

Ana I. Pereira · Florbela P. Fernandes ·  
João P. Coelho · João P. Teixeira ·  
Maria F. Pacheco · Paulo Alves ·  
Rui P. Lopes (Eds.)

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# Optimization, Learning Algorithms and Applications

First International Conference, OL2A 2021  
Bragança, Portugal, July 19–21, 2021  
Revised Selected Papers

 Springer



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### *Editors*

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Maria F. Pacheco   
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Florbela P. Fernandes   
Instituto Politécnico de Bragança  
Bragança, Portugal

João P. Teixeira   
Instituto Politécnico de Bragança  
Bragança, Portugal

Paulo Alves   
Instituto Politécnico de Bragança  
Bragança, Portugal

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# Preface

The volume CCIS 1488 contains the refereed proceedings of the International Conference on Optimization, Learning Algorithms and Applications (OL2A 2021), an event that, due to the COVID-19 pandemic, was held online.

OL2A 2021 provided a space for the research community on optimization and learning to get together and share the latest developments, trends, and techniques as well as develop new paths and collaborations. OL2A 2021 had more than 400 participants in an online environment throughout the three days of the conference (July 19–21, 2021), discussing topics associated to areas such as optimization and learning and state-of-the-art applications related to multi-objective optimization, optimization for machine learning, robotics, health informatics, data analysis, optimization and learning under uncertainty, and the Fourth Industrial Revolution.

Four special sessions were organized under the following topics: Trends in Engineering Education, Optimization in Control Systems Design, Data Visualization and Virtual Reality, and Measurements with the Internet of Things. The event had 52 accepted papers, among which 39 were full papers. All papers were carefully reviewed and selected from 134 submissions. All the reviews were carefully carried out by a Scientific Committee of 61 PhD researchers from 18 countries.

July 2021

Ana I. Pereira

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# Contents

## Optimization Theory

Dynamic Response Surface Method Combined with Genetic Algorithm to Optimize Extraction Process Problem .....	3
<i>Laires A. Lima, Ana I. Pereira, Clara B. Vaz, Olga Ferreira, Márcio Carochó, and Lillian Barros</i>	
Towards a High-Performance Implementation of the MCSFilter Optimization Algorithm .....	15
<i>Leonardo Araújo, Maria F. Pacheco, José Rufino, and Florbela P. Fernandes</i>	
On the Performance of the ORTHOMADS Algorithm on Continuous and Mixed-Integer Optimization Problems .....	31
<i>Marie-Ange Dahito, Laurent Genest, Alessandro Maddaloni, and José Neto</i>	
A Look-Ahead Based Meta-heuristics for Optimizing Continuous Optimization Problems .....	48
<i>Thomas Nordli and Nouredine Bouhmala</i>	
Inverse Optimization for Warehouse Management .....	56
<i>Hannu Rummukainen</i>	
Model-Agnostic Multi-objective Approach for the Evolutionary Discovery of Mathematical Models .....	72
<i>Alexander Hvatov, Mikhail Maslyayev, Iana S. Polonskaya, Mikhail Sarafanov, Mark Merezchnikov, and Nikolay O. Nikitin</i>	
A Simple Clustering Algorithm Based on Weighted Expected Distances .....	86
<i>Ana Maria A. C. Rocha, M. Fernanda P. Costa, and Edite M. G. P. Fernandes</i>	
Optimization of Wind Turbines Placement in Offshore Wind Farms: Wake Effects Concerns .....	102
<i>José Baptista, Filipe Lima, and Adelaide Cerveira</i>	
A Simulation Tool for Optimizing a 3D Spray Painting System .....	110
<i>João Casanova, José Lima, and Paulo Costa</i>	



Optimization of Glottal Onset Peak Detection Algorithm for Accurate Jitter Measurement .....	123
<i>Joana Fernandes, Pedro Henrique Borghi, Diamantino Silva Freitas, and João Paulo Teixeira</i>	
Searching the Optimal Parameters of a 3D Scanner Through Particle Swarm Optimization .....	138
<i>João Braun, José Lima, Ana I. Pereira, Cláudia Rocha, and Paulo Costa</i>	
Optimal Sizing of a Hybrid Energy System Based on Renewable Energy Using Evolutionary Optimization Algorithms .....	153
<i>Yahia Amoura, Ângela P. Ferreira, José Lima, and Ana I. Pereira</i>	
<b>Robotics</b>	
Human Detector Smart Sensor for Autonomous Disinfection Mobile Robot ....	171
<i>Hugo Mendonça, José Lima, Paulo Costa, António Paulo Moreira, and Filipe Santos</i>	
Multiple Mobile Robots Scheduling Based on Simulated Annealing Algorithm .....	187
<i>Diogo Matos, Pedro Costa, José Lima, and António Valente</i>	
Multi AGV Industrial Supervisory System .....	203
<i>Ana Cruz, Diogo Matos, José Lima, Paulo Costa, and Pedro Costa</i>	
Dual Coulomb Counting Extended Kalman Filter for Battery SOC Determination .....	219
<i>Arezki A. Chellal, José Lima, José Gonçalves, and Hicham Megnafi</i>	
Sensor Fusion for Mobile Robot Localization Using Extended Kalman Filter, UWB ToF and ArUco Markers .....	235
<i>Sílvia Faria, José Lima, and Paulo Costa</i>	
Deep Reinforcement Learning Applied to a Robotic Pick-and-Place Application .....	251
<i>Natanael Magno Gomes, Felipe N. Martins, José Lima, and Heinrich Wörtche</i>	
<b>Measurements with the Internet of Things</b>	
An IoT Approach for Animals Tracking .....	269
<i>Matheus Zorawski, Thadeu Brito, José Castro, João Paulo Castro, Marina Castro, and José Lima</i>	

Optimizing Data Transmission in a Wireless Sensor Network Based on LoRaWAN Protocol .....	281
<i>Thadeu Brito, Matheus Zorawski, João Mendes, Beatriz Flávia Azevedo, Ana I. Pereira, José Lima, and Paulo Costa</i>	
Indoor Location Estimation Based on Diffused Beacon Network .....	294
<i>André Mendes and Miguel Diaz-Cacho</i>	
SMA Covid-19 – Autonomous Monitoring System for Covid-19 .....	309
<i>Rui Fernandes and José Barbosa</i>	
<b>Optimization in Control Systems Design</b>	
Economic Burden of Personal Protective Strategies for Dengue Disease: an Optimal Control Approach .....	319
<i>Artur M. C. Brito da Cruz and Helena Sofia Rodrigues</i>	
ERP Business Speed – A Measuring Framework .....	336
<i>Zornitsa Yordanova</i>	
BELBIC Based Step-Down Controller Design Using PSO .....	345
<i>João Paulo Coelho, Manuel Braz-César, and José Gonçalves</i>	
Robotic Welding Optimization Using A* Parallel Path Planning .....	357
<i>Tiago Couto, Pedro Costa, Pedro Malaca, Daniel Marques, and Pedro Tavares</i>	
<b>Deep Learning</b>	
Leaf-Based Species Recognition Using Convolutional Neural Networks .....	367
<i>William Oliveira Pires, Ricardo Corso Fernandes Jr., Pedro Luiz de Paula Filho, Arnaldo Candido Junior, and João Paulo Teixeira</i>	
Deep Learning Recognition of a Large Number of Pollen Grain Types .....	381
<i>Fernando C. Monteiro, Cristina M. Pinto, and José Rufino</i>	
Predicting Canine Hip Dysplasia in X-Ray Images Using Deep Learning .....	393
<i>Daniel Adorno Gomes, Maria Sofia Alves-Pimenta, Mário Ginja, and Vitor Filipe</i>	
Convergence of the Reinforcement Learning Mechanism Applied to the Channel Detection Sequence Problem .....	401
<i>André Mendes</i>	



# Optimizing Data Transmission in a Wireless Sensor Network Based on LoRaWAN Protocol

Thadeu Brito<sup>1,2,3</sup> , Matheus Zorawski<sup>1</sup> , João Mendes<sup>1</sup> ,  
Beatriz Flávia Azevedo<sup>1,4</sup> , Ana I. Pereira<sup>1,4</sup> , José Lima<sup>1,3</sup> ,  
and Paulo Costa<sup>2,3</sup>

<sup>1</sup> Research Centre in Digitalization and Intelligent Robotics (CeDRI),  
Instituto Politécnico de Bragança, Campus de Santa Apolónia,  
5300-253 Bragança, Portugal

{brito,matheuszorawski,joao.cmendes,beatrizflavia,apereira,jllima}@ipb.pt

<sup>2</sup> Faculty of Engineering of University of Porto, Porto, Portugal

<sup>3</sup> INESC TEC - INESC Technology and Science, Porto, Portugal

<sup>4</sup> Algoritmi Research Centre, University of Minho, Campus de Gualtar,  
Braga, Portugal

**Abstract.** Internet of Things, IoT, is a promising methodology that has been increasing over the last years. It can be used to allow the connection and exchange data with other devices and systems over the Internet. One of the IoT connection protocols is the LoRaWAN, which has several advantages but has a low bandwidth and limited data transfer. There is a necessity of optimising the data transfer between devices. Some sensors have a 10 or 12 bits resolution, while LoRaWAN owns 8 bits or multiples slots of transmission remaining unused bits. This paper addresses a communication optimisation for wireless sensors resorting to encoding and decoding procedures. This approach is applied and validated on the real scenario of a wildfire detection system.

**Keywords:** Internet of Things · LoRaWAN · Wireless sensor network · Fire detection · Transmission optimisation

## 1 Introduction

The project “SAFe: Forest Monitoring and Alert System” proposes an intelligent system for monitoring situations of potential forest risk. This system combines sensor nodes that will collect various parameters, namely temperature, humidity and data related to infrared sensors that identify the presence of flame. This collection of information, combined with a system based on artificial intelligence and other collected data (such as weather forecasting), will allow an efficient and intelligent data analysis, allowing the creation of alerts of dangerous situations to alert the different actors (for example, firefighters, civil protection or city

council). These alerts will be parameterized and presented in a personalised way and tailored to each actor. Thus, it is intended to minimise the occurrence of ignitions, monitor fauna and flora, and contribute to the environmental development of the region of Trás-os-Montes, particularly in the district of Bragança. In addition to the capacity for early detection of forest fires, the accurate estimation of the fire hazard, using fire risk indices in real-time, on a sub-daily and local scale, in meteorological data, in the availability of fuels and moisture content vegetation is of utmost importance.

The SAFe's implementation is localised to the Serra da Nogueira area, belonging to the Natura 2000 Network, with unique characteristics in the Bragança region and with an extension of approximately 13 km. Considering the region's size to be monitored, data transmission is carried out by LoRaWAN, it is based on low power wide area network, which will guarantee full coverage of the area in question [1–4]. This type of communication is increasingly used, and it is estimated that by 2024, the IoT industry will generate revenue of 4.3 trillion dollars [5]. LoRaWAN networks are composed of end devices, gateway and network server in the midst of a rising state. The data is acquired from the modules (end devices) that send the data to the gateway, which in turn relays the messages to the servers through non-LoRaWAN networks, such as Ethernet or IP over cellular [6].

The LoRaWAN communication itself uses the Chirp Spread Spectrum (CSS) modulation, where chirp pulses modulate the signal. It uses an unlicensed sub-gigahertz ISM band that varies in width depending on the region of application, and for Europe, it is 863 MHz to 870 MHz [7]. Within the communication settings, three parameters allow a network adjustment between bit rate and robustness of the transmission: Bandwidth (BW), Spreading Factor (SF), and Coding Rate (CR) [8]. The format of the Long Range (LoRa) message is subdivided into several layers, the physical layer, which is composed of five fields, the preamble, the physical header (PHDR), the physical header Cyclic Redundancy check (PHDR0\_CRC), the physical payload (PHY Payload) and an error detection tail (CRC). The Medium Access Control (MAC) layer, in turn, is a subdivision of the PHY Payload field and consists of three fields, the MAC Header (MHDR), MAC Payload and the Message Integrity Code (MIC). Within the MAC Payload is the message itself, which consists of three fields, the Frame header (FHDR), the port and the frame payload (FRM Payload) [9].

At this stage of the SAFe project, while the testing period is still underway in an urban environment, the configuration being used is a 125 kHz Bandwidth, an SF7 and a Coding Rate of 4/5, which allows transmission of about 242 Bytes. However, taking into account that when the project passes the test phase in a rural environment, it will be necessary to cover a larger area. It is probably required to use a larger SF to improve the communication range, thus reducing transmission capacity. Using an SF12 and maintaining the BW and CR, the LoRaWAN communication will allow transmission of about 51 Bytes. Therefore, there is a necessity to optimise the use of all bits' packages as much as possible and not lose data essential for the project's realisation. But some sensors have

a 10 or 12 bits resolution while LoRaWAN owns 8 bits, which means, in some cases, multiples slots of transmission remaining unused bits. In this way, this paper addresses a communication optimisation for Wireless Sensors Network (WSN) resorting to encoding and decoding procedures.

The rest of paper is organized as follows. After the introduction, Sect. 2 presents the related work. In Sect. 3, the system architecture is addressed where the focus of this paper is highlighted. Section 4 presents the algorithm used to optimise the data sent though the network and Sect. 5 stresses the results of the proposed encoding/decoding. Last section, concludes the paper and points future work.

## 2 Related Work

Real-time data collection and analysis had new possibilities with the emergence of the Internet of Things (IoT). Typically, IoT devices are manufactured to use in short-range protocols, such as Bluetooth, WiFi, and ZigBee, among others, which may have limitations, such as the maximum number of devices connected to the gateway and high power consumption. On the other hand, mobile technologies that have an extensive range (LTE, 3G and 4G) are expensive, consume a lot of power and have a continuing cost [10]. The gap that these technologies have been filled in part by Low Power Wide Area Network (LPWAN) technologies, which cannot cover the critical applications (e.g. remote health monitoring), but fit very well for the requirements of IoT applications. Although low data rates limit LPWAN systems, there are reasons for their use, especially after the emergence of technologies such as SigFox, NB-IoT, and LoRa [10, 11].

LoRa network is highly used in applications requiring low power consumption and long-range, in which LoRa Alliance proposed the LoRaWAN, defining the network protocol in the MAC and network layers [10, 12]. Different scenarios of typical LoRa networks applications are pointed out in [12], besides providing documentation of LoRa networks design and implementation, searching for a better cost-benefit for LPWAN applications. Also, Baiji [13] showed in 2019 an application using LoRaWAN in an IoT monitoring system aimed to minimise the non-productive time for Oil and Gas companies through the detection of a leak.

Researches are being carried out seeking improvements to the LoRaWAN system, in which numerous of these are being carried out to seek optimisation of data rate, airtime, and energy consumption through an Adaptive Data Rate (ADR), as in [14], where a review of research on ADR algorithms for LoRaWAN technology is presented. Also, in [15] a method for the use of ADR is implemented, seeking to provides an optimal throughput with multiple devices connected to LoRaWAN. Furthermore, a study was carried out looking for maximising the battery lifetime through different LoRaWAN nodes and scenarios in [16], demonstrating relationships between different payload sizes with the power

consumption and air time using various SF and CR, indicating an increase in power consumed per bit of payload used.

As cited by [11], all the advantages of the LoRa network of achieving a long transmission range, low power and enabling low-cost IoT applications to exist with the downside of low data throughput rate. The LoRaWAN data transfer must respect a protocol, and it is possible to send the Payload in several forms, as long as it is within the limit of the number of Bytes. Considering the article [9], the LoRa message format is composed of several fields that have specific or limited sizes. Therefore, it is necessary to optimise the use of bytes in Payload as much as possible. Some techniques have been used over the years, such as selecting the best packet size by adjusting the bit rate to achieve maximum throughput as demonstrated by the authors of the article [17], to optimise the payload sizes in voice and video applications. Still, within this dynamic, the authors of the article [18] used throughput as the optimisation metric and based their study on wireless ATM networks to maximise the efficiency of transfer by adapting the packet size with the variation of the channel conditions. Also, Eytan Modiano in [19] presented a system to dynamically optimise the packet size based on the estimates of the channel's bit error rate, being particularly useful for wireless and satellite channels.

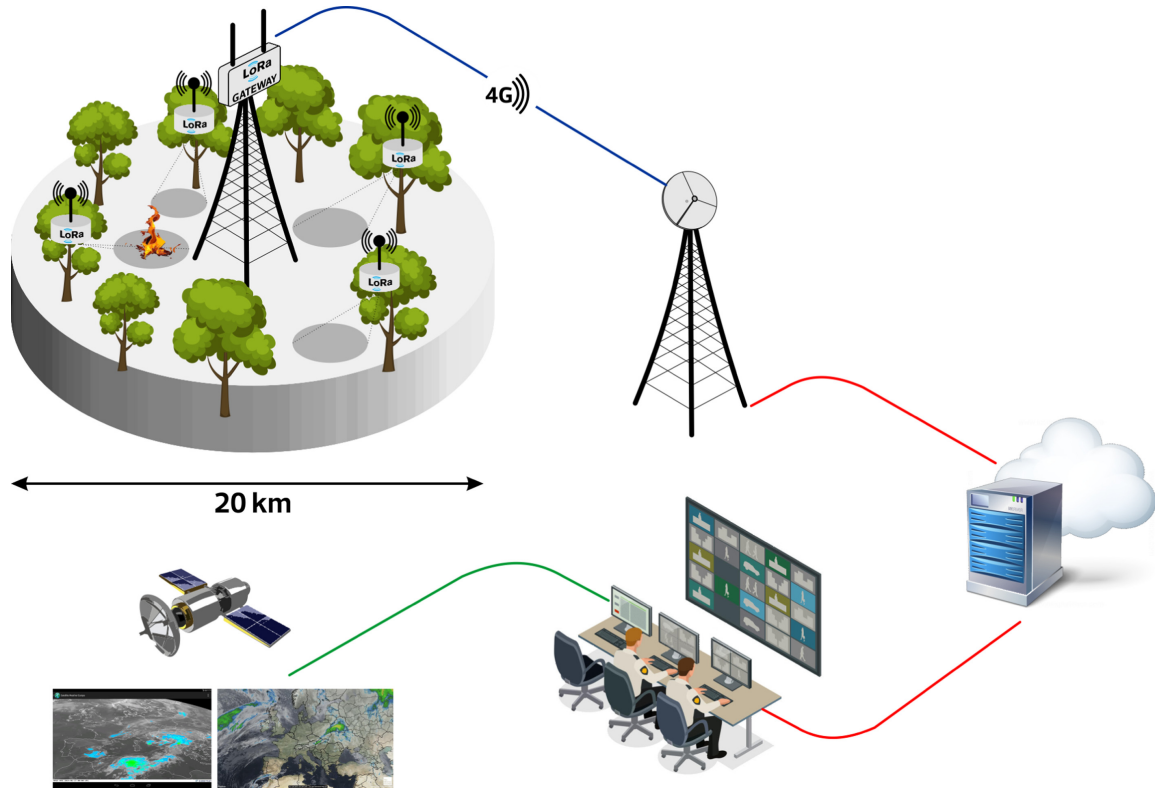
The importance of packet size optimisation in WSN are debated in the article [20], presenting several packet size optimisation techniques and concluding that there is no agreement on whether the packet should be fixed or dynamic. The packet size optimisation technique to be used for WSN is defended in article [21], proposing an energy efficiency factor evaluation metric, being a fixed packet size for the case studied. Furthermore, the authors of the article [22] present a variation of the packet size depending on the network conditions to increase the network throughput and efficiency. The method of payload recommended by The Things Network (TTN) when using LoRaWAN has described in [23] only the specifications of how to send each type of data. Suppose the user needs to use the minimum Bytes possible for a shorter transmission time and being within the limit of 51 Bytes. One way to implement data encoding for the LPWAN network as LoRaWAN is through Cayenne Low Power Payload (LPP), which allows sending the payload dynamically, according to the LoRa frequency and data rate used [24]. Nevertheless, besides the various optimisations that have been mentioned above, it remains crucial to find the form to optimise the data that will be transferred for better use of the LoRa network.

### 3 System Architecture

The presented system has the main ideology of monitoring a determined area with a reduced application cost. In this way, it is expected to obtain constant surveillance to identify any ignition that may exist. This surveillance will be achieved using the various sensors present in the modules (end devices). Each module consists of six sensors, with three distinct values types: five flame sensors and a sensor that provides data on relative humidity and air temperature. The



entire operation of the safe project combines the collection of data from the modules present in the forest; this data collection mustn't fail. Consequently, the proper functioning of data transmission is guaranteed because if the data does not reach the competent entities, they will be of no use. This transmission will be handled using LoRaWAN technology, which allows us to connect several modules to different gateways. Thus, guaranteeing low-cost and efficient coverage of the area. Figure 1 exemplifies the whole system architecture of the SAFE project.

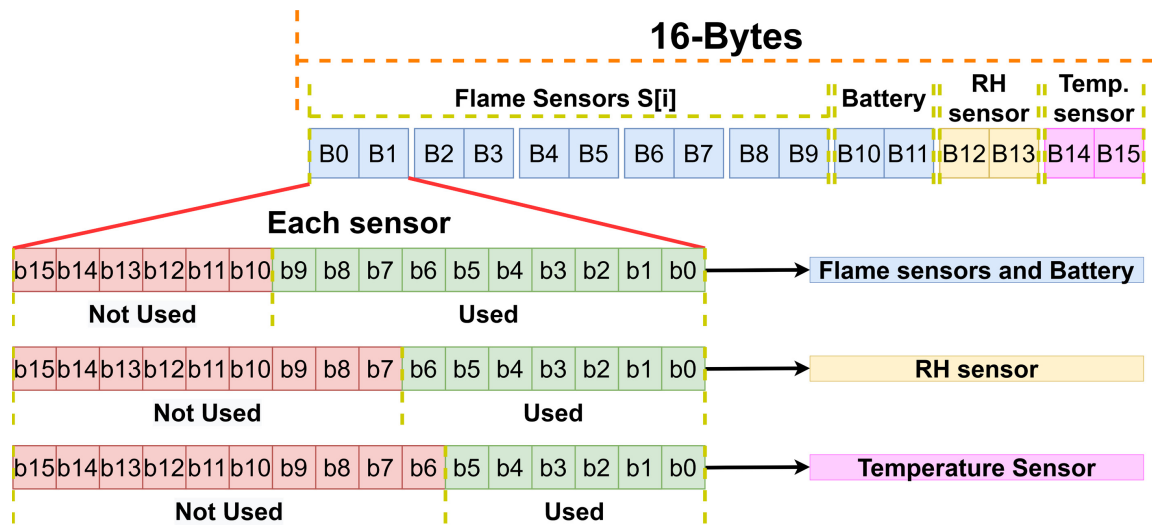


**Fig. 1.** System architecture of SAFE project [2].

Due to the specificity that each of these elements is defined and developed (Fig. 1), this work is concentrated only on the LoRaWAN's Transmission and the proposed arrangement bits protocol. Other SAFE's approaches can be seen in [1–4]. In this way, considering the high number of modules (end devices) used in this project, there is a requirement to ensure no collisions between their messages; the whole modules' description is explained in [2,4]. Thereupon, it is necessary to optimise the size of the data package sent, thus guaranteeing the minimum necessary occupation of bytes to reduce the chances of these collisions and consequent data losses. The normal sending process (recommended by TTN service) will be exemplified below to explain the process, for this information will be used related to the six sensors present in the modules as well as the battery information, resulting in:

- **Flame sensors:** as mentioned before, each node has five flame sensors. These sensors make 10 bits reading, which means they will generate values between 0 to 1023. To send just one value from one of them is demanded to use 2-Bytes (16 bits), consequently, to send all of them is required a total of 10-Bytes.
- **Relative humidity sensor:** this sensor produces a reading between 0 to 99, but 2-Bytes are required to transmit the message.
- **Temperature sensor:** gives a reading between 0 to 50 °C, which results in the use of 2-Bytes to transmit.
- **Battery level:** the battery level is converted from a 10 bits ADC and a total of 2-Bytes will be required to transmit.

Analysing the sending process recommended by TTN service, it is possible to observe that the total package has 16-Bytes per interval of sending data (Fig. 2). However, the messages resulting from all sensors have unused bits because 16-Bytes has 128 bits and the data produced by all sensors in each module has a total of 73 bits. Thus, this work aims to optimise the use of these unfilled bits. Using the example of flame sensors, it is possible to notice that the 2-Bytes sent correspond to a message with 16-bits, and only 10-bits are used to transmit the value with a range of 0 to 1023. In this case, 6-bits are left to optimise with the data from another sensor's value. This example can be seen looking for  $B0$  and  $B1$  in the following Fig. 2.



**Fig. 2.** 16-Bytes (B) of data necessary to send. The unused bits (b) are red coloured. Where  $i \in \mathbf{R} | 0 \leq i \leq 4$ .

This methodology of taking advantage of the unused bits will be explained in the following section in more detail.



## 4 Algorithm

From the previous section, it is possible to notice that for each flame sensor data ( $S[i]$ ), there are some unused bits ( $x$ ) available to receive data to be carried. In this way, the humidity sensor ( $RH$ ) and the temperature sensor ( $T$ ) will be stored on a Byte type and the battery ( $B$ ) level will be stored in 10 bits resolution. Table 1 presents the transmission package without optimise the encoding procedure (using the recommendation in [23]).

**Table 1.** Transmission without optimise encoding.  $S[i]$  are the flame sensors, H, T and B are the Humidity, Temperature and Battery Voltage sensors, respectively. The ‘x’ are the unused bits.

<b>S[0]</b>									
x	x	x	x	x	x	s9[0]	s8[0]	...	s0[0]
<b>RH</b>									
x	x	x	x	x	x	x	RH7	...	RH0
<b>S[1]</b>									
x	x	x	x	x	x	s9[1]	s8[1]	...	s0[1]
<b>T</b>									
x	x	x	x	x	x	x	T7	...	T0
<b>S[2]</b>									
x	x	x	x	x	x	s9[2]	s8[2]	...	s0[2]
<b>B</b>									
x	x	x	x	x	x	B9	B8	...	B0
<b>S[3]</b>									
x	x	x	x	x	x	s9[3]	s8[3]	...	s0[3]
<b>S[4]</b>									
x	x	x	x	x	x	s9[4]	s8[4]	...	s0[4]
<b>Legend</b>									
	Used by 10 bits sensors								
	Used by Relative Humidity								
	Used by Temperature								

This approach intends to sort the bits using a less value as possible in bytes protocol used to transmit the data by LoRaWAN. The main idea is distribute each bit from each sensor during the encoding process, in a manner that is necessary use the minimum Bytes. In order to optimise these *Not Used* bits (b15, b14, b13, b12, b11 and b10, see Fig. 2) for each sensor  $S[i]$  (where  $i \in \mathbf{R} | 0 \leq i \leq 4$ ), the word (2-Bytes) of Humidity (RH), Temperature (T) and Battery (B) will be separated and combined with the unused bits at the encoding procedure. A bit manipulation operation is required to use the previous unused bits. Table 2 presents the previous “ $x$ ” assuming bits from other sensors. Moreover, even including the H, T and B values, there are 4 bits that remains unused. They will be appropriated to hold four auxiliary bits (AUX) for future applications.

**Table 2.** Proposed encoding map.

<b>S[0]</b>										<b>RH</b>									
T5	T4	T3	T2	T1	T0	s9[0]	s8[0]	...	s0[0]	x	x	x	x	x	x	x	x	...	x
<b>S[1]</b>										<b>T</b>									
RH3	RH2	RH1	RH0	T7	T6	s9[1]	s8[1]	...	s0[1]	x	x	x	x	x	x	x	x	...	x
<b>S[2]</b>										<b>B</b>									
B1	B0	RH7	RH6	RH5	RH4	s9[2]	s8[2]	...	s0[2]	x	x	x	x	x	x	x	x	...	x
<b>S[3]</b>										<b>Legend</b>									
B7	B6	B5	B4	B3	B2	s9[3]	s8[3]	...	s0[3]		Used by 10 bits sensors								
<b>S[4]</b>											Used by Relative Humidity								
AUX4	AUX2	AUX1	AUX0	B9	B8	s9[4]	s8[4]	...	s0[4]		Used by Temperature								
											Used by AUX								

This encoding operation is carried by an embedded system that is programmed on C language. For the encoding, the proposed C code is detailed on Listing 1.1. The message will go through the network and will be decoded by Listing 1.2.

**Listing 1.1.** Data encoder

```

S[0] = S[0] + (T & 0X3F) << 10;
S[1] = S[1] + (T & 0XC0) << 10 + (H & 0X0F) << 12;
S[2] = S[2] + (H & 0XF0) << 10 + (B & 0X003) << 14;
S[3] = S[3] + (B & 0X0FC) << 10;
S[4] = S[4] + (B & 0X300) << 10 + (AUX & 0X0F) << 12;
//TRANSMIT array S

```

An opposite operation of decoding will allow to extract the Flame values, Humidity, Temperature and Battery voltage from the message. The C code of decoding procedure can be found on Listing 1.2.

**Listing 1.2.** Data decoder

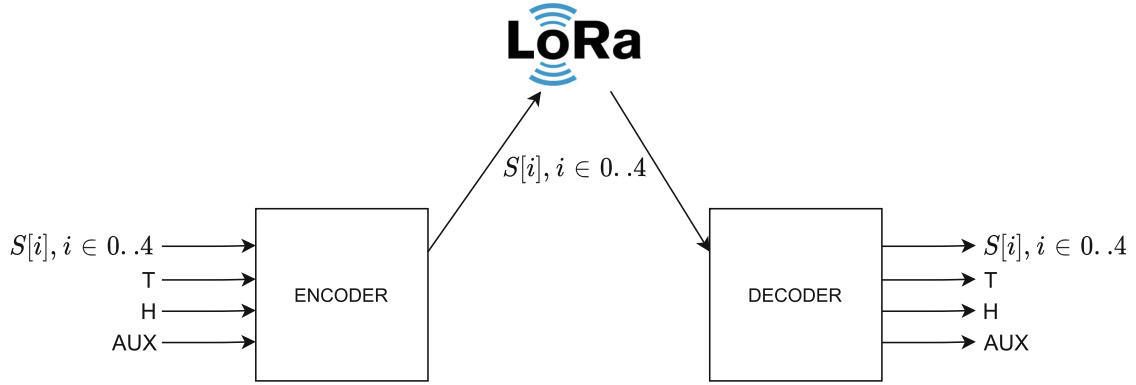
```

//RECEIVE array S
T = (S[0] & 0XFC00) >> 10 + (S[1] & 0X0C0) >> 4;
H = (S[1] & 0XF000) >> 12 + (S[2] & 0X3C00) >> 6;
B = (S[2] & 0XC000) >> 14 + (S[3] & 0XFC00) >> 10
  + (S[4] & 0X0C00) >> 2;
AUX = (S[4] & 0XF000) >> 12;

S[0] = S[0] & 0X03FF;
S[1] = S[1] & 0X03FF;
S[2] = S[2] & 0X03FF;
S[3] = S[3] & 0X03FF;
S[4] = S[4] & 0X03FF;

```

This optimised message can be sent to the LoRaWAN network and at the destination, a decoder procedure will construct the original message. Figure 3 presents the encoding, transmission and decoding procedures. As it can be seen, the packet to be sent keeps the size of produced by the flame sensors  $S[i]$  (where  $i \in \mathbf{R} | 0 \leq i \leq 4$ ):



**Fig. 3.** Encoding and decoding procedures. Size of  $S[i]$  is 16 bits whereas T and H are 8 bits. AUX is an auxiliary variable that was created with the 4 bits that remain free.

By this way, a reduction of transmission from 8 words (16-Bytes) to 5 words (10-Bytes) is obtained. The following section demonstrates the difference between the procedure of recommendation by TTN and this approach.

## 5 Results

To demonstrate the difference during transmission using the LoRaWAN network with the algorithm proposed in this work, a comparison test is performed with the method recommended by TTN. Therefore, five sensor modules are configured for each transmission method: five nodes with the TTN transmission technique and another five nodes with the algorithm shown in the previous section. Figure 4a shows the ten modules distributed on the laboratory bench. They are configured according to the mentioned algorithms and simulate a small WSN. All of these modules are configured to transmit at a specific frequency within the range that LoRaWAN works. In this way, it is possible to probe the behaviour of the duty cycle in the settings of the gateway used. The gateway used is RAK 7249, which is attached to the laboratory's roof shown in Fig. 4b.



(a) Five modules configured for each algorithm.



(b) LoRaWAN gateway used for the test.

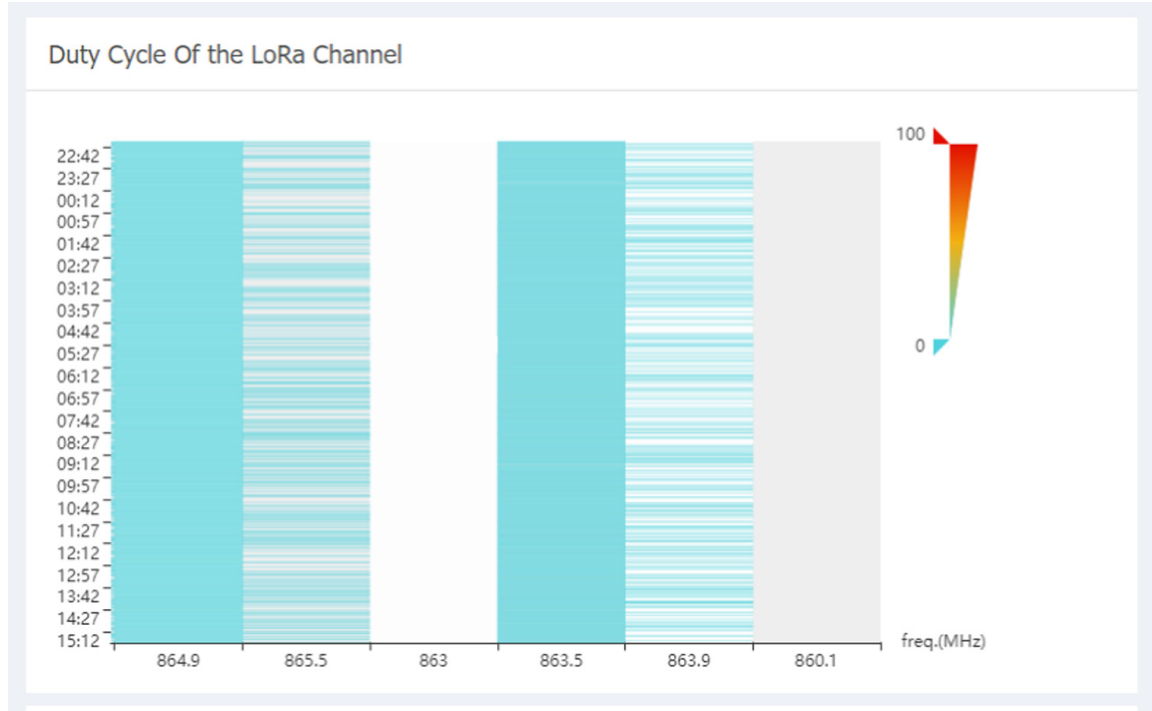
**Fig. 4.** Structure used for the comparison between the two algorithms.

The five modules responsible for transmitting the data from the sensors with the TTN's recommendations were configured to send using the frequency 864.9 MHz, 865.5 MHz and 863 MHz (this frequency is exclusive for FSK). The other five modules, those configured with the algorithm proposed in this work, carried out transmissions on 863.5 MHz, 863.9 MHz and 860.1 MHz (this frequency is exclusive for FSK use). To ensure that no other frequencies were used, the RAK 7249 was still configured to only work on these six frequencies mentioned. In addition, it was also chosen to enable a modulation concentrator for each frequency range used.

All ten modules have the same sending values; that is, all of them must send the five values of the flame sensors, the battery level, humidity and relative air temperature. They have also been configured to communicate in an interval of 60 s. Therefore, after all the necessary configurations for the test, the modules were deposited under the laboratory bench. Where they are under the same climatic circumstances, it is also possible to guarantee that they will send corresponding temperature and humidity values. Also, note that all batteries are fully charged.

After 24 h, the generated duty cycle graph was verified in the gateway system, and a screenshot is shown in Fig. 5. Through this graph, it is possible to notice the difference between the two algorithms based on the lines generated when the gateway performs transmissions at a specific frequency. In this sense, it is observed that the primary frequencies are the most used (864.9 MHz and

863.5 MHz), as the firmware will always choose them as the first attempt to send. However, when analysing the secondary frequencies (865.5 MHz and 863.9 MHz), a difference is noted between the amount of duty cycle occupation. This graphical difference demonstrates the optimisation of bytes during data transmission through the LoRaWAN network. Thus, using the algorithm proposed in this work, it is possible to insert more modules under the same facilities.



**Fig. 5.** A screenshot is obtained from the RAK 7249 gateway system after 24 h of use.

## 6 Conclusion and Future Works

The limited resources of the LoRaWAN requires a reduction of the transmission data. In this paper, an optimised transmission is proposed availing the unused bits of the 16 bits values. It is a lossless bit manipulation that will fit the temperature, humidity and battery level data bits in the unused bits of the sensors data where each sensor occupies 10 bits. The remaining 6 bits are used to perform this procedure. Also, 4 free bits will allow to use an auxiliary value from 0 to 0xF that can help on synchronisation and packet sequence verification. The results showed that the reduction of the duty cycle of the LoRaWAN while maintaining the data integrity allow to verify that the proposed methodology is able to be installed on the wireless sensor network, as future work.

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