, the connectivity, the relative size and the price. Tables 1 and 2
197 both represents the different values for the attributes retain respectively for the micro controllers and
198 the Raspberry Pi. The first thing that jump out from these tables is the increase in both processor
199 frequency and memory with 16 MHz and 1 kB of RAM for the Arduino USB to a dual core 240 MHz
200 with 512 kB of RAM for the ESP 32. And the phenomenon is the same for the different Raspberry Pi
201 with 700 MHz and 512 MB of RAM for the Pi1 to 1 GHz and the same amount of memory but with
202 half the size of the Pi Zero W and four cores at 1.2GHz with 1 GB of RAM for the Pi3 but with the
203 same form factor. Moreover, it is pretty obvious that the embedded connectivity became a must in
204 this period with the integration of Wi-Fi and both Bluetooth and BLE on the ESP 32 and Wi-Fi/BLE
205 for the Pi Zero W and Pi3. Finally, it must be noted that the price of these platform stay the same of
206 decrease drastically even if the platforms increase in power and connectivity.

207 Technological evolution since the beginning of the 2000s was impressive. The democratization
208 of the SoC permits an increase of power for such piece of hardware while reducing costs and power
209 consumption. In parallel, SoC integrate much more advanced features like Wi-Fi and Bluetooth
210 communication. Subsequently, it seems now possible to create powerful applications on embedded
211 hardware and one of these applications is the creation of more intelligent transducers as depicted in
212 the IEEE 1451 standard.

Table 1. Arduino USB and ESP 32 comparison

	Arduino USB	ESP32 Thing
Released Year	2005	2016
Processor Frequency	16 MHz	2 x 240 Mhz
Memory	1 kB	512 kB
Connectivity	None	Bluetooth + BLE and WiFi
Relative size	1	0.5
Price (USD)	35	7

Table 2. Raspberry Pi platform over time

	Pi 1	Pi Zero W	Pi 3
Released Year	2012	2017	2016
Processor Frequency	700MHz	1 GHz	4 x 1.2GHz
Memory	512 MB	512 MB	1 GB
Connectivity	None	BLE/WiFi	BLE/WiFi
Relative size	1	0.5	1
Price (USD)	35	9	35

213 4. Proposed solution

214 We saw that existing smart homes had some weaknesses in both reliability and scalability. Yet
 215 these kinds of weak points are not bearable in the assistance domain. Here, we propose a new kind
 216 of architecture using the latest technological advances to provide a reliable and scalable distributed
 217 environment to safely run the assistance.

218 The main concept behind our solution is that the only non-removable elements of a smart
 219 environment are the transducers themselves. And if we think about it, they represent a vast number
 220 of distributed entities. With the latest hardware innovations it is feasible to equip each of them with
 221 an intelligent entity with both communication and processing capabilities creating a huge network
 222 with highly distributed computation potential and no single point of failure. In this vision, the
 223 generic smart entities have to answer the three main objectives firstly formulated by Lewis [31]. It
 224 means that they must allow to move the artificial intelligence to the closest sensing point, provide
 225 methods to easily install, configure and maintain this smart network and finally ease the interfacing
 226 between many different sensors. The first issue that such an architecture has to deal with is the
 227 difference between all the intelligent entities we can use. Indeed, if we want our solution to be
 228 the most generic possible we have to cope with the most different hardware and operating systems.
 229 Thus, we want to make feasible the integration of sensors based on different operating systems (e.g.
 230 Linux or FreeRTOS) implemented on different hardware but also to be able to interface different
 231 communication protocols (e.g. ZigBee, Wi-Fi or BLE). In order to answer this problem, we had to
 232 think of a new way to communicate between all these entities.

233 4.1. Communication protocol

234 There are many ways to communicate by using messages in the literature or industry. As far
 235 as we know the most popular ones are MQTT, RabbitMQ and ZeroMQ [8,9,33]. The first of them is
 236 mainly used in the Internet of Thing application by its high portability and its reduce footprint in
 237 terms of memory and power. It's a publish/subscribe protocol where clients connect to a centralized
 238 instance named broker. It supports different type of quality of service which affects the reliability
 239 of communication (message is delivered at most once, at least once or exactly once). Finally, it can
 240 support a "Last will and testament" (LWT) which allows to send a specific message on a specific
 241 subject when the entity disconnect in an abnormal way from the network. RabbitMQ, on the other
 242 hand, is a leading messaging protocol mainly use in distributed architectures. It implements the

243 Advanced Message Queuing Protocol (AMQP) and consequently has a broker architecture which
244 provides ease of development in favor of scalability and speed since the broker adds latency and
245 treatment and the message exchanged are pretty big. Finally, ZeroMQ is a messaging system which
246 allows developers to create themselves the architecture including brokerless ones. The main problem
247 with this approach is the portability since ZeroMQ relies on POSIX sockets which are only supported
248 in Unix and Windows operating systems. To conclude, none of the leading messaging protocol fit
249 our application since we want one without a centralized unit like a broker and heavily portable in
250 order to deal with the most part of the possible entities in a Smart Home which can be composed of
251 embedded systems running on top of different Real Time Operating Systems (RTOS) like FreeRTOS
252 or RiotOS.

253 The contribution we make in this paper is a new communication protocol with two main
254 characteristics. First, it can be embedded on any device from computers to microcontrollers as long
255 as they implement an IP stack (over Wi-Fi, 6LowPan or ZigBee IP). Second, our protocol does not
256 have the need for any main server also known as a broker. This last point was an issue in the most
257 popular solutions (e.g. MQTT, NATS, etc.). To build our solution, we made two basic assumptions.
258 The first one is that all our messages will stay in the smart home network. The second is that UDP is
259 the minimum requirement for any device that wants to communicate over a network as it is the base
260 of many network configuration protocols (e.g. DHCP or DNS).

261 One of the first issues we ran into is the fundamental difference between configuration and
262 data streams. The first one has to be based on a reliable delivery system allowing point to point
263 communication without the urge of the highest data speed. The second one, have to be able to stream
264 a huge quantity of information in a minimum of time without the highest reliability to many different
265 listeners. In order to answer this problematic, we propose to use two different channels like the FTP
266 protocol [34]. We will now explain how these two channels work in order to offer all the features we
267 want.

268 4.1.1. Configuration channel

269
270 As said sooner, the configuration of a smart entity has to be distributed over a reliable
271 communication. In order to achieve this objective, we propose to use CoAP, a well-known IoT
272 protocol, already implemented in many platforms [35]. It allows us to use HTTP-like request to get
273 or change values represented by URI in a fail-safe manner based on an acknowledgment system for
274 important messages (i.e. messages with high reliability). We propose to use this URI representation
275 for the entity configuration. In order to facilitate the understanding of such a concept, we present a
276 simple example in Table 3. In this table, each line represents a possible configuration variable with
277 its CoAP URI, the methods allowed in order to get or set the information and the type of data that
278 is asked by the entity. The first one is pretty obvious and represents the update frequency used by
279 a potential sensor. It is a simple integer, represents by the URI /rate and that can be obtained or
280 modified by using respectively GET or POST request. The usage of a HTTP-like protocol allows us
281 to define read-only values like the version which is a string that is only reachable via a GET request
282 on the /version URI. Moreover, we can exchange much more complex data types like JSON to clearly
283 define hardware configurations and interaction. Another interesting feature of CoAP that we use
284 in our configuration sample is the block-wise extension of the protocol that permit the transfer of
285 large binary file like updates for the embedded software. Finally, the last feature of CoAP we use
286 in this configuration channel is the ability to encrypt the communication by the usage of DTLS. To
287 demonstrate the utility of such a feature we propose to change symmetric encryption keys on the
288 device. This kind of operation is critical since it has to be highly secured in order to guarantee the fact
289 that nobody can intercept these keys to listen and speak over a secure network. With our method, we
290 simply exchange keys by using CoAP protected by SSL which guarantees both confidentiality and

Table 3. A configuration example based on CoAP

	URI	Methods allowed	Data type
Update rate	/rate	GET/POST	Integer
Version	/version	GET	String
Hardware	/hardware	GET/POST	JSON
Update	/update	POST	Binary
Encryption key	/keys	POST	Binary

291 authentication. Now that the configuration channel operation is explained, we can think about how
 292 to exchange information through the data channel.

293 4.1.2. Data channel

294

295 In addition to a reliable communication channel to ensure the good configuration of any
 296 device, we have to provide a method to exchange data messages between the smart entities in our
 297 architecture. We want to be able to transfer huge data in a minimal amount of time, to discover smart
 298 sensors connected to the network and to give the ability to encrypt sensitive information. To attain
 299 these aims, we propose a simple publish/subscribe messaging protocol based on UDP multicast
 300 chosen for its ability to transfer to one or many listeners at once in addition to its low-latency. To
 301 sum up the protocol work-flow, users join the multicast group, they register to any topic, represented
 302 by a simple character string, and will receive any messages labeled by this topic. They may also ask
 303 for any sensors connected to the network group by sending a discovery packet with a request in it
 304 (e.g. sensors with service "temperature") and those sensors will answer with their IP address and the
 305 different topics they expose. We are now going to explain the three different modes of our messaging
 306 protocol which are data, discovery and encrypted data.

307 4.1.3. Data message

308

309 Figure 4a represents a single and unencrypted data message in our solution. Every packet begins
 310 with an options byte, divided in two parts. The first three most significant bits (MSB) compose the
 311 version number of this packet and is all set to 0 for the version we present here. The other bits
 312 are reserved and unused except for the two least significant bits (LSB) who represent flags used in
 313 discovery and encrypted mode and have to be set to 0. These options are followed by the topic length
 314 which can go from 1 to 255 characters encoded on a single byte. Any message without topic has to
 315 be considered as an error and forget. Next come the topic with a variable length defined previously
 316 followed by the data length encoded on two bytes (the protocol tolerate empty data packages) and
 317 the data associated. Finally, a checksum is computed with the whole packet by using the CRC-32
 318 algorithm already used in the Ethernet frame which is a fast and lightweight hash algorithm. It has
 319 to be noted that the maximum size of one packet is limited by the theoretical limit of UDP which is
 320 65,535 bytes. When a client receives a packet, the first operation he has to do with it is the checksum
 321 validation. If this fails, the packet has to be dropped as the protocol does not have any mechanisms
 322 to send the packet again. Otherwise, the packet can be split by using the different sizes and the data
 323 and topic can be easily read.

324 4.1.4. Discovery message

325

326 A really interesting feature we have to provide is a mechanism to discover entities in the network.
 327 To achieve this, each of them can register custom key-value pairs in addition to any readable (i.e.
 328 allowing GET method) configuration values that will be exposed to any discovery request. Figure 4b

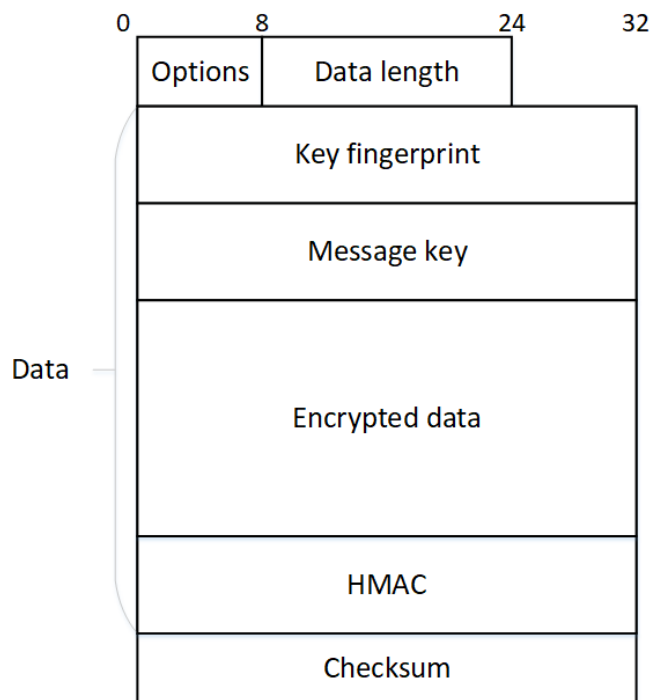
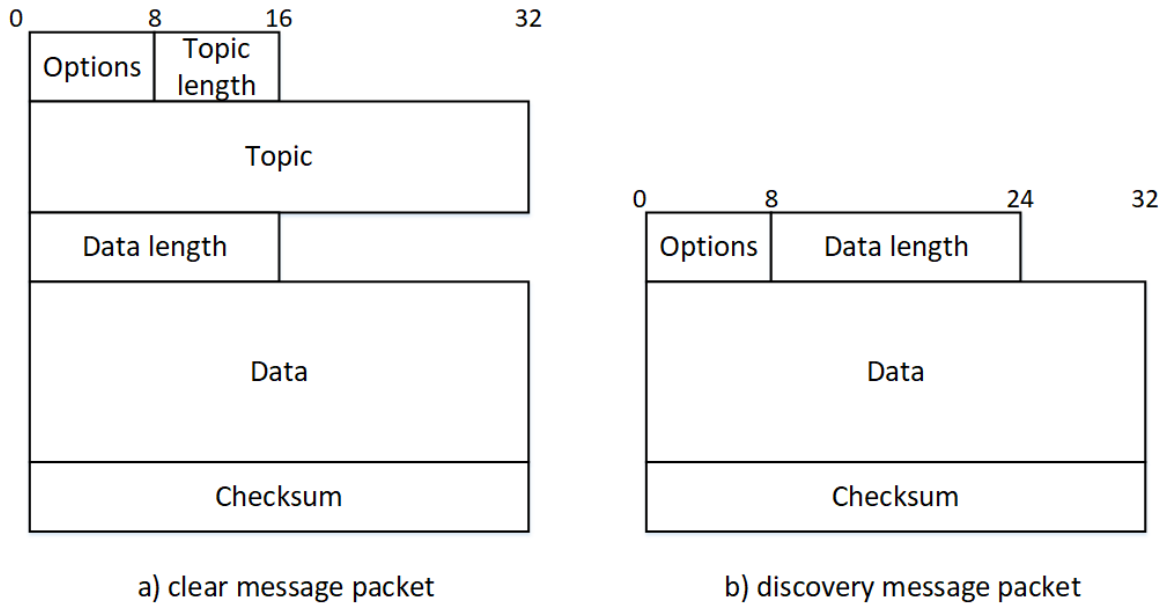


Figure 4. Representation of the three packets present in our protocol.

Table 4. An example of the discovery decision process

Key	Requested	Exposed	Conclusion
version	"1.0.1"	"1.0.1"	OK
name	"temp2"	"temp32"	NOK
units	"C"	["C", "F"]	OK
units	"K"	["C", "F"]	NOK
units	["C", "F"]	"C"	OK
units	["C", "F"]	"K"	NOK
units	["C", "F"]	["C", "K"]	OK
units	["C", "F"]	["K", "R"]	NOK

329 represents a single discovery packet. As the data packet, the options come first. The only difference is
 330 the LSB of this particular byte which is set to 1 representing a discovery packet. Next come the length
 331 of the data containing the request, a simple key-value data structure represented in JSON format.
 332 For security reasons, data has to be set in order to limit the number of answers. Consequently, any
 333 discovery packets with a zero-length data have to be ignored and consider as an error. When the
 334 frame is received, receivers will have to decide if they have to answer or not. To achieve this, they
 335 check every key in the request if they do not have one of them they can drop the packet they must
 336 not answer. Regarding the associated values, two cases are possible : either the key contains a single
 337 value or an array of values. Consequently, four situations can occur depending on the combination of
 338 two different data type on the receiver and the sender. Examples of each case are presented in Table
 339 4. The first case shown here is when both requested and exposed are single value. In this case, the
 340 two values have to be exactly the same as presented in the first two lines of the table. Next, come the
 341 case where one has an array and the other a single value here, the single value has to be in the array
 342 like in the lines 3 to 6 in the Table 4. Finally, when both have arrays as values, at least one value in the
 343 requested array has to be present in the exposed one.

344 4.1.5. Encrypted data message

345

346 As we deal with sensitive information inside a smart environment, our protocol has to provide
 347 an easy way to secure the communications. We adapt a well-known secure chat found in the Telegram
 348 [36] application to achieve this goal.

349 The encryption process starts by adding random padding to the message. Indeed, Advanced
 350 Encryption Standard (AES), the encryption algorithm we use, is a block cipher so our data have to
 351 be a multiple of the block size B_s . The number of random bytes P_s we have to add is defined by
 352 the formula 1. We always add B_s bytes in order to always put some randomness in the original
 353 message then, we add enough bytes to be a multiple of B_s (16 in the case of AES). In formula 1 we
 354 use $DataLength + 2$ because the last step in the packet preparation is the prepending of the data size
 355 a two-byte long number.

$$P_s = B_s + (B_s - (DataLength + 2) \% B_s) \quad (1)$$

356 When the packet is ready to be encrypted, we create the message key used to generate the AES
 357 key and IV by computing the SHA1 of the packet to encrypt. This hash is given to a Key Derivation
 358 Function (KDF) with the Secret Key preconfigured by using the configuration channel. The KDF
 359 presented in algorithm 1 is a sequence of different SHA1 hash and will result in two values, the
 360 128-bit AES Key and AES IV used to encrypt the packet with AES. Next, in order to identify the secret
 361 key used to encrypt, we compute a unique fingerprint of it by using the Base64 representation of the
 362 key SHA256. Finally, the hash message authentication code (HMAC) is computed by using the secret
 363 key with the SHA1 hashing algorithm.

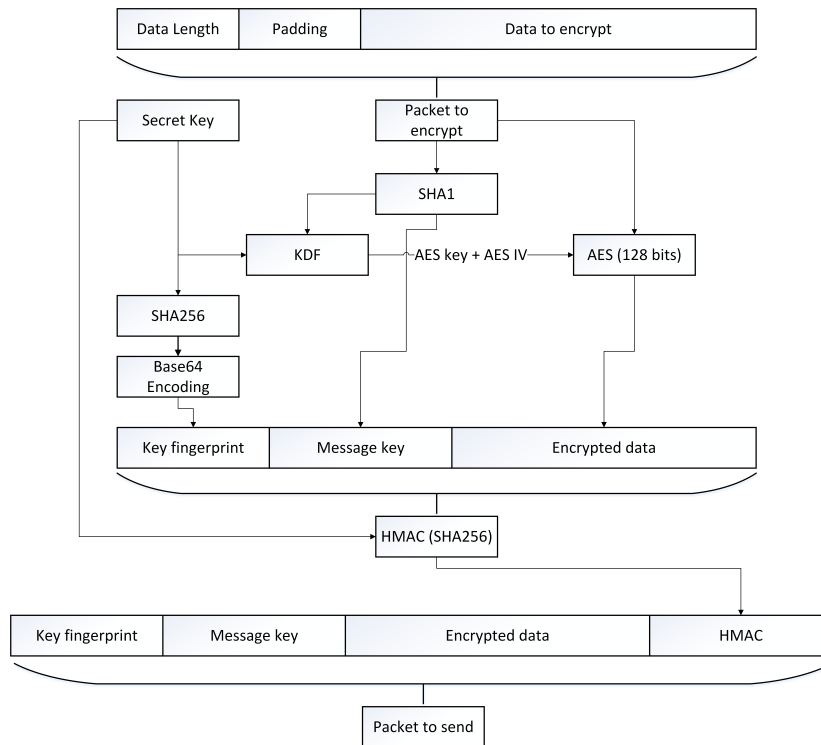


Figure 5. The encryption process.

Data: secret_key[16], msg_key[20]

Result: aes_key[16], aes_iv[16]

sha1_a = sha1(msg_key + secret_key[0...3]);

sha1_b = sha1(secret_key[4...5] + msg_key + secret_key[6...7]);

sha1_c = sha1(secret_key[8...11] + msg_key);

sha1_d = sha1(msg_key + secret_key[12...15]);

aes_key = sha1_a[0...3] + sha1_b[0...7] + sha1_c[4...7];

aes_iv = sha1_a[12...15] + sha1_b[12...19] + sha1_d[0...3];

Algorithm 1: The Key Derivation Function

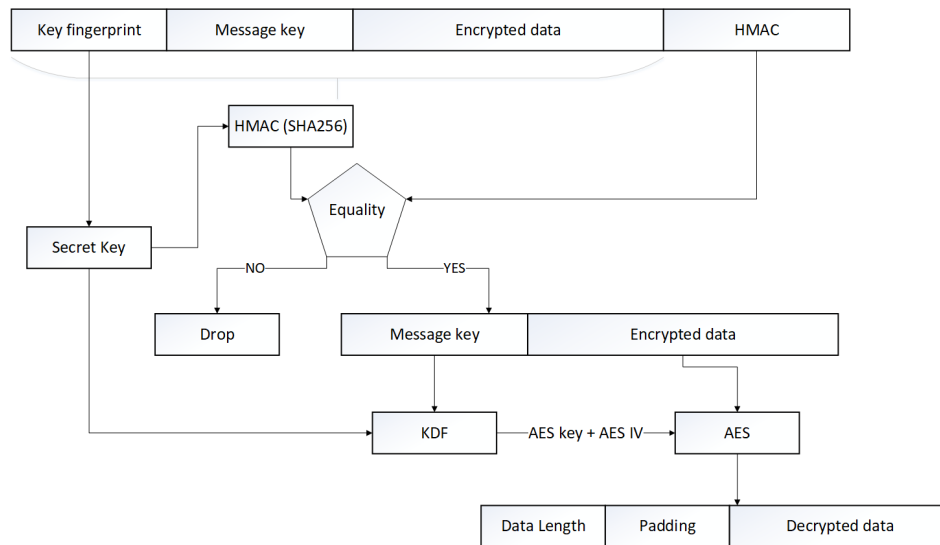


Figure 6. The decryption process.

364 In order to decrypt an encrypted packet (identified by a 1 at position 1 in the option byte of the
 365 header) the receiver has to do some operations sum up in Figure 6. The first task to accomplish is to
 366 verify the HMAC to ensure the authentication and the integrity of the packet. To accomplish this, we
 367 first have to retrieve the preconfigured secret key used to encrypt identified by its fingerprint. Next,
 368 we can compute the HMAC of the packet (without the sender HMAC) and check for equality. If
 369 the two hash message authentication codes are different, the packet must be dropped otherwise, the
 370 decryption process can begin. First, we have to reconstruct the AES key and IV used to encrypt the
 371 message by passing the preconfigured secret key and the message key received in the KDF algorithm
 372 presented earlier. Next, we can use AES 128 algorithm in decrypt mode with the generated key and
 373 IV to finally decrypt the whole message. The final step is to remove the padding bytes which can be
 374 easily done with the message size store in the first two bytes.

375 4.2. Tests and discussion

376 In order to validate the proposed protocol, we have made three tests on it. The first one is about
 377 bandwidth and try to maximize the number of messages per second exchanged to demonstrate the
 378 speed of our protocol. The second one answers a question about the UDP protocol. Indeed, UDP does
 379 not provide safety mechanisms about lost, corrupted or disordered network packet. Consequently we
 380 decide to show the number of packets we did not receive because of this lack. Finally, we want to
 381 demonstrate the fact that even with the same message and secret key, our encrypted packet is always
 382 totally different in order to prevent semantic attack since our Initialization Vector is not randomly
 383 generated. To accomplish that, we will compute a similarity measure between encrypted packets
 384 containing the same message with the same encryption key. The hardware used in our tests was a
 385 laptop (MSI GT62VR), a Raspberry Pi 3 and a Raspberry Pi Zero W. The first one was connected to a
 386 Gigabit wireless router (LinkSys WRT1900AC) through its Gigabit wired and wireless (AC Wi-Fi) card
 387 and the other ones over a simple Wi-Fi connection. Concerning the implementation of our solution,
 388 we used C++ with the libraries Boost ASIO (for the network) and Crypto++ (for the encryption
 389 algorithms).

390 In order to test the bandwidth capabilities of our protocol, we first generate 9 random messages
 391 with different sizes from 16 bytes to 60 kibibytes (kiB). Then, we bound a listening process on the
 392 wired network card on the laptop in order to monitor the packets transiting through the network
 393 while we use wireless connections to send 20 000 packets for each different sizes with 10 000 encrypted
 394 and another 10 000 not. Results from this test are summed up in Figures 7 to 9 where the upper plot

395 represents the messages per second (Msg/s) and the lower one the data rate in mebibytes per second
 396 (MiB/s) transmitted by our protocol both depending on the message size. Figure 7, 8, 9 respectively
 397 show the laptop, Raspberry Pi 3 and Raspberry Pi Zero W results. The first thing we can note is the
 398 important drop in both data rates and message per second occurring for packet larger than 2kiB on
 399 every platform and for both encrypted and non-encrypted messages. With respectively 10.9, 13.1, 3.8,
 400 4.6, 2.2 and 2.78 times fewer messages sent depending on the platform and encryption, our solution
 401 seems to present a limit concerning high-frequency large messages (greater than 1000 per second). We
 402 investigate the reason for such a decreasing in performances and it seems to be fragmented Ethernet
 403 packets that appear when the length of the transported data is greater than the maximum data size
 404 of an Ethernet packet which is 1522 bytes. The second observation is pretty obvious and is the fact
 405 that message rate and data rate decreased when using encryption. This is due to the computation
 406 of different cryptographic algorithms like AES or SHA but we can say that for a packet under the
 407 2kiB limit the encryption process does not impact too much our protocol. Finally, we can say that our
 408 protocol does not overload the Raspberry Pi Zero W, the smallest platform, since both encrypted and
 409 clear tests give nearly the same results in terms of data rates which can be explained by a fully used
 410 network adapter.

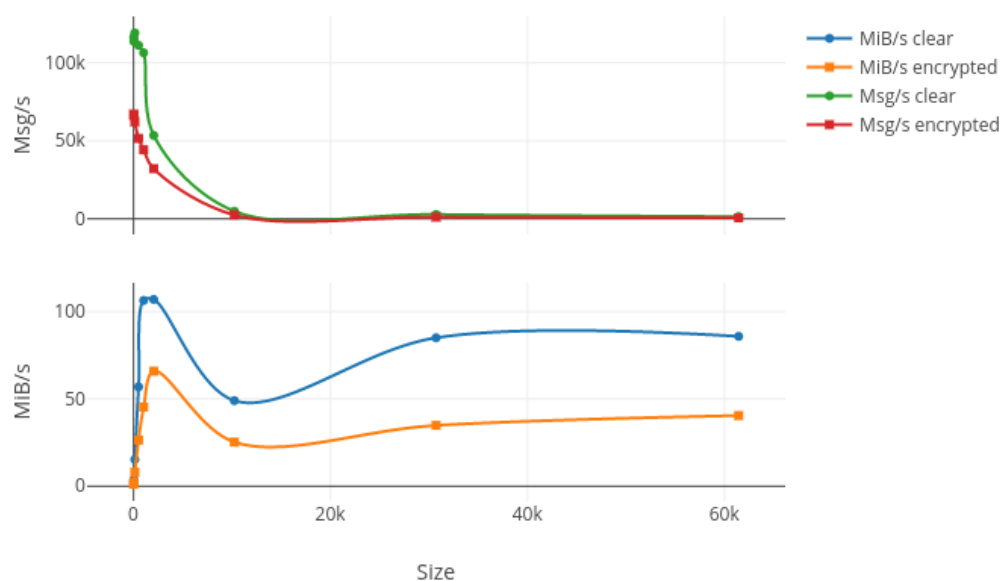


Figure 7. Bandwidth results in terms of Msg/s and MiB/s for 10,000 send on the laptop of an increasing message size in both encrypted and clear mode

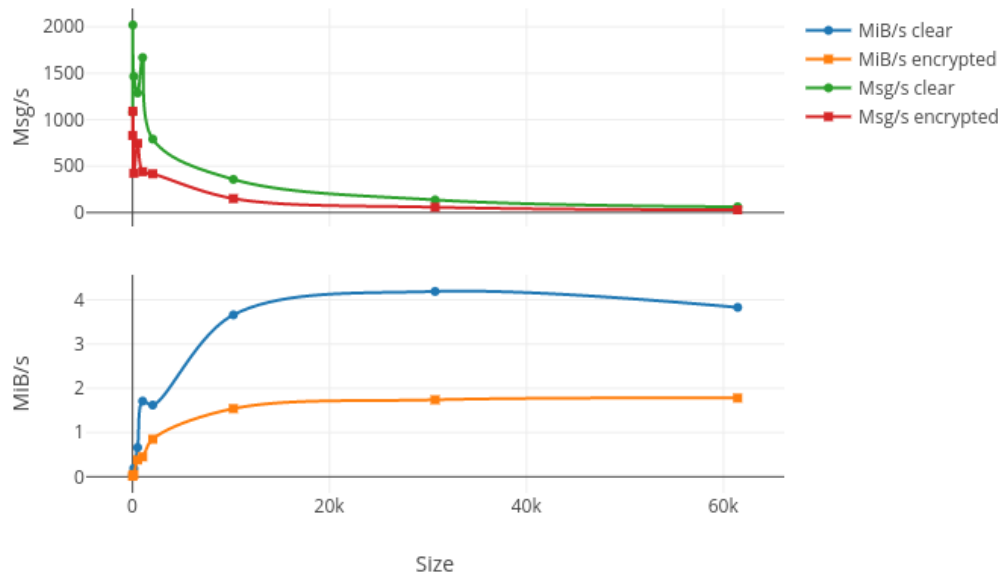


Figure 8. Bandwidth results in terms of Msg/s and MiB/s for 10,000 send on the Raspberry Pi 3 of an increasing message size in both encrypted and clear mode

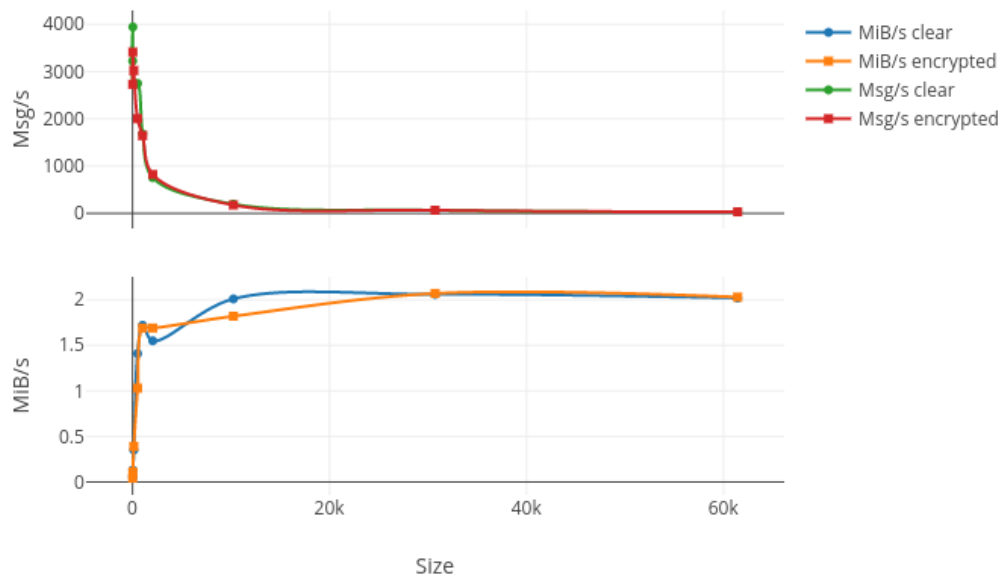


Figure 9. Bandwidth results in terms of Msg/s and MiB/s for 10,000 send on the Raspberry Pi Zero W of an increasing message size in both encrypted and clear mode

411 UDP is a protocol without any reliability mechanisms. It means that packets can be lost or
 412 corrupt and the protocol does not support any means to retrieve this packet unlike TCP. As our work
 413 is based on UDP and does not provide such mechanisms either, we wanted to show and quantify
 414 the risk of data loss. To do so, we create 10 000 messages of 1kiB composed by a two-bytes long
 415 sequence number and some random padding. While the wired interface on the laptop monitor the
 416 packet arrival (order, corruption using the embedded checksum or loss) the wireless connections
 417 send 10 000 packets in order. The results of this test are summed up in Table 5 where each line
 418 represents a different platform (laptop, raspberry Pi 3 and raspberry Pi zero W) and the three columns
 419 are respectively the number of lost packet, corrupted ones and finally the number of sequencing
 420 problems. As we can see, on the same network, we did not lose or receive corrupted packets but we
 421 had some problem in the sequence (2 for the laptop and Raspberry Pi 3 and 3 for the Raspberry Pi
 422 Zero W). That means that over 10 000 messages only 2 or 3 arrive before the previous one a result that
 423 tends to prove the relative reliability of our protocol even if we use UDP as our transport protocol.

Table 5. Number of lost, corrupted or misplaced packets over 10,000 send of a 1kiB message

Sender	Packets lost	Packets corrupted	Sequence problem
Laptop	0	0	2
Pi 3	0	0	2
Pi Zero W	0	0	3

424 The last test we did is a similarity test. Indeed, while we ensure the confidentiality of the data in
 425 encrypted mode, we did not randomly generated our Initialization Vector (IV) for the AES encryption.
 426 Instead, we use an algorithm (KDF described in algorithm 1) to be able to compute the IV based on
 427 information found in the encrypted packet and the secret key. And yet, the randomization of this IV
 428 guarantees the semantic security of AES. Consequently we wanted to know if our algorithm using
 429 random padding bytes is random enough to ensure a non-similarity between encrypted packets with
 430 the same secret key and containing the same data. To validate this point, we compute the Euclidian
 431 distance between 1000 secured packets containing the same 1kiB message of random data. The results
 432 of this test are summed up in table 6. In this last one, the first line represents the maximum distance
 433 we can obtain by having one packet with all bytes set to 0 and the other 255 this maximum distance
 434 will be used as a reference to compute the percentage of difference between packets. Next comes the
 435 mean distance computed on all the packets, with a value of 3458.79, together with a maximum of
 436 3775.54 and a minimum of 3175.23, we can say that our algorithm guarantee a relative non-similarity
 437 between those packets. Finally, we report the mean difference percentage of 42.39% in the last line
 438 of the results table. This means that for the exact same packet encrypted with our algorithm we
 439 generate encrypted packets that are on average 42.39% different. Consequently, we can say that our
 440 encryption process, even if it does not use a full random one, compute sufficiently different IV for the
 441 same packet in order to protect it from semantic and similarity attacks.

Table 6. Distance between 1000 packets containing the same message of 1kiB on the same topic and encrypted with the same encryption key

Variable	Value
Maximum theoretical distance	8160
Mean distance	3458.79
Maximum distance	3775.54
Minimum distance	3175.23
Mean difference percentage	42.39%

442 In conclusion, we can say that our tests were divided in three. The first one was to compute the
 443 speed capabilities of our protocol. In this case the results demonstrate that for data under 2kiB we

444 assure a very high speed with nearly a thousand messages per second on the lightest platform we
445 executed the test on. Another conclusion that we made thanks to this test is the relative low-impact
446 of the encryption even on light platform. The second test was realized to ensure the reliability of
447 UDP multicast on a single network. Indeed, this communication protocol does not provide reliability
448 insurance mechanisms and we wanted to put a number on the risks inherent in its usage. With 0
449 packet lost or corrupted and a maximum of 3 order problems over 10 000 messages sent we can safely
450 say that even if we use UDP as our base communication protocol, our method seems to be reliable
451 enough for data streaming inside a Smart Home. Finally, the last test we executed was to compute
452 the similarity between secured packets containing the same message and encrypted with the exact
453 same secret key. Indeed as our Initialization Vector is computed instead of randomly generated we
454 had to prove that the semantic security of AES is guaranteed. Our results show that our algorithm to
455 derive the AES key and IV from the message itself and the secret pre-shard key is random enough to
456 an average of 42.39% difference between 1000 encrypted packets containing the same data.

457 5. Conclusions and future works

458 In this paper, we introduced a new distributed way to communicate between smart entities
459 distributed in an environment. Unlike MQTT or RabbitMQ, well-known protocols, we don't need
460 to have a centralized broker instance and unlike ZMQ, a well-known framework to implement
461 messaging protocols, we don't rely on POSIX sockets which are hard to embed on tiny devices
462 without a Linux or Windows operating system. Our protocol, like FTP, relies on two channels. The
463 first one is for the configuration of every entity in the network and is based on COAP, a well-known
464 protocol in the field of the Internet of Thing, for its reliability. The other one is a data channel, based
465 on multicast messages with a protocol entirely define in this paper. This last one permits to send
466 discovery requests to the network as well as messages encrypted or not. For the encrypted way, we
467 adapt the Telegram protocol a well-known secure instant messaging protocol in order to work on tiny
468 devices and provide an easy authentication with a HMAC.

469 In this paper, we realized three different tests to ensure our capabilities in terms of speed,
470 reliability and security. In the light of the results, we can say that for data under 2kiB we ensure
471 a very high speed with nearly a thousand messages per second on the lightest platform we executed
472 the test on. Moreover, with no packet lost or corrupted and a maximum of 3 order problems over
473 10 000 messages sent we can safely say that our method seems reliable enough for data streaming.
474 Finally, our results show that our algorithm to derive the AES key and IV from the message itself and
475 the secret pre-shard key is random enough to generate an average of 42.39% difference between 1000
476 encrypted packets containing the same data ensuring the security against semantic attacks.

477 Ultimately, we can say that our protocol with its two channels allows to make a difference
478 between configuration values which need a high reliability but won't be changing every millisecond
479 and data values or streaming which need higher data rates but less reliability. Moreover, our protocol
480 support a native encryption mode which provides security for sensitive information we can easily
481 find in a Smart Home. In addition, we designed our innovation to be both brokerless, which is
482 the main difference between many existing messaging protocol, and easily portable as it only relies
483 on UDP, a communication protocol found in every network applications. Lastly, as a future work,
484 we want to realize more tests including tests on embedded environment, like a Real Time OS on a
485 microcontroller and tests with a lot more entities in the network.

486 Abbreviations

487 The following abbreviations are used in this manuscript:

488 AES: Advanced Encryption Standard

489 AI : Artificial Intelligence

490 Amb. I: Ambient Intelligence

491 AMQP: Advanced Message Queuing Protocol
492 BLE: Bluetooth Low-Energy
493 CoAP : Constrained Application Protocol
494 CRC: Cyclic Redundancy Check
495 EEPROM : Electrically Erasable Programmable Read-Only Memory
496 GHz: gigahertz
497 HMAC : keyed-Hash Message Authentication Code
498 HTTP: Hypertext Transfer Protocol
499 IEEE: Institute of Electrical and Electronics Engineers
500 IoT: Internet of Things
501 IV: Initialization Vector
502 KDF : Key Derivation Function
503 LNCF: Lght Node Communication Framework
504 LIARA: Laboratoire d'Intelligence Ambiante pour la Reconnaissance d'Activités
505 LSB : Least Significant Bit
506 LWT: Last Will Testament
507 kB: kilobyte
508 kiB: kibibyte
509 kiB/s: kibibyte/second
510 MHz: megahertz
511 MiB: mebibyte
512 MiB/s: mebibyte/second
513 MSB : Most Significant Bit
514 Msg/s: messages/seconds
515 OSGI: Open Service Gateway Initiative
516 RAM: Random Access Memory
517 RTOS : Real Time Operating System
518 SDSH: Software Defined Smart Home
519 SHA: Secure Hash Algorithms
520 SoC: System on Chip
521 SPoF: Single Point of Failure
522 UDP: User Datagram Protocol
523 XMPP : eXtensible Messaging and Presence Protocol
524 ZMQ : ZeroMQ

525

526 **Author Contributions:** All the authors contributed equally to this work.

527 **Conflicts of Interest:** The authors declare no conflict of interest.

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